

Windows and Classrooms: A Study of Student Performance and the Indoor Environment



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PREFACE

The Public Interest Energy Research (PIER) Program supports public interest energy research and development that will help improve the quality of life in California by bringing environmentally safe, affordable, and reliable energy services and products to the marketplace.

This document is one of 33 technical attachments to the final report of a larger research effort called *Integrated Energy Systems: Productivity and Building Science Program* (Program) as part of the PIER Program funded by the California Energy Commission (Commission) and managed by the New Buildings Institute.

As the name suggests, it is not individual building components, equipment, or materials that optimize energy efficiency. Instead, energy efficiency is improved through the integrated design, construction, and operation of building systems. The *Integrated Energy Systems: Productivity and Building Science Program* research addressed six areas:

- ◆ ***Productivity and Interior Environments***
- ◆ ***Integrated Design of Large Commercial HVAC Systems***
- ◆ ***Integrated Design of Small Commercial HVAC Systems***
- ◆ ***Integrated Design of Commercial Building Ceiling Systems***
- ◆ ***Integrated Design of Residential Ducting & Air Flow Systems***
- ◆ ***Outdoor Lighting Baseline Assessment***

The Program's final report (Commission publication #P500-03-082) and its attachments are intended to provide a complete record of the objectives, methods, findings and accomplishments of the *Integrated Energy Systems: Productivity and Building Science Program*. The final report and attachments are highly applicable to architects, designers, contractors, building owners and operators, manufacturers, researchers, and the energy efficiency community.

This Windows and Classrooms (Product #2.4.10c) is a part of the final report within the Productivity and Interior Environments research area and presents the results of a study into relationships between the indoor classroom environment and student performance.

The Buildings Program Area within the Public Interest Energy Research (PIER) Program produced these documents as part of a multi-project programmatic contract (#400-99-413). The Buildings Program includes new and existing buildings in both the residential and the non-residential sectors. The program seeks to decrease building energy use through research that will develop or improve energy efficient technologies, strategies, tools, and building performance evaluation methods.

For other reports produced within this contract or to obtain more information on the PIER Program, please visit www.energy.ca.gov/pier/buildings or contact the Commission's Publications Unit at 916-654-5200. All reports, guidelines and attachments are also publicly available at www.newbuildings.org/pier.

ABSTRACT

This study investigates whether daylight and other aspects of the indoor environment in elementary school student classrooms have an effect on student learning, as measured by their improvement on standardized math and reading tests over an academic year. The study uses regression analysis to compare the performance of over 8000 3rd through 6th grade students in 450 classrooms in the Fresno Unified School District, located in California's Central Valley. Statistical models were used to examine the relationship between elementary students' test improvement and the presence of daylight in their classrooms, while controlling for traditional education explanatory variables, such as student and teacher demographic characteristics. Numerous other physical attributes of the classroom were also investigated as potential influences, including ventilation, indoor air quality, thermal comfort, acoustics, electric lighting, quality of view out of windows, and the type of classroom, such as open or traditional plan, or portable classroom. The study also utilized on-site observations of classrooms and surveys of teachers to provide additional insight into comfort conditions. The study did not replicate the findings of a previous study when using the same form of the statistical models. However, this study did find that various window characteristics of classrooms were had as much explanatory power in explaining variation in student performance as more traditional educational metrics such as teacher characteristics, number of computers, or attendance rates. The study provides a range of likely effect sizes for environmental variables that other researchers can use to refine the needs of future studies.

Author: Lisa Heschong, Heschong Mahone Group

Keywords: Daylight, Productivity, Student Performance, Window, Skylight, Absenteeism, Attendance, Health, Classroom Condition, School Design, Views

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EXECUTIVE SUMMARY

This study investigates whether daylight and other aspects of the indoor environment in elementary school student classrooms have an effect on student learning, as measured by their improvement on standardized math and reading tests over an academic year. The study uses regression analysis to compare the performance of over 8000 3rd through 6th grade students in 450 classrooms in the Fresno Unified School District, located in California's Central Valley. Statistical models were used to examine the relationship between elementary students' test improvement and the presence of daylight in their classrooms, while controlling for traditional education explanatory variables, such as student and teacher demographic characteristics. Numerous other physical attributes of the classroom were also investigated as potential influences, including ventilation, indoor air quality, thermal comfort, acoustics, electric lighting, quality of view out of windows, and the type of classroom, such as open or traditional plan, or portable classroom.

Previous Studies

This study is the third in a series of studies looking at the relationship between daylighting and student performance. The first, *Daylighting in Schools*,¹ which was completed for Pacific Gas and Electric in 1999, examined school districts in three states. In Seattle, Washington and Fort Collins, Colorado, where end-of-year test scores were used as the outcome variable, students in classrooms with the most daylighting were found to have 7% to 18% higher scores than those with the least. In San Juan Capistrano, California, where the study was able to examine the improvement between fall and spring test scores, we found that students with the most daylighting in their classrooms progressed 20% faster on math tests and 26% faster on reading tests in one year than in those with the least.

A second study, the *Daylighting in Schools Reanalysis Report*² completed for the California Energy Commission in 2001 further investigated the results from the Capistrano school district. We investigated whether better teachers were being stationed in more daylit classrooms, and thereby inflating the importance of the daylight variable. In that district, we found that there was no assignment bias of better teachers to more daylit classrooms. Furthermore, the addition of information about teacher characteristics to the original student performance models did not reduce the significance or magnitude of the daylight variables. Among twelve models considered in that study we identified a central tendency of a 21% improvement in student learning rates from those in classrooms with the least amount of daylight compared to those with the most.

¹ Heschong Mahone Group (1999). *Daylighting in Schools*. An investigation into the relationship between daylight and human performance. Detailed Report. Fair Oaks, CA.
(http://www.h-m-g.com/Daylighting/daylighting_and_productivity.htm)

² Heschong Mahone Group (2001) *Re-Analysis Report, Daylighting in Schools*, for the California Energy Commission, published by New Buildings Institute, www.newbuildings.org

Fresno Study

This study's primary goal was to examine another school district, one with a different climate and curricula, to see whether the original methodology and findings would hold. We collected more information about the lighting and daylighting conditions in the classrooms, to allow us to test which attributes of a daylit classroom were more likely to contribute to a "daylight effect," if any. We also wished to understand how other aspects of the indoor environment affected student performance and interacted with daylight. To accomplish these goals, this study gathered detailed information about classroom conditions, including lighting and daylighting, HVAC, ventilation, windows, surface coverings, view, and indoor air quality. Whereas we had done on-site surveys only a sample of classrooms for the previous studies, for this study we went on-site to measure attributes in every classroom, surveying a total of 500 classrooms in 36 schools.

The preliminary statistical analyses replicated the structure of the models used in the previous studies. They used a holistic variable called the *Daylight Code* to rate classrooms by the amount of daylight available throughout the school year. In these replication models, the *Daylight Code* was **not** significant in predicting student performance for Fresno. It had the least explanatory power of the variables considered, and lowest significance level. Thus, we could not replicate the Capistrano findings based on a similar model structure. We proceeded with more detailed statistical analysis to see if we could identify specific influences of school or classroom design on student performance, and perhaps gain some insight as to why the *Daylight Code* was not significant in Fresno as it had been in Capistrano, Seattle and Fort Collins.

We used multi-linear regression analysis to test a wide variety of variables to see which provided the best explanation of student performance. Of the variables describing the physical conditions of classrooms and schools, characteristics describing windows were generally quite stable in their association with better or worse student performance. Variables describing a better view out of windows always entered the equations as positive and highly significant, while variables describing, glare, sun penetration and lack of visual control always entered the models as negative.

In addition, attributes of classrooms associated with acoustic conditions and air quality issues followed a similar pattern. Those variables representing sources of internal noise, such as a loud HVAC system or a loud ballast hum from the lighting system, were consistently associated with negative student performance, while increasing the amount of carpet (which reduces acoustic reverberance) in the classroom was associated with better student performance in reading. Variables related to indoor air quality showed that in Fresno automatically controlled mechanical ventilation (*No Teacher Control of Fan*) was positive, while visible water damage or a surveyor assessment of musty air in the classroom was negative.

Summary of Study Findings

The findings of regression models in this study support the general conclusions that:

- **The visual environment is very important for learning.**
 - **An ample and pleasant view** out of a window, that includes vegetation or human activity and objects in the far distance, supports better outcomes of student learning.
 - **Sources of glare** negatively impact student learning. This is especially true for math learning, where instruction is often visually demonstrated on the front teaching wall. Per our observations, when teachers have white marker boards, rather than black or green chalk boards, they are more likely to use them and children perform better in math.
 - **Direct sun penetration** into classrooms, especially through unshaded east or south facing windows, is associated with negative student performance, likely causing both glare and thermal discomfort.
 - **Blinds or curtains** allow teachers to control the intermittent sources of glare or visual distraction through their windows. When teachers do not have control of their windows, student performance is negatively affected.
- **The acoustic environment is also very important for learning.** Situations that compromise student focus on the lessons at hand, such as reverberant spaces; annoying equipment sounds, or excessive noise from outside the classroom, have measurable negative effects on learning rates.
- **Poor ventilation and indoor air quality also appear to negatively affect student performance.** However, in FUSD these issues are almost hopelessly intertwined with thermal comfort, outdoor air quality and acoustic conditions. Teachers often must choose to improve one while making another aspect of the classroom worse.
- **Physical characteristics of classrooms are just as likely to affect student learning as many other factors commonly given much more public policy attention.** Variables describing the physical conditions of classrooms, most notably the window characteristics, were as significant and of equal or greater magnitude as teacher characteristics, number of computers, or attendance rates in predicting student performance.

Problems with Daylit Classrooms

We tested each statistical model with and without the *Daylight Code*. When we added the *Daylight Code* the other variables remained essentially the same, but the *Daylight Code* always came in as significant and negative, telling us that there was some characteristic of classrooms sorted by the *Daylight Code* that was associated with a negative effect. Examination of the performance of individual classrooms, considering all of their window characteristics plus the *Daylight Code*, showed that there were three types of classrooms in Fresno that were performing particularly well in relationship to their daylight characteristics—finger plan classrooms, grouped plan classrooms and portables—as long as they had no glare or other undesirable window characteristics. Thus, classrooms with both the highest and the lowest *Daylight Code* were seen to support better student performance.

Many potential explanations for the negative influence of the *Daylight Code* were considered, and we went back on site to see if there were any systematic reasons why students in classrooms with a higher Daylight Code would perform worse, or those in classrooms with a low Daylight Code would perform better. In this second phase of the study, detailed examination of a number of potential confounding variables, including view-related distractions, glare, operable windows, radiant thermal comfort, indoor air quality and acoustic performance were considered. To better understand the results of the regression analysis, we visited 40 classrooms while they were in operation and surveyed 116 teachers about their assessment of and operation of their classrooms.

Overall, the daylit classrooms in Fresno had some consistent problems that might have degraded student performance, and which we believe did not exist in the previous districts studied. The most compelling of these were the acoustic problems created in the daylit classrooms. We found the classrooms with high daylight codes to have reverberation levels above current national recommendations, while classrooms with low daylight codes typically met or exceeded those recommendations. This reverberation problem tended to be aggravated by the presence of teaching assistants who provide in-class tutorials for individuals or small groups. In low *Daylight Code* classrooms these tutorials were often held outside of the classroom in conveniently adjacent common areas, while in the high *Daylight Code* classrooms they took place in the back of the classroom, raising the background noise level and making the teacher's voice less intelligible.

In addition, we noted teachers in classrooms with a high *Daylight Code* were more likely to teach with their windows open, primarily to compensate for poor temperature control and to improve ventilation. These open windows allowed in more noise from the outside, exacerbated by crowded schools running on multiple lunch and recess schedules. We noted from the various regression models that, on the one hand, continuous mechanical ventilation seemed to improve student performance, while on the other hand, a higher percentage of operable windows were associated with lowered performance. We hypothesize that the poor outdoor air quality in Fresno¹, combined with the epidemic of asthma in school children, suggests the preferred use of mechanically filtered air rather than natural ventilation in FUSD.

We also considered whether the problems we detected with daylit classrooms could be rectified, and calculated the value of potential energy savings if daylit classrooms were operated to reduce reliance on electric lighting. Acoustic analysis of the daylit classrooms showed that the reverberance problem could be corrected with the use of more sound-absorbing surfaces, such as carpet and high quality acoustic tile. The use of dual pane low-e glazing on the windows could simultaneously improve both the acoustic conditions in the classrooms and thermal comfort. Energy analysis showed substantial potential savings (1.1 kWh/sf) for retrofitting existing FUSD daylit classrooms with photocontrols. California could achieve an additional 3300 to 4800 megawatt-hours (0.6 to 0.9 kWh/sf) of energy savings statewide for each year that all new school construction included good daylighting design with photocontrols. This would accumulate to 33,000 to 48,000 megawatt-hours per year savings after ten years.

¹ Fresno has nationally high levels of small particulate pollution associated with lung damage, per J Raloff "Air Sickness" in Science News, Vol. 164, No 5.

The Importance of School Design Choices

These findings suggest the importance school planners should give to the architectural design of schools. The statistical models repeatedly demonstrate that physical condition of classrooms and schools are just as likely to affect student learning as many other factors commonly given much more public policy attention. Variables describing the physical conditions of classrooms, most notably the window characteristics, were as significant and of equal or greater magnitude as teacher characteristics, number of computers, or attendance rates in predicting student performance. The partial R^2 of the different variable types is also very informative. The one variable which is specific to the individual—their fall test score—predicts about 10% of the variation in the gain from fall to spring. The demographic variables, which describe generic groups to which the individual belongs, predict performance with an order of magnitude less precise, or about 1% each. The physical characteristics of the schools again drop another order of magnitude in predictive power, each significant variable describing on the order of 0.1% of the variation in student performance.

However, even though the physical characteristics of a classroom have a very minor potential influence on the performance of a given individual, they will reliably affect hundreds or thousands of students over the life of the building, typically 50 years. Since the design of classrooms is entirely within the control of the school district, much more so than student or teacher demographics, optimized design of schools should be a central concern for all new school construction.

1. INTRODUCTION

This study investigates whether daylight and other aspects of the indoor environment in elementary student classrooms have an effect on student learning, as measured by their improvement on standardized math and reading tests over an academic year. Statistical models were used to examine the relationship between elementary students' test improvement and the presence of daylight in their classrooms, while controlling for traditional education explanatory variables, such as student and teacher demographic characteristics. Numerous other physical attributes of the classroom were also investigated as potential influences, including ventilation, indoor air quality, thermal comfort, acoustics, electric lighting, quality of view out of windows, and the type of classroom, such as open or traditional plan, or portable classroom.

The study uses regression analysis to compare the performance of over 8000 3rd through 6th grade students in 450 classrooms in the Fresno Unified School District, located in the Central Valley of California. This effort is part of a suite of studies funded by the Public Interest Energy Research program of the California Energy Commission to look at the effect of daylight on human performance in three workplace environments: retail, schools and offices. This study was designed primarily to test whether the findings of a previous study completed for Pacific Gas and Electric in 1999, "Daylight and Student Performance," could be replicated in a new environment. That study examined the Capistrano Unified School District along the coast of southern California, and found that children in classrooms with the most daylight were learning 20-26% faster on reading and math curricula, as evidenced by their progress on standardized fall and spring tests, compared to children in classrooms with no daylight. Two other districts also studied, in Seattle and Fort Collins Colorado showed similar results. A second, follow-up study added information about teacher credentials to the Capistrano analysis, and found that this effect could not be explained by "better" teachers being assigned to more daylight classrooms. It confirmed a central tendency of a 21% improvement in test scores from fall to spring in fully daylight classrooms compared to non-daylight classrooms.

We chose to study elementary schools since children at that age spend most of their school time in one physical environment—their assigned classroom—whereas students in middle schools and high schools tend to move from classroom to classroom throughout the day. We reasoned that if the physical environment affects learning, it should be easier to identify any effects at the elementary level where we could characterize a given student's environment with some certainty.

Since this is an interdisciplinary study, there are readers of many disciplines who have interest in its findings, including architects and engineers, school administrators, educational researchers, public health officials and statisticians. We have attempted to satisfy the concerns of a wide range of readers, and so

have perhaps included more detail than any one of these readers may find useful.

In the introduction we discuss the background context that motivated and informed the study. We then present our rationale for a choice of a study site, and our data collection and analysis methodologies. The findings of the statistical analysis are presented in a more reader-friendly format in the body of the text, and as formal statistical tables in the Appendix. The Appendix also includes specific details of our data collection and analysis methods. After discussing the findings of the regression analysis, we describe the second phase of the study, when we observed classrooms while in operation and spoke with teachers in an attempt to achieve a deeper understanding of the regression findings. Finally, we discuss the implications of this study for school design and the potential value of daylighting in schools in terms of energy savings for the Fresno district and the state of California as a whole.

1.1 Background

The impact of daylighting on the performance of school children has been a subject of interest for many years. Before fluorescent lighting became prevalent, it was generally assumed that all school rooms would be daylit as a matter of course. The California Department of Education had a rigorous review process for the architectural design of classrooms to ensure that daylighting standards were met. As a result, California classrooms built in the 1950s and early 1960s remain excellent examples of daylighting practice. The finger plan with multiple rows of single classrooms, each with large windows along the north and south sides, became a standard for California K-12 campuses.

However, starting in the late 1960s, a number of forces came into conflict with the daylit design of classrooms. Engineers, facility managers and educational theorists all recommended a more compact, grouped building shape with few if any windows. As states like California tried to accommodate an exploding school age population, portable classrooms also proliferated, to accommodate as many students as quickly as possible. As a result of these various pressures, the finger plan school was largely abandoned in California, and a vast experimentation in school design was undertaken. Many of the classrooms built since the 1960s have little daylighting. Windows were commonly built with heavily tinted glass that allows a view out but no useful daylight in. Numerous schools have been built with no windows at all.

Similar trends occurred nationally, and internationally, though perhaps without such a dramatic shift in design practice as in California. Concerned about the trend towards schools, and all types of buildings, without windows, Belinda Collins of the National Bureau of Standards conducted a major literature review

on the study of windows in 1974¹. Collins found that the many researchers of the time were dismissive of the importance of windows, citing lack of hard evidence of their benefits and easy proof of cost savings. She concluded that research completed as of 1974 was suggestive of the importance of windows, but inconclusive:

“Much, though not all, of the evidence from the windowless classroom studies is inconclusive, or inadequate, while that from windowless factories is circumstantial, based on hearsay, rather than research. As a result, only tentative conclusions can be drawn about the qualities of windowless spaces that make them somewhat less than desirable.”

More recently interest has revived in the importance of windows for both the provision of daylight and the value of a view through a window, especially of nature. Studies conducted by Heschong Mahone Group, described in the following sections, were a first step to demonstrate and quantify an association between presence of daylight and better student performance. At the same time, recent research in physiology and photobiology has been underscoring the fundamental importance of circadian rhythms in health and mental function. These circadian rhythms, inherent in all forms of life on earth, evolved to respond to natural patterns of bright light during the day and complete darkness at night. Wavelengths of light in the blue region of the spectrum, very similar to the spectrum of the blue sky, have been shown to interact with the production of the hormone melatonin that controls much of our cycles of sleep and mental alertness². Researchers are just starting to sort out the relative importance of timing, duration, intensity and spectrum in our needs for light exposure during the day to maintain a healthy circadian pattern³.

The inclusion of daylight in classrooms has become a leading feature of a movement for “high performance schools,” i.e. school buildings that can potentially improve student performance, reduce operating costs and minimize negative impacts on the environment.⁴ At this point in time there seems to be more interest in promoting high performance schools than reliable information about what the best choices for school design truly are in a given context.

¹ Collins, B. "Windows and People: a Literature Survey, Psychological Reaction to Environments With and Without Windows", National Bureau of Standards, June 1975

² G C Brainard , et al. “Action spectrum for melatonin regulation in humans evidence for a novel circadian photoreceptor,” *The Journal of Neuroscience*, august 15, 2001. 21(16): 6405-6412

³ M Rea “Light – Much More Than Vision” in the proceedings of the 5th International LRO Lighting Research Symposium, Lighting Research Office of EPRI, November 2002

⁴ See information on the Coalition for High Performance Schools at www.chps.net and The Energy Smart Schools Program of the US Department of Energy, <http://www.eren.doe.gov/energysmartschools/>

1.1.1 The Complexity of the Indoor Environment

This study of elementary school classrooms in Fresno Unified School district attempts to untangle some of the many complex issues in school design. By looking at more than just daylight, we have attempted to understand the interrelationship between multiple indoor environmental issues, such as lighting quality, thermal comfort, ventilation and acoustics. These issues are clearly intertwined in the classrooms in Fresno, and often teachers must make choices to improve one condition at the expense of another. In the classrooms we surveyed, we saw considerable challenges created for teachers with poor thermal comfort, air quality and acoustic conditions that degraded their ability to teach well.

Windows are perhaps one of the most complex aspects of the classroom environment. They can provide a classroom with daylight, views, ventilation and a communication conduit with the outside world. They can also allow thermal discomfort, glare, noise and distractions into the classroom. In our previous studies we attempted to control for the complex nature of windows by including top-lit spaces in the study that would introduce “pure” daylight into a classroom without all of the issue of view, distraction, and communication presented by windows. In this study, Fresno Unified School District did not have any classrooms with toplighting. All of the daylight that we considered was provided by windows with all of these other complicating factors. Thus, in order to better understand the way windows might be influencing student performance, we collected much more information about the various characteristics of the windows in each classroom. We considered orientation, size and location of the windows, glazing tint, presence of blinds or curtains, glare potential, amount of operable area, and view out of the window. Of all of these characteristics of windows perhaps the most interesting and most controversial is the importance of view.

1.1.2 The Importance of View

Educational researchers have theorized that views out of windows cause unnecessary distractions for children in the classroom. Removal of distractions was one of the prime motivations for the windowless classroom of the 1960s and 70s. However, at the same time educational psychologists have been studying the importance of a stimulating visual environment in the learning process. Many studies have shown that rats learn faster and have better memories in more stimulating environments. Ophthalmologists, concerned about the prevalence of eye strain in the modern work environment, have stressed the importance of the availability of distant views to offer relaxation to the eye engaged in close work on a computer or other near task. Young children, whose eyes and visual processing capabilities are still developing, may be especially sensitive to these issues¹.

¹ ML Wolbarsht “The Development of Myopia in Relation to the Lighting Environment” in the proceedings of the 5th International LRO Lighting Research Symposium, Lighting Research Office of EPRI, November 2002

Researchers looking for real-world impacts windows discovered that views of nature potentially improve people's health and well-being¹. For example, prisoners with windows facing the surrounding hills instead of the interior prison courtyard visit the infirmary less frequently and report fewer stress-related ailments.^{2,3} A famous study of heart surgery patients found that patients whose window overlooked a field edged with trees healed faster and required less pain medication than those with a view of a brick wall.⁴

Others interested in landscape and the natural environment have postulated theories that views of nature reduce stress or improve attention. One theory to explain the importance of views to nature suggests that natural elements trigger quick, positive emotions that help reduce physiological stress. This theory, bolstered by various laboratory and field studies, suggests that urban dwellers might constantly be experiencing low-level stress reactions which impact their physical health and behaviors, and that might be alleviated by exposure to natural scenes.⁵

Attention Restoration Theory offers a different mechanism to explain the benefits of exposure to views of nature. This theory suggests that views of natural scenes have the potential to restore the directed attention capabilities of the brain after extended cognitive activity has drained a person's ability to focus and concentrate. Once the mind's ability to suppress distractions and impulses has become exhausted, people perform more poorly on tests requiring concentration. It also impacts people's ability to suppress urges for inappropriate behavior in favor of thoughtful consideration. Finally, this capacity affects emotion; people whose attention is exhausted show irritability and impatience. In various studies, natural window views have been shown to restore or maintain peoples' ability to concentrate over extended periods.⁶

Given this body of research, we hypothesized that the quality of the view from a classroom may influence student learning. While we considered view only one of

¹ Summarized from C Knecht "Urban Nature and Well-Being: Some empirical support" Berkeley Planning Journal, 17, in publication

² West, M.J. (1985). Landscape and stress response in the prison environment. M.L.A. thesis. Department of Landscape Architecture, University of Washington, Seattle, WA.

³ Moore, E.O. (1982). A prison environment's effect of health care service demands. Journal of Environmental Systems, 11, 17-34.

⁴ Ulrich, R.S. (1984). View through a window may influence recovery from surgery. Science, 224, 420-421.

⁵ Ulrich, R.S., Simons, R.F., Losito, B. D., Fiorito, E., Miles, M. A. and Zelson, M. (1991). Stress recovery during exposure to natural and urban environments. Journal of Environmental Psychology, 11: 201-230. The overall theory is explained in Ulrich, R. S. (1983). Aesthetic and affective response to the natural environment. In I. Altman and J. F. Wohlwill, Eds. Human Behavior and Environment: Advances in Theory and Research, 6, 85-125. NY: Plenum.

⁶ Kaplan, S. (1995). The Restorative Benefits of Nature: Toward an Integrative Framework. Journal of Environmental Psychology, 15, 169-182.

many potential characteristic of windows that could be an important factor, we decided to include simple metrics view content and distance in our analysis.

1.1.3 Air Quality

The quality of indoor air has also become a driving issue in research on the quality of our schools. In California a number of recent projects have investigated indoor air quality in schools, with particular attention to problems caused by mold and by poor construction in portable classrooms.¹ Concerned with the prevalence of asthma, the California Department of Health Services has also undertaken a long-term study of children in Fresno, looking at their exposures both at home and at school². Our own 1999 study indicated that operable windows were associated with better student performance in Capistrano, implying that there might be benefits to natural ventilation in schools.

Given the growing concern with indoor air quality, and especially with the high pollution levels in Fresno, we coordinated with related studies in California. Our surveys included simple, observable features that might reflect on the indoor air quality in the classrooms, while other studies included simple information on the lighting and windows characteristics in the classrooms studied.

1.1.4 Acoustics

Recent research into classroom acoustics has shown that children need much quieter, less reverberant spaces in order to hear and understand spoken words than adults do. Children are immature listeners and have a difficult time mentally separating the verbal signal from background noise. While adults have developed the ability to understand most familiar words when the noise level and speech are approximately equal, normal healthy children younger than 13 need background noise level that are significantly quieter than the signal they are trying to understand³. Children who have temporary hearing loss due to ear infections, who are trying to learn a new language, or have auditory or attention problems need even more favorable acoustics in order to successfully understand speech. In response to this new understanding of the needs of children as they learn to process auditory information, the American National Standards Institute (ANSI) has issued new standards that significantly lower the sound level limits for background noise in classrooms, to 35 dBA and 55 dBC.

Given that Fresno Unified School District has an enormously high proportion of children learning English, we recognized that the acoustic conditions of classrooms could potentially be one of the most critical factors in learning

¹ RTI International, California Portable Classrooms Study, for California Air Resources Board and Department of Health Services, contract 00-317, May 2003

² I Tager "Responses to Short-term Fluctuations in Particulate Air Pollution in Asthmatic Children", (F.A.C.E.S.) on going study for the California Air Resources Board, contract 99-322.

³ P Nelson, "Sound in the Classroom, Why Children Need Quiet" in ASHRAE Journal, February 2003

success for that district. Thus, we attempted to include indicators of noise and acoustic conditions in our study of classrooms.

1.1.5 Energy Efficiency

Energy efficiency in schools is a major concern nationwide, since every dollar spent on maintaining a school is one dollar less that is available for curriculum materials and support. The use of daylight to illuminate classrooms is an obvious approach to reducing energy use, since almost all K-12 schools operate during the daytime when there is plenty of daylight available. However, it is also a complex issue, since energy savings from daylight is not only a function of daylight availability, but also how efficient design of the daylight and electric lighting system is, whether and when the electric lights are turned off, and how much extra heat loss or heat gain is introduced into the buildings via the openings provided for daylight. In order for architects and engineers to optimize these issues, they need to understand the positive and negative characteristics of windows and skylights, and how they interact with other systems in the building. Thus, we have attempted to characterize some of the component parts of a classroom daylighting system in order to provide more detailed guidance on classroom design.

1.2 Summary of Original 1999 Study

The original Daylighting in Schools study,¹ completed for Pacific Gas and Electric in 1999, found a compelling statistical correlation between the amount of daylighting in elementary school classrooms and the performance of students on standardized math and reading tests. For that study we identified three study sites of large school districts that had a range of daylighting conditions in their classrooms. We specifically selected districts that had a number of classrooms lit from above with skylights or roof monitors, to examine “pure” daylighting without the complications associated with window daylight, such as view.

The three districts were located in San Juan Capistrano, (Southern) California; Seattle, Washington; and Fort Collins, Colorado. These three districts have very different climates, different school building types, different curriculums and different testing protocols. We collected test scores and demographic information for all second through fifth graders in the district, and classified their classrooms for the amount and quality of daylight available. The districts also provided us with information about student demographic characteristics, special school programs, size of schools, etc.

We added information to these data sets about the physical conditions of the classrooms to which these children were assigned. We reviewed architectural

¹ Heschong Mahone Group (1999). Daylighting in Schools. An investigation into the relationship between daylight and human performance. Detailed Report. Fair Oaks, CA. www.h-m-g.com

plans, aerial photographs and maintenance records and visited a sample of the schools in each district to classify the daylighting conditions in over 2000 classrooms. Each classroom was assigned a series of codes on a 0-5 scale indicating the size and tint of its windows, the presence and type of any skylighting,¹ and the quality and quantity of overall daylight expected from windows and toplighting combined. This last value, a qualitative evaluation of classroom daylight which we termed the *Daylight Code*, was used again in this study and will be further explained in Section 6.2.

Ultimately the study analyzed test score performance for 8000 to 9000 students per district. We looked at both math and reading scores in all three districts, and analyzed each separately, alternately using the holistic Daylight Code and the separate window and skylight codes, for a total of twelve statistical models.

