

Work Disruption, Worker Health, and Productivity: Evidence from Teaching

Mariesa A. Herrmann
Columbia University

Jonah E. Rockoff
Columbia Business School and NBER

October 2009*

Abstract

We use data from New York City to examine the impact of work disruptions and worker health on productivity in teaching, as measured by student achievement. Extended work disruptions due to leaves of absence and mid-year resignations have large negative effects on math and English achievement, roughly equivalent to moving from the 50th to the 30th percentile of the teacher productivity distribution, or substituting a novice teacher for a teacher with four years of work experience. Work disruptions due to daily absences also have significant negative effects, though to a smaller degree. Additionally, we use the timing of student tests to identify the impact of worker health, above and beyond any relationship it bears with work disruption. We find some evidence that student achievement is lower in years when their teachers miss work for serious illness after students are already tested. Consistent with our empirical strategy, no effects are found for work disruptions that occur after exams for reasons other than health shocks.

* Correspondence should be sent to jonah.rockoff@columbia.edu. We thank Jacob Vigdor and Dick Murnane for their comments on an early draft of this paper. Financial support was provided by the Educational Finance Research Consortium. Mariesa Herrmann's work is also supported by a National Science Foundation Graduate Research Fellowship. Any opinions, findings, conclusions or recommendations expressed in this study are those of the author(s) and do not necessarily reflect the views of the National Science Foundation. All errors are our own.

Recent estimates suggest that about two percent of work time in the U.S. is lost annually due to worker absence, the vast majority of which is caused by poor worker health (Bureau of Labor Statistics, 2008a). Worker absence can have important disruptive effects on production, as labor inputs are unlikely to be perfectly substitutable and contracts are unlikely to require workers to compensate employers fully if they are absent. However, few studies convincingly demonstrate the impact of work disruption on labor productivity.¹

Moreover, many worker health shocks may not result in absence but still significantly reduce productivity. While a large body of literature examines the links between worker health and wages, earnings, labor force participation, and education (see Currie and Madrian (1999), Smith (1999), Currie (2009)), there has been far less work on the impact of worker health on worker productivity, particularly in developed countries such as the U.S.² While a literature in social psychology exists on the impact of health on productivity at work (dubbed “presenteeism”), these analyses use cross-sectional data and self-reported measures of productivity (e.g., Goetzel et al. (2004), Pauly et al. (2008)). Our study aims to begin to fill this gap through a careful analysis of the impact of work disruptions and worker health on the productivity of public school teachers.

In addition to being one of the largest occupations in the U.S., teaching is an attractive setting in which to study these issues. A substantial body of research documents a sizeable impact of teacher quality on students’ academic achievement.³ These studies have generally

¹ The only recent examples we have found are Kleiner et al. (2002), Krueger and Mas (2004) and Mas (2008), who present convincing evidence on the productivity impacts of work disruption caused by labor disputes at, respectively, a large U.S. commercial aircraft manufacturer, the tire manufacturer Bridgestone/Firestone and the construction equipment manufacturer Caterpillar.

² Strauss and Thomas (1998) cite several experimental studies from the 1970s on worker health and worker productivity among agricultural workers in developing countries.

³ See Rockoff (2004), Rivkin et al. (2005), and Aaronson et al. (2008). While there is debate regarding causal impacts of teachers (see Kane and Staiger (2008) and Rothstein (forthcoming)), our identifying assumptions concerning variation within teachers are weaker than those needed to identify variation in quality between teachers.

assumed teachers work continuously over the school year, but, in practice, nearly all teachers are absent at some point each year, and a significant fraction take extended periods of leave due to health shocks. A substitute teacher will be present in the classroom during periods of work disruption, but he or she may provide lower quality instruction. Finally, while lost work time due to absence is roughly the same for teachers as for all workers nationwide (Bureau of Labor Statistics, 2008b), there is evidence that school teachers show up to work despite health problems more frequently than workers in other occupations (Aronsson and Dallner (2000)).

In this paper, we use detailed panel data on teachers and students in New York City that contains precise information on the timing and causes of work disruptions in teaching. By disruption, we mean any period of time when a student's regularly assigned teacher is absent, either for a single day or for an extended period of time, or a teacher's departure during the school year. In order to separately identify the impacts of work disruption and worker health, we take advantage of the timing of student exams, which are taken several months prior to the end of the school year. Work disruptions that occur after students have taken their exams cannot have a causal effect on test scores via work disruption. However, disruptions due to serious health conditions are likely to bear a relationship with worker productivity prior to exams via correlation with prior health status. In contrast, work disruptions for reasons other than teacher health that occur after exams should have no impact on test scores.