The Capistrano Unified School District proved to be our most interesting study site for a number of reasons. The District administers standardized tests both in the fall and spring, allowing us to compare the change in students' math and reading test scores while they spent the year in one classroom environment. Because the District has a number of standardized portable classrooms at every elementary site, we were able to use these portables as a standardized condition controlling for the influence of individual school sites or neighborhoods. We also collected additional information at this district about the HVAC and ventilation conditions of the classrooms, which was also included in the analysis.

In Capistrano, using a regression equation that controlled for 50 other variables, we found that students with the most daylight in their classrooms progressed 20% faster on math tests and 26% faster on reading tests in one year than those with the least daylight. Similarly, students in classrooms with the largest window areas were found to progress 15% faster in math and 23% faster in reading than those student in classrooms with the least window area. Students that had a well-designed skylight in their room, one that diffused the daylight throughout the room and which allowed teachers to control the amount of daylight entering the room, also improved 19-20% faster than those students without a skylight. Classrooms with a skylight that allowed direct beam sunlight into the classroom and did not provide the teacher with a way to control the amount of daylight were actually seen to have a negative association with student performance. In addition, in three of the four Capistrano models, the presence of an operable window in the classroom was also seen to have a positive effect on student progress, associated with 7-8% faster learning. These effects were all observed with 99% statistical certainty.

The Seattle and Fort Collins school districts administer only one standardized test at the end of the school year. In these districts, the study used the final scores on math and reading tests at the end of the school year and compared

¹ In Capistrano, the skylights were given a variable type (A, AA, B, C, D) rather than a scalar.

the results to the district-wide average test score. In both of these districts we also found positive and highly significant (99%) effects for daylighting. Students in classrooms with the most daylighting were found to have 7% to 18% higher scores than those in classrooms with the least.

1.3 Findings of Capistrano Re-Analysis Study

Reviewers of the original school study specifically asked if “better” teachers were more likely to be assigned to the more daylight classrooms, thus influencing the results. Thus, “Re-Analysis Report, Daylighting in Schools,” a follow-on study¹, was conducted to address that and other concerns by re-examining our most detailed models for the Capistrano district. This reanalysis of the original study data was intended to answer key questions raised by the peer review of the earlier study, and expand our understanding of methodological choices for further work. For the reanalysis study we conducted four tasks:

The **Teacher Survey** collected information from a sample of teachers in the Capistrano school district about their education and experience levels, preferences for classroom features and operation of those features. While the teachers we surveyed generally had a preference for windows, daylight and views in their classrooms, these preferences were not found to be driving classroom preferences. Far more important was an almost universal desire for more space, a good location, quiet, lots of storage and water in the classroom.

The Teacher **Bias Analysis** further examined information from the Teacher Survey. The goal of the Bias Analysis was to discover if the original study had over-inflated the effect of daylight on student learning by not accounting for a potential “assignment bias” of better teachers to more daylit classrooms. We conclusively found that there was not an “assignment bias” in Capistrano influencing our results. When we added the teacher characteristics from the survey to the original student performance models, the daylight variables were not reduced in significance. Further analysis of other sub-populations repeated these findings and identified a central tendency of a 21% improvement in student learning rates from those in classrooms with the least amount of daylight compared to those with the most.

In the **Grade Level Analysis**, we re-analyzed the original student test score data for both Capistrano and Seattle by separate grade level, instead of aggregating the data across the four grade levels (2-5). Our goal was to determine if this method would more accurately explain the relationship of student performance to daylighting. The data did not show any significant patterns between a daylight effect and the separate grade levels. Allowing the results to vary by grade did not noticeably improve the accuracy of the models. Therefore, we conclude that looking at data across grade levels is a sufficiently accurate methodology.

¹ Heschong Mahone Group (2001) “Re-Analysis Report, Daylighting in Schools,” for the California Energy Commission, published by New Buildings Institute, www.newbuildings.org

In the **Absenteeism Analysis**, we used absenteeism and tardiness data in the original Capistrano data set as dependent variables and evaluated them against the full set of explanatory variables from the original study, plus the new information on teacher characteristics. Student attendance data are certainly not the best indicator of student health. Yet to the extent that attendance data do reflect student health, our findings do not suggest an obvious connection between physical classroom characteristics and student health. Notably, daylighting conditions, operable windows, air conditioning and portable classrooms were not found to be significant in predicting student absences.

1.4 Goals for This Study

For this study, we collected a vast amount of information about the schools and classrooms. While our primary interest was investigating the potential effect of daylight and windows on student performance, we also considered the influence of other aspects of the physical environment. There were a number of reasons for that wider view. First, it was likely that there might be interactive effects between daylight and/or window characteristics and other physical characteristics of the classroom, such thermal comfort and acoustic environment. Secondly, it was possible that the other qualities of classrooms might have a greater influence on student performance than our variables of interest. If we excluded them from the analysis, we would not be aware of their influence and could potentially overestimate the daylight effects. Finally, both we and the study's sponsor, the California Energy Commission, are interested in what makes a better classroom, especially those systems that affect the quality of the indoor environment, and how decisions for a better learning environment affect the energy use of schools.

The primary goal for this study was to test whether the methodology and findings of the previous Daylighting and Schools Study, completed in 1999, would hold at another school district, ideally one with a different climate, administrative and curriculum style. For any scientific study, regardless of the strength of the initial findings, replication is the acid test of validity:

- Will we find similar results, that more daylight in classrooms is associated with faster student learning, using the same methodology, but with a different study population?

In conducting a replication study there are also opportunities to improve upon the first by adding additional information to the analysis. On the one hand, we wanted to investigate whether potentially confounding variables which we could not control for in the original study, such as the mild coastal climate of Capistrano, or neighborhood effects in Seattle or Fort Collins might influence our findings. On the other hand, given a positive finding the pilot study, it is worthwhile to increase the investment in data collection for a replication study to provide further analytic power in trying to understand the mechanisms of a

possible effect. Thus, one of our secondary goals in this study was to collect much more information about the schools, especially their lighting and daylighting conditions, in order to test which attributes of a daylit classroom were more likely to contribute to a possible “daylight effect.” With further detail, and assuming similar positive results, we hoped to be able to answer questions such as:

- Does view from a window contribute to positive performance, independent of the daylight illumination?
- How important is lighting quality, such as lack of glare or uniform distribution of light, compared to lighting intensity?
- Are there discernable negative effects associated with daylit classrooms, such as increased noise in the classroom or thermal discomfort?

The indoor environment of any building is highly complex and interactive, and the various environmental systems—illumination, thermal comfort, acoustic comfort, ventilation—should not be studied in isolation from each other. Human response to all these environmental conditions is integrated by our bodies and brains into overall comfort and performance. There are many ways these systems interact. For example, windows not only provide daylight but may also provide ventilation, while letting in more noise and heat from outside. Similarly, electric lighting fixtures may “hum,” and mechanical ventilation systems may be noisy. Many other studies have looked at environmental systems in isolation and failed to account for the possible influence of other systems. Thus, to properly study our variable of interest—daylight—we needed to also characterize other environmental systems in the classrooms that could potentially be interacting with daylight to increase, or negate, a “daylight effect.”

Once we recognized that we would be collecting additional information about the other environmental systems or conditions in the classrooms in order to control for our variable of interest, it was natural to add another set of goals for the study: to consider the impact of those environmental conditions on learning. There were a number of other controversial issues in California school design that could possibly be addressed in the same project, given sufficient data collection. These include:

- Is the quality of ventilation and indoor air quality in classrooms having an impact on student health or performance?
- Are portable (relocatable) classrooms having an impact on student health or performance?
- Do the acoustic qualities of classroom impact student learning?
- Does centrally managed control of lighting, ventilation and/or temperature provide a more conducive environment for learning, or should teachers be provided with local controls for their own classroom?

Our ultimate goal is to provide useful guidance to architects and school district planners in shaping the design priorities for schools. They need not only to know whether certain aspects of the school environment affect student performance,

but also to understand which design decisions will best achieve those goals. As much as possible, they need to understand how specific design characteristics of a school or classroom are associated with impacts on students, and what the trade-offs are for incorporating those characteristics into the design of schools. Will it cost more money to construct? Will it reduce operating costs? Are there other benefits? While most of these critical design decisions are far beyond the scope of this study, we did hope to provide as much specific guidance as possible that could be directly useful to school designers and planners.

Since this project is funded by the California Energy Commission, resulting changes in energy use in buildings due to any design changes is also a primary concern. Thus, one of our final goals was to answer the question:

- What are the potential energy savings of more extensive use of daylighting design in California K-12 classrooms?

1.5 About Statistical Analysis

This study relies on statistical analysis of the performance of thousands of students to detect very subtle effects of the physical environment on student learning.

However, individual human behavior is not highly predictable. We can only account for influences that we can easily and reliably measure across very large groups of people. This limitation tends to reduce statistical models of behavior to very simplistic explanatory variables such as generic group membership, like a student's ethnic group, their socio-economic or bi-lingual status. While this is the best information we have, it still does not do a very good job of explaining individual performance. The models in this study account for less than 25% of the influences on how well a student masters the standardized math and reading curriculum at Fresno Unified School District. The other 75% of variation in student scores remain unexplained, and may be a function of a given individual's motivation, their relationship to their teacher, what they had for breakfast, how they slept the night before, or may be just completely random.

There are often suggestions made that statistical analysis is just an elaborate form of deception. There is a common joke that "there are lies, damn lies, and statistical lies." Statistical analysis is indeed a very complex and relatively arcane subject that can easily be obfuscated to avoid scrutiny. However, it also provides enormous power in understanding huge trends in the world that are beyond our perception as individuals. Statistical analysis allows us to see patterns in large data sets that are not otherwise readily apparent. Multivariate regression analysis allows one to isolate the certainty and magnitude of a given effect, while simultaneously controlling for competing influences that inevitably occur. Thus, it is an enormously powerful tool to understand behaviors in the real world.

Sometimes the findings of complex analysis are fully in line with common sense expectations. At that point, proving the obvious might seem a waste of time, but

even then, there is very useful information provided by the statistical analysis, such as the magnitude and certainty of the effect. On the other hand, when the findings are not consistent with expectations, they provide an important impetus to look further and try to understand underlying mechanisms.

In this study, one of our measures of validity for an explanatory variable has been consistency across multiple models with different outcome variables and different explanatory variables. We have used two outcome variables—performance on reading and math tests—and tested dozens of models as another way to see which explanatory variables were consistent in predicting student performance. We have also included as many other valid explanatory variables as possible to control for other influences and make sure that we were not overestimating the effect of interest. This approach results in very complex models, with lots of explanatory variables.

Even when our analysis points to useful and credible information about the relationship between classroom design and student performance, it is just the beginning. The statistical models provide evidence that a relationship exists, with a certainty and magnitude of that relationship, but cannot tell us why the relationship exists. Other kinds of research, either laboratory or field experiments, are better suited to testing theories about why the relationship exists and proving causal mechanisms. What is new in our work is applying an epidemiological approach to understanding and quantifying the relationship between the built environment and human performance.

We have made an effort to make the findings of the statistical analysis accessible to the average reader. Please see the Appendix for definitions of statistical terms used in this report. It is our hope that this kind of scientific investigation into how building design decisions impact human performance will provide useful information leading to a better and more healthful built environment.

2. STUDY DISTRICT

The site selection process was perhaps the most critical factor in determining the structure and outcome of this research project. In order to replicate the previous study, it was important to find another elementary school district with a range of daylighting conditions. In the previous study, the difference between fall and spring test scores proved to be the most sensitive metric of student progress; here we hoped to use a similar outcome metric. Easy access to the school sites for data collection detailed information about physical conditions was also an essential criterion of selection.

Given the goals discussed above, the following criteria guided our selection process:

- Have a **variety of classroom daylighting conditions**, with some classrooms receiving very little daylight and others receiving a great deal, and ideally, a range of conditions in-between.
- Be a **large district**. Statistical analysis requires a large number of data points, thus, a district with more elementary schools and higher enrollment would provide more data and more likely yield more statistically significant results. A larger district also generally has more staff resources, such as a designated research department, and sophisticated data collection to support such a study.
- Use **fall and spring standardized tests**. We had determined from the previous school study that the difference between fall and spring tests provided the most sensitive metric of student progress within a given classroom per year.
- Maintain **electronic databases** of demographic information. To control for other factors affecting student performance, such as student socioeconomic status or teacher qualifications, a wide variety of demographic information is required.
- Avoid **confounding factors**. We hoped to find a district in which daylighting conditions were not strongly associated with other variables that might influence student performance, such as building age or neighborhood socioeconomics. For example, if low-income neighborhoods consistently had schools with low daylight levels while high-income neighborhoods consistently had schools with high daylight levels, separating the effects of daylight from these confounding factors would be difficult.
- Have **different climate and architectural conditions** from those in Capistrano, the primary district studied in the previous Daylighting in Schools Study. By testing various climates and building types we should

be able to determine whether the effects of daylighting occur consistently or are limited to certain environmental conditions.

- **Be located in California.** The goal of the PIER program is to fund research that will enhance overall energy efficiency and quality of life in the state of California. Research on in-state schools was most likely to be persuasive to state policymakers, and more cost effective for the research project.
- **Be willing to participate.** We obviously needed the district's cooperation in providing data and access to facilities, and we hoped to find a district that would be interested in the study results. Enthusiasm would be more likely to facilitate the study.

2.1 Selection of District

After identifying the selection criteria, the process of locating an appropriate district began. We reviewed information about California school districts available on various governmental web sites, especially the enrollment sizes of elementary and unified districts.

To find a school with the same or similar testing protocol as the previous study, we enlisted the help of the testing association that developed the standardized tests used in Capistrano, the Northwest Evaluation Association (NEA) of Portland Oregon. NEA gave us a list of California districts administering their customized tests in both fall and spring. We included only districts that had at least two continuous years of experience with the NEA tests, to ensure that any "first year anomalies" would be excluded. This generated a list of about eight potential districts in California.

The three largest school districts on this list happened to all be adjacent to each other: Fresno, Clovis, and Visalia. These three districts also fit the criteria of having climatic conditions very different from Capistrano's mild coastal weather, since they are all located in the southern end of California's Central Valley, with a hot dry summer and colder winters. We prioritized the districts on the basis of their size. Fresno is the largest with an enrollment of 79,461, compared to 32,000 for Clovis, and 24,000 for Visalia. Therefore, Fresno was the tentative first choice and Clovis second.

Next, we interviewed district officials in both Fresno and Clovis to determine the districts' appropriateness for the study and interest in participating. This initial screening suggested that Fresno Unified School District (FUSD) would have sufficient classroom variety to support the study, while Clovis, as a newer district, was less likely to have such variety. A proposal was sent to the FUSD Director of Research, who was enthusiastic about the study. After circulating the proposal through various departments, especially the Facilities Department, the District formally agreed to participate in the study.

2.1.1 Verification Site Visits and Initial Meeting

Upon receiving an agreement to participate, we made a preliminary trip to Fresno to discuss the research approach with district staff and verify that sufficient data could be made available. We also used this trip to briefly visit a sampling of FUSD elementary schools to verify that a sufficient range of daylighting conditions was present. In preparation, we mapped the location of all FUSD elementary schools and the District's socioeconomic gradients (free and reduced lunch, percent English speaking), so that we could sample a range of conditions.

We met with representatives of both the FUSD Research and Evaluation Department and the Facilities Department to determine the feasibility of the study. An initial test of the student database suggested that about one-half of the District's third through sixth grade elementary students would meet our criteria for inclusion in the study, or about 10,000 students. This number seemed adequate given the sample sizes of previous successful studies.

We also discussed what information could be made readily available for the study, how the study could be helpful to the District and any limitations the District wished to place on the study methodologies. Most importantly, the District requested that any on-site studies be conducted when school was out of session, either during the summer, or after school or on weekends during the school year.

Preliminary drive-by visits of twelve schools showed a fairly wide range of daylighting conditions and architectural styles, independent of neighborhood conditions. The district contained a mix of classrooms with substantial daylight, classrooms designed for views only, and classrooms with little daylight presence or views. Thus, the initial verification convinced us that the Fresno Unified School District met our criteria and would be suitable for the study.

2.1.2 No Skylit Schools in Fresno

One of the clear challenges, however, presented to us in the selection of FUSD as the study site is that the District contained no skylit or toplit schools that would allow us to distinguish the effects of daylight as illumination source from all of the other characteristics of windows that might influence student comfort and performance. In our previous study, we had specifically selected districts that included a substantial number of toplit schools for inclusion in the study. Fresno did have well-daylit classrooms, but these were all of one type—the classic sidelit finger plan classroom from the 1950s and 60s. We did find some variety in these classrooms, primarily due to renovation over the years. The classic finger plan classroom has a high ceiling with a continuous wall of north-facing windows, and a strip of high, well-shaded south-facing windows. Renovations we encountered included lowering the ceiling, removing or obscuring some of the windows. We hoped that these variations in the classic Finger Plan classroom would provide us with sufficient variety of daylighting conditions for our analysis.

2.2 Description of Participant District

FUSD is quite a large district, with 61 elementary schools. It is the fourth largest district in California. In 2002 it enrolled 80,600 total students of which 46,000 elementary school students were in grades K-6.¹ The school district has a very diverse ethnic population, with over 100 languages spoken at home, and children born in 84 countries. Students classified as primary English speakers represent 56% of the elementary school population, and 32% are classified as learning English. The elementary school population is classified as 17% White, 56% Hispanic, 12% African American, 15% Asian and 2% other. Of these, 73% are classified as economically disadvantaged, and 10% are classified as Special Education students. Students in grades 3-6 ranked in the 30th-36th percentile in state standardized reading tests, and in the 38th-50th percentile on math tests. Thus, Fresno Unified is a very large, urbanized school district with a majority low-income population, and a substantial population of non-English speakers.

2.2.1 The City of Fresno

The city of Fresno is located in the southern portion of California's inland Central Valley, where it serves as a marketing, financial and industrial center for the region's predominantly agricultural economy. The city has a small downtown area, many industrial regions along the railway yards, and huge sprawling residential areas dating from the 1920's onward to brand new subdivisions being built in former orchards at the outer edges of the city. Indeed, Fresno has been identified as one of America's cities with the lowest population density

The climate of Fresno is considered somewhat extreme by California standards, with a very hot dry summer and winters that can be much cooler than coastal areas. Still the winter temperatures rarely drop below freezing, and almost any month in wintertime can see occasional balmy 70 degree days. The dry season, with uninterrupted blue skies and temperatures that range from the 80s into the high 100s, typically lasts for seven months from May through October.

Fresno has a high level of air pollution, and often exceeds state air quality standards. The prevalence of poor air quality is due to many contributing factors, including the long, rain-free summers, the low-lying valley location, its agricultural and industrial neighbors, and especially the heavy automobile use by the city's ever expanding population. Asthma rates in children are especially high and have been an area of concern for the California Public Health Services.

¹ Statistics are taken from reports posted on the District's website for 2002. www.fresno.k12.ca.gov

2.2.2 District Organization and Testing Protocols

Being a very large district, FUSD allows a great deal of local autonomy in administrative and curriculum decisions. The elementary schools are “neighborhood based,” with no busing or integration plans specifically designed to mix children from different neighborhoods. The District does allow transfers between schools, if the parents provide their own transportation. It also runs a few special district-wide schools, included a gifted and talented magnet school and three charter schools. A concern of the District is unstable school population, due to the large number of students that transfer between schools and in and out of the district. Many parents are migratory workers, or frequently relocate to new households due to economic conditions. The District tracks the degree of instability in its student population by school site with various indices.

With a quickly growing population, like so many other California school districts, FUSD has difficulty providing enough classrooms for all of its students. In response, about one-third of the elementary schools have been converted to “year-round” schools that can support 25% more students by running the schools all summer. Almost every school site is also provided with “portable” or “relocatable” classrooms—modular classrooms, similar to mobile homes—that can be moved between sites if needed.

The state of California requires all districts to administer standardized tests, “STAR-9” to all students to assess performance, and to report results and progress by student, school and district level demographic categories. In support of this requirement, and for on-going assessment of district performance, FUSD maintains a centralized Research and Evaluation Department (REA) that was our primary contact for this study. In addition, FUSD recently joined the Northwest Education Association (NEA) and instituted the administration of additional standardized tests which are customized to the FUSD curriculum, called ABC tests by the District, they are know more generically as RIT tests. These RIT tests are equivalent to the core curriculum tests used in the Capistrano analysis, and also form the basis of this study.

2.2.3 FUSD School Building Types

FUSD has long had a policy of neighborhood schools, and in the early years of the city, small elementary schools were located about every two miles so that children could easily walk to school. These schools were originally designed for a population of about 200-300 students. Most schools built before 1950 were later closed and rebuilt to new earthquake safety standards. Growth continued through the 1970s, then tapered off. So the district has many schools dating from the 1950s through the 1970s. After the late 1970s most population growth was met with the addition of portable classrooms to existing school sites, substantially expanding their student population, often to 600-800. Only recently has the district undertaken new permanent building projects, along with substantial remodeling of existing schools.

Like the majority of California schools of the time, the schools were planned as one-story buildings with outside walkways and large grassy playgrounds surrounded by large shade trees. Later educational policy encouraged the development of “open plan” or “pod” schools that featured interconnecting classrooms, and/or shared multi-purpose spaces. These open plan schools were typically set in the middle of a paved playground area, limiting the amount of vegetation around the buildings. Once a school plan was developed, it was often repeated a number of times, so there are a number of nearly identical schools within the District. There is only one two-story elementary school in the district, but it has been identified as the model for future school construction, primarily to reduce the space needs for new school sites.

All classrooms include some form of air conditioning. It is original in all classrooms built since the 1970s and retrofitted in earlier buildings. Many classrooms were being modernized during the study period, including painting, new lighting, carpets and new wiring. We only included classrooms in our study where the modernization was in progress before or after the testing period.

For the purposes of this project, HMG classified the FUSD elementary school classroom buildings into five basic plan types that captured the key differences in layout and daylight availability:

- **Finger Plan:** Wings one classroom wide with exterior entrances, all facing south
- **Double Loaded:** Wings of back-to-back classrooms with exterior entrances, either north or south facing
- **Grouped Plan:** Classrooms with an interior corridor, often open to one another, facing any direction
- **Pinwheel:** A variation of grouped plan with radiating classroom wings
- **Pod:** Non-orthogonal grouped classrooms, with many shared internal spaces
- **Portables:** Modular classrooms with exterior entrances, typically lined up in north or south facing rows

These plan types, and their associated classroom configurations, are described in more detail below.

Finger Plan

A finger plan school has rows of classroom buildings that are one classroom wide, with exterior entrances on the south. The long axis runs east to west, and the fingers run parallel to one another, with large windows on the north and south walls. A planting strip, typically with large shade trees, runs between each wing. This design was common in the 1950s and early 1960s.

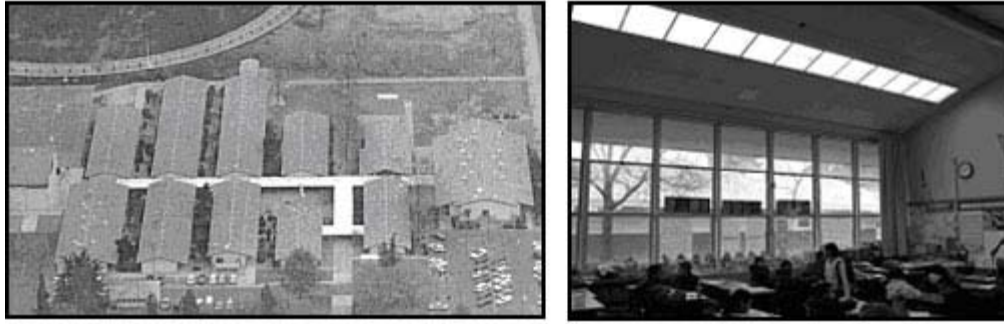


Figure 1: Aerial View and Classroom Interior of a Finger Plan School

A typical finger plan classroom has a rectangular floor plate with a high, sloped ceiling (10' to 14' height). The interior surfaces were originally wood paneled walls with hard wood flooring, but many have since been remodeled with acoustic paneling on the upper walls and carpet on the floor. The classrooms include operable, clear glass windows and blinds or shades on the inside. The south windows are well shaded by overhangs that extend over the exterior walkway on the south side.

Although originally designed with extensive windows on the north and south sides, many finger plan schools have had their window areas reduced. Often, all south-facing windows have been removed except for a single high strip. At some schools, the north-facing glazing area has also been reduced with opaque metal or insulating panels.

Double-Loaded Plan

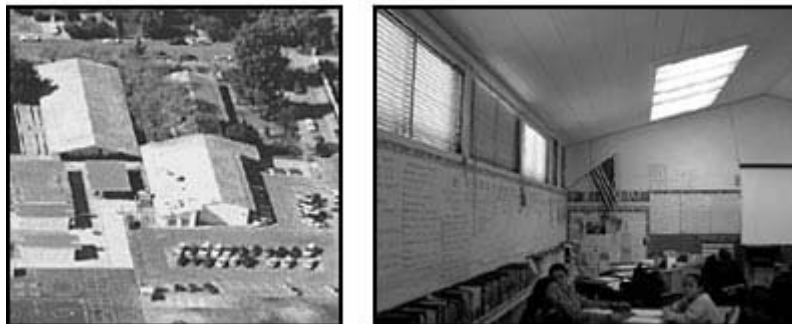


Figure 2: Aerial View and Classroom Interior of a Double Loaded School Plan

Buildings in the double loaded plan are two classrooms wide, with classrooms facing either due north or due south. The classrooms share an interior wall, and each classroom has only one exterior wall with a door and clear glass windows. Typically the windows on this one wall are smaller than the windows in the finger plan classrooms, and often include exterior shading louvers that greatly reduce both the view and the available daylight. The double-loaded classroom has a

sloped ceiling similar to the finger plan classrooms, and the two classroom types share similar construction materials and surface finishes.

Grouped Plan

This category includes various school plans where the classrooms are grouped around a central corridor or work area. Classrooms can face in any direction. These are generally of a later era than either the finger plan or double-loaded plan schools.

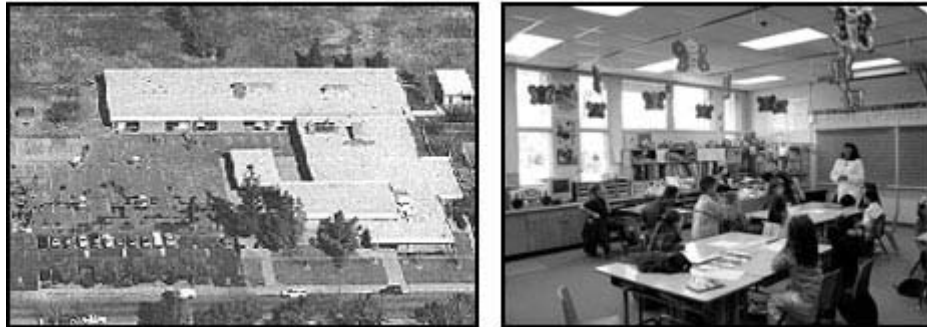


Figure 3: Aerial View and Classroom Interior of a Grouped Plan

Classrooms in a grouped plan layout typically have a rectangular floor plate, with wood frame or concrete wall construction and carpeted floors. The ceiling is typically a hung ceiling at about 10 feet high with recessed troffer luminaires. Classrooms typically have one external wall, with high strip or view windows. The area of windows varies largely depending on the school type and era built. Typically the windows are tinted, but not well shaded and rarely operable.

The classrooms share walls with other classrooms on the other three sides. In many schools, separations between classrooms are either permanently open, or can be opened with movable partitions.

Pinwheel Plan

The pinwheel plan is a variation of the grouped category. Three of these schools, built in the 1970s, are included in our study. Classrooms face in all directions, and have an exterior door that opens directly to the outside play area, and an interior door that opens into the central activity space. The classrooms are typically smaller than the finger plan or grouped classrooms, since some of the allotted classroom space was redistributed to shared “learning areas” in the central corridor.

The classrooms typically do not receive much daylight, as the only classroom windows are single, tinted but unshaded windows adjacent to the exterior doorways. The construction is either wood frame or concrete blocks, and the ceiling is sloped from 10’ to 16’ in height, with the greatest height inwards. The classrooms have a rectangular floor plate with acoustic tiles on the upper walls

and ceiling, and carpet on the floor. Electric lights are suspended fluorescent indirect fixtures.



Figure 4: Aerial View and Classroom Interior of a Pinwheel Plan

Pod Plan

The pod plan as seen in Figure 5, arose from an education theory encouraging open classrooms, where students could move between learning areas. Many of these schools have subsequently been retrofitted with acoustic partitions or permanent walls. The classrooms are typically non-orthogonal in footprint and a hipped roof covers all the classrooms. Ceiling and surface conditions vary, but most classrooms are carpeted.

Classrooms in some pod schools are arranged two-deep from the perimeter with the remaining central space used as a common area for library and other purposes. The classrooms that are along the perimeter have a small, deeply tinted window alongside a door facing the exterior. Classrooms on the interior do not have any exterior windows or any other means of getting natural light.



Figure 5: Aerial View and Classroom Interior of a Pod Plan

Portable Classrooms

Portable classrooms are prevalent in the district, comprising a total of 54% of all classrooms in this study (as compared to 42% in Capistrano). Two kinds of portables are included in this study:

Bungalows—the original portable classroom, these modular 30' X 32' classrooms were designed to mimic finger plan classrooms, with high ceilings and large windows on one side which should be, but are not always, oriented north. They are being phased out of use and were fairly rare in our sample. They are typically set at grade on wooden joists supported on movable concrete footings.



Figure 6: Bungalow Exterior View and Interior View

Portables—newer 24' x 40' portables, typically have a hung ceiling at 9', and a 4' x 8' tinted and operable window front and back. Most commonly the portables are arranged in rows facing north or south, sharing a common exterior walkway. However, due to site constraints, many also face east or west. They are typically air-conditioned from a roof- or wall-mounted unit, and set at or near grade on steel I beam. These exist at most of our school sites.