A general concern in studies of the impact of teachers on students is that observable teacher characteristics are correlated with important omitted variables. For instance, teachers who are absent less often may also work harder outside of the classroom and be more effective in general. For this reason, we control for teacher-school-grade fixed effects, thereby estimating the impact of work disruptions and health shocks by comparing the productivity of a teacher in

the same school and grade over time. Additionally, we control for a wide set of observable student characteristics, including students' prior achievement.

We find clear evidence that work disruptions in teaching have a significant negative impact on student achievement. We estimate that an extended work disruption prior to student exams causes achievement to fall by 0.07 and 0.04 standard deviations in math and English, respectively. These effects are roughly equivalent to moving from the 50th to the 30th percentile of the teacher quality distribution, or substituting a novice teacher for a teacher with four years of experience.⁴ We also estimate that an additional 10 teacher absences prior to student exams reduce achievement by 0.022 and 0.008 standard deviations in math and English, respectively.⁵

We also find support for the importance of teacher health in determining productivity. Specifically, there is a significant negative relationship between student achievement and teacher absences after the exam for illnesses that are certified by a doctor. Extended disruptions and daily absences occurring after the exam for other reasons are not significantly related to student achievement, supporting the interpretation that teacher health issues have important causal impacts on student learning, above and beyond the effects of work disruption.

In Section 2 we describe the data and provide descriptive statistics, and in Section 3 we describe our methodology and present our empirical estimates. Section 4 concludes.

2. Data and Descriptive Statistics

The New York City Department of Education (hereafter “the DOE”) is the largest school district in the U.S. Importantly for our study, the DOE keeps careful records of the type, timing,

⁴ Estimates of the variation in teacher quality and returns to experience are based on an analysis of New York City data by Kane et al. (2008). Note that their analysis excludes teachers with an extended work disruption.

⁵ A number of previous papers investigate the impact of teacher absence. Two recent studies (Miller et al. (2008), Clotfelter et al. (2009)) that use a teacher fixed effects approach find estimates similar to ours: 10 additional teacher absences reduce student achievement in math by 0.017 to 0.033 standard deviations. There is also a considerable literature on teacher absences in developing countries (e.g., Chaudhury et al. (2006), Duflo and Hanna (2005)).

and duration of teachers' absences, extended leaves, and departures from teaching. Our data come from several files. First, we have information on the complete "service history" of all full-time teachers employed during the school years 1999-2000 through 2007-2008. A teacher's service history file consists of a set of observations representing a period of time in the teacher's employment during which his/her work activity did not change. Most important for our analysis are variables for work status (e.g., active duty, retired, on maternity leave), and the start and end dates of each service period. We use work status codes to identify extended work disruptions, which we initially group into 11 categories: Maternity Leave, Child Care Leave, Medical Leave, Sick Family Member Leave, Personal Leave, Sabbatical, Resignation or Retirement, Involuntary Termination, Certification Termination, Death, and Other Leave.⁶ "Other leaves" include events such as military deployment, teaching abroad or in a charter school or university, or working for the teacher's union. Certification Termination refers to termination of teachers who lacked required credentials. These occur primarily just before the school year 2003-2004, when certification requirements were strictly enforced in New York State (see Kane et al. (2008)).

Extended work disruptions can span school years and impact students when they end as well as when they begin. We therefore match each extended disruption to the years during which it began and ended. If the event began or ended in the middle of the school year, we consider it disruptive. We also link events that start in the summer months to teachers from the preceding school year so that we can test for any relationship between student achievement and these "non-disruptive" events.

⁶ Although there are many work status codes, many represent variations on the same type (e.g., "Sabbatical-doctoral Study 1 Year" vs. "Sabbatical-doctoral Study 6 Months.") We classify sabbaticals for "restoration of health" as medical leaves. In about 10 percent of cases, one extended leave is followed immediately by one or more additional types of leave (e.g., maternity leave frequently turns into child care leave.) When this occurs, we aggregate them into a single leave of absence and use the category of the first initial leave to classify the entire sequence.

Annual files on teacher absences (as opposed to extended disruptions) provide us with the date and reason for each absence. We group absences into three categories: Self-Treated Sickness / Personal Days, Certified Sickness / Injury, and Other Absences. The “Other” category includes conferences and other school related activities, death of a family member, funerals, jury duty, military service, attendance of a child’s graduation, religious holiday, grace period, any unauthorized absence, and partial absences of at least half a day.⁷ The most common type of absence is Self-Treated Sickness, and teachers are allowed to use up to 10 of these days per year. However, if a teacher presents proof of illness from a doctor, the absence is coded as Certified Sickness and does not count against the annual 10 day cap. Thus, teachers with serious health shocks may have a strong incentive to submit certification. Teachers are allowed to use three Self-Treated Sickness days for personal business (which are coded as Personal Days), but it is highly unlikely that teachers designate all days taken for personal business. Support for this notion can be found in tabulations of absence by the day of the week. As one might expect, 40.2 percent of absences for Certified Sickness take place on a Monday or Friday. In contrast, 47.4 percent of absences for Self-treated Sickness (and 51.2 percent of Personal Days) take place on these days. We therefore group Self-Treated Sickness and Personal Days together.