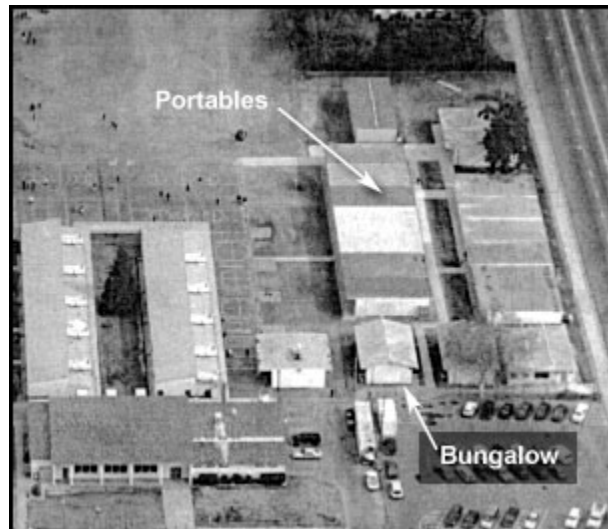


Figure 7: Aerial View of Portables and Bungalows



Figure 8: Portable External View and Internal View

3. DEMOGRAPHIC INFORMATION

The District agreed to provide us extensive data on student performance on standardized tests, student demographics and information about their associated teachers and schools. We agreed to keep all individual information confidential, and to avoid the use of any identifying information about individuals or groups.

Our first task was to understand the limitations of the testing protocol for our proposed study, and to select an appropriate study population. Once we received information from the district's databases, we needed to understand its coding so that we could select appropriate explanatory variables for our models and understand their potential meaning. This process is described below.

3.1 Testing Procedures

The District administers two sets of standardized tests: the state-mandated SAT9, administered near the end of the school year; and a RIT "level test" specifically customized to the FUSD curriculum by the Northwest Evaluation Association. The RIT tests are administered twice, in fall at the start of the school year and in the spring, near the end. We only used the RIT fall-spring tests for our study, following our experience with similar tests by the same association in the Capistrano District.

Students typically take the tests in their own classroom, unless they were absent, they were in a particular special education program, or something was wrong with their home classroom. In most schools even special education students test in the classroom. Teachers have a two-week window within which to test all of their students.

If a student missed the test-taking window, they were flagged in the data set for "retaking," and were given the test during a make up period. Other students who were found to test outside of their expected grade level, either higher or lower, were also flagged for "retaking." These students were then readministered a make-up test at their appropriate grade level. Make-up tests might be given at another location, but this was not indicated in the data set. Students who were retested were included in the study sample with a flag variable indicating that special status.

3.2 Study Population

In meeting with the District we initially identified approximately 10,000 students attending that fit our analysis profile. We included all students who met the following criteria:

- **Were in grades 3 to 6 in an elementary school.** At FUSD second graders may take the standardized tests, but are not required to do so.

Thus, we began our population with third graders. Most sixth graders are still located in elementary schools, but some 6th graders are located at middle schools. We only included sixth graders that were located in traditional elementary schools where they would have one primary classroom and one teacher.

- **Took both a fall and spring RIT test.** Since we wanted to study the difference between fall and spring RIT test scores, a student needed to have completed both tests.
- **Were assigned to the same reference classroom for both time periods.** FUSD has a high transience rate; it is fairly common for students to move between schools when their parents relocate, or to be taken out of school for a long period and then reassigned to a new classroom. We also excluded students who had been temporarily relocated to another classroom during the year to accommodate any remodeling or repair of their home classroom.
- **Were not in a multi-track program.** This is a generic term for year-round schools, where the school year is variously broken into four or five sessions. In these programs “classrooms” of children and a teacher may move to a new room at the start of any new session. Different “tracks” take different periods of the year as a “summer vacation.” Thus, we had no assurance that a child in a multi-track program would have spent the ten month study period in one physical classroom.

More of the multi-track schools are in the lower income areas of Fresno, where overcrowding of school sites is a greater problem. While dropping these multi-track schools from our study would skew the demographics of our study away from a balanced profile of the Fresno District, we agreed with the District administrators that this would not be a problem, since we were not trying to understand the Fresno District so much as the effect of daylighting in classrooms. Thus, while we needed a balance of demographics within the schools that we did study, we did not need to have a representative population of the Fresno District.

- **Spoke English with a minimal fluency.** FUSD classifies its students’ command of English on a scale of 1-6, where 1 is no English, and 6 is full fluency. We included students with a fluency level of 3 and above.
- **Were typically in their primary classroom 80% of the school day.** We removed students from the study population who were in special education classes that removed them from their “home” classroom for more than two hours a day for special programs. Special education students who remained in the study sample were flagged with a variable.
- **Were not in a charter school.** No charter schools were included in the study database, due to their non-standard curricula.

This population of students were identified to be attending 37 of the 61 elementary schools in the district, in a total of 500 classrooms.

Once the list of candidate school sites and classrooms was developed, we mapped the schools and compared their locations to district maps showing population distribution by economic status and language status. This initial check showed that our school sites were fairly evenly distributed around the district and suggested that we would have sufficient population and geographic diversity in our sample.

3.3 Student Level Information

In addition to standardized test scores for the 2001-2002 school year, the District provided us with information about each student qualified for inclusion in the study, including educational and language status, ethnic and socio-economic information.

Following is a brief summary of the type of information about student demographic characteristics available to the study. The final list of all variables used in the analysis, and their descriptive statistics, are included in the Appendix.

- **Gifted and Talented Education (GATE).** High achieving students are qualified for special GATE programs. Most students designated as GATE remain in the classroom and receive enrichment activities before, after or during school. Commonly, they make up less than 25% of a class, at times only 1 student per class. In some schools, one entire classroom may be a GATE class. One school in our study, Manchester GATE, is a special GATE school, where all students receive GATE instruction.
- **Special Education Programs:** Students with physical or educational disabilities are qualified for a variety of Special Education programs. All Special Education students' activities are governed by their IEP (Individual Education Plan) and so can be highly variable.

We excluded any special education student who was believed to be in a special program outside of the home classroom for more than 80% of the school day. The remaining special education students were grouped together with a single indicator variable.

- **English Learners (EL).** A third of district elementary students were English language learners. Divided into levels of 1-5, with a 6th level when a person is considered to graduate to Fully English Proficient (FEP) status.
- **Grade Level, Ethnicity and Gender.** All students were assigned to a grade level (3-6) and their gender and ethnic origin noted.
- **Free or Reduced Lunch.** Free or reduced lunch status was used as a proxy of economic status. Students qualify for a free or reduced lunch when their family income is below a certain poverty threshold. However, schools where more than a certain threshold (85-90%) of students qualify, the entire school receives free lunch.

- **Attendance.** We processed student attendance information into a percent attendance variable that described days present divided by days registered in district.

In addition we chose not to use other data which the district did not consider to be reliably categorized, such as number of tardies, number of suspensions or reason for absences.

3.4 Teacher Level Information

FUSD was able to provide us with information about the teachers of each student included in the study. This information was given a randomized ID and cross-matched by room number to maintain confidentiality. In the earlier Capistrano study we did not have access to teacher information and so needed to collect it directly in surveys. We believe the information included in the Fresno study to be far more accurate since it was from district records. Information provided about the teachers included:

- **Years teaching in the District.** Tenure is based on number of years employed in the district.
- **Salary level.** Salary is determined by a uniform salary schedule based on years in the district and education level (four classes – BA+30, BA+45, BA+60, BA+90), 12 or 15 steps total. Since we already had information on years in the district (above), this data became more indicative of education level.
- **Credential status.** In addition to fully credentialed teachers, we were able to note if a teacher was in pre-credential status or had been designated a mentor teacher. Fresno has very few teachers with emergency credentials, and there were none in our study population.
- **Gender and Ethnic status.** We had this information available, but did not use it in the analysis as it did not seem educationally relevant.
- **Multi-grade classroom.** A small number of teachers were assigned to teach children in a mixed grade level classroom, such as third-fourth or fifth-sixth.

3.5 School Level Demographic Information

It is generally agreed that the school environment, created by the whole population of students attending a school and their families, can affect a given individual's performance. Thus, we also strove to include indices that would describe the demographic variation in school site populations. The Research, Evaluation and Assessment Department (REA) of the District processes student and parent socioeconomic information into a number of indices for each school in the district.

This school-level data were not provided in the initial data set, but rather added at the end of the analysis period after FUSD staff suggested that our models might be strengthened by accounting for the larger socio-economic influences in a school or neighborhood. We selected the five indices described below to provide a balanced picture of the socioeconomic status of the overall population of each school in the study.

- **Student Mobility.** There is a great deal of transience between schools within the District, mostly because of farm labor families. One-third of district elementary students (with some schools as high as one-half) attend more than one school in a single school year. REA created a mobility index that expressed the ratio of the number of new students and exiting students in a school 60 days after the start of the session relative to the average enrollment of the school. This varied from a low of 104% to a high of 168%.
- **Parent Education:** Parent education level is required by the state to be reported by the teacher for SAT9 reporting. On a scale of 1=no high school degree to 5=advanced degree, parent education level was typically rated by teacher, or it might also have been self-reported by student. Thus, it is not considered a very reliable variable at the student level, but becomes more reliable as an average at the school and district level, indicating the general educational environment of the students. It varied from 1.6 to 3.9 in our sample schools.
- **English Learning.** This was based on the same information as the classification of each student's English fluency level, however, here it was processed for all the children in the school, not just those included in our study. It indicates the percentage of students who are not yet English proficient. It varied from 1% to 50% in the schools in our study.
- **Free and Reduced Lunch.** This index shows the percentage of a school's student population that is low-income and hence qualifies for a free or reduced lunch. It varied from 9% to 98% in the schools in our study.
- **CalWorks.** A final index captures the percentage of students who have a parent or guardian who is qualified for CalWorks, the state economic welfare program, another indicator of the percentage of low income families in the school. It varied from 0% to 50% in our sample schools.

This school-level data were not provided in the initial data set, but rather added at the end of the analysis period after FUSD staff indicated some concern that we needed better ways to account for the larger socio-economic influences in a school or neighborhood.

4. PHYSICAL DATA COLLECTION

In addition to all of the information about student, teacher and school demographics, the District provided us with initial information about the physical conditions of the schools and classrooms from their various facility databases. This consisted primarily of the initial construction data of a school; total school population and square footage; the number of traditional and portable classrooms; other resources on site such as libraries or multipurpose rooms; and a remodeling schedule, indicating both recent past and future remodeling schedules.

Remodeling

During the study period, including the year before and after the data collection phase, modernization was scheduled for approximately half of all FUSD schools. Typical modernization consisted of painting, removing asbestos floor tile and replacing it with carpet and vinyl, adding tack boards and white boards, adding additional electric outlets and Ethernet intranet computer connections, and modernizing lighting. This was a concern for two reasons. First, we needed to exclude any students from the study who did not occupy one classroom continuously during the study school year. However, often temporary moves were only known at the school level. Secondly, we needed to know the physical condition of the classrooms during the study period, but it was very likely that those conditions might be changed by the time we able to visit the classrooms.

Maintenance

The district did not maintain a central tracking system for the type of equipment installed in a classroom or for the overall maintenance status of each classroom. Thus, we did not have a centralized source of HVAC or lighting status for the classrooms. The District does operate an Energy Management System for most schools which records operating conditions, but these records were too inconsistent to be useful in our study.

Plan Room

FUSD maintains a plan room with all construction documents filed by school site, which we were allowed to examine. We were also given aerial photos and floor plan diagrams for each site. We used these to classify room types and orientation for each classroom. We developed a notebook with aerial photos, plans and list of classrooms for each candidate school.

On-site Visits

Once we evaluated the information available in District facility databases, we determined that it would be necessary to go on site to verify conditions in each classroom in the study. Our primary concern was to verify daylighting and lighting

conditions in each classroom. While on site, we were also able to collect considerable information about other physical characteristics of the classrooms, such as geometry, surface materials, and some limited information about the condition of other equipment.

Two Phases of Data Collection

The data collection during August 2001 is hereafter referred to as Phase 1 data collection. Information was collected on 500 classrooms in 36 schools. The information from these surveys was used to create variables for the statistical regression models of student performance. Phase 2 of on-site data collection was conducted during February of 2002. During this second phase, we went back to the district to observe a small subset of classrooms in operation. We made observations and took measurements in 40 classrooms in 14 schools. The information from Phase 2 was used to inform the understanding and interpretation of the regression models.

The methodologies used and the general observations of each data collection phase are described below.

4.2 Phase 1 Data Collection

A data collection plan was prepared in consultation with FUSD officials that outlined the data to be collected and the methodology to collect the data. Our primary limitation was a request by the district to conduct all site visits while the schools were unoccupied. This constraint limited primary data collection to the month of August 2001. Some survey schedules were specifically juggled to avoid summer school.

4.2.1 Survey Protocol

The surveyors were onsite primarily between the hours of 7:00 AM and 3:30 PM when custodial crews were available to unlock classrooms and answer questions. The survey protocol was designed to allow all data collection at each school to be completed in about two hours. Teams of two architecturally-trained surveyors made observations and took measurements for each school, its neighborhood, and the classrooms included in our study sample.

4.2.2 Survey Forms

The survey instruments were developed following interviews with FUSD facility personnel about expected conditions, brief reconnaissance visits to a dozen FUSD schools, and review and comments by the project Technical Advisory Committee. The primary goal was to isolate some of the potential mechanisms for a daylight effect, such as illumination level and control, sunlight penetration and glare potential. In addition, we attempted to collect information about other potential influences on student comfort and performance, such as ventilation, acoustics and air quality. The Advisory Committee made a strong

recommendation to look into both ventilation/air quality issues and how view quality might affect student performance. The surveys collected considerably more information than the previous PG&E study, to support both more detailed analysis and insight about the findings.

Two survey instruments were developed and are shown in the Appendix to this report. One form was created to collect school site information and a second form was created to collect information for each classroom. Definitions and protocols were developed, practiced and discussed to ensure consistent methods and interpretations by all surveyors.

Some of the more interesting challenges of our survey methodology are discussed below. A complete explanation of survey definitions and methodology is included in the Appendix.

4.2.3 Survey Challenges

We encountered a wide variety of window conditions in the classrooms and somewhat modified our survey procedures to try to capture the variety of conditions encountered that might influence either the amount of daylight or the visual ambience of the classroom. Our goal was to collect sufficient information about the daylighting conditions in each classroom to support analysis that could distinguish between different aspects of daylight quality. Thus, we collected information about window geometry, tint, shading, view, and internal and external coverings and controls. We were clearly disadvantaged by observing the windows while the classrooms were unoccupied. Some classrooms looked like they were still occupied, with bulletin boards covered with pictures and student work. Some of these classrooms had decorative curtains or paper covering the windows, clearing indicating efforts by the teacher to control light or view. Other classrooms had clearly stripped down to their bare elements, with no personal effects, so it was impossible to know how they had been operated during the previous school year which would be the subject of our study. Some of these had their windows covered over with butcher paper, just as a temporary security measure for the summer.

Daylight Characteristics

Describing daylight illumination conditions was challenging since daylight illumination is highly dynamic, while we were making observations in only one point in time. The amount and distribution of daylight in a classroom could easily be altered by such dynamic variables as:

- Daily and seasonal movement of the sun in relation to orientation of the window, shading elements, and reflections off of adjacent buildings
- Seasonal changes in shading and reflection from deciduous vegetation

Thus, we relied on architects trained in daylight assessment who could visualize the effect of such movements and changes over time. Sunlight penetration into the class and glare sources on the teacher board were also assessed by

imagining the yearly movements of the sun and how it would interact with classroom orientation.

The variety of conditions blocking windows also presented a challenge to catalog. We encountered interior blinds, curtains, windows obscured with paper or translucent glazing, windows with operable or fixed exterior louvers, and windows with security bars. We created categories for all of these conditions.

View

Given the variety of conditions encountered on site, perhaps the most challenging task was creating a consistent surveyor assessment of view. We settled on three parameters of view:

- Size of the view window, in square feet, defined as the window area between desk and door height
- Distance of the view, in three categories, defined as near (within 25 feet), mid (between 30 and 65 feet), and far (more than 70 feet away).
- Elements of the view, either including human activity, such as a view of a playground, lunch area or parking lot, or including vegetation, primarily trees or bushes somewhere in the field of view.

Daylight Illumination Levels

Illumination readings were taken with both the electric lights on and off at a number of set locations to determine current daylight and electric illumination levels. However, since these values would change over time, this was not considered a particularly reliable assessment of daylight illumination conditions.

Electric Lighting

The FUSD classrooms had a variety of electric lighting conditions, including luminaire types, fluorescent lamp types and color, and condition of the luminaires. Electric illumination levels generally met Illuminating Engineering Society of North America (IESNA) recommendations, providing fairly uniform illumination on walls and desks with about 20 to 50 foot candles. We determined electric light levels by subtracting light levels readings with the lights off from those taken with the lights on. Lights were allowed to warm up for 10 minutes before readings were taken. We also counted the number of fixtures and working lamps and did lumen method calculations to verify our findings.

HVAC System

The surveyors determined whether the classroom was part of a central HVAC system or it had its own HVAC unit, both by observation and by interviewing the site contact. They checked the delivery system of both heating and cooling (roof or wall) and determined the type of use of the HVAC controls, if any. The fan was turned on to check for operation and noise level. If local control of the fan was not available, that was noted.

Ventilation and Air Quality

Since the classrooms were unoccupied and typically closed up for the summer, it was difficult to assess air quality conditions that might have prevailed while school was in session. The surveyors did note any conditions that were obvious at the time of survey. Upon entering the classroom for the first time, surveyors noted if the air seemed especially stale or musty. A number of the classroom carpets were being shampooed, and often those classrooms were especially musty if they had been closed up before the shampoo had fully dried. Whenever possible, we noted if the ventilation fan was working, and under the teacher's control. We noted the presence of any portable fans in the classrooms, pet cages and any visible presence of mold or water damage. We examined the foundations of portable classrooms and noted any presence of rodent tracks or holes (where ground squirrels are especially likely to take up residence).

Acoustic Conditions

Surveyors were able to easily note the presence of sound absorbing materials, such as acoustic tile and carpets, and sound transmission opportunities, such as operable windows or lack of doors between classrooms. We were also able to turn on the ventilation system and lighting system to note the presence of especially loud equipment noise or an annoying hum from the lights. We also noted the presence of obvious load noise sources outside, such as a nearby freeway, airport, railroad, or construction site. It was more difficult to assess the transmission of noise through walls, or intermittent sources of noise such as PA systems or activities within the classroom.

4.2.4 Phase 1 Data Collection Limitations

The Phase 1 data collection was greatly limited by using only one-time measurements taken during an unoccupied period. It was done during the summer when the schools were not in session due to the project scheduling constraints. As a result the surveyors were forced to make many assumptions. The teachers were assumed to be using blinds/curtains and local HVAC control to optimize their comfort in the classroom. Where we observed any actions to block daylight or prevent local control of HVAC, such as cover plates over thermostats, these overrides were assumed to be present during the school year. (We later learned that many teachers had discovered ways to override centrally controlled HVAC systems, by jiggling wires or using a screw driver to change settings.)

Many of our observations might have been influenced by conditions of a classroom closed up for the summer. For example, the surveyor assessment of "stale air" in the classroom was likely due to HVAC systems that had not actively been operated for a number of weeks. Surveyor assessment of "musty/moldy" air was possibly due to recent carpet cleaning that had introduced moisture into a warm, closed up classroom. It was impossible to know how teachers operated

their windows or blinds during the school year based on the conditions of the windows we observed during the summer.

In our analysis based on the Phase 1 survey data, we assumed that the classrooms were being operated basically as observed during the summer. However, in order to understand our regression findings, we wanted more information about how the classrooms were actually operated and the kinds of decisions teachers were making about the environmental controls available in their classrooms. We therefore decided to do a follow-on study to analyze a sub-sample of the classrooms during a period when the classrooms were occupied. This effort is described below as Phase 2 Data Collection.

4.3 Phase 2 Data Collection

The Phase 2 data collection effort was conducted in February 2003 to observe the classrooms during normal school hours. The intent was to overcome some of the limitations of the Phase 1 analysis by observing the classrooms in operation. This would enable us to see how teachers modified their classrooms during the school year and make some observations about how they actually used the classroom controls available to them. We also took this opportunity to survey a subset of the teachers about their attitudes and actions, and to take some on-site measurements of comfort conditions while the class was operating, such as illumination, temperature and sound levels.

The Phase 2 data collection effort has the limitation of small samples providing only a snap shot of information at one point in time. However, taken together the August and February data collection efforts complement each other and help to give a more complete picture of classroom conditions in the District. The Phase 2 data also allows us to make some predictions about the accuracy of collecting data in unoccupied classrooms.

4.3.1 Selection of Sample Classrooms

The site survey team had a limited time window to carry out the school observations. The total number of classrooms that could be surveyed in the given time was estimated to be 12 schools (one-third of 36 schools surveyed in Phase 1) and 24 classrooms (about 5% of the classrooms of the 500 classrooms visited in Phase 1).

We selected schools from the Phase 1 sample that provided the greatest diversity of classroom types, while also making sure that we had representatives of at least two examples of each daylight classification level (*Daylight Code* described in Section 5.2) in different neighborhood economic conditions.

Ultimately, we visited and collected data on 40 classrooms in 14 schools, and collected survey forms from 114 teachers. We conducted informal interviews with about twenty five teachers.

4.3.2 Survey Protocol

After scheduling the classroom visits, and reviewing procedures with the school administrators, we verified on site that the classrooms in our data base were still occupied by 3rd- 6th grade classes. In some cases, they were empty or occupied by the wrong grade level. In those cases, we attempted to identify a nearly identical classroom with the correct grade level.

The surveys were conducted by groups of two surveyors. One surveyor was constant in all surveys and had also led one of the August survey teams, in order to provide continuity across datasets.

4.3.3 Classroom Observations

In each of the classrooms, the surveyors briefly introduced themselves to the teacher and explained the nature of their observations. The surveyors conducted their observations in the back of the classroom quietly for 10-15 minutes, noting the operation of windows, lights, mechanical system, and making subjective assessments of the acoustic, thermal and lighting environment. The surveyors also carried a notebook with information previously recorded about the classroom in Phase 1 such as dimensions, surface materials, etc. and confirmed the accuracy of these observations. In certain cases, the surveyors demonstrated their tools and explained their observations to the students at the request of the teacher.

The primary intent of the onsite observations was to observe the classroom operational status and receive feedback from the teachers on the classroom comfort conditions. A secondary objective was to confirm the information gathered during the Phase 1 surveys.

The surveyors confirmed the classroom type, grade level and the name of the teacher. The surveyors then checked the classroom geometry and construction information previously noted. The percentages of different classroom surface materials, such as paper and acoustic tiles, were recorded for each surface. Also, the window operation and window coverings were recorded in detail.

The surveyors noted the state of the HVAC and lighting controls, with special attention to the level of local controls available to the teachers. In many classrooms with no local HVAC controls, teachers had found ways to “cheat” the system to turn them off when needed. Apart from these quantitative measurements, the surveyors also recorded a subjective rating of the classrooms for their air quality, water damage, thermal comfort, and acoustic comfort on standard rating scalars.

4.3.4 Classroom Measurements

Apart from the observational data collected above, the surveyors also took quantitative measurements to assess the environmental conditions in the classrooms. The surveyor's carried a handheld illuminance meter that recorded the ambient light levels (both horizontal and vertical) in the classroom at various

locations. Similar to the lighting measurements, the surveyors recorded the thermal comfort of the classrooms with two tools – a digital thermometer to record the ambient air temperatures in the space and an infrared thermometer to measure the radiant or surface temperatures of the various surfaces in the classroom. Surveyors also recorded the acoustic decibel levels in the classroom while the class was in session, and then again during an unoccupied period to estimate the amount of noise present in the classroom, as well as sound penetration from other sources. A handheld decibel level meter was used for this purpose.

During a time period when the classroom was not occupied (typically during recess), the surveyors also measured the daylight and lighting levels (using a handheld Minolta light meter), acoustic decibel levels (using a handheld decibel meter), carbon dioxide levels in the air (using a handheld CO₂ sensor) and radiant temperatures of various surfaces in the classroom (using a handheld radiant temperature “gun”, which uses a laser beam, similar to a presentation pointer, to assess the radiant temperature of surfaces).

4.3.5 Teacher Interview

Wherever it was convenient, the surveyors interviewed the classroom teachers about their experience of the lighting, thermal, ventilation and acoustic conditions in the classrooms. The teachers were very enthusiastic in their responses and provided important insights into the operation of classrooms. The teachers also gave their opinions on the positive and negative aspects of their classrooms, especially comfort complaints and their impact on the students. Insight gleaned from these interviews later helped inform our interpretation of the statistical findings.

4.3.6 Teacher Questionnaire

HMG prepared a two page questionnaire comprised of multiple choice questions aimed at understanding teachers opinions on classroom comfort and how they interact with the various controls and amenities in the classrooms. This questionnaire was distributed to all 3rd-6th grade teachers in each school visited via the school secretary. A self addressed stamped envelope was left at each school for returning the completed questionnaires.

5. DATA PROCESSING AND VARIABLE DEFINITION

The following sections describe the methodology that was used to process the data collected during onsite surveys and the data received from the district. The sections also identify the variables that were used for statistical models and analysis.

5.1 Data Entry and Quality Control

The data from the site visits were collected on paper survey forms, then entered into electronic databases, with standard error bounds testing and validation features. The data were checked and processed within Microsoft Access, and then transferred into SAS for statistical analysis. All of the site data was examined to make sure that it was reliable and provided a sufficient range of conditions for useful analysis.

Some classrooms were included in the final demographic dataset, but had not been visited on-site. When ever possible, we matched these classrooms to a similar classroom which had been surveyed in the same school. Calls to the school administration and photographs of the school were used to confirm these judgments.

Eventually, of the 500 classrooms for which HMG collected on-site data, there are 45 classrooms that are not included in the analysis described in later sections of this report. The primary reason for dropping surveyed classrooms was an inability to match (or map) the room numbers observed on-site with the room numbers associated with the student and teacher data provided by FUSD.

5.2 Final Study Population

	Math	Reading
Total student records received from FUSD	10423	10423
Disqualified or incomplete records	-929	-1120
Records not mapped to surveyed classroom	-731	-721
Students in renovated or moved classroom	-245	-172
Final study population	8518	8410

Figure 9: Final Study Population

We received data for a total of 10, 483 students. About 10% of these records were removed because they were incomplete or did not fully match our criteria for qualification in the study. Another group, about 7%, was removed because we could not successfully match them with surveyed classroom data, typically because of different room naming methods in the various databases. Finally,

another 2% were removed after we determined they had been temporarily moved during the study period, typically because their classroom underwent some renovation or repairs. These changes to the study population are detailed in Figure 9.

5.3 Assignment of *Daylight Code*

Since one of the objectives of the project was to attempt a replication of the Capistrano study, it was necessary to assign a *Daylight Code* to each classroom following the previous methodology. The *Daylight Code* was assigned after all of the classrooms have been surveyed in order to understand the balance between highest and lowest conditions observed.

The primary criteria for the code were as follows:

Daylight Code 5:	Even and balanced daylight allowing operation of classroom without electric lights for a large portion of the school year. This might translate to approximately 45-75% potential electric lighting savings during daylight hours.
Daylight Code 4:	More asymmetrical daylight allowing operation of classroom without electric lights occasionally in all or frequently in parts of the classroom. This might translate to approximately 20-40% potential electric lighting savings.
Daylight Code 3:	Daylight in part of the classroom, which would allow occasional turning off of part of the electric lights. This might translate to approximately 5-15% potential electric lighting savings.
Daylight Code 2:	Some daylight in classroom, but insufficient for normal operation without electric lights.
Daylight Code 1:	Minimal daylight.
Daylight Code 0:	No daylight in classroom.

Figure 10: Daylight Code Criteria

The *Daylight Code* is meant to be a simple and crude, but holistic, assessment of the amount of daylight available in a classroom over the course of the school year. Two daylight experts worked together to assign the *Daylight Code* to classrooms following a similar method to that used in the Capistrano study, but with a much higher level of information available to make judgments about classrooms. While the Capistrano study was based on review of plans and site visits to a sub-sample of the schools, the Fresno assignments were made with surveyed information from all classrooms, including area, tint and shading of windows, orientation, classroom proportions and measure illumination levels. In addition to information from the surveys, photographs and plans of the

classrooms were also available for a cross check. Classrooms that seemed to fit between two criteria were assigned a half code rating, such as 2.5 or 4.5.

Below an example of a classroom typifying each *Daylight Code* is presented with a photograph and brief description.



*Figure 11: Photos of Daylight Code 5 Classrooms
North Windows (left) and South Windows (right)*

In application, *Daylight Code 5* was applied to Finger Plan classrooms that were in well maintained and original condition, with high ceilings, fully shaded south clerestory windows, and continuous desk-to-ceiling clear windows to the north. These classrooms were likely to have 12' to 14' ceilings.



*Figure 12: Photos of Daylight Code 4 Classrooms
North Windows (left) and South Windows (right)*

Daylight Code 4 was typically applied to finger plan classrooms that had been renovated with lowered ceilings and/or reduced window area to the south and/or north. *Daylight Code 4* classrooms had more asymmetric daylight distribution than *Daylight Code 5* classrooms.



Figure 13: Photo of Daylight Code 3 Classroom

Daylight Code 3 was applied to double loaded classrooms that have large windows with modest to high visible light transmittance. These classrooms typically had ten foot ceilings. A few portable classrooms with excellent north-south orientation and nearby reflective surfaces that would increase the amount of reflected daylight in the classroom were also considered *Daylight Code 3*.



Figure 14: Photo of Daylight Code 2 Classroom

Daylight Code 2 was applied to most portable classrooms and traditional classrooms with modest view windows and heavily tinted glass. These classrooms typically had low 9' ceilings and one 4' x 8' window on either side of the room.



Figure 15: Photo of Daylight Code 1 Classroom

Daylight Code 1 was applied to any classroom with one small window area, typically next to an exterior door. Some *Daylight Code 1* classrooms had high narrow strip windows above the door height which provided little daylight or view. *Daylight Code 1.5* was also assigned to a few portable classrooms that were heavily shaded by nearby structures.



Figure 16: Photo of Daylight Code 0 Classroom

Daylight Code 0 was applied to any classroom with no windows at all. There were six of these in the study, in the interior of two open-plan schools.

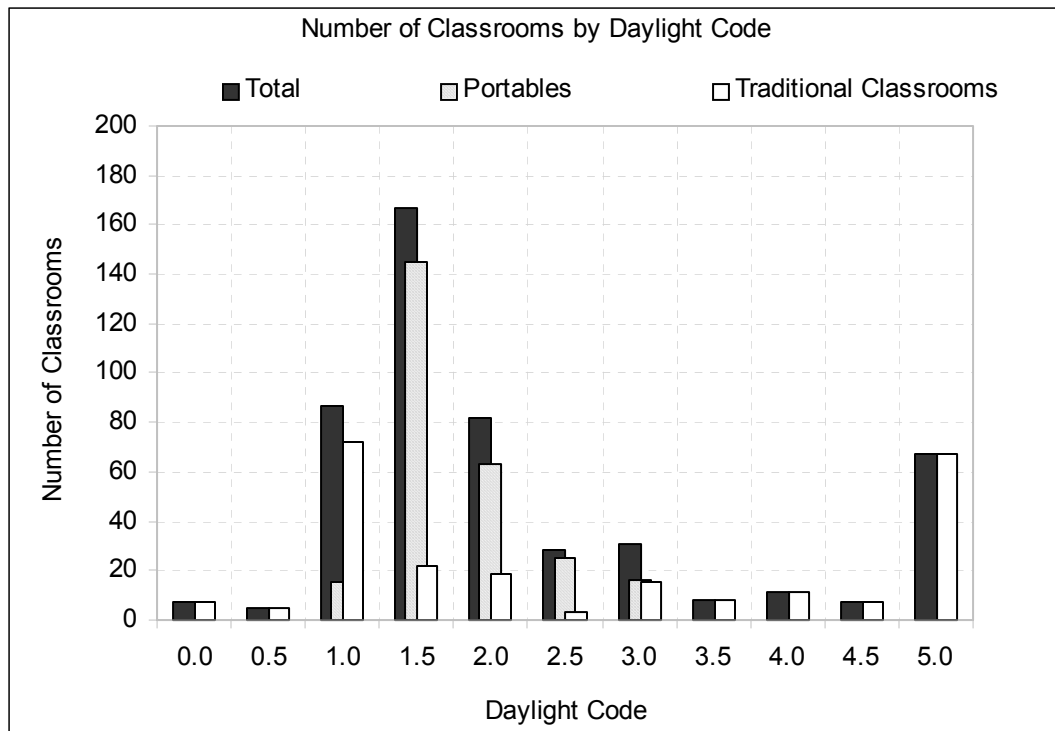


Figure 17: Distribution of Classrooms by Daylight Code

Figure 17 shows the distribution of classrooms by their *Daylight Code*. The vast majority of classrooms were classified in the lower end of the scale, from 1 to 2. Most of these were portable classrooms, which constituted 54% of our dataset.

We ended up with far fewer high *Daylight Code* classrooms than we had hoped from our initial survey. There seemed to be at least two reasons for this. First of all, younger children seem to be given preference for traditional classrooms in FUSD, so our population of 3rd to 6th graders was more likely to be assigned to the portables at a school than the traditional classrooms. Secondly, many finger plan classrooms were under renovation, and so were removed from our dataset.

We also ended up with less diversity in *Daylight Code* classrooms than we had hoped. There was a very small number of classrooms that were classified from *Daylight Code* 3 or 4, and only six with no windows at all given *Daylight Code* 0. Thus, the final data set did not present as wide or diversified range of daylight conditions as we had found in Capistrano.

Once the *Daylight Code* was assigned, we attempted to develop an equation that would predict the *Daylight Code* from information about the windows or survey readings. Given that the illumination readings were taken at different times during the day, and that the classrooms had a variety of orientations, the survey illumination readings were a poor predictor of the *Daylight Code*. Of all the illumination measurements taken, we found that the average of the three horizontal readings was the most stable predictor of the *Daylight Code*, but even this was not a reliable predictor of the assigned *Code*. We concluded that it

would be best to test individual components of the windows and classrooms against the *Daylight Code* in the regression analysis, to see which was a more precise predictor of student performance.

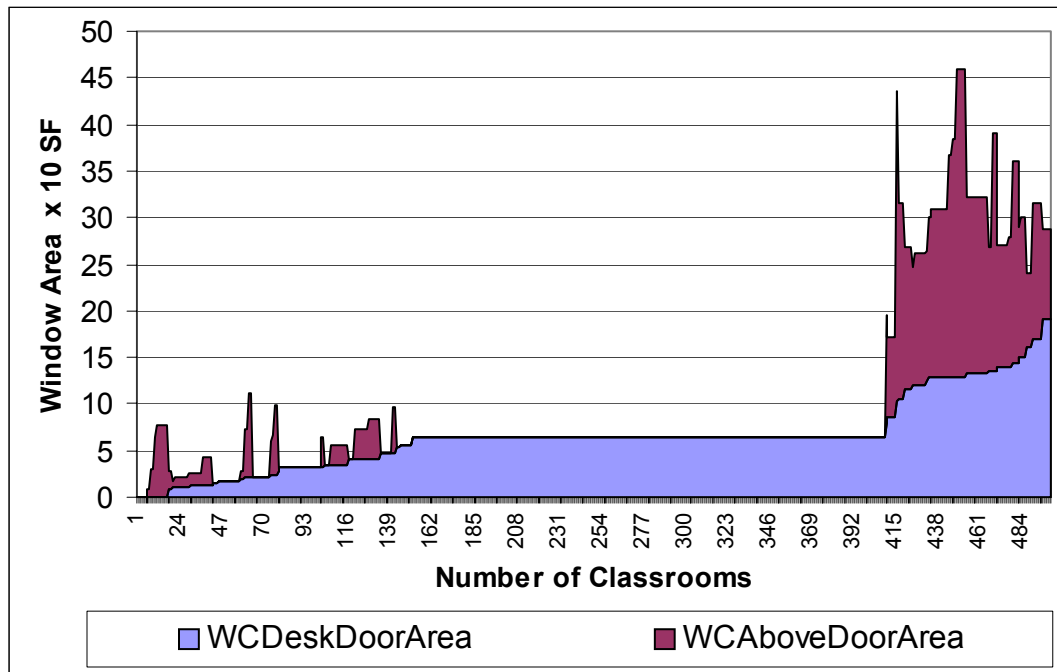


Figure 18: Distribution of Window Areas

Figure 18 illustrates the distribution of window area by the number of classrooms with each condition. The window area is broken into two types, view window (WCDeskDoorArea) and high windows located above door height (WCAboveDoorArea). The data is sorted by view area. This graph shows that at low *Daylight Codes* there is a combination of view and high window area, and that at high daylight codes there is almost a 50/50 balance between the two types of windows. The long flat stretch in the middle of the graph are all the portable classrooms in the study, with nearly identical 4' x 8' windows at the two ends of the classroom.

5.4 Definition of Analysis Variables

The onsite data collected were distilled into meaningful variables. Some information was grouped together. Some variables were observed to have too little variation to be significant in analysis, and so were dropped. For example, almost all classrooms had a phone, so we dropped that indicator variable.

There are seldom obvious ways to transform continuous data into variables for analysis. Our world rarely comes in discrete parts, but for analysis we need to describe the characteristics of that world in some mathematical fashion. Typically we describe information by what we can easily measure, or by names that we have all agreed have some common meaning. In the case of these

studies looking at how the physical environment affects student performance, we do not necessarily know the correct way to define the data into variables. This study is largely an effort to try to understand what is significant about the physical environment, and how to best measure or describe that so that it can be more precisely related to outcomes.

It is always an estimate on our part as to how those component parts are best measured and described. For example, is “sun penetration” best measured or estimated by the surveyor? Should it be defined as a simple yes/no variable or a categorical scalar variable, such as bad to good on a scale of 1-10? For dynamic conditions, like sun penetration, measurement on-site would not describe the severity of the condition over the course of the school year. In general, if characteristics could not be quickly and reliably measured on site, we relied on surveyors to make educated judgments about the severity of problems and rate them on a simple scale of 0-3 or 1-5, and accepted these judgments as a categorical scalar variable.

Multiple dimensions of any characteristic can theoretically be combined into an “index variable,” if one actually understands the relative influence of various components of the information. However, without good information to back up our relationship of the component parts, an index variable is likely to mask our understanding by blending too much information together.

The *Daylight Code* is essentially an index variable, attempting to capture multiple dimensions of the window and lighting conditions in a classroom in a single scalar. While it was developed and applied by daylight experts, it is still quite a crude assessment of the lighting and window conditions in a classroom. In this study, we wanted to understand the components of a daylighting or window index in greater detail, and so we have preserved as much information about the window characteristics as possible. We also retained the *Daylight Code* as a test variable, to see if there was some additional information better captured in this holistic assessment

5.5 Selection of Analysis Variables

Approximately 150 explanatory variables and two dependent variables were ultimately defined and considered at some point in the preliminary analysis. This group was then refined and reduced to 15 demographic variables which formed a base demographic model, and about 80 variables which described various physical conditions of the schools and classrooms. The demographic variables have been described earlier, in Sections 3.3 through 3.5. Here we describe the physical variables derived from the survey data. They are grouped by issue category for convenience, but many variables apply to more than one category.

School Site Characteristics

- ♦ School Age (number): age of school based on year 2000, ranging from 20-60 years
- ♦ Student population (number): number of students per school

- ♦ Location (yes/no): four conditions of location of school – near freeway or airport flypath, near agriculture, near boulevard, or near construction site
- ♦ Neighborhood type (yes/no): Three types – residential, commercial, industrial
- ♦ Neighborhood vintage (yes/no): four conditions of age of neighborhood – prewar vintage (1900-1940), 40s/50s vintage, 60s/70s vintage, 80s/90s vintage
- ♦ Neighborhood economic status (yes/no): three conditions – lower, mid or upper/affluent economic status
- ♦ School conditions (scalar 1–5): five variables based on surveyor subjective rating of status of site construction, paint, grass, asphalt, and trees

Window and Daylight characteristics

- ♦ Daylight Code (scalar 0–5): 0=no daylight, 5= maximum condition. *Daylight Code* assignment discussed in earlier Section 0
- ♦ Window orientation (yes/no): five orientations of windows – primary window facing east, west, north, south, or no window
- ♦ Window area (number): two conditions – area of view window between desk to door, and high window area, higher than door.
- ♦ Window tint (scalar 0–2): 0=clear glass, 1= slight tint, 2= heavy tint (VLT<.40)
- ♦ Sun penetration (scalar 0–4): surveyor estimate of amount of sun entering the classroom over the course of the school year; from “never to major problem”
- ♦ Glare (scalar 0–4): surveyor estimate of potential glare on teaching wall from windows over the course of the school year; from “never to major problem”
- ♦ Window view (scalar 0–3): 0=no view, 1=near view (<25’), 2=mid view, 3=far view (70’+)
- ♦ View Quality: (yes/no): two categories – either vegetation or human activity
- ♦ Security measures on windows (yes/no): Bars, mesh or lexan on windows
- ♦ No blinds or curtains (yes/no): blinds or curtains not available at the windows
- ♦ Operable windows (yes/no): operable windows
- ♦ Exterior doors (yes/no): two yes/no variables – no exterior doors or two exterior doors.

Classroom Characteristics

- ♦ Classroom size (scalar 1–3): 1=<950 sf, 2=950-970 sf, 3=>970 sf
- ♦ Classroom type (yes/no): seven classroom types based on classroom layout: single loaded, double loaded, interior corridor, no doors (open passageway), operable walls, common room, portables
- ♦ Teaching board type (yes/no): three types – black board, white board or green board
- ♦ Amenities (yes/no): four types – presence of sink, built in storage, internal bathroom, phone
- ♦ Equipment (yes/no): three types – presence of TV, aquarium or pet cages
- ♦ Computer (number): number of computers in classrooms

Indoor Air Quality

- ♦ Floor type (yes/no): two conditions of floors- slab on grade and wood at grade
- ♦ Room indoor air condition (yes/no): five types – room having stale air, musty/moldy air, water damage, rodents observed under portables, new condition in classrooms
- ♦ HVAC systems (yes/no): two types – central or unit system, wall or roof mounted
- ♦ HVAC controls (yes/no): Thermostat controls accessible in classrooms
- ♦ No teacher control of fan (yes/no): teacher does not have control of HVAC ventilation fan in classroom
- ♦ Portable fan (yes/no): presence of portable fan in the room
- ♦ Percentage of flooring (percent): percent of carpet on floor, versus hardwood or vinyl

Noise

- ♦ Ballast hum (yes/no): noisy ballast hum
- ♦ Loud HVAC (yes/no): noisy HVAC system
- ♦ Percentage acoustic wall (percent): % of acoustic wall tile surface for classroom

Electric light

- ♦ Indirect luminaire (yes/no): indirect or direct/indirect, or other
- ♦ Luminaire condition (scalar 0–3): surveyor subjective rating of condition of luminaire, 0= deteriorated, 1=aged, 2=average 3=good/brand new
- ♦ Ballast type (yes/no): whether electronic ballast or not
- ♦ Lamp color (yes/no): four types of color conditions – lamp color <3500 °K, 3500 °K, or >3500 °K, or mixed fluorescent (mixture of various colors of fluorescent lamps)
- ♦ Electric illuminance (number): average horizontal electric illuminance, computed from three readings in classroom at 4' above floor at center of room, and 5' from exterior wall and opposite wall
- ♦ Lamp type (yes/no): T8 lamps or T12 lamps

6. STATISTICAL METHODOLOGY

The heart of this study was the statistical analysis of the data collected. This analysis entailed developing statistical models that sought to explain how physical conditions at schools might affect student learning rates. The statistical methodology involved various stages of preliminary investigations before arriving at a final comprehensive model.

Thus, given this holistic approach, the number of variables we wished to consider as explanatory variables was very large. We necessarily had to follow a rather complex process to hone the variables down to those which would best capture the relevant influences without overstating the case or unduly complicating the models. We still ended up with very complex models. The Fresno District did not lend itself to simple explanations. We considered this an investigative procedure, testing each variable for consistency and explanatory power.

The discussion in this section explains the statistical standards and methodology employed, and the many steps taken towards developing a final model.

6.1 Modeling Standards

All of the analysis was pursued using multivariate regression models run in SAS using a variant of backwards step-wise regression to eliminate the least significant variables. The analysis used $p \leq 0.10$ as the threshold criteria for inclusion of explanatory variables in the models¹. Our prejudice was to include as many control variables as possible so that we would be less likely to inflate the effect of the variables of interest.

There are 3 stepwise variable selection procedures that are often employed in linear regression: forward selection, stepwise selection, and backward elimination. The forward selection procedure starts with an equation that contains only the constant term and successively adds explanatory variables one-by-one, until the last variable added to the model is insignificant. Stepwise selection is essentially a forward stepwise procedure, with the exception that at each iteration, the possibility of deleting a variable is also considered.

The backward elimination method first calls for fitting a model using all potential explanatory variables and calculating the t-statistic associated with each variable. The explanatory variables are then deleted from the model one-by-one, until all variables remaining in the model are associated with a significant t-statistic. During each iteration, the variable with the least explanatory power is identified and deleted from the model.

¹ See Appendix section 8.3 for an explanation of “p-values”.

The RLW variable selection method,¹ used in this study, is a variant of the backward elimination method. Similar to the backward elimination method, the RLW variable selection method begins with calculating a model using all potential explanatory variables and the associated t-statistics. However, the RLW method allows for the deletion of multiple variables during each iteration, whereas the backward elimination method does not. This procedure helps to identify co-linearities between insignificant variables, which might otherwise be dropped without first understanding how such co-linearities could potentially influence results. Specifically, the RLW method consists of the following steps:

1. Calculate a “full” linear regression model including all potential explanatory variables.
2. Identify all insignificant variables from the model resulting from step 1.
3. Perform an F-test to test whether the set of individually insignificant variables are statistically significant as a group. Specifically, the null hypothesis of the F-test is that the beta coefficients of each of the variables in the group are zero, while the alternative hypothesis is that there is at least one variable in the group whose beta coefficient is not zero. If the F-test shows the set of variables are not statistically significant as a group, all variables identified in step 2 are also identified for deletion. If the set of variables tested is statistically significant as a group, this indicates a collinear relationship between the variables is affecting the model. In this case, a reduced set of variables is defined for the F-test and deletion from the model.
4. Calculate a reduced model including all explanatory variables that were not identified for deletion.
5. If any previously significant variables become insignificant in the reduced model, calculate an F-test for all variables previously deleted from the model and the newly insignificant variables under the guidelines provided in step 3.

Regression models try to fit lines that best describe a plot of data points. Multivariate models consider more than one dimension at once. Linear models try to fit straight lines through the data. It is also possible, but far more complex, to consider curved, or non-linear, relationships. In this study, we restricted ourselves to simple linear relationships.

Models were judged based on their R^2 , the parsimony (minimum explanatory variables for maximum explanatory power), and consistency of explanatory variables between the two models, math and reading. Whenever possible we used variables based on measured data, and preferred to have more variables describing discrete bits of information rather than forming index variables that combined multiple dimensions of classrooms or school characteristics.

¹ The RLW variable selection methodology was developed by Dr. Roger Wright, lead statistician of this study.

6.2 Preliminary Investigations

Figure 19 diagrams the methodology taken in developing a final model. We followed a series of discrete steps that allowed us to observe the behavior of variables relative to each other. First we created a stable base model that used all the available demographic information about students and teachers to predict student progress on the ABC tests. Once we felt that this model was doing the best job possible of modeling student performance, we then used it as a base to form a replication model designed to approximate the same set of variables used in the earlier Capistrano study.

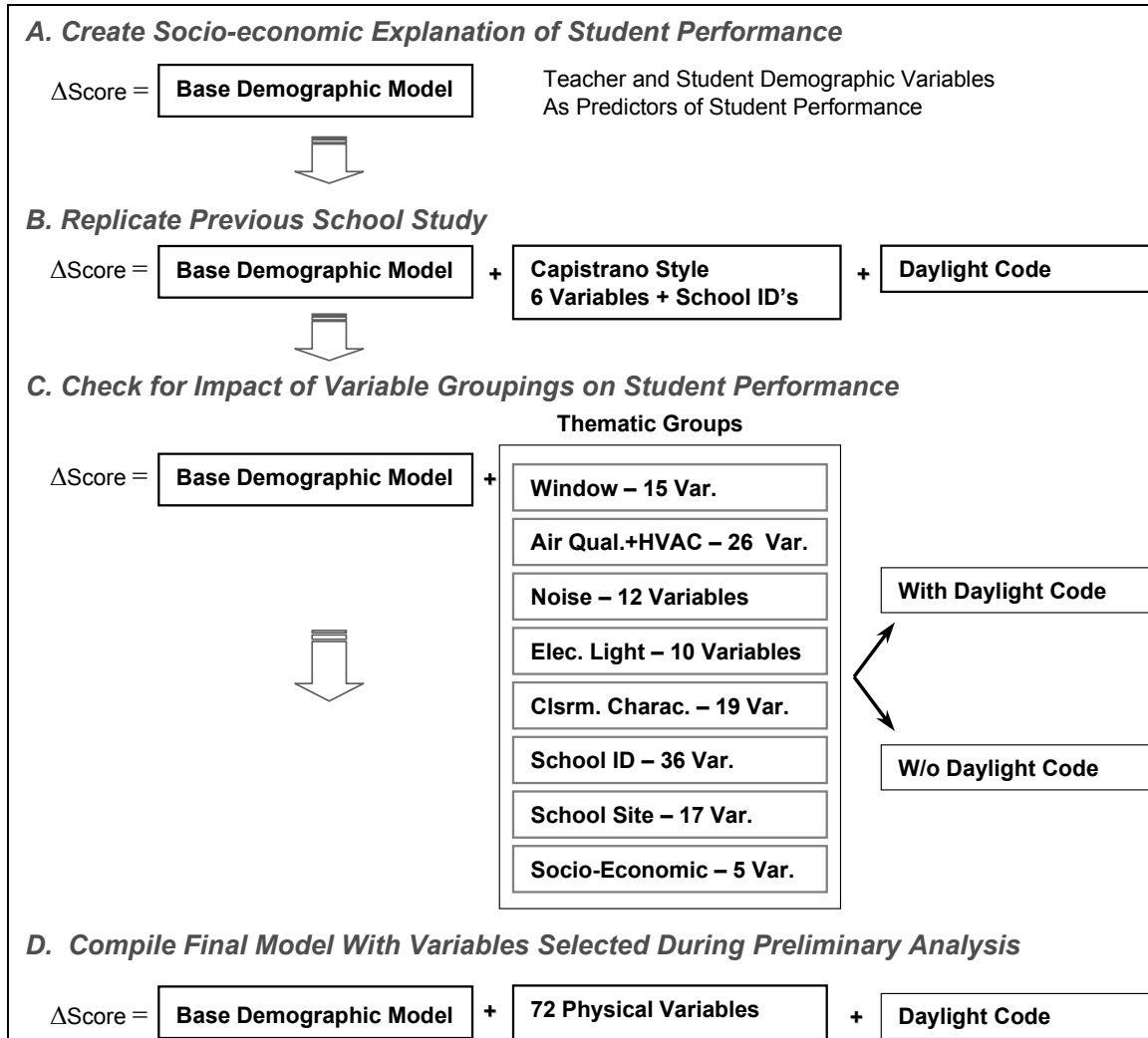


Figure 19: Methodology Flowchart

Our next step was to test a set of thematic groups of the school and classroom characteristic variables using all the detail available in the Fresno school survey database. Investigations with the thematic groups helped us to refine our final models. In addition we also ran a separate set of regressions using the *Daylight Code* as the dependant, or outcome, variable. This approach helped us to identify potentially troublesome collinear variables and investigate them in more

detail. Our final step was to develop some preliminary final models and review them with our contacts in the Fresno district for interpretation and comment. This process resulted in adding a few variables to the final models.

Each of these steps, and the information learned from them, are described in more detail below.

6.2.1 Defining the Base Demographic Model

We began by creating a base demographic model that would explain the progress in student test scores using only student and teacher level information. This is similar to the approach taken by most educational researchers in trying to understand the influence of personal characteristics, such as student English language proficiency, on learning. This model would then serve as the base against which we would later test the physical variables of interest.

A preliminary model was created to identify any outliers. In this step, all demographic variables as well as a subset of key classroom physical characteristics variables were included in the calculation of the model. Any students whose standardized residual of the predicted value was large were deemed to be outliers. The typical outlier had a gain (or loss) from fall to spring which was three to four times the standard deviation for the dataset. Outliers were kept in the models throughout the process, identified with their own indicator variables that absorbed all of the effect of that particular student.

Once all the outliers were identified and declared, a base demographic model was created that included any statistically significant demographic variables and indicator variables for the outliers. At this stage, only demographic variables were allowed to compete for inclusion in the model. Both the reading and math models were calculated using both the RLW method, described earlier, and forward variable selection techniques as a way of verifying and validating the stability and robustness of the models.

Ten student characteristics and five teacher characteristics were consistently significant in both the reading and math models. This model proved to be highly stable in subsequent tests.

Variables Selected for Base demographic model

Teacher characteristics

- Annual salary of teachers per \$1000 (salary number): based on years in the district and education level
- Number of years at FUSD (number)
- Mentor teacher (yes/no): teacher identified for special leadership role
- Pre-tenure teacher (yes/no): has been at FUSD less than three years
- Multi-grade classroom (yes/no): teacher responsible for more than one grade level.

Student characteristics

- Grade level (yes/no): student grade level from 3rd to 6th grade
- Percentage attendance (number): number of days attended versus number of days enrolled
- Qualified for/enrolled in GATE (yes/no): students designated as GATE (Gifted and Talented Education)
- Special Education student (yes/no): all Special Education categories included in the data were grouped together.
- English Language development level (yes/no): from 3 = modest level, to 6= full proficiency
- Free lunch (yes/no): Lowest income students
- Reduced lunch (yes/no): Low income student
- Non-standard living situation (yes/no): homeless, foster care, group care, or other non-nuclear family situations
- Student gender (yes/no): a flag variable indicating gender of student
- Ethnic student (yes/no): including seven classifications

6.2.2 Replication Model

Once the demographic model was complete, we created a set of physical variables that were similar to those used in the previous study, to see if we could replicate the results of the previous study. This model contained a limited set of variables which were similar to those used in the Capistrano analysis such as Daylight Code, operable windows, portable classroom, open classroom, room area, school population and school age. In these initial runs, the FUSD daylighting code did **not** show any significance and the model R^2 value remained almost unchanged from the base demographic model.

The *Daylight Code* had the least explanatory power of all variables considered, and lowest significance level. For math, the estimated beta coefficient was positive at $B=0.02$, with a significance level of $p=0.82$. For reading, the estimated beta coefficient was positive at $B=0.01$ significance level of $p=0.90$. The model was basically telling us that the *Daylight Code* as defined was not a useful predictor of student performance in the Fresno District, considering only those variables that were included in the Capistrano model.

Even though the findings of this replication model did not support the hypothesis that daylight has a positive influence on student learning, we decided to proceed with our analysis to see if we could learn anything more specific about the mechanisms of school design on student performance, and perhaps why the *Daylight Code* was not significant in this simple model as it had been in Capistrano, Seattle and Fort Collins.

6.2.3 Comparison of R² between Models of Two Districts

As part of the preliminary model investigations, we also considered the option of using the fall test score as an explanatory variable to predict the amount of progress made from fall to spring. This approach basically accounts for the “learning curve,” or the common experience that greatest progress is made when one is first learning new information. Thus, the higher the fall test score, relatively less progress will be expected by spring.

When the fall test level was added, student variables which would seem to be a function of initial testing level became more intuitive. For example GATE became positive and special education became negative. The fall test score did not, however, affect the results of any of the other variables. We chose to include it since it makes interpretation of many variables affected by the learning curve more intuitive.

In the Fresno data, the fall test variable had by far the most explanatory power of any student or teacher level variable considered, explaining about 10% of the variation in the data. Given the power of the fall test score, we felt compelled to also test it in the previous CUSD model to see what effect it had on the Capistrano data. Adding the fall test variable to the original CUSD data showed it to be similarly powerful in that district, again predicting about 10% of the variation in student test scores. As discussed above, the GATE and grade level variables switched signs, but we were relieved to find that the *Daylight Code* remained in the model essentially unchanged, both positive and significant.

Figure 20 below shows the model R² for the original Capistrano models with and without the fall test scores and the Fresno replication models, also with and without the fall test scores. Just below, is a column summarizing the change in R² due to the addition of fall test variable. It shows that the fall test score was responsible for explaining roughly 10% of the variation in student progress fall to spring in both districts and both tests.

District		Reading Model	Math Model
Capistrano	R ² w/o fall test	0.256	0.246
	R ² with fall test	0.339	0.357
	<i>Fall test difference</i>	<i>0.083</i>	<i>0.111</i>
Fresno (replication model)	R ² w/o fall test	0.065	0.101
	R ² with fall test	0.174	0.236
	<i>Fall test difference</i>	<i>0.109</i>	<i>0.135</i>

Figure 20: Comparison of R² in CUSD and FUSD models

Figure 20 can also be used to understand the difference in the explanatory power (model R²) of the models between the two districts. For example, for the reading model without fall tests the equivalent Capistrano model is able to explain about

four times as much of the variation in the data as the Fresno models (25.6% versus 6.5%). From this exercise, it can be concluded that there is more inherent variation in the population of FUSD students and/or teaching methodologies since basically the same set of variables explain substantially less of the variation in the data in Fresno than in Capistrano. We will discuss this issue later in the findings section. With more inherent variation in the data, FUSD is less likely to have significant findings.

6.2.4 Daylight Code as Outcome Variable

We used a regression model with the *Daylight Code* as the outcome, or Y variable, to understand the collinearities among the data. With this type of model, the significant explanatory variables tell us which variables we are considering best predict the *Daylight Code* of a classroom. Those variables with very high significance or partial R^2 could potentially confound our final results if they were significant in both this model and the final student performance models.

Two models were run with the *Daylight Code* as the outcome variable. In the first model, "*Daylight Code – Physical*", the potential explanatory variables were all physical characteristics variables except those which were obviously associated with the *Daylight Code*, such as window characteristics. This exercise revealed that there were a few variables of concern. *No Teacher Control of Fan* was most strongly associated with the *Daylight Code*, and was also showing up as significant in predicting student scores. Other variables of concern were Percent Carpet, which showed a slight negative association with the *Daylight Code*, and T-8 Lamps and Warm Lamp Color, which both showed a positive association.

In the second model, the "*Daylight Code – Demographic*" model, the potential explanatory variables were all the demographic variables. The model with physical variables resulted in a model $R^2=0.87$, while for the models using demographic explanatory variables $R^2=0.087$, showing that demographic information about the teachers and students were only one-tenth as powerful in predicting the *Daylight Code*. None of the teacher characteristics were very strongly associated with the *Daylight Code*. The variable with by far the largest magnitude B-coefficient was *GATE*, indicating that students designated "Gifted and Talented" were more likely to be located in daylight classrooms. We already suspected this would be true, since one of the schools with the most daylight classrooms was operated as a 100% GATE school. This was a serious concern for us, since GATE students tend to make more progress per year, once the level of the fall test score has been considered.

We used information from the thematic group models, described below, to study these collinearities and potential interactions further.

6.3 Thematic Group Models

As the next step in our investigations, we added related groups of variables describing the physical conditions of the classrooms into the base demographic

model. The Fresno District had much more detailed survey data available describing the schools and classrooms than did the Capistrano study. In order to organize the study and select the best variables for the final analysis, we grouped the variables into seven groups, called thematic groups, each including all the known characteristics about a school or classroom that might influence a particular area of concern. Each thematic group contained 12-20 variables, which overlapped if they might influence more than one issue. For example, the variable *No Operable Windows* is included in three groups: windows, indoor air quality (IAQ) and noise.

The use of thematic groups of variables allowed us to investigate many more variables than the Capistrano Unified School District study. The thematic groups are as follows:

- ***Classroom characteristics***: This consists of classroom types, amenities and construction characteristics.
- ***Window characteristics***: This consists of all classroom features that define window orientation, area, operability, etc.
- ***Electric lighting characteristics***: These include characteristics that define the quality and quantity of electric light like illuminance values, types of luminaires, etc.
- ***Air quality characteristics***: All characteristics of the school and classroom that contribute to indoor air quality of the classroom.
- ***Noise characteristics***: All characteristics that contribute to noise levels in the classrooms.
- ***School site characteristics***: All characteristics associated with school site like school location, vintage, neighborhood, etc.
- ***School site ID***: An indicator variable for each school.

The list of variables considered in each model in the thematic groups is shown in the Appendix.

6.3.1 Refining Variables

The thematic model approach also allowed us to test alternative ways to describe the school characteristics. For example, would it be better to enter an indicator variable for each school site into the model, or to group schools by neighborhood and site characteristics? One approach might provide more precision in the modeling, but provide less interpretable data and vice versa. Significant but erratic variables were investigated and reconsidered. An example is *Stale Air* which entered the models as positive or negative depending on the presence of other variables. We considered combining it with related yes/no variables *Moldy/Musty* and *Water Damage* into a scaled index for air quality. However, when we did so the R^2 of the model dropped. So ultimately, we concluded that

Stale Air was a poorly defined variable, derived from the subjective experience of surveyors. We dropped it while retaining *Moldy/Musty* and *Water Damage*, which had shown consistent performance.

At least two model runs per thematic group were created—one model that included the Daylight Code and the other without the code—to test if the *Daylight Code* was interacting with other variables in that group and affecting results. Generally the *Daylight Code* had negligible effect on the R^2 of the models, and rarely caused any variables to shift in magnitude or move in or out of the models.

6.3.2 Thematic Models Findings

All of the variables considered in the thematic groups added fairly little to the explanatory power of the models. The increase in model R^2 attributable to each group is listed in Figure 21. The strongest thematic group was the School ID group (adding 1.1% to 1.5% to the model R^2), followed by Classroom Characteristics (adding 0.3% to 1.1% to the model R^2). The next strongest groups were Window Characteristics, Air Quality/HVAC Characteristics and School Site Characteristics (adding 0.6% to 0.7%). The Noise and Electric Lighting groups were the least successful at predicting student performance (adding 0.1% to 0.4%). The strength of the various thematic groups could possibly be a function of the intrinsic importance of that issue, but could also be a function of the number of variables considered in each group and our precision in measuring and defining useful variables to explain these issues.

Models	Total variables considered	Math R^2	Math R^2 addition from thematic group	Reading R^2	Reading R^2 additoin with thematic group
Base demographic model		0.170		0.235	
<i>School ID</i>	36	0.185	0.015	0.246	0.011
<i>Classroom</i>	19	0.181	0.011	0.238	0.003
<i>Air Quality & HVAC</i>	27	0.176	0.006	0.242	0.007
<i>Window</i>	14	0.176	0.006	0.242	0.007
<i>School Site</i>	17	0.177	0.007	0.242	0.007
<i>Noise</i>	12	0.174	0.004	0.238	0.003
<i>Electric Lighting</i>	10	0.172	0.002	0.236	0.001

Figure 21: Thematic Models, R^2 contribution

The findings of the Window Characteristics model were most relevant to this study, and are discussed in detail in Section 7.2 below. We will discuss findings about other physical variables in the discussion of the final model, in Section 7.3.

6.4 Final Model Specification

The following actions were taken based on the lessons learned during all the preliminary model investigations. The list of variables selected to be included in

the final models are described in the descriptive statistics for each model included in the Appendix.

6.4.1 Investigation of Collinear Variables

Tests for collinearity among the classroom physical characteristics variables using the singular value decomposition methodology¹ was done for all thematic groups, and later for the full model. The collinearity test was performed to observe whether some school physical characteristics showed collinearity with the variables in the demographic model and among each other.

During the modeling phase where all potential explanatory variables were considered, we discovered some collinearities. Variables that were individually insignificant were found to be significant predictors of student performance when considered as a group. Specifically, whenever a set of variables was identified for possible removal from the model, we used an F-test where the null hypothesis is that the beta coefficients of each of the variables in the group are zero, while the alternative hypothesis is that there is at least one variable in the group where the beta coefficient is not zero. For example, using this approach, floor types and HVAC characteristics were found to be highly collinear with one another. We decided that it was more important for the interests of this study to retain information about the HVAC system than the floor types, so to clarify the group we dropped all the floor type variables from later analysis.

In addition, running two thematic groups together allowed us to test collinearity among variables in different groups. Variables that changed sign or dropped significance in the presence of other variables were considered possibly collinear, and thus identified for further investigation. Further investigation included a redefinition of the variable, and/or looking at the source classrooms for other explanations. An example of this is the *Stale Air* variable, which entered various models as significantly positive or negative, depending on which other explanatory variables were considered. We eventually removed it from the models given this erratic behavior and since it was based on the subjective experience of the surveyor when the classroom was not in use.

From the exercise in which we used the *Daylight Code* as the outcome variable, we had identified a number of suspicious variables possibly collinear with daylight that might influence our final results. These included *GATE* students, *No Mechanical Ventilation Control*, *Percent Carpet*, *T-8 Lamps* and *Warm Lamp Color*. As we ran the thematic group models with and without the *Daylight Code*, we observed how these variables behaved. If one of them shifted dramatically in magnitude or significance once the *Daylight Code* was inserted into the model, then we would know that it was indeed interacting with the *Daylight Code* in predicting student performance. Upon examination, the *GATE* variable, which was our biggest concern, showed very little variation among all the models. The

¹ Belsley, David A., Kuh, Edwin, and Welsch, Roy E. *Regression Diagnostics: Identifying Influential Data and Sources of Collinearity*. John Wiley & Sons, Inc, 1980. Chapter 3.

suspect variable that showed the greatest changes was *No Mechanical Ventilation Control*, which sometimes dropped out of the models, depending upon which other variables were considered. We considered dropping it, or adding an interaction variable between it and the *Daylight Code*, but decided that since it entered the final models after the *Daylight Code* and all other window characteristics, that it was not absorbing any of their effects. Indeed, in the final math model, it was the last significant variable to enter the model.

6.4.2 Drop School ID Variables

The *School ID* variable (a unique indicator variable for each school) was clearly very collinear with the School Site Characteristics group of variables (general characteristics about a site, such as the condition of the paint or the neighborhood). We needed to choose one group or the other to clarify the model. We had used *School ID* in the Capistrano models because we did not have consistent information about all of the sites in that district. In our thematic tests the *School ID* models had a higher R^2 , but also had many more significant variables. We realized that there would be more variation in other variables across groups of schools, than within a given school, so we were likely to get more precision in our other variables of interest by using the shared school site characteristics rather than the *School ID*. Furthermore, information about the quality of the sites would be more informative in the models, allowing us to assess the impact of shared characteristics, rather than treating each school as an indicator variable, but without the ability to interpret why it might be performing above or below norm. Thus, we decided to drop the *School IDs* in favor of the school characteristics.

As part of this investigation we realized that almost one-half of all FUSD school sites proved significant as explanatory variables, compared to less than one-third of CUSD school sites. This suggested that individual school sites are likely to be more important in explaining student performance than in Capistrano, and thus the difference in educational performance between schools is much greater in Fresno than in Capistrano. Indeed, the Fresno District has a site-based curriculum model, where the principal and teachers at a given school have more latitude in determining instructional styles and testing protocols than in Capistrano. For example, if a teacher can choose when to teach certain elements of the curriculum, but that choice is out of sync with District testing schedules, then students in the classroom might perform less well on the tests, but still master the subject area. The majority of Fresno schools are also considered neighborhood schools, where children attend the school nearest their home. Capistrano had far more magnet programs that encouraged students to travel across the district to attend special schools. As a result, Capistrano may have had more mixing of student demographics among schools.

6.4.3 Add School Level Socioeconomic Variables

As a final step, preliminary models were presented to the FUSD representatives for discussion and comment. They suggested that school level socioeconomic conditions might be influencing results, and so they recommended that we add variables to account for these effects. They created a set of socioeconomic indicators for each school in our study that we then included in the final models. These five new variables described the overall student population average of each school, rather than the individual socioeconomic conditions of the individual students included in the study. They are described in Section 3.5.

7. REGRESSION ANALYSIS FINDINGS

This section describes the findings of the replication, thematic and final models, and offers some interpretation.

To facilitate interpretation, the findings are presented in a variety of formats. The primary comparison between models is done via percentage effects, derived from the B-coefficient for each variable, as described in Section 7.2. For the final models, we also present information about the significance, order of entry and partial R^2 of each variable. The order of entry discussion includes possible mechanisms to explain the behavior of each physical variable.

7.1 Replication Model

The simple model meant to replicate the data format used in Capistrano showed that the *Daylight Code* was **not** significant in predicting student performance. Thus, we could not replicate the Capistrano findings based on a similar model structure.

This model had substantially less explanatory power than the equivalent Capistrano model, even with the addition of more precise information about the teachers and schools. One possible explanation for this was the spread of the gain in scores was less for FUSD than CUSD. The standard deviation for the gain in math scores in FUSD was 81% of that in Capistrano, while the standard deviation for the gain in reading scores was 72% of that in CUSD. This reduction in spread is likely to be partly a function of the different age groups studied; since the younger students had a much wider spread in learning rates than the older students. The CUSD data included grades 2-5, while the FUSD data included grades 3-6.

As mentioned earlier, even though the findings of this replication model did not support the hypothesis that daylight has a positive influence on student learning, we decided to proceed with our analysis to see if we could learn anything more specific about the mechanisms of school design on student performance, and perhaps why the *Daylight Code* was not significant as it had been in Capistrano, Seattle and Fort Collins.

There were many possibilities to consider. It could be that for some reason daylight is not as useful or benign in Fresno as the other locations. It could be that there was something negative associated with daylight in Fresno classrooms that was countering any positive effects. It could be that the *Daylight Code* we had created for Fresno did not correctly reflect actual operating conditions, such as if the teachers always kept their window blinds closed during school. It could be that the previous findings were a fluke, and that daylight does not have any reliable correlation to student performance. It could be that the Fresno models had too many other collinear conditions that were affecting results. The rest of

this report describes our effort to understand why the *Daylight Code* was performing differently in Fresno from our previous studies.

7.2 Window Characteristics Thematic Model

The thematic model which tested just Window Characteristics against the base demographic model was found to be as strong a model as those assessing general Classroom Characteristics or Air Quality and HVAC Qualities. Overall the group of window characteristics added 0.6% and 0.7% to the explanatory power of the base demographic models. This is more than the 0.1% to 0.3% added by the various Window, Skylight and *Daylight Codes* to the Capistrano models. This implies that, even though the overall models were not doing as well at predicting student performance, we had refined the description of window characteristics so that they were doing a better job than what we had done in Capistrano.

To facilitate interpretation, the findings below are presented as percentage effects, along with the significance (p) of the variable. The percentage effect shows how much the outcome variable would change over a range of that variable, if all other factors considered in the regression equation were held constant. The percentage effect is calculated using the B-coefficient multiplied by a specified range and then divided by the mean of the outcome variable, the change in fall to spring scores. To make the reported percentage effect more useful, we have tried to choose ranges that might be meaningful to the reader, such as minimum to maximum condition, 100 square feet, or 10 computers, rather than basing the range on the means and standard deviations of the data.

MATH MODEL

Variable Description	Range	% Effect		Sig.
Daylight Code	None to most		-9%	0.019
Primary window wall faces east	If yes		-12%	0.000
No secondary window wall	If yes		-4%	0.080
Sun penetration	None to most		-9%	0.002
View distance	<25 ft to 75ft or more	7%		0.014
No operable windows	If yes	6%		0.011
No blinds or curtains	If yes		-5%	0.001

Figure 22: Window Characteristics Thematic Math Model

READING MODEL

Variable Description	Range	% Effect		Sig.
Daylight Code	None to most		-16%	0.032
Primary window wall faces east	If yes		-8%	0.004
Primary window wall faces south	If yes		-7%	0.000
Window area above door (high)	100 sqft more		-7%	0.003
Window area desk-door (view)	100 sqft more	21%		0.000
View distance	<25 ft to 75ft or more	5%		0.093
Not operable windows	If yes	9%		0.000
No blinds or curtains	If yes		-7%	0.000

Figure 23: Window Characteristics Thematic Reading Model

Figure 22 and Figure 23 present a summary of the significant window variables in the Window Characteristics thematic models. Below we discuss a possible interpretation of each variable, starting with those that are shared between both models, and thus are assumed to be the most stable.

Significant for both Reading and Math:

Primary Window Wall faces East: Children are performing worse in both the math and reading models when their classroom's primary window faces east. Most likely there is low-angle sun coming in through the windows in the morning, while class is just starting. The low-angle sun is likely to cause extreme glare, and possibly some thermal discomfort, or alternatively motivate the teacher to block the windows in order to avoid these problems.

View: Surveyors rated view as near, mid or far. The farther away the view, the better children are doing in both math and reading.

No Operable Windows: Children are doing better in both math and reading when their classroom does not have windows that can be opened. Operable windows in Fresno seem to be a cause of increased noise and poor air quality from outside sources. Teachers open their windows for ventilation, often to overcome discomfort from poorly functioning HVAC systems. But when they do so, they are trading off more noise from outside for better thermal comfort and ventilation.

No Window Control: Without blinds or curtains at the windows, children are doing worse in both math and reading. Teachers who do not have any blinds or curtains at their windows cannot make adjustments to deal with temporary glare or distractions from outside the windows.

Significant for Just Math:

Window Tint: As window tint decreases, children are making less progress in Math. This would seem to be another indication of how important glare is in math learning.

Sun Penetration: The more often direct sun is likely to get into the classroom, based on the surveyors' assessment of window orientation and shading, the worse children are doing in math.

Significant for Just Reading:

No Primary Window Wall: Six classrooms in two open plan schools (open passages between classrooms) are in the interior and have no windows at all. According to the model, under these conditions children are doing better in reading. It is possible that these classrooms suffer less disruption than their neighbors, which have windows and doors, since there is less noise from outside and also less traffic through the classroom by other classes to get to the outside.

In general, children in open plan classrooms in Fresno seem to perform better than the norm.

Window Area above Door: This represents the amount of high window area in the classroom, higher than door height. The greater this area, the worse children are doing in reading. The classrooms with the greatest amount of this type of windows are bungalows, the older version of portable classrooms, based on the finger plan design. The next largest area is found in original finger plan classrooms that have never been retrofitted with a lowered ceiling or reduction of window area concurrent with HVAC improvements. There are many possible reasons for this negative effect, such as glare or thermal discomfort. Based on our Phase 2 investigations and analysis, our hypothesis is that this variable is a marker for a classroom with a higher ceiling, and therefore a more reverberant space that interferes with listening and language arts instruction.

Window Area from Desk to Door: This condition represents the primary view area of the window at eye level. The larger the area of view window, the better children are doing in reading. This attribute would seem to be consistent with the earlier finding discussed above that more distant views positively influence both math and reading scores.

7.3 Final Math and Reading Models

In the final modeling process, 72 variables describing physical conditions at the schools were considered as potential explanatory variables. Twenty one variables describing physical conditions of the schools or classrooms proved significant in the math model, of which seven, or one-third, were window characteristics. Twenty seven were found significant in the reading model, of which eight were window characteristics. This is a lot of information. In Capistrano we considered far fewer explanatory variables, so fewer came into the models. Also, here in FUSD, we have broken down information, such as the *Daylight Code*, into constituent characteristics, again increasing the detail considered, and reported. The full detail on each model with descriptive statistics are available in the Appendix.

The criteria for acceptance within the models were $p \leq 0.10$, or greater than 90% certainty that this was a true effect. By allowing the slightly more generous standard of $p \leq 0.10$ instead of $p \leq 0.05$, we insured that all variables that might influence results were considered, and that we did not unintentionally exclude a variable that might influence the variables of interest. In the final models, almost all variables exceed the more strict $p \leq 0.05$ criteria, with the exception of one (*Student Gender*) in the math model and three physical variables in the reading model.

Below, first we explain the findings of the final models relevant to just the window characteristics, and compare those to the earlier findings of the Window

Characteristics thematic model. We then discuss the full models, with all significant variables.

7.3.1 Window Characteristics in Final versus Thematic Models

When in competition with all other variables, seven window characteristics still entered the final models as highly significant. This is essentially the same number that entered the much simpler Window Characteristics thematic group models. Furthermore, the collective R^2 of the Window Characteristics changed very little, even when considered in relationship to so many other explanatory variables. (see discussion below on Order of Entry and Partial R^2 in Section 7.3.4)

MATH MODEL

Window Characteristics	Range	% Effect		Sig.
Daylight Code	None to most		-22%	0.002
Primary window wall faces east	If yes		-12%	0.000
Window area above door (high)	100 sqft more	7%		0.010
Glare from windows	None to most		-7%	0.011
Security measures on windows	If yes		-9%	0.001
No blinds or curtains	If yes		-5%	0.007
View vegetation	If yes	10%		0.000

Figure 24: Window Characteristics in Final Math Model

READING MODEL

Window Characteristics	Range	% Effect		Sig.
Daylighting Code	None to most		-29%	0.000
Two exterior doors	If yes	10%		0.022
Primary window wall faces east	If yes		-8%	0.008
Primary window wall faces south	If yes		-9%	0.000
Window area desk-door (view)	100 sqft more	14%		0.006
No blinds or curtains	If yes		-5%	0.010
Security measures on windows	If yes		-8%	0.007
View activity	If yes	6%		0.050

Figure 25: Window Characteristics in Final Reading Model

Figure 24 and Figure 25 show the percentage effects for window characteristics found significant in the final models. Those that are consistent with the Window Characteristics thematic models are shaded (we treated view vegetation or view activity as consistent with view distance, as they are all related). A very consistent picture emerges:

- ♦ Any characteristic having to do with glare is negative.
- ♦ Any characteristic having to do with a better view is positive.

Sun Penetration. *Primary window wall faces east* is once again significant with a negative effect for both reading and math. Likewise, *No blinds or curtains* is negative for both. In the reading model, *Primary window wall faces south* is also

negative. All of these imply that uncontrolled sun penetration into the classroom and associated heat and glare are serious negative effects on learning.

Glare. Glare is a clear negative influence here. The variable *Glare from windows*, which is negative in the math model, describes the likelihood of glare from windows on the teaching surface. Elsewhere in the math model, a white teaching board turns up as a positive effect on math learning (and is discussed further in Section 8.5). This could possibly be a glare effect, as window reflections are less debilitating on a white surface than a dark surface such as black or green. The negative effects of the *No blinds or curtains* variable could also reflect a teacher's inability to respond to outside glare sources by controlling blinds or curtains.

View. The quality of the view through the windows is also showing highly significant and positive, as it did earlier in the Window Characteristics thematic models. This time, instead of the distance of the view it is the content of the view which is significant. The models find that being able to see vegetation (in math) or human activity (in reading) out of the window is a positive influence on learning. We feel that all of these measures of view quality are quite crude, since they were subjective measures judged by a number of surveyors. However, the consistency of the findings on the positive effects of view on learning is certainly good reason to look into this association further. This finding is further reinforced by the variable *Window area desk-door* (view windows) also showing strongly positive in the reading model.

One additional window characteristic shows up as highly significant and negative in the final models that was not included in the original Windows Characteristics thematic group—*Security measures on windows*. We had originally assumed that this variable would best describe a condition of security threats to the classroom. But upon interviews with the teachers in the Phase 2 data collection efforts we realized that it was typically a historical remainder of some past situation. Since there were only a few of these per campus we do not believe that they reflect general neighborhood trends. Typically, the classroom seems to have held computers or other valuables at one point, but is now being used as a general purpose classroom. Thus, we now interpret this measure to be an indicator of a very bad view. This variable consistently entered all full models we tested as highly significant and negative.

Window Code. The *Window Code* is also significant and negative in both models. But it is modified by another powerful variable in each model. In the math model, more *Window area above the door* is positive. High *Daylight Code* classrooms generally have substantial window area above door height. Thus, while being a *Window Code 5* classroom implies a negative effect of -22%, most of these classrooms also have an additional 100 sf or more of high glass area, which adds a positive 6.8% effect. In the reading model, larger view window area and having two exterior doors are also positive. Most high *Daylight Code* classrooms also have large view area windows, adding 200-300 sf more view area, or a positive effect of 27-41%, and two exterior doors, adding a positive

10%. As a result of these cumulative effects, most of the high *Daylight Code* 4 and 5 classrooms are predicted to have a net positive effect on learning.

Thus, it is difficult to understand the implications of the *Daylight Code* without accounting for related variables that are obviously collinear with the *Daylight Code*. Later, in Section 8, we describe those actual classrooms in Fresno that were predicted to have the highest and lowest learning effects, as predicted by their various combination of window characteristics.

We choose to include the *Daylight Code* in the models, even though it is obviously collinear with other variables, because when we excluded it the other variables remaining in the model did not shift appreciably. Thus, we judged that the *Daylight Code* was adding important information to the model, telling us either that there is something negative associated with the high end of the *Daylight Code*, or alternatively something very positive about the low end. Section 8, discussing the findings of the Phase 2 analysis, attempts to understand why the *Daylight Code* proved consistently negative in the models.

7.3.2 Percentage Effects

Figure 26 and Figure 27 present the percentage effects for all the variables in the full models. These tables allow the reader to compare the relative magnitude of effects found for all the variables. For example, being identified as a *GATE* (gifted and talented) student predicts that a student will make 36.7% more progress per year than norm, while being identified as a *Special education* student predicts 27.9% less progress.

The percentage effect needs to be interpreted relative to the range used to describe it. Sometimes the range is binary as in no to yes, or on a scale of 0-5 as best to worst. In variables with an extended scale, such as number of students in a school or percent attendance, we used a simple range that is readily understood.

The reader should be cautioned, however, that predicted magnitudes are the least reliable output of a regression equation. Magnitudes of predicted effects are likely to shift as different variables are considered or different populations studied. Far more stable information is derived simply the sign of the B-coefficient. Thus, it is more informative and reliable to note if a significant effect is positive or negative, rather than concentrating on the size of the predicted effect.

Variable Description	Range	% Effect	Consistent?
Fall math RIT score	10% above average	-36%	
Re-test for fall math	If yes	39%	
Student Level Variables			
Third grade	If yes	-15%	
Fourth grade	If yes	-31%	Yes
Fifth grade	If yes	-11%	
Percentage attendance	10% increment	9%	
Enrolled in GATE	If yes	37%	Yes
Special Ed student	If yes	-28%	Yes
Student English development	scalar 3 - 6	12%	Yes
Free lunch	If yes	-5%	Yes
Student gender	If yes	-10%	Yes
Ethnic student (Type 12)	If yes	-10%	Yes
Ethnic student (Type 13)	If yes	-17%	Yes
Ethnic student (Type 15)	If yes	-13%	
Ethnic student (Type 16)	If yes	20%	
Teacher Level Variables			
Multi-grade classroom	If yes	-14%	Yes
Annual salary	\$ 10,000 more	4%	
Number of years at FUSD	10 years	-3%	
Mentor teacher	If yes	8%	
Pre-tenure teacher	If yes	13%	
School Socio-economic Characteristics			
School English learner (EL)%	10% increment	18%	Reverses
School parent education	Least to best	25%	Yes
School Characteristics			
Age of school in 2000	10 years more	-4%	
Neighborhood is lower economic status	If yes	-13%	
Neighborhood is prewar vintage	If yes	16%	Yes
Neighborhood is 40s/50s vintage	If yes	7%	
Paint condition	Worst to best	7%	
Classroom Characteristics			
Interior corridor classroom	If yes	-30%	
Operable walls classroom	If yes	14%	
White teaching board	If yes	8%	
Computers	10 more	17%	Yes
Security measures on windows	If yes	-9%	Yes
Window Characteristics			
Daylight Code	None to most	-22%	Yes
Primary window wall faces east	If yes	-12%	Yes
Window area above door	100 sf more	7%	
Glare from windows	None to most	-9%	
No blinds or curtains	If yes	-5%	
Vegetation in view	If yes	10%	
Air Quality & HVAC Characteristics			
Pets in classroom	If yes	-21%	
Central HVAC system	If yes	-7%	
Wall mounted heating unit	If yes	5%	
No teacher control of fan	If yes	7%	
Acoustic Characteristics			
Loud HVAC system	If yes	-17%	
Model Summary:			
RMSE		5.81	
R ²		19.2%	

Figure 26: Percentage Effects of Final Math Model .

Variable Description	Range	% Effect		Consistent?
Fall reading RIT score	10% above average		-46%	Yes
Re-test for fall reading	If yes	30%		Yes
Student Level Variables				
Fourth grade	If yes		-13%	Yes
Fifth grade	If yes		-9%	
Percentage attendance	10% increment	4%		Yes
Enrolled in GATE	If yes	16%		Yes
Special Ed student	If yes		-27%	Yes
Student English development	scalar 3 - 6	11%		Yes
Free lunch	If yes		-5%	Yes
Non-standard living situation	If yes		-16%	
Student gender	If yes		-3%	Yes
Ethnic student (Type 12)	If yes		-4%	Yes
Ethnic student (Type 13)	If yes		-11%	Yes
Teacher Level Variables				
Multi-Grade classroom	If yes		-7%	Yes
Socio-economic Characteristics				
School mobility	10% increment	10%		
School English learner	10% increment		-9%	Reverses
School free/reduced lunch	10% increment	3%		
School parent education	Least to best	27%		Yes
School CalWork	10% increment		-7%	
School Characteristics				
Students in school	100 more		-5%	
School near blvd	If yes	6%		
School near construction noise	If yes	13%		
Neighborhood is residential/commercial	If yes	17%		
Neighborhood is upper economic status	If yes	14%		
Neighborhood is prewar vintage	If yes	11%		Yes
Grass condition	Worst to best	13%		
Classroom Characteristics				
Room area	Small to large	7%		
No doors classrooms	If yes		-12%	
Number of computers	10 more	10%		Yes
Security measures on windows	If yes		-8%	Yes
Window Characteristics				
Daylighting Code	None to most		-29%	Yes
Two exterior doors	If yes	10%		
Primary window wall faces east	If yes		-8%	Yes
Primary window wall faces south	If yes		-9%	
Window area desk-door	100 sf more	14%		
No blinds or curtains	If yes		-5%	
Activity in view	If yes	6%		
Air Quality Characteristics				
Water damage visible	If yes		-15%	
Musty/Moldy air in classroom	If yes		-10%	
No teacher control of fan	If yes	10%		
Percentage carpet	0% to 100%	8%		
Electric Light Characteristics				
T8 lamps	If yes	12%		
Lamp color is warm (CCT<3500)	If yes		-16%	
Mixed fluorescent (poor lighting maintenance)	If yes		-6%	
Acoustic Characteristics				
Loud ballast hum	If yes		-19%	
Model Summary:				
RMSE		5.64		
R ²		25.5%		

Figure 27: Percentage Effects of Final Reading Model

These tables also note whether the variable found significant in one model was also found significant in the other. Variables that are consistent in both models are considered the most robust in predicting overall student performance, since they apply to both math and reading learning rates. If we applied the stringent criteria that variables must be significant in both models, the models would be reduced to just seventeen variables. Only one variable was found to reverse signs between the two models—*School English learner %*—which was positive in math and negative in reading. The higher the percentage of the school population that is learning English, the worse students are doing in reading relative to norm, but the better they are doing in math.

Very many of these variables are likely significant only for the Fresno district, or perhaps only for the specific population that we studied. Which variables enter a model as significant are very much a function of the context, and which other variables are being considered simultaneously. Variables may also serve as a proxy for some associated condition. For example, *Pets in classroom* shows up as negative in the math model. In previous thematic models, *Pets in classroom* often showed up positive and significant in the reading models. Thus, it does not seem to be a consistently negative characteristic. Perhaps rather than having a direct effect, *Pets in classroom* may be an indication of the type of teacher running the classroom. For example, it could be that teachers who are likely to keep pets in their classroom are more focused on language arts than mathematics. Or perhaps having pets in a classroom causes a distraction during timed math tests, but provides reassurance and creative inspiration for language learning.

Many variables are subject to multiple interpretations. We provide only a brief discussion of possible interpretations for the various physical characteristics that proved significant in the models other than the window characteristics, since they are not the focus of this study. These are summarized in Figure 29 and Figure 30.

7.3.3 Significance Level

MATH MODEL

Variable Description	B	p
Constant	29.51	0.000
DEMOGRAPHIC VARIABLES	0.001	or less
Fall math RIT score	-0.16	0.000
Enrolled in GATE	3.32	0.000
Fourth grade	-2.80	0.000
Re-test for fall math	3.54	0.000
Special Ed student	-2.52	0.000
Student gender	-0.91	0.000
Percentage attendance	0.08	0.000
Multi-grade classroom	-1.23	0.000
Ethnic student (Type 13)	-1.54	0.000
Ethnic student (Type 12)	-0.91	0.000
Student English development	0.37	0.000
Fifth grade	-1.01	0.000
Third grade	-1.39	0.000
Annual salary (per \$1000)	0.04	0.000
School parent education	0.97	0.000
Free lunch	-0.47	0.001
	0.01	or less
Pre-tenure teacher	1.15	0.003
Mentor teacher	0.76	0.005
Number of years at FUSD	-0.03	0.008
	0.05	or less
School English learner (EL)%	3.30	0.016
Ethnic student (Type 16)	1.80	0.044
	0.10	or less
Ethnic student (Type 15)	-1.17	0.078
PHYSICAL VARIABLES	0.001	or less
Primary window wall faces east	-1.12	0.000
Number of computers	0.15	0.000
Age of school in 2000	-0.03	0.000
White teaching board	0.75	0.000
Operable walls classroom	1.26	0.000
Neighborhood is 40s/50s vintage	0.63	0.000
Loud HVAC system	-1.52	0.000
Vegetation in view	0.93	0.000
Neighborhood is lower economic status	-1.16	0.000
Interior corridor classroom	-2.73	0.000
Neighborhood is prewar vintage	1.48	0.001
Security measures on windows	-0.82	0.001
	0.01	or less
Pets in classroom	-1.88	0.001
Daylight Code	-0.40	0.002
Wall mounted heating unit	0.44	0.004
No blinds or curtains	-0.42	0.007
Window area above door	0.06	0.010
	0.05	or less
Glare from windows	-0.20	0.011
Central HVAC system	-0.64	0.011
No teacher control of fan	0.63	0.011
Paint condition	0.22	0.030

READING MODEL

Variable Description	B	p
Constant	37.59	0.000
DEMOGRAPHIC VARIABLES	0.001	or less
Fall reading RIT score	-0.20	0.000
School English learner (EL)%	-7.77	0.000
Re-test for fall reading	2.53	0.000
Fourth grade	-1.09	0.000
Fifth grade	-0.74	0.000
Enrolled in GATE	1.33	0.000
Special Ed student	-2.27	0.000
Ethnic student (Type 13)	-0.97	0.000
School mobility index	8.69	0.000
Student English development	0.30	0.001
	0.01	or less
Free lunch	-0.45	0.001
Multi-grade classroom	-0.62	0.006
School CalWork%	-6.09	0.008
Percentage attendance	0.04	0.012
Non-standard living situation	-1.32	0.013
School free/reduced lunch %	2.69	0.022
Ethnic student (Type 12)	-0.33	0.024
School parent education	1.02	0.027
	0.05	or less
Student gender	-0.22	0.079
PHYSICAL VARIABLES	0.001	or less
Loud ballast hum	-1.59	0.000
Primary window wall faces south	-0.76	0.000
Neighborhood residential/commercial	1.42	0.000
School near construction noise	1.08	0.000
Daylighting Code	-0.49	0.000
No teacher control of fan	0.87	0.000
Grass condition	0.37	0.000
Students in school	0.00	0.000
Musty/Moldy air in classroom	-0.85	0.001
	0.01	or less
Neighborhood upper economic status	1.18	0.002
Number of computers	0.09	0.002
Window area desk-door	0.12	0.006
Security measures on windows	-0.71	0.007
No doors classrooms	-1.04	0.008
Primary window wall faces east	-0.65	0.008
No blinds or curtains	-0.40	0.010
	0.05	or less
Water damage	-1.29	0.012
Two exterior doors	0.86	0.022
Lamp color is <3500	-1.33	0.022
Percentage carpet	0.01	0.025
Neighborhood is prewar vintage	0.94	0.032
Mixed florescent	-0.47	0.033
Activity in view	0.52	0.050
	0.10	or less
School near blvd	0.52	0.054
Room area (SQFT)	0.31	0.088
T8 lamp	1.00	0.090

Figure 28: Final Math and Reading Models sorted by Significance of Variables

The significance level of variables is perhaps the best way to assess its strength in the model and likelihood of consistently appearing in other models. Figure 28 shows all the variables in the math and reading models sorted by their significance level. The highest significance level, $p \leq 0.0001$, expresses that there is a 99.99% certainty that the effect does indeed exist, or is not zero. A significance level of $p \leq 0.10$ expresses that there is a 90% certainty of a valid effect.

The lowest criteria for entry into these models is $p \leq 0.10$. Had we run a model with higher criteria for entry for the physical variables, such as $p \leq 0.05$, only a very few of the physical variables would have dropped out, three for the Reading Model, and none for the Math Model.

We can see that for the math model, *Primary window wall faces east* is just as significant as both *Number of computers in classroom*, a condition which is widely believed to improve math education, and *Percent attendance*, a very important concern of all parents and administrators. In the reading model this window characteristic also has a similar level of significance as *Number of computers in classroom* and is even more significant than *Percent attendance*.

We all know that there are massive state and national efforts, involving multi-million dollar programs, aimed at putting more computers into classrooms or improving attendance in schools. These models tell us that, for Fresno, there is an equal likelihood of improving student performance by avoiding building classrooms that face east as there is by adding more computers to the classroom or by reducing absenteeism. Furthermore, once in place, those non-east facing classrooms are likely to stay put for forty or fifty years, continuing to support better student performance at no additional yearly cost.

It is interesting to note that for the reading performance *Loud Ballast Hum* is the most significant physical variable in predicting performance. This is the high pitched sound made by some poorly functioning magnetic ballasts for the fluorescent lighting system. It is easily fixed by replacing the older fluorescent system with new electronic ballasts with a good sound rating (A). This variable is as significant as *GATE*, and even has a larger magnitude of effect: -19%, versus a positive 16% for *GATE*. Thus, they could be considered to cancel each other out. Thus, according to the model, a gifted and talented student in a classroom with humming ballasts is likely to make no more progress than the average student when located in a classroom with a quiet lighting system.

7.3.4 Partial R^2 and Order of Entry

In this section we present the findings of the same full models, but in another format. Here we look at the variables by their order of entry into the models, and the partial R^2 contributed by each variable to the overall R^2 of the model. This view is another way to assess the strength of the variables. Those with the highest significance and largest effects are likely to enter first. If the variables were completely uncorrelated, they would sort perfectly by their partial R^2 . If two variables are collinear, they can influence when the other one enters the model.

Interpreting R^2

The partial R^2 attributed to a variable can be interpreted as “the amount of variance in the data that is explained by that variable.” Thus, in Figure 29 below, the Fall math RIT score enters the math model first, with an R^2 of 0.043. It could be said that this variable is explaining 4.3% of the variation in the students’ math progress.

The partial R^2 of the variable might be interpreted as the “precision” of the variable. A variable with an R^2 of 1.0 would perfectly predict the outcome. A variable with an R^2 of 0.5 influences only 50% of the outcome. Something else influences the other 50%. An explanatory variable might predict a 20% difference in performance (all other things held constant) with 99% certainty that this is indeed a true effect, but it is still only 50%, or 5% or 0.5% of the equation.

Here, in these models, all the physical conditions of the schools and classrooms together are judged to influence about 1.5% to 2% of overall student performance. Each physical characteristic by itself tends to influence about 0.3% to 0.1% of the outcome. These are, of course, very small numbers, and need to be set into perspective. As we saw in the earlier discussion on the Fall test RIT score, information about a specific individual is seen to predict about 5-10% of their performance. When we move down a notch to more generalized information, as in which generic socio-economic or ethnic group an individual belongs to, the explanatory power of the variables drop one order of magnitude, to about 1% per characteristic. With the physical variables, our precision has dropped another order of magnitude, to 0.1%.

So why would such a small effect be interesting and valuable to know? Perhaps the most compelling reason is that the physical conditions of the environment are completely within our human control when we make design decisions about new buildings. We typically have no control of our demographic characteristics, such as age or ethnic background. And it requires enormous and persistent political will to change social conditions, like the transience of the student population or the education level of parents. But design decisions about the physical environment are completely within our control, and once made, have very long term effects. A school building in California is likely to have about a fifty year life span. Thus, a decision about the physical environment, even though it has a relatively small amount of influence on individual performance, will continue to

have an effect for fifty years and will influence hundreds or thousands of individuals over its lifetime, which greatly multiplies its importance.

We might think of the life-cycle value of various effects. For example, buying ten more computers for a classroom is predicted to improve math student performance by 10%, with a partial R^2 of 0.2%. However, the computers may only last for 5 years. Thus, they have five years worth of influence at 0.2% precision. Providing a window with a pleasant view of trees and grass from a classroom is also predicted to improve math performance by 10%, with a partial R^2 of 0.1%, but will probably last for fifty years. The view may have slightly less precision in achieving the desired goal, but will have ten times as long to influence performance. With this perspective, the view has a bigger long term impact on student learning and so should be a more important investment decision.

Order of Entry and Partial R^2 Tables

Order of Entry	Variable Description	Partial R^2	Pos.	Neg.	Issues	Possible Interpretation
1	Fall math RIT score	0.043		neg		
2	Enrolled in GATE	0.028	pos			
3	Fourth grade	0.015		neg		
4	Re-test for fall math	0.012	pos			
5	School English learner (EL)%	0.010	pos			
6	Special Ed student	0.010		neg		
7	Student gender	0.005		neg		
9	Percentage of attendance	0.003	pos			
10	Multi-grade classroom	0.003		neg		
13	Primary window wall faces east	0.003		neg	Glare	Low-angle morning sun causing glare?
14	Ethnic student (Type 13)	0.002		neg		
15	Ethnic student (Type 12)	0.004		neg		
18	Number of computers	0.002	pos			
20	Security measures on windows	0.002		neg	View	Bars on windows provide negative view?
21	Age of school in 2000	0.002		neg		
22	Student English development	0.002	pos			
34	Mentor teacher	0.001	pos			
36	Free lunch	0.001		neg		
37	White teaching board	0.001	pos		Glare, IAQ	Less glare, less dust from chalk? More use?
38	Fifth grade	0.001		neg		
39	Third grade	0.003		neg		
40	Operable walls classroom	0.001	pos			
41	Neighborhood is 40s/50s vintage	0.001	pos			
42	Wall mounted heating unit	0.001	pos		IAQ	More control of temp.? Portables and finger plan?
43	Loud HVAC system	0.001		neg	Noise	Makes hearing teacher difficult?
44	Pets in classroom	0.001		neg	IAQ	Possible allergies? Teacher type?
45	Pre-tenure teacher	0.001	pos			
46	Annual salary (per \$1000)	0.001	pos			
47	Number of years at FUSD	0.001		neg		
48	School parent education	0.001	pos			
49	Vegetation in view	0.001	pos		View	View of outside vegetation is relaxing?
50	Glare from windows	0.001		neg	Glare	Too much glare on teaching surface?
51	Neighborhood-lower economic status	0.001		neg		
52	Interior corridor classroom	0.001		neg		
53	Neighborhood is prewar vintage	0.001	pos			
54	No blinds or curtains	0.000		neg	Glare	Teacher cant prevent glare/distraction from windows?
55	Ethnic student (Type 16)	0.000	pos			
56	Paint condition, worse to better	0.000	pos		Site	Better image=more motivation?
57	Ethnic student (Type 15)	0.000		neg		
58	Daylight Code	0.000		neg	Daylight	See Phase 2 analysis discussion
59	Window area above door (high)	0.001	pos		Daylight	Less glare, but more daylight?
60	Central HVAC system	0.000		neg	IAQ	No individual control over thermostat?
61	No teacher control of fan	0.001	pos		IAQ	Mechanical ventilation always on?
16-35	18 Outlier Students	0.021				
	Total R^2	0.192				

Figure 29: Order of Entry and Partial R^2 in Final Math Model

The tables in Figure 29 and Figure 30 are sorted by variable order of entry into each model. We have highlighted the physical variables, those which might constitute a design decisions for a school or classroom, in bold. We have also noted the direction of their effects, whether positive or negative, and added a column of possible interpretations for the meaning of that finding. The outliers are combined at the bottom for simplicity.

The combined partial R^2 for the window characteristics variables remained comparatively high, at 0.5% for the reading model, compared to 0.7% for the thematic model. The math model combined R^2 was 0.7%, compared to 0.6% in the thematic model. This is more than the 0.1% to 0.4% contributed by the *Daylight Code* or Window and Skylight Codes in the Capistrano models. Thus, it can be concluded that information about the window characteristics of classrooms is indeed robust and influential on student learning, even competing with all the other aspects of schools and classrooms that we considered as explanatory variables.

Order of Entry	Variable Description	Partial R^2	Pos.	Neg.	Issues	Possible Interpretation
1	Fall reading RIT score	0.183		neg		
2	School English learner %	0.011		neg		
3	Special Ed student	0.009		neg		
4	Re-test for fall reading	0.007	pos			
5	Enrolled in GATE	0.004	pos			
6	Fourth grade	0.004		neg		
7	Fifth grade	0.004		neg		
8	School near construction noise	0.002	pos		Noise, IAQ	Improving neighborhood??
9	Loud ballast hum	0.002		neg	Noise	Annoying hum creates distracting noise?
10	Ethnic student (Type 13)	0.002		neg		
16	Security measures on windows	0.001		neg	View	Bars on windows provide negative view?
17	Primary window wall faces south	0.001		neg	Glare, Heat	Sun on south window causing glare, overheating?
21	Free lunch	0.001	pos			
24	Neighborhood residential & commercial	0.001	pos			
25	Student English development	0.001	pos			
26	Percentage attendance	0.001	pos			
27	Non-standard living situation	0.001		neg		
28	Daylighting Code	0.001		neg	Daylight	See Phase 2 analysis discussion
29	No blinds or curtains	0.001		neg	Glare	Teacher can't prevent glare/distraction from windows?
30	Primary window wall faces east	0.001		neg	Glare	Low-angle morning sun causing glare?
31	Multi-grade classroom	0.001		neg		
32	Musty/moldy air in classroom	0.001		neg	IAQ	Likely indicator of poor air quality?
33	School free/reduced lunch %	0.000	pos			
34	Ethnic student (Type 12)	0.000		neg		
35	School near blvd	0.000	pos			
36	Water damage	0.000		neg	IAQ	Possible source of poor air quality? Poor maintenance?
37	View activity	0.000	pos		View	More stimulating view of people?
38	Student gender	0.000		neg		
39	Window area desk-door (view area)	0.000	pos		View	Larger view area?
40	Mixed florescent or can't tell	0.000		neg	Lighting	Poor lighting maintenance?
41	No teacher control of fan	0.000	pos		IAQ	Mechanical ventilation always on?
42	No doors classroom (open clsm)	0.000		neg	Noise	Room can't be isolated from neighbors' noise?
43	Grass condition	0.000	pos		Site	Lush vegetation = better play area? Better image?
44	School mobility	0.000	pos			
45	Number of computers	0.000	pos			
46	Number of students in school	0.000		neg		
47	Percentage of floor carpet	0.000	pos		Noise, IAQ	Reduced reverberance? Less dust?
48	School parent education	0.000	pos			
49	School CalWork%	0.000		neg		
50	Neighborhood upper/affluent economic status	0.001	pos			
51	Neighborhood is prewar vintage	0.000	pos			
52	Two exterior doors	0.000	pos		IAQ, Daylight	Cross ventilation? Finger plan classroom?
53	Lamp color is warm (CCT<3500)	0.000		neg	Lighting	Older lighting system? Poor maintenance?
54	Room area	0.000	pos		Room	More room for students and teachers?
55	T8 lamps	0.000	pos		Lighting	Newer, better quality lighting system?
11 to 22	8 Outlier Students	0.012				
	Total R^2	0.255				

Figure 30: Order of Entry and Partial R^2 in Final Reading Model

These values are even more impressive when compared to the R^2 of variables typically considered in educational policy such as % attendance (0.003 math, 0.001 reading), eligible for free or reduced lunch indicating low income status (0.001 math and reading), or the number of students in the school (0.000 reading only).

7.3.5 *Daylight Code* versus Predicted Effect for All Window Characteristics

With the replication model, we learned that the *Daylight Code* was not significant in FUSD when considered against the same type of variables used in the Capistrano study. In the thematic and final models, when we added information about the window and classroom characteristics, the *Daylight Code* entered the models as significant and negative. However, in these models each classroom not only has a *Daylight Code*, but also many other window characteristics that influence learning. It is the net effect that matters. Thus, we calculated the net effect for each classroom of all window characteristics plus its *Daylight Code*. We then plotted this net effect against the *Daylight Code* of that classroom in Figure 31 and Figure 32. These plots present an interesting story. The polynomial trend lines suggest that the window arrangements of classrooms are having more positive effects at the top and bottom end of the scale. This suggests that, in FUSD, there is something positive about classrooms with a very low *Daylight Code* and those with a high *Daylight Code*.

In order to understand this pattern better, we went into the data and looked at which classrooms were predicted by the model to have the most positive, and the most negative effects, as determined by their window characteristics.

The following pattern emerged:

For math, the classroom with the best performance based on their window characteristics (+2% to +8%), are either finger plan classrooms with a *Daylight Code* of 5 and a view of vegetation out of a north window with blinds, or they are grouped classrooms in a pinwheel or pod school with a *Daylight Code* of 0.5-1, with good window control and no glare caused by the window. The worst performing classrooms (-20% to -30%) are either portables or low window code classrooms with east facing windows with no view and no controls at the window. Classroom with any security measures on the windows (bars, mesh) also tend to rate very low.

For reading, the best performing classrooms (+15 to +25%) are mostly *Daylight Code* 4 classrooms with a north view and two doors, or portables facing north, also with a view and good window controls. The worst performing classrooms are east facing portables with no view and no window controls, or east or south facing traditional classrooms with a medium *Daylight Code* (2-3) with no window (or shading) controls, and often no door to the outside.

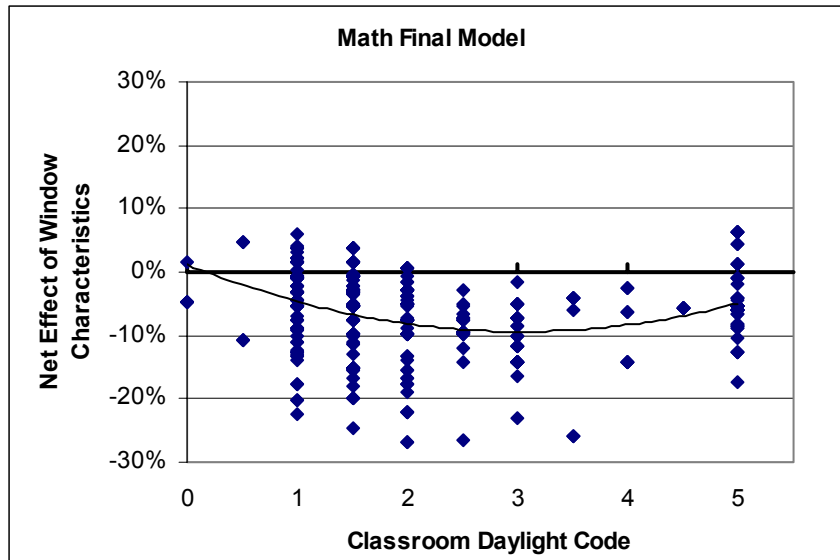


Figure 31: Plot of Percentage Effects for All Window Characteristics versus Daylight Code, Math Model

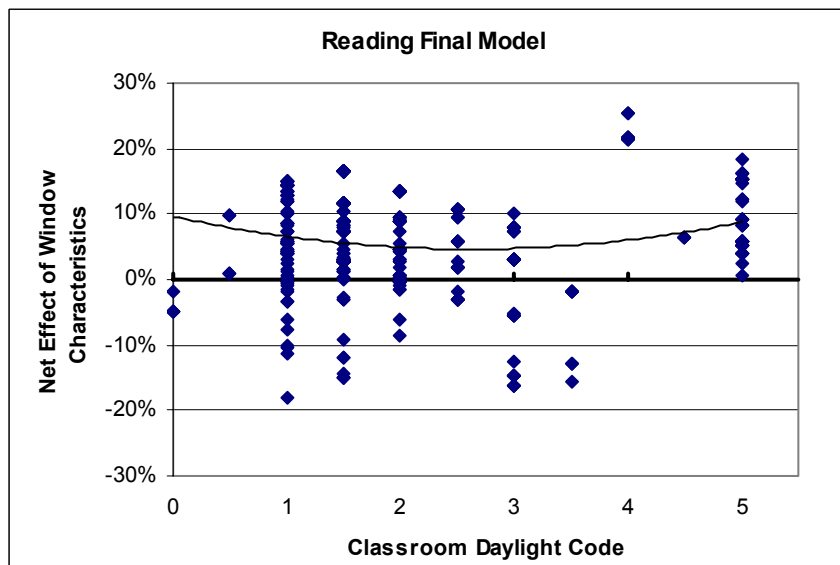


Figure 32: Plot of Percentage Effects for All Window Characteristics versus Daylight Code, Reading Model

Thus, all other things being equal, finger plan, portable and grouped classrooms all seem to be performing quite well as long as they have a good view and sun control. Classrooms at the worst end of the scale tend to be either poorly oriented portables without a view or window controls, or traditional classrooms with modest sized windows that have poor sun control and poor views.

From this exercise, we concluded that one of the reasons we saw no significance or positive trend for the *Daylight Code* in Fresno was that there were classrooms at both the low and the high end of the *Daylight Code* that were performing well.

7.4 Regression Analysis Conclusions

In the replication model the *Daylight Code* was not significant. We then added additional information to the models about details of the window characteristics of the classrooms, accounting for area, orientation, view and glare. We tested each statistical model with and without the Daylight Code. When we added the Daylight Code the other variables remained essentially the same, but the Daylight Code always came in as significant and negative. This told us that there was some additional characteristic(s) associated with the *Daylight Code* that was either very negative for the high *Daylight Code* classrooms or very positive for the low *Daylight Code* classrooms. Our final step of calculating the net effects of actual combinations of window characteristics told us that many of high *Daylight Code* classrooms were indeed performing very well, but so were some of the lowest daylight code classrooms.

The regression analysis was all based on data that were collected during August, when the classrooms were not occupied. We reasoned that it was possible that there were ways that the classrooms were being operated during the school year that we could not have observed in August that might be influencing our findings. For example, if teachers tended to close their blinds or paper over their windows more often in high *Daylight Code* classrooms, then our scale might have been misapplied. Alternatively, the negative affect attributed to the *Daylight Code* by the regression models could actually be caused by some operational condition systematically associated with the *Daylight Code*. For example, if high *Daylight Code* classroom were more likely to have poorly functioning HVAC systems, then the thermal discomfort caused by the HVAC system might cause poor student performance but be attributed to the *Daylight Code* by the regression models.

We decided to go back on-site to observe a sample of classrooms in operation to see if there were any obvious operational issues which were systematically associated with the *Daylight Code* that might be influencing our results. The February Phase 2 data collection and analysis was designed to try to understand if there was a quality of the *Daylight Code* that we had left out from consideration, or if the *Daylight Code* had been wrongly applied. Our observations and the findings of the Phase 2 data collection are discussed in the next section.

8. PHASE 2 ANALYSIS

We used the Phase 2 data collection to try to answer the following questions raised from the regression analysis:

- Did the *Daylight Code* reflect actual operating conditions in the classrooms?
- Was there some aspect of lighting quality in the daylit classrooms that might negatively affect student performance?
- Was there some aspect of thermal comfort in the daylit classrooms that might negatively affect student performance?
- Was there some aspect of air quality in the daylit classrooms that might negatively affect student performance?
- Was there some aspect of acoustic conditions in the daylit classrooms that might negatively affect student performance?
- Were there any other systematic problems associated with more daylit classrooms that might be responsible for a negative effect?
- Alternatively, was there some aspect of the non-daylit classrooms that would positively affect student performance?

While the on-site observations were one-time observations, they did provide a standardized method of analyzing the classroom environment during operational conditions. Overall, the surveyors' observations matched well with the teachers' assessment of the classrooms based on their survey responses. The surveyors also conducted informal interviews with the teachers. The interviews targeted the teacher's choices in controlling the classroom environment via doors, windows, blinds, thermostats, fans and lighting controls, as well as the teacher's opinions on the classroom comfort conditions year round. The teachers also provided valuable insights into some problems found consistently in various classrooms.

8.1 Study Population and Methods

We analyzed classroom characteristics for 104 classrooms by combining data from Phase 1 onsite, Phase 2 onsite and teacher surveys into a common database in order to facilitate comparison. We used Pearson's correlation coefficients to study relationships between the teacher survey responses and the Daylight Code. In addition, information from the February surveys was also analyzed with simple linear regressions between two variables.

Of the 40 classrooms visited in Phase 2, 38 were also visited in Phase 1; two comparable classrooms were added to the Phase 2 sample to get more data on certain classroom features.

We asked to receive teacher surveys from all third through sixth grade teachers in the 14 schools we visited in Phase 2. We received teacher surveys for 116

classrooms, of which 87 classrooms were included in the Phase 1 database and assigned a *Daylight Code*. Figure 33 summarizes the relationship of the three datasets used in the Phase 2 analysis.

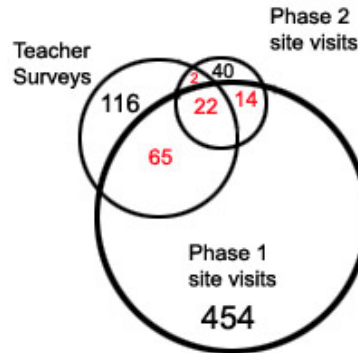


Figure 33: Venn Diagram of Phase 2 Study Population

8.2 Reduction in Daylight

Are the teachers covering the windows or closing the blinds more in daylight classrooms than in non-daylit classrooms, so that there is actually much less daylight than would be expected from the August survey?

Yes, a little bit, but not enough to affect regression findings.

In the teacher survey we did not see any significant difference in the amount of time teachers claimed to close blinds or paper over windows between the more and less daylit classrooms.

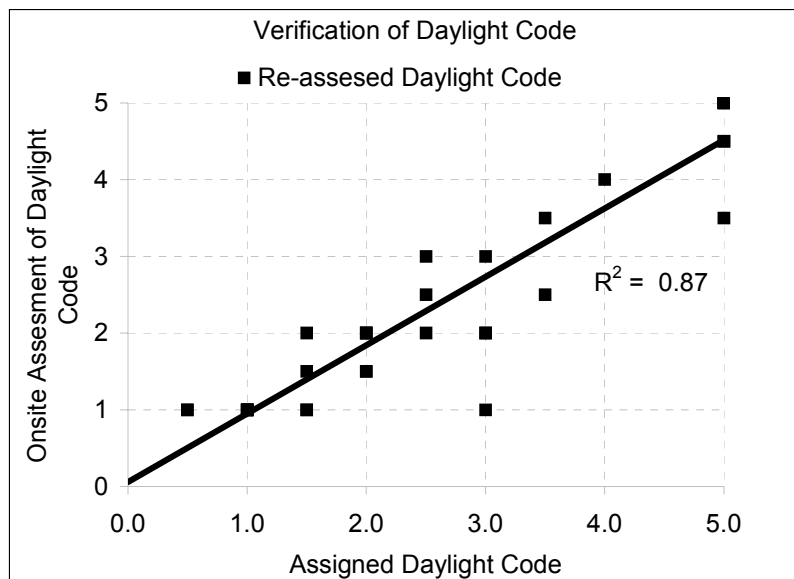


Figure 34: Assigned Daylight Code versus Operating Conditions

When we visited classrooms we checked the ranking of the classrooms by *Daylight Code*, given the conditions that we found during operation. In all cases the *Daylight Code* was rated at the same or lower code than before. Figure 34 represents a plot of originally assigned *Daylight Code* versus surveyors' observations in February. It shows that on average, a *Daylight Code* 5 classroom would be reclassified to a 4.5 code during operation, or a ten percent reduction. Thus, a *Daylight Code* 3 classroom would be assigned a new 2.7 code. However, this modest shift in the fitted line would not affect our regression analysis.

NOTE: In Figure 34 and similar following graphs, the points shown plotted can represent more than one occurrence. The Pearson's correlation (reported in parenthesis in the text) was used to judge the correlation. The graph serves merely as a visual representation of the relationship.

8.3 Classroom Lighting and View

Are there pervasive lighting quality problems in the daylit classrooms that might be interfering with student performance?

Probably not

In general the teachers preferred the lighting quality in the more daylit classrooms, although this difference was not significant ($p=.33$).

The most highly significant finding from the teacher survey regarding lighting quality is the lower the *Daylight Code*, the more likely teachers reported that they "did not have enough natural light" ($p=.001$). This was actually the strongest correlation and largest magnitude effect in the teacher survey, and somewhat reassuring in our assessment of daylight presence in the classrooms.

8.3.1 View

To a lesser extent, but still significantly, teachers were more likely to report distraction from the windows the higher the *Daylight Code* ($p=.01$). We also interviewed the teachers whenever possible about the distraction issue. The teachers who were most impassioned about the distraction problem were those in classrooms where the exterior circulation path moved children directly outside of windows, such in the portable, pod and pinwheel classrooms and some finger plan classrooms that still had low south facing windows. They mentioned individual students peering in the windows looking for their friends and columns of classes passing close-by causing a distraction. Teachers in classrooms with only low north facing windows that looked out on to a landscaped strip, with a circulation path at some distance (20' +/-) from the window, did not complain about the window distractions. This implies a rule-of-thumb for school designers—exterior circulation paths should be kept at some distance from classrooms, and when that is not possible, at least there should not be a low view window between the class and pathway.

As might be expected, teachers were also more likely to agree with the statement “I wish we had a better view of the outside” as the *Daylight Code* decreased (but with low significance, $p=.39$). The slopes between the two questions about view are almost perfect inverses of each other, as shown in Figure 35. Since the black view line always stays below the center value, it also suggests that teachers are more dissatisfied with their view, or lack thereof, than they are bothered by the distractions from the windows.

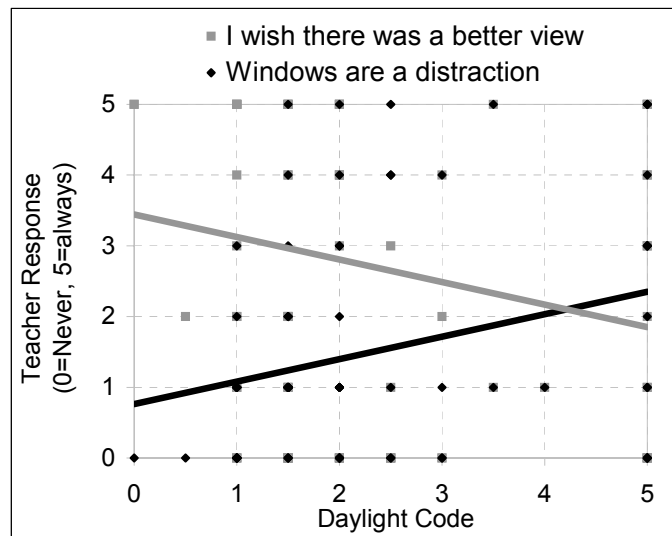


Figure 35: Teacher Rating of Window View and Distraction

In our February observations we also noted the position of blinds and curtains. There were a few clear patterns. Blinds or curtains on north or west facing windows were almost always fully or partially open, while those on unshaded south-facing windows were almost always fully closed. Blinds or curtains on low, view windows that were directly adjacent to a student circulation path were almost always closed. If these windows did not have blinds or curtains, often the teacher would paper-over the windows, or place high furniture against the window to block the view of the pathway. These observations reinforce the school design suggestion made above, and also strongly suggest that south-facing windows will provide little benefit of daylight or view unless they are shaded.

8.3.2 Glare

There was absolutely no correlation, positive or negative, between the *Daylight Code* and teachers' responses to the statements “Some areas of the room are too dim” or “There is not enough control of the lighting conditions.” However, teachers in daylit rooms were slightly more likely to report problems with glare from both electric lights and sunlight, and more reflections on the teaching board, although none of these trends were significant either ($p=.10-.40$).

Surveyors rated the classrooms for potential glare from reflections on the white board or black board. This rating was found to increase for the classrooms with higher *Daylight Codes* (Figure 36).

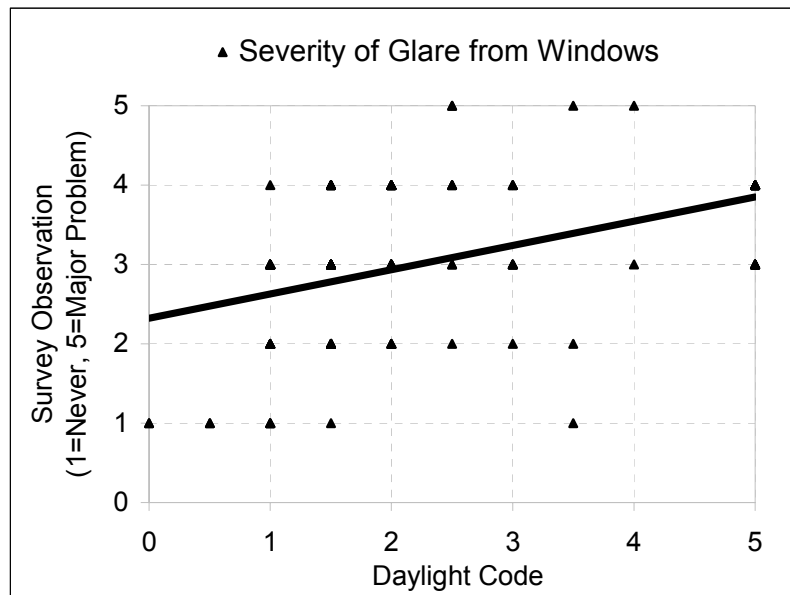


Figure 36: Surveyor Rating of Glare Potential from Windows

This is primarily due to large window areas with minimal shading on the north side, receiving glare from the sky and adjacent buildings. This glare rating was included in the regression analysis and proved significant and a negative influence on math performance. Since this glare rating was included in the model, it should have controlled for the effects of window glare, and allowed the *Daylight Code* to operate independently. Thus, even though daylight classrooms are indeed associated with more glare, we do not believe that this is the reason for the negative association with daylight in the models.

During the informal interviews, many teachers expressed a desire to have more daylight in their classrooms provided the glare and distraction concerns were resolved. A teacher in a school with a south-facing, unshaded window volunteered that even though the sun comes into the classroom and can be glaring and hot, she leaves her black-out curtains open most of the time because she believes that “daylight is good for the kids” and “I need to see outside.” One solution often mentioned by teachers in the pinwheel schools was adding clearstory windows or skylights to the high ceiling that could address both the issues. During the February survey one pinwheel school had a two-hour electricity blackout the previous morning. We asked the teachers how they coped with the blackout. One took her class outside to read, but found that it was too cold to sit still, so she decided to let them run around until the power came back. Two others opened the curtains to the little windows in their rooms, and asked their students to read quietly by the meager daylight available.

Thus, overall, we found that teachers appreciate the presence of daylight and a view, and feel the overall lighting quality of the daylit classrooms is slightly better than those with less daylighting. There are clearly visual quality problems associated with the daylit classrooms, but they are not driving issues of the teachers.

Are these conditions significantly different in Fresno compared to the previous districts studied?

Perhaps slightly.

Classrooms in Fresno with a higher *Daylight Code* are universally of the finger plan or double-loaded category. In Capistrano, classrooms with a high *Daylight Code* had more variety in classroom and school plan types, including three skylit plan types with aggressive daylight but modest view windows. Thus, since there was a much greater differentiation between the presence of view and the presence of daylight in Capistrano, Capistrano was more likely to have good daylight conditions without distractions or glare. Furthermore, in Capistrano's coastal climate, morning fogs are likely to reduce the problem of glare from sunlight during the start of the school day.

8.4 Classroom Thermal Comfort

Are the more daylit FUSD classrooms less thermally comfortable than the less daylit classrooms?

In some cases, maybe.

This question has a number of possible mechanisms, which we attempted to answer using a variety of analysis techniques. Were the classrooms too cold or too hot? When were they too cold or too hot? Were they being operated differently, or was there an inherent problem due to the design of the classrooms that would cause them to be less thermally comfortable?

From the teacher survey, we found that the teachers had a very slight tendency to rate the daylit classrooms as more thermally comfortable ($p=.29$), although those in more daylit classrooms tended to consider them more on the warm side than the cool side.

When we took temperature measurements on-site, the surveyors found some classrooms with very high supply air temperatures, and overall the temperature was higher as the *Daylight Code* increased (Figure 37). This was observed during mild weather that should have required minimal heating, and implied that the thermostat controls for the air delivery were seriously out of adjustment in those rooms. It was also observed in those rooms with high air delivery temperatures that the teachers did not have access to the thermostat setting to correct the problem. Where the teachers had local control, they were observed to use the systems on a "need to use" basis, keeping the temperatures within an acceptable comfort range. Where the teachers did not have control of the

thermostat they solved the temperature problem by opening the windows and/or door to vent the hot air.

In our informal interviews we asked the teachers when their classroom was too hot and when it was too cold. Counter to expectations, they almost always answered that their classroom was too hot in the winter and too cold in the summer. Indeed, many teachers, especially in double-loaded classrooms, spoke of instructing their students to leave a warm winter jacket in the classroom throughout the school year, so that they could put it on when the air conditioning made the classroom too cold. These comments clearly imply that the discomfort in the classroom was caused by over-zealous heating and air-conditioning and not by weather conditions influencing the classroom thermal environment.

As part of our February observations, we observed the student's clothing and counted the number of students wearing T-shirts, long sleeved shirts, sweaters or sweat shirts, and puffy winter jackets. In every room, in every school, we observed a mix of all clothing levels, and could not distinguish a pattern of one type of school or classroom being warmer or cooler, as judged by student clothing levels. We did however note that in the one two-story school, with large unshaded windows facing north and south, a few students sitting next to the north-facing windows were wearing puffy jackets and gloves, while students next to south-facing classroom were wearing primarily t-shirts and no jackets.

8.4.1 Operable Windows and Local Thermostat Controls

The observation that teachers lacked control of temperature settings and were likely to open their windows to adjust temperature led us to investigate whether there was a systematic difference in the availability of local thermostat controls versus centralized temperature controls or in the way the classrooms were operated.

Data from the August surveys showed that daylit classrooms were just about as likely to have local controls as non-daylit classrooms (Figure 37, right axis). However, there was a substantial trend towards higher air temperatures delivered by the heating system while we were there in February (Figure 37, left axis).

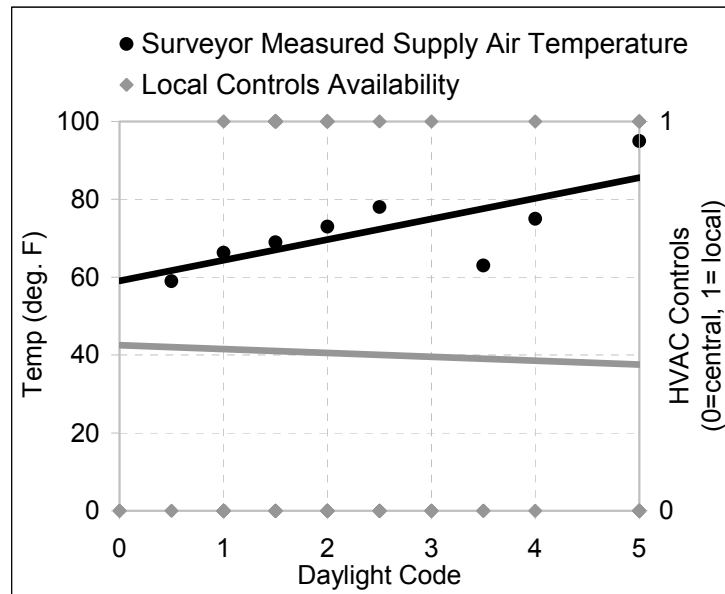


Figure 37: Measured Air Delivery Temperature and Local Temperature Controls

When the temperature becomes uncomfortable in a classroom without local thermostat control, the first line of defense seems to be to “jimmy” the system somehow to turn off the supply of offending conditioned (and ventilation) air. The February surveyors discovered that most of the teachers in classrooms that had centralized controls had figured out clever workarounds to turn the HVAC off, or to turn down the fan speed in order to compensate for the unusually high supply temperatures. This was in spite of FUSD policies that forbid the teachers from doing any local adjustment to these centralized controls. Typically the teachers would first explain the policy and their understanding that it was necessary for energy conservation, but then apologize that they simply could not teach in an uncomfortable classroom.

Even though we found little difference between the number of daylit versus non-daylit classrooms with and without local controls, there may be an important difference in how the lack of local controls affects the self-contained daylit classrooms compared to the grouped or open-plan classrooms with little daylight. This is because the more daylit classrooms, especially those classrooms with windows on only one side, typically rated *Daylight Code* 3 or 4, are more likely to experience unequal radiant loads, depending on which orientation they are facing. The grouped and open plan classrooms with low *Daylight Code* ratings of 0-2, on the other hand, have less exposure to radiant effects and re-circulate the conditioned air among all classrooms, so that there is more mixing of air across conditions.

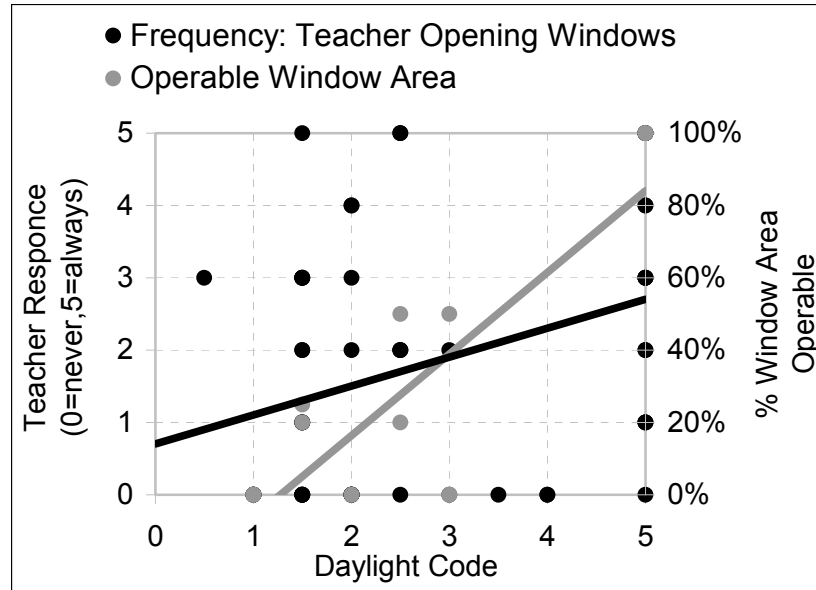


Figure 38: Frequency of Classroom Operable Window Operation

Faced with thermal discomfort, the teachers in high *Daylight Code* classrooms opened windows more often, as discussed above. The teacher surveys clearly indicated that the teachers use operable windows more often in classrooms with higher *Daylight Code*. It is to be noted that the classrooms with higher *Daylight Codes* also have larger operable window areas (Figure 38) while the classrooms with *Daylight Codes* 0-2 rarely have any windows that are operable. In our interviews teachers told us they tend to use the windows to compensate for over-heating or over-cooling from the HVAC system.

8.4.2 Radiant Temperature Analysis

One of the important factors that determine the thermal comfort in classrooms is the radiant temperature of the various surfaces, which affects the students and teachers independently of the classroom air temperature. Typically, automatically controlled HVAC systems are adjusted based on the inside air temperature. However, a person's thermal comfort is a function of many variables, including air temperature, the velocity of the air, their clothing and activity levels, and importantly, the amount of heat radiating at them (or away from them). A person sitting near to a fire will be warmed by the radiant heat. A person sitting next to a cold window may experience a "cooling draft" when in reality they are losing heat primarily by radiating their body's warmth to the cold window surface. Thus, if a classroom has very warm surface temperatures because it is being heated by the sun, or cooler temperatures because it has poorly insulated surfaces exposed to cold outside air temperatures, such as single pane glass or uninsulated concrete block walls, then the people inside can be very uncomfortable in spite of a comfortable air temperature.

During the February survey, we took radiant temperature readings of classroom surfaces, using a type of laser gun designed for that purpose. The survey was conducted when temperatures outside were mild (high temperatures of ~65°F) and as a result the radiant effects observed on the site were not dramatic. Overall, the classrooms with higher *Daylight Codes* had lower surface temperatures due to the larger single pane window areas that stay close to the temperature of the cool outdoor air. The wall, ceiling and floor surface temperatures also trailed the outdoor air temperature in this mild weather. As would be expected, the highest surface temperatures were for surfaces in direct sun, especially unshaded, tinted windows that absorbed the sun's heat.

8.4.3 Radiant Temperature Analysis

We were concerned that the onsite measurements did not give enough variation in temperature conditions that would reflect the conditions during the more extreme months in Fresno. It was therefore decided to generate a computer simulation model that would predict the classroom radiant temperatures for various seasonal conditions. By doing a computer simulation of radiant temperature balances in two classroom types, we hoped to see whether different levels of thermal comfort in daylit and non-daylit spaces could be the source of the negative daylight influence.

In this study, we have used mean radiant temperature (MRT) to assess comfort due to radiant heat transfer. To get an hourly assessment of MRT, simulation software called RadTherm was used. The data collected from the site visits such as construction, geometry, materials and surface coverings were used in the generation of the radiant simulation models. A full discussion of the RadTherm analysis and assumptions are included in the Appendix.

Another method of assessing comfort due to radiant heat transfer is to calculate the Predicted Mean Vote (PMV) on a thermal sensation scale of a large population of people exposed to a certain environment. PMV predicts the comfort level as “felt” by a large population in any given thermal situation. Calculation of PMV takes into account various factors such as activity level, clothing level, metabolic rate and evaporative heat transfer between people and their surroundings. We decided not to expand our analysis to include calculation of PMV for the classrooms, as it would have involved too many variables where we did not have adequate data from the site visits. We instead used MRT which would represent the thermal environment surrounding the students in each classroom.

Analysis Assumptions

We modeled two types of classrooms in RadTherm— a typical finger plan and a typical pinwheel plan. These two classroom types exemplify the extremes of conditions found in FUSD. The finger plan classroom typifies the FUSD classroom with the maximum amount of daylight, while the pinwheel plan typifies those classrooms with minimal daylighting.

For modeling purposes, we duplicated the classroom geometry, materials and orientation as observed during our onsite surveys. We made some simplifying assumptions to provide a common baseline for assessing classroom conditions across the different climates. It was assumed that air temperature in the classroom will be maintained constant by the HVAC system at a comfortable temperature (72 °F) and hence the contribution of conventional heat transfer within the classroom would be minimal and will be driven primarily by surface temperatures. We also modeled only a single student located eight feet from a window, rather than a classroom full of students.

Additional weather variables such as solar radiation, outdoor dry bulb temperature, sky cover, etc., were also input in the RadTherm models using typical year weather files(TMY2) for Fresno, Seattle and Capistrano.

Simulations were run for the first three days of February, May, August and October for each of the two classroom types. These four months were chosen to consider the extreme winter and summer conditions (February and August) and to capture swing-season conditions for fall and spring months during which tests are given to students (May and October).

Analysis Findings

The MRT analysis showed that the finger plan classrooms have noticeably higher radiant temperatures in the spring (May) and summer (August) months than the pinwheel classroom, in all three climates. The three day running room temperatures for the two classroom types for May are plotted in Figure 40. As would be expected, the classrooms in Fresno have the highest temperatures in the spring and summer, followed by Capistrano, then Seattle. Since the classrooms selected for our study are not in session during the summer months, the impacts of higher temperatures in Fresno would not have an adverse impact on the student performance in our study population.

In general, the differences in mean radiant temperatures between the finger plan and pinwheel classrooms for the three locations are not significant during the fall and winter months. During spring and summer, the finger plan classrooms were found to be hotter, with peak mean radiant temperatures about 5-7 deg F greater. Thus, this analysis does not support the hypothesis that finger plan classrooms have likely to have significantly less thermal comfort due to radiant temperature extremes than the comparison pinwheel classroom.

Somewhat surprisingly, Capistrano classrooms are the hottest in the fall and winter months, with Fresno and Seattle in second and third places respectively. This is due to coastal climate conditions, where Capistrano experiences its hottest and clearest weather in fall and spring, while in the summer time ocean fogs tend to reduce radiant temperatures. Figure 39 illustrates the basic difference between the three climates we have considered in our analysis—Fresno, Capistrano and Seattle. It shows the percentage of daytime that skies are clear and sunny, and the average daily peak temperature for each month of the year. Fresno has the greatest seasonal extremes, while Capistrano has the

most uniform conditions year round. These charts are derived from Typical Meteorological Year (TMY2) data.

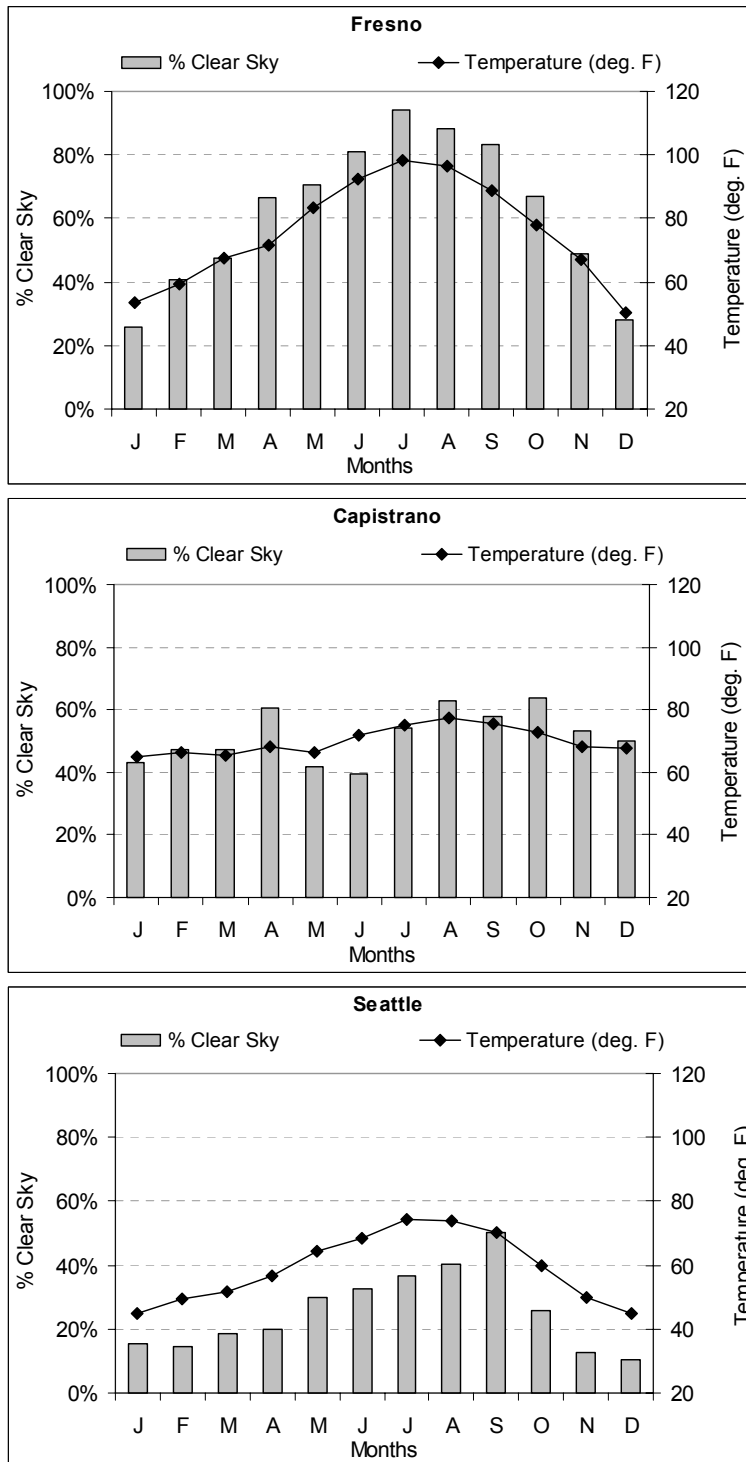


Figure 39: Comparison of Three Climates

Another interesting observation was that the profile of MRT for the south facing pinwheel classrooms shows a pointed peak in temperature during the middle of

the day, compared to a more rounded profile of the finger plan classrooms, as shown in Figure 40. This sharp peak is clearly caused by the heat radiated through an insulated, but unshaded south wall in the pinwheel classrooms. In addition, a slower afternoon cooling trend is attributable to the additional heat reradiated by the playground black-top paving directly adjacent to the exterior wall. The south walls of the finger plan classrooms are well shaded by both overhangs and vegetation and so avoid much of this heat pulse through their walls during the middle of the day. Most of their rise in MRT is instead attributable to the large expanses of single pane glass that conduct heat from the rising air temperature.

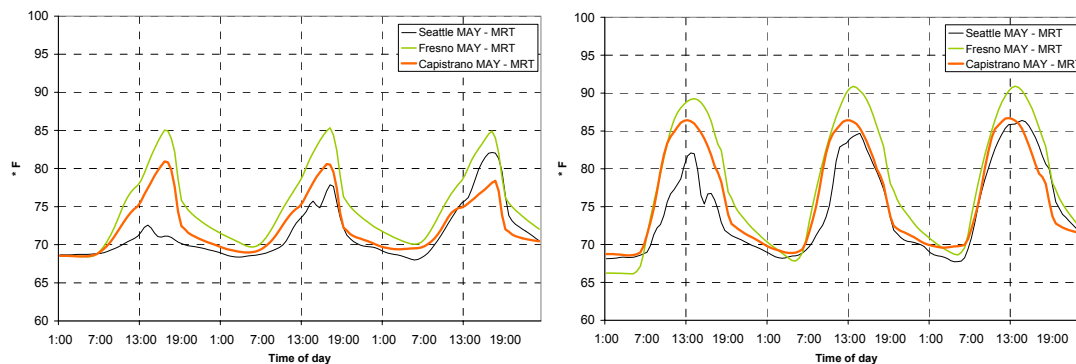


Figure 40: Mean Radiant Temperatures, pin wheel (left) versus finger plan (right) classrooms, for the first three days of May

Are these conditions significantly different in Fresno compared to the previous districts studied?

Probably not.

We attempted to compare the effects of mean radiant temperature for the finger plan classrooms across three climates, and found that during the school year there is not a significant difference in effects of the three climates. Thus, we conclude that the finger plan type of classroom does not pose greater thermal discomfort due to radiant effect in Fresno than in Seattle or Capistrano, nor does it create obviously worse conditions than the pinwheel classroom (facing south).

It is more likely that differences in operation and management of the HVAC system are responsible for any thermal discomfort in the classroom. We did not study HVAC operation in Capistrano or Seattle, or survey those teachers about their behavior in operating the classrooms, so we cannot compare the three districts on this score.

Simple Radiant Comfort Improvements

There is a simple school planning approach that could improve thermal comfort in all classrooms in hot climates like Fresno, by reducing radiant heat transmitted through exterior walls—all east, south and west facing classrooms should have a landscaped strip located directly outside with bushes and trees to shade the walls from the hot sun, as opposed to extending the black-top paving right up to

the walls of the school. Higher levels of insulation in the walls would also reduce this effect, but without providing the other advantages that a vegetation strip can provide by reducing noise (see discussion on acoustics about children banging on walls) and providing a pleasant view (see discussion on view).

We also tried to assess the thermal comfort impacts of improving the window glazing in the finger plan classrooms. We wanted to see how much of a difference high performance glass could make for radiant comfort in a finger plan classroom. Changing the single pane, clear window glass to double pane low-e glass would greatly reduce thermal transfer through the large window areas and keep the interior surface of the windows much closer to the conditioned air temperature of the classroom. We found that the addition of high performance glass to the finger plan classrooms did indeed reduce the mean radiant temperatures significantly, with an average drop of around 7 degrees in the summer peak and 5 degrees in the spring peak.

The MRT for the modified finger plan classroom became similar to that in the pinwheel classrooms, minus the noon peak effect for the pinwheel classrooms due to lack of wall shading discussed above. This analysis confirms that with high performance glass HVAC air delivery temperatures can be more moderate (thus both saving energy, and reducing the HVAC temperature problems we observed on site) while still maintaining overall thermal comfort in finger plan classrooms.

8.5 Classroom Indoor Air Quality

Do the daylight classrooms in FUSD have worse air quality than the non-daylit classrooms?

Probably not.

The classroom indoor air quality includes various factors such as air movement, quality of air, amount of pollutants in the air, smells, mold, dampness, etc. The teachers rated their classrooms individually for these factors, and also gave an overall rating of satisfaction with the classroom air quality.

There was no significant difference in how teachers rated the ventilation quality of their classrooms, although there was an indication that teachers in the more daylit rooms were slightly more likely to report good ventilation quality ($p=.22$), and those in more daylight classrooms were also more likely to consider them drafty than stale.

Some teachers raised a concern with the black and green boards in the classrooms, which they felt might trigger student asthma attacks from excessive chalk dust. Indeed, the surveyors observed that most of the black and green boards were papered over and not used. The white boards, in contrast, mostly showed evidence of recent use. Since the type of teaching surface was controlled for in our models, we do not believe that any negative air quality influence associated with teaching surface type would affect the *Daylight Code*

results. In our regression findings, the presence of a white board was a positive influence on math learning. Based on our February observations we would interpret this to mean that white boards are more actively used by the teachers in math instruction.

The surveyors also measured the CO₂ levels in twenty five classrooms during their visits. The CO₂ levels that we measured were generally higher than the current national standards prescribed by the American Society of Heating Refrigeration and Air-conditioning Engineers (ASHRAE)—ASHRAE standard 62-1989 recommends a maximum CO₂ level of 1000 ppm (parts per million). The average CO₂ level recorded was 1257 ppm. The lowest value was 680 ppm and the maximum value was a somewhat alarming 3147 ppm. The worst offenders in terms of CO₂ tended to be the portable classrooms where the surveyors also noted stale air or smells.

There were two obvious air quality problems noted by the February surveyors. One, a number of teachers in the southern end of the district commented on the need to close their windows occasionally to avoid dust and noxious smells from outside the school—from industrial and agricultural sources, street and sewer repair projects, and/or leaf blowers operated by landscaping crews. Thus the teachers who relied on operable windows for ventilation were faced with a dilemma: reduce ventilation in the classrooms or put up with the dusty or foul smelling air. Generally they chose to reduce ventilation.

Two, a few teachers in portable classrooms had noise problems that they attempted to solve by keeping the windows closed and the ventilating fan turned off—resulting in very stuffy and humid classrooms. Close proximity to the playground necessitated keeping windows and doors closed to reduce noise transmission from outside. But very noisy fan units also made it difficult to hear in the classrooms, so the teachers would choose to teach with the windows and doors closed and the fan off. Humidity and CO₂ would then build up while the class was in session, and then the teacher would ventilate the classroom when the students left for recess by turning the fan on or opening a window.

These problems illustrate the tangled knot of problems created by ventilation, noise, and thermal comfort problems. Teachers often have to choose one criteria at the expense of the other two—they can have thermal comfort but only with more noise, or they can have quiet, but only with poor ventilation.

Overall, we found that classrooms with noisy or poorly controlled HVAC systems were more likely to have poor indoor air quality as the teachers took extraordinary measures to try to overcome these other problems. But portables (with a low *Daylight Code*) were just as likely to experience these problems as high *Daylight Code* classrooms. Therefore, we conclude that there is no relationship between daylighting and poor indoor air quality that might explain the result of the negative daylight effect.

Are these conditions significantly different in Fresno compared to the previous districts studied?

Probably.

We did not study the relationship of classroom type to indoor air quality issues in Capistrano or Seattle, so we cannot compare the districts on that level.

However, most of Fresno's air quality problems seem to come from outside the buildings. Fresno is known for poor air quality, both due to smog build-up and various industrial and agricultural sources of air pollution. Capistrano and Seattle both have coastal climates, where clean air is continuously blown in from the ocean. Thus, it is likely that daylit classrooms with operable windows in Fresno might experience more air quality problems than a similar classroom in Capistrano and Seattle.

8.6 Classroom Acoustic Performance

Do the daylit classrooms in FUSD have more acoustic problems than the non-daylit classrooms?

Very likely.

There was no difference in teacher assessment of the acoustic quality of their classrooms, and especially no difference in their assessment of how well their students can hear. There was however, a significant difference in the teacher's assessment of the source of the noise. As might be expected, in more daylit classrooms the noise is more likely to come from outside of the building (presumably through the windows) while in the less daylit classrooms the noise is more likely to come from other classrooms.

The surveyors rated the outside noise on an intensity scale and found that the sound intensity did not change between the types of classrooms during their observations. Thus it is likely that the problem is intermittent. In this context it is interesting to note that the teachers also indicated that they tend to use operable windows more often in daylit classrooms compared to non-daylit classrooms. The opening of these windows could lead to increased sound penetration into the classrooms from playground and traffic noises.

The surveyors also observed that the daylit classrooms tended to be reverberant. The surveyors rated reverberance in the classrooms during the regular classroom activities on a subjective 0-4 scalar. Reverberance measures the delay time for sound reflected within a space to die down. Spaces that reflect more sound due to hard surfaces, and that have a longer delay time, due to larger distances between reflective surfaces, will have a higher reverberance rating. As reverberance increases, especially within the sound wavelengths represented by human speech, intelligibility is reduced.

The February surveyors also observed that many classrooms had a teacher's assistant working with one or two students in the back of the classroom, often

explaining lessons to them in Spanish. This secondary set of voices, speaking quietly while the teacher was instructing the class, created additional sound interference at the same human speech wavelengths. A reverberant classroom would make such additional speech sounds even more troublesome. In the bungalow classrooms (*Daylight Code* 4 or 5), a free-standing furnace was typically located inside the classroom—another source of internal noise which would be especially problematic in a reverberant room.

Teacher complaints about the less daylit classrooms focused on noise transmitted from hallways or other classrooms. Many of these classrooms are in a grouped plan, with shared internal spaces, and many also have open passageways so that the classroom can never be fully acoustically isolated. Teachers in the pinwheel schools voiced the most noise complaints to the surveyors, with two interesting explanations. One, a number of teachers explained that they have problems with loud sounds created by children banging on the exterior walls of the classrooms, since the outside circulation path and the play area are directly adjacent to the classroom. Two, children in the interior shared spaces often become quite noisy, since the closed doors to the classroom do not make them aware that other classes are in session. In contrast, surveyors observed that students were remarkably quiet in open plan classrooms, where there was an obvious visual connection to adjoining classes. Teachers in those schools explained that the student body had been trained over the years to use “indoor voices” inside the building, and thus they rarely had problems with excessive noise from other classes.

8.6.1 Classroom Reverberance Analysis

Given the findings above suggesting that there might be problems with the reverberance in more daylit classrooms in FUSD, we decided to analyze the acoustic performance of FUSD classrooms in order to quantify the difference in reverberance levels. We used a calculation method known as “the Sabine formula” for a simple analysis of reverberation time using room dimensions, absorption coefficients for materials and the absorbing surface area. While extremely detailed three-dimensional computer simulations of acoustic performance are available, we felt that the two-dimensional Sabine estimate was sufficient for simple, rectilinear classrooms.

We compared the predicted reverberance time for two extreme conditions in our sample of FUSD classrooms: a typical finger plan classroom (higher *Daylight Code* classroom) and a typical pinwheel classroom (lower *Daylight Code* classroom). A modified finger plan classroom with an improved acoustic design was also compared. Following is a summary of the materials input in the Sabine formula for each of the classroom types

Building Component	Finger Plan			Sound Absorption Coefficient (alpha)*	Pin Wheel Plan			Sound Absorption Coefficient (alpha)*
	% of Total Building Area	Material			% of Total Building Area	Material		
Walls	31%	Vinyl	55%	0.08	47%	Vinyl	32%	0.08
		Paper	25%	0.08		Paper	40%	0.08
		Bulletin board	20%	0.74		Bulletin board	20%	0.74
		--	--	--		Plaster on lath	8%	0.05
Window	10%	Glass		0.03	1%	Glass		0.12
Door	1%	Plywood		0.12	1%	Plywood		0.12
Ceiling	25%	Acoustic Tiles		** 0.56	26%	Acoustic Tiles		** 0.56
Luminaires	5%	Acrylic		0.12	--	--		--
Floor	29%	Vinyl		0.03	26%	Carpet		0.37
	100%	<< Total			100%	<< Total		

* Source: Architectural Acoustics by M David Egan, McGraw-Hill, 1988

** Source: Acoustical Surfaces Inc. http://www.acousticalsurfaces.com/acoust_ceilings/ss_ceiling.htm?d=20

Figure 41: Acoustic Analysis Inputs

Findings of the acoustic analysis

The American National Standards Institute (ANSI) “Acoustical Performance Criteria, Design Requirements and Guidelines for Schools” ANSI S12.60-2002, recommends maximum reverberation time for three sizes of classrooms as shown in Figure 42. For the finger plan classroom, the volume is 12,000 cu.ft, hence maximum reverberation time recommendation is 0.7 sec. For pin-wheel plan classroom, the volume is 9972.4 cu.ft, hence maximum reverberation time recommendation is 0.6 sec.

Learning space ^{a)}	Maximum one-hour -average A-weighted steady background noise level ^{b,c)} dB	Maximum reverberation time for sound pressure levels in octave bands with midband frequencies of 500, 1000, and 2000 Hz s
Core learning space with enclosed volume < 283 m ³ (< 10 000 ft ³)	35	0.6
Core learning space with enclosed volume > 283 m ³ and ≤ 566 m ³ (> 10 000 ft ³ and ≤ 20 000 ft ³)	35	0.7
Core learning spaces with enclosed volumes > 566 m ³ (20 000 ft ³) and all ancillary learning spaces	40 ^{d)}	e)

Figure 42: Table from ‘ANSI Acoustical Performance Criteria Design Requirements and Guidelines for Schools’ ANSI S12.60-2002,

From our calculations (discussed in the Appendix) based on the assumptions above, it was found that the reverberation time for finger plan classrooms is **0.77 sec** and for pin-wheel plan classrooms is **0.48 sec**. This supported the surveyor assessment that the finger plan classroom is a more reverberant space than the pinwheel plan classroom. Per this analysis, the finger plan classrooms would fail the recommendation for reverberation time, while the pinwheel plan classroom exceeds them.

From this analysis, we concluded that the FUSD finger plan classrooms may indeed disadvantage the students' ability to listen to the teacher. With larger volumes and more reverberant surfaces, minor distractions within the finger plan classroom could be amplified in significance with a longer reverberation time. The in-class tutoring we observed in Fresno is likely to be more disruptive in the finger plan classrooms, since there is no alternative for the small study groups to work except inside the classroom. In the grouped and open plan classrooms, the hallway or shared workroom is often available for small group work.

Are these conditions significantly different in Fresno compared to the previous districts studied?

Probably.

Capistrano also had many similar finger plan classrooms, which are likely to have similar acoustic challenges. However, it is less likely that these acoustic problems would have been collinear with the high *Daylight Code* in Capistrano, since there were many other types of high *Daylight Code* classrooms, including skylit classrooms with full carpeting and hung acoustic tile ceilings. In Fresno, the older, un-renovated finger plan and bungalow classrooms dominated the high *Daylight Code*, and thus are more likely to have had consistent acoustic problems.

It is also likely that the school overcrowding and in-class tutoring in Fresno exacerbates the acoustic problems of the finger plan classrooms. Overcrowded schools tend to move to multi-session lunches and recesses, so that one group of students is scheduled to have recess while another is in the classroom studying. With large single pane windows, which are frequently opened for ventilation or to reduce the overheating from a poorly controlled HVAC system, much noise is introduced from outside. We did not observe in-class tutoring nearly as much in Capistrano, with a smaller immigrant population than in Fresno—so operation of the classrooms may play a role also.

Is it possible to design daylit classrooms without acoustic problems?

Yes.

There are three opportunities to reduce some of these negative acoustic effects of the finger plan design. First, changing the windows to double glazing, which would also improve thermal comfort and reduce heating and cooling costs, would significantly reduce noise transmission from outside when the windows are closed. Second, offering the teachers control of the HVAC thermostat would reduce the number of times that teachers feel compelled to open the windows as a form of temperature control. Third, increasing the area and quality of sound absorbing surfaces in the classrooms could greatly improve reverberance time.

In the acoustic analysis we considered improvements to the traditional finger plan classroom that would reduce the reverberation time. Higher grade acoustical tiles on the ceiling reduced the reverberation time to **0.69 sec**, just under the maximum recommendation of 0.7 sec. By changing only the flooring from vinyl to a carpet, the reverberation time reduced to **0.54 sec**. A combination of both

measures reduced the finger plan classroom reverberation to **0.50 sec**, essentially the same as the pin-wheel classroom.

Thus, while sub-optimal acoustic performance seems to be associated with the FUSD finger plan classrooms, it is not inherent to the design of that classroom, and can be improved with choice of surface materials.

8.7 Other Observations

While we were on site in February, we also looked for systematic differences between high and low *Daylight Code* classrooms that might not be related to illumination, thermal or acoustic comfort.

From the teacher survey we learned that there might be a systematic difference in the number of years teachers spend in different classroom types. On average the teachers in non-daylit classrooms had been assigned to their current classrooms for 3-4 years while teachers had remained in daylit classrooms an average of 7-8 years. Our regression analysis controlled for years of service in FUSD, but not length of time assigned to a particular classroom. In general, teachers hate to move to a new classroom, since they must personally move all of their supplies and set up new bulletin boards, etc. This finding from the teacher survey implies that FUSD teachers may indeed be better at “holding on to” daylit classrooms, or for some reason there is less administrative shuffling in finger plan schools than pinwheel or pod schools. This is an indication of a slight “assignment bias” whereby some teachers are more likely to be assigned to a daylit classrooms than others.

The February surveyors were also somewhat surprised by two observations about the open plan classrooms. First of all, they found the students in open plan classrooms were surprisingly quiet. Teachers in some of these schools explained that the students are trained from kindergarten to speak quietly indoors and respect the other classrooms nearby. As a result, they seemed to have much less boisterous behavior than students in classrooms that could be physically closed off from each other with doors and walls.

Secondly, the surveyors were impressed with how often teachers in open plan or grouped plan schools mentioned their fellow teachers in our interviews or were seen walking the hallways and conversing in groups. They frequently mentioned “covering for each other” or coordinating testing periods. In the traditional schools with isolated classrooms, we observed teachers conversing together in the workrooms or lunch rooms, but not in the classrooms or (the generally non-existent) hallways. As a result, we hypothesized that the teachers in the open plan schools might have developed especially collegial and supportive relationships. Given that the most successful skylit school plan type in Capistrano (Skylight Type A) is also a grouped plan, with interior corridors and shared workspaces, it is possible that the high student performance associated with that classroom type is partially a function of daylighting and partially a function of the school plan type.

These observations are obviously not systematic, and from a very small sample, and thus could easily be skewed. However, they do suggest two possible mechanisms describing why the open plan schools, which have all low *Daylight Code* classrooms, might be performing better than the high *Daylight Code* classrooms.

In addition, we noted a systematic difference in the use and layout of traditional versus portable classrooms. The rectangular portable classrooms seemed to have a layout advantage over the traditional, square classrooms. Since the teaching wall is along the long 40' wall, student desks can only be arranged three rows deep. As a result, students have better acoustic conditions than if they were sitting in the back fourth row of desks in a square classroom. It is difficult for a student to be more than 20' from a teacher in a portable classroom, while they can easily be 25' to 30' away in a square classroom with the same square foot area. Another potential advantage of the rectangular layout plan for younger children¹ often is that it leaves about an eight foot zone at the far ends of the classroom that are typically set up as small group work areas, tutorial or study areas, whereas in the square plans it is more difficult to carve separate work or study areas out of the classroom layout. Thus, the rectangular portable plan may be more supportive of current teaching approaches that involve small group learning and in-classroom bi-lingual instruction.

8.8 Phase 2 Analysis Conclusions

Based on our Phase 2 observations and analysis, we did not find a “smoking gun” that could easily explain why the *Daylight Code* was consistently negative in the regression models. Thermal comfort and lighting quality issues were not found to be more pervasive in the more daylit versus the less-daylit classrooms. We did find, however, a number of trends that together might tip the balance in favor of better performance in low *Daylight Code* classrooms, once all other window characteristics had been accounted for.

The most obvious trends associated with the *Daylight Code* are the acoustic problems created in the daylit classrooms as a result of more reverberant spaces, combined with high levels of noise transmission from outside due to open windows. We know that high *Daylight Code* classrooms have more operable window area and a greater likelihood that the teachers will open the windows. These structural problems are combined with increased noise outside due to multiple recess schedules and more noise introduced inside the classrooms due to in-class tutoring.

We also observed some potential advantages that open plan and portable classrooms may have over traditional classrooms in the FUSD context. Any one

¹ Since older children are larger, their desks take up more space, and combined with more students per teacher, a sixth grade class will typically fill a portable classroom, leaving no extra small group space.

of these observations, or a combination of them, could be the reason why the *Daylight Code* entered the regression models as negative.

9. ENERGY SAVINGS POTENTIAL

The main focus of this project was to understand the comfort- and productivity-related issues with daylighting in classrooms. However there are energy related issues with daylit classrooms that are equally important for the school district. School budgets are inevitably very tight, and any money saved in energy costs can be redirected to student learning supplies. While the energy efficiency potential of the classrooms was not dealt with specifically in the onsite observations and teacher surveys, we conducted a separate engineering analysis of the FUSD classrooms to understand the potential for reducing energy consumption by using daylighting features in the classroom.

We then used this analysis to extrapolate the potential energy savings for two scenarios. The first scenario is the approximate energy saving potential if FUSD retrofitted all of their existing elementary school classrooms with daylighting controls and better window glazing and electric lights. The second scenario is the approximate energy savings if all new construction of schools in California optimized the classroom design to take advantage of daylighting. These estimates are very rough approximations, intended to give order-of-magnitude answers to these questions.

9.1 Classroom Types Analyzed

Since a variety of site conditions exist within each school and each classroom, analysis of each classroom was not feasible. Hence we conducted the analysis on three prototypical classrooms found in FUSD:

- Finger plan classroom – *Daylight Code 5.0* – This classroom type represents the highest *Daylight Code* among FUSD classrooms. The typical classroom is oriented along the east-west axis with large north windows and high, narrow, fully-shaded windows on the south wall. All glass is clear, single pane.
- Pinwheel classroom – *Daylight Code 1.0* – This classroom type represents one of the lowest *Daylight Codes* in FUSD classrooms, with minimum daylight penetration in the classrooms. These classrooms share three walls with other classrooms, and have one exterior wall with a small tinted window. The typical classroom is analyzed with the external wall facing west.
- Classrooms along an interior corridor – *Daylight Code 2.5/3.0* – These classrooms are arranged along a central corridor, and have one external wall with large, unshaded windows. The single pane, tinted windows have vertical blinds for controlling glare and sun penetration. We analyzed two orientations for the typical classroom, north and south, since with unshaded windows, there is potentially a large difference in their performance. This classroom type, in a two-story building, has been adopted as the prototype for future elementary school construction in FUSD.

9.2 Energy Analysis Methodology

We conducted the analysis using an energy simulation software tool called eQuest, a simplified front-end for the DOE-2 calculation engine. eQuest allows detailed building geometry, construction materials and HVAC and lighting systems. Some modeling assumptions were standardized in order to facilitate comparison between types. For example, the models were run with the same lighting system and same lighting power density in all four classroom types, even though they were not observed to be the same on-site. For the lighting controls analysis we assumed a target illumination level of 50 foot-candles at the desk level, the observed norm in the district.

We used a standard ten month school year for analysis, and used schedules for the HVAC equipment, lighting and occupancy specified by the California energy code. No occupancy was assumed during the summer months or on weekends and school holidays in the classrooms. The analysis was conducted on an hourly basis for the whole year, using a standard (TMY) weather tape for California climate zone 13.

Figure 43 offers a brief summary of the classroom characteristics input in the eQuest models.

Classroom Type	Area (sf)	Window Orientation	Window Glass	Lighting Power		Lighting System	Construction		
				Installed	Modeled		Wall	Roof	Floor
Finger Plan	960	North + South	Single Pane Clear	1.1	1.25	T12	Wood Frame R-11	Built-up R18	Concrete Slab w. carpet
Pinwheel	855	West	Single Pane Bronze	2.1	1.25	T12	Wood Frame R-11	Built-up R18	Concrete Slab w. carpet
Interior Corridor – North	870	North	Single Pane Bronze	1.25	1.25	T12	Wood Frame R-11	Built-up R18	Concrete Slab w. carpet
Interior Corridor – South	870	South	Single Pane Bronze	1.25	1.25	T12	Wood Frame R-11	Built-up R18	Concrete Slab w. carpet

Figure 43: Classroom Characteristic Input in eQuest Models

9.3 Measures Analyzed

We analyzed the impact of three types of three energy efficiency strategies related to the daylighting features in the classrooms: 1.) using of automatic lighting controls to turn off the electric lights when there is sufficient daylight available 2.) replacing the existing lighting with a more efficient system with a lower lighting power density (LPD) and, 3.) changing single pane windows to double pane with a high efficiency, low-e coating. For the lighting controls we looked at three options: a.) a simple on/off system b.) a slightly more complex system that turns off the lights in two steps, 50% and 100% and c.) a high end system that would provide continuous dimming in response to changing daylight levels.

We also analyzed combinations of these measures to find the optimum combination. While we present the savings potential of these measures, we did not attempt to calculate construction costs or to provide a cost-benefit analysis of these measures within this report. Figure 44 summarizes the measures considered in the analysis.

-- Energy Efficiency Measures
 1 -- Automated Lighting Control - ON/OFF
 2 -- Automated Lighting Control - ON/50%/OFF
 3 -- Automated Lighting Control - Continuous Dimming/OFF
 4 -- Improved Window Glass - Single Pane to Double Pane Low-e
 5 -- Lower LPD - Replace T12 lamps with T8

Figure 44: Classroom Energy Efficiency Measures Analyzed

9.3.1 Analysis Findings

Based on assumptions made in the model, the finger plan classroom would benefit most from the combination of measures, with a 22% reduction in total energy use. The south-facing interior corridor classroom comes next, with 19% savings. Figure 45 summarizes the total (combined lighting and HVAC) savings for the four classroom types in Fresno's climate.

Classroom Type	Finger Plan		North Window		South		PinWheel	
Daylight Code	5.0		3.0		2.5		1.0	
Base Case, Lighting + HVAC Energy Use (kWh)	7.06		6.73		7.35		4.99	
Energy Savings v. Base Case:	kWh/sf	%	kWh/sf	%	kWh/sf	%	kWh/sf	%
Measure 1 Lights auto off	1.20	16%	0.28	3%	0.70	8%	0.11	2%
Measure 2 Lights auto 1/2 or full off	1.33	18%	0.60	8%	1.01	12%	0.23	4%
Measure 3 Lights auto dimming	1.43	19%	1.20	15%	1.41	17%	0.44	8%
Measure 4 Improved glass	0.40	5%	0.09	1%	0.21	3%	0.04	1%
Measure 5 More efficient lights	0.29	4%	0.57	7%	0.60	7%	0.44	8%
Measures 4 + 5 + 2	1.59	22%	1.01	13%	1.59	19%	0.64	11%

Figure 45: Classroom Total Energy Savings Estimate

The daylighting controls provide the greatest level of savings. In the finger plan classrooms the simple ON/OFF controls have the potential to save about 62% of the lighting energy consumption over the base case classroom, and the 2 step and dimming controls savings about 74% and 84% of the lighting energy respectively. These lighting energy savings correspond to about 16% to 19% of the classroom total energy consumption. Replacing the T12 lamps with more efficient T8 lamps saves about 4% of the annual total energy of the finger plan classroom by itself.

Replacing the window glass with a higher performance glass reduces the lighting energy savings from daylighting controls slightly due to the lower visible transmittance of the higher performance glazing, but produces higher total energy savings due to reduction in the heating and cooling loads on the HVAC system. The double pane glass has two additional important advantages. It

would improve thermal comfort by stabilizing the mean radiant temperatures within the classrooms, reducing the need for heating and cooling in both summer and winter (not reflected in these calculations). And, it would keep the classrooms quieter by reducing sound transmission from outside noises when the windows are closed.

The interior corridor classroom facing south benefits more (19%) than the north facing classroom (13%) from these efficiency strategies due to the greater reduction in solar transmission through the windows by using higher performance glass. The model accounts for the occupants closing the blinds or curtains whenever there is direct sun on the windows. The pinwheel classroom shows the lowest savings from any of these measures due to its minimal window area. The most cost effective retrofit for these classrooms would seem to be the lighting retrofit measure that saves about 8% of the annual total energy on its own.

9.3.2 Retrofit Savings Estimates for FUSD

The annual savings estimates developed above are for prototypical classrooms in *Daylight Code* 5, 3, 2.5 and 1 respectively. These savings would not be identical in all of the classrooms in the respective *Daylight Code* due to different site conditions across the school district. However, to extrapolate savings for other *Daylight Codes*, we fitted a line to predict total energy savings (MWh) for classrooms in each *Daylight Code* (shown in *Figure 46*). Applying this equation, we calculated approximate total energy savings for the population of 500 classrooms analyzed during the onsite surveys, and estimated a potential savings of 576 MWh for the 500 classrooms combined.

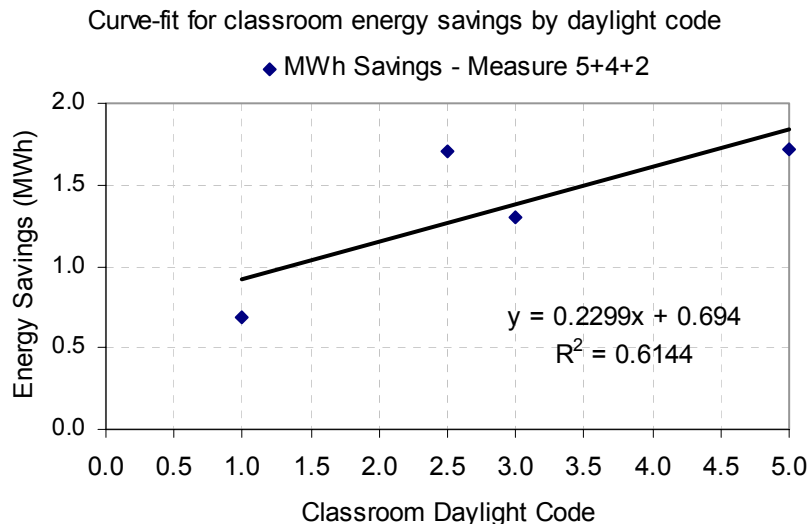


Figure 46: Estimate of Classroom Energy Savings by Daylight Code

To further extrapolate this value to the entire FUSD elementary school population, we assumed that the relative distribution of *Daylight Code*

classrooms was the same throughout the district as in our particular sample. Using the total 2002 elementary student population, and assuming an average of 27 students per classroom, then the approximate annual energy savings for the district to retrofit all elementary classrooms with the three energy efficiency measures would be 1,950 MWh/yr for all of FUSD elementary school classrooms.

Thus, the average energy savings per retrofitted classroom is about 1.15 MWh, or an estimated average power reduction of approximately 1.1 kW per classroom (for combined lighting, heating and cooling effects) over the 10-month traditional school year. Since many of the District's classrooms are operated during the summer months, the energy savings would be considerably higher for those classrooms. However, we did not model summer time effects in our analysis, and so did not include them in our estimates.

9.3.3 Savings Estimate for Statewide New Construction

It is not easy to estimate statewide energy savings or power reductions from daylighting strategies since the energy savings accrued is highly dependant on local climate variation and individual building design and operation. Typically, energy savings from daylighting strategies for schools will be highest in milder, coastal climates or those greatly dominated by cooling loads where the reduction in lighting energy use helps reduce internal heat generation. Daylighting energy savings are generally reduced as heating loads increase. Fresno, in California's Central Valley, has cloudier, cooler winter weather than many other locations in California, especially the heavily populated coastal zones, thus would be expected to have less daylighting energy savings than many other areas with milder winters. In addition, our estimate is only for the traditional school year, ignoring summer energy savings. Thus, if we can assume that the estimated Fresno energy savings represent a reasonably conservative example of the energy savings potential of daylighting in schools, we can use the Fresno analysis to project potential energy savings for the state.

We assumed that all new classrooms would meet the 2005 proposed lighting standard of 1.2 Watts/sf, and would achieve the maximum potential daylighting savings represented by the *Daylight Code* 5 classrooms of Fresno. With this as the new construction base case, the addition of high performance glass and 2 step lighting controls to the *Daylight Code* 5 classrooms saves 0.61 kWh/sf for the ten month period.

Data from the Dodge new construction database shows that educational spaces constitute about 8% of the 84.8 million sf annual commercial construction market.¹ This amounts to roughly 6.8 million sf of annual new construction in education sector. Assuming that roughly 80% of this construction could be daylight using windows, or sidelighting, and applying the savings estimate of 0.61 kWh/sf

¹ Brook, Martha. 2002. *California Electricity Outlook: Commercial Building Systems*. Presentation at PIER Buildings Program HVAC Diagnostics Meeting, Oakland, CA on April 16

derived above, to this construction area, we estimate that the annual savings from using daylighting controls in all new educational spaces would be about 3330 MWh annually, worth about \$0.5 million dollars per year. That value accumulates yearly as new buildings are added each year, so that after 10 years the value would increase ten-fold, to approximately \$5 million dollars per year savings for the state's school districts (in 2003 energy prices and dollars). This represents a very rough estimate of the sidelighting energy savings technical potential for new schools in California.

We estimated the savings possible from using skylights, or toplighting, and daylighting controls for the same area of classrooms separately¹. We estimated savings of 0.88 kWh/sf from top lighting in classrooms. Applying this estimate to the estimated annual new construction in classrooms per above, we estimate total annual savings from daylighting controls with skylights to be 4781 MWh, worth about \$0.7 million per year, or \$7 million in ten years.

Daylighting System	Location	Retrofit Savings		New Construction Savings	
		kWh/sf	MWh	kWh/sf	MWh
Side-Lighting	FUSD	1.1	3195	--	--
	Statewide	--	--	0.61	3300
Top-Lighting	Statewide	--	--	0.88	4781

Figure 47: Retrofit and New Construction Savings from Daylighting Measures

¹ L. Heschong and J. McHugh. *Skylights: Calculating Illumination Levels and Energy Impacts*. Journal of the Illuminating Engineering Society, Winter 2000, Vol.29, No.1, pp. 90-100

10. CONCLUSIONS

We ended up with a highly complex final model. The Fresno data base did not lend itself to simple explanations. There are no definitive answers here, or “proof positive” of any hypothesis. There are however some consistent suggestions about the importance and value of good classroom design, with an assessment of the magnitude of its influence on student performance.

Our studies of the classrooms showed that windows and the resulting lighting quality in classrooms are very much a key issue in learning, and can have both positive and negative impacts on student performance. The surveys show that teachers have strong desire for more daylight and better views, while the regression analysis shows that glare, sun penetration and lack of window controls can negatively impact learning. The regression findings clearly support the theory that interesting window views enhance rather than detract from student learning.

In summary, the findings of this study support the conclusions that:

- The visual environment is extremely important for learning.
 - An ample and pleasant view out of a window, that includes vegetation or human activity and objects in the far distance, support better outcomes of student learning.
 - Sources of glare negatively impact student learning. This is especially true for math learning, where instruction is often visually demonstrated on the front teaching wall. Per our observations, when teachers have white marker boards, rather than black or green chalk boards, they are more likely to use them and, as the regression analysis indicates, children perform better in math.
 - Direct sun penetration into classrooms, especially through unshaded east or south facing windows, is associated with negative student performance, likely causing both glare and thermal discomfort.
 - When teachers do not have control of their windows, student performance is negatively affected. Blinds or curtains allow teachers to control the intermittent sources of glare or visual distraction through their windows,
- The acoustic environment is also extremely important for learning. Situations that compromise student focus on the lessons at hand, such as reverberant spaces; annoying equipment sounds, or excessive noise from outside the classroom, have discernable negative effects on learning rates.
- Poor ventilation and indoor air quality are correlated with lower student performance. However, in FUSD these issues are almost hopelessly intertwined with thermal comfort, outdoor air quality and acoustic conditions.

Teachers often must choose to improve one while making another aspect of the classroom worse.

We also found no evidence that portable classrooms are inherently bad for student learning. Indeed, some portables seem to be performing very well. The problems associated with portables seem to arise when portables deteriorate, and then they become very bad. Given the crowding and budget pressures most school districts in California face, even bad portables will be kept in service by districts that are struggling to provide enough housing for their ever-growing student populations.

We did not find any evidence that higher levels of daylight illumination or more hours of useful daylight per year, as potentially indicated by the *Daylight Code*, are associated with better student performance in Fresno. We did observe, however, that finger plan classrooms in Fresno with high *Daylight Codes* were performing above average, largely attributable to their better views and better sun control. We also noted consistent problems associated with the high *Daylight Code* classrooms, most notably acoustic problems causing more background noise both inside and outside of the classrooms.

These problems can be addressed with better classroom design and material selection. Based on our observations, we would recommend the following:

- ♦ provide quiet, continuous mechanical ventilation in Fresno combined with local teacher control of the thermostat in order to avoid reliance on operable windows for ventilation and temperature control
- ♦ Add more sound absorbing surfaces in finger plan classrooms to help reduce background noise levels from inside the classroom
- ♦ Add dual pane low-e glass to reduce sound transmission from outside the classroom and improve overall thermal comfort
- ♦ Shade all south or east facing windows from the direct sun
- ♦ Add planting strips with trees outside of classrooms to improve both radiant comfort and reduce noise transmitted by students banging on the walls as they pass or play nearby

The addition of automatic daylight controls that reduce electric light use when daylight is available could also save the Fresno district a good deal of money. If the state encouraged their use in new schools statewide, the savings could accumulate to about \$5 to \$7 million dollars per year and 3,330 to 4800 megawatt-hours of energy after ten years of new construction. The energy savings, combined with the positive effects of view out of windows observed in Fresno, or the positive effects of increased daylight observed in Capistrano, create a win-win situation for daylighting design in classrooms. Designers and school officials are advised to avoid designs that create glare or allow direct sun into classrooms, while optimizing the opportunities for interesting views and energy savings with their school designs.

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