We measure student achievement using data on test scores in math and English covering all students in New York City from 3rd to 8th grade in the school years 1998-1999 through 2007-2008. Because we use prior test scores as a control variable, our analysis focuses on students in grades 4 to 8 and the school years 1999-2000 to 2007-2008. In addition to test scores, our student data include information on student demographics: receipt of free and reduced price

⁷ “Grace period” applies to teachers who have exhausted their sick days and are therefore not paid for their absence. A grace period is capped at 30 days and typically applies to teachers who are absent prior to an extended leave (e.g., maternity). Absences that are consecutive to teaching disruptions are also counted as part of those disruptions. In addition to grace period absences prior to maternity leave, medical leaves are typically preceded by several days of absence where the teacher is certified to be sick.

lunch, special education, and English Language Learner services. The data also contains information on student absences and school suspensions, and we use students' prior year values for these outcomes as control variables. Because many students receiving special education services in New York are taught in classrooms (or even schools) containing solely students with disabilities, and most of these students are tested using methods that differ dramatically from the standardized statewide exams, we limit our analysis to students in regular education classrooms, excluding classrooms where the portion of students receiving special education exceeded 25 percent. Because of potential coding errors, we also exclude classrooms with less than 7 or greater than 45 students.

Importantly, the student file also provides identifiers that can be used to link students with their reading and math teachers. Students in elementary grades (4, 5, and some 6th graders), typically have the same teacher for both subjects, while older students are taught by two different teachers. One shortcoming of our data is that it is limited to regular classroom teachers, and we do not have information on substitute teachers who fill in for those who go on leave.

Summary statistics on extended disruptions and similar events that are non-disruptive, by type, are shown in Table 1. Approximately 10,693 events of either type occur in our data, and about one third of these (3,106 events) are disruptive. The variation in the fraction of events that are disruptive gives us a sense of the ability of teachers and administrators to control the timing of different event types. For example, maternity and medical leaves—where we do not expect much control over timing—are disruptive 89 and 92 percent of the time, respectively. In contrast, resignation and retirement, sabbatical, and involuntary termination are disruptive roughly 10 percent of the time. Note that about three quarters of extended disruptions are in the categories of maternity leave, medical leave, and resignation/retirement.

For each type of extended disruption, we calculate the fraction that begin and/or end in the middle of the same school year and the number of instructional days the teacher misses due to the event. Overall, 48 percent of disruptive events begin and end in the same school year, while 41 percent begin in the middle of the year but end after the school year, and the remaining 11 percent are events that began before the school year but end sometime during the year. The average number of instructional days missed is 58, and the averages of individual types of events are quite similar, varying between roughly 40 and 70 days. The sole exception are disruptions for sabbatical (where teachers take one semester of leave), which averages 93 days but varies little—the 25th and 75th percentiles are 92 and 93 days, respectively. For all other types of events, there is considerable variation in the number of days missed within leave types, with 25th percentiles around 5 to 35 days and 75th percentiles around 55 to 100 days.

In order to examine the characteristics of teachers who experience extended disruptions, we use data from the payroll records which cover every full-time teacher on the DOE payroll at the end of September, November, and May of each school year 1999-2000 through 2007-2008. The payroll file includes teachers' gender, ethnicity, age, and type of certification, and information on each teacher's salary schedule and position, which are functions of teaching experience and graduate education.

These summary statistics are presented in Table 2.⁸ For purposes of comparison, we also display the average characteristics of teachers who do not experience an extended work disruption. There are several interesting differences in the characteristics of teachers across types of events, some of which are not surprising. For example, 99 percent of teachers taking

⁸ When calculating these statistics we consolidate events of the same type for teachers with multiple events in the same year. For example, a teacher who goes on medical leave and (after returning briefly) goes on personal leave in the same year will generate two observations, as would a teacher on medical leave twice, but in different years. In contrast, only one observation is generated by a teacher who goes on medical leave twice in the same school year. Thus, the numbers of observations by leave/departure type in Table 2 are similar to Table 1 but not exactly the same.

maternity leave and 97 percent taking child care leave are coded as female, relative to 78 percent for teachers with no disruptive event. Teachers taking maternity and child care leave also tend to be younger and have fewer years of teaching experience. In contrast, teachers taking sabbatical have much higher levels of experience (7 years on average, as opposed to 4.4 for teachers without any disruption event). While it is not surprising that nearly all terminations for certification occur for teachers who have been coded as lacking certification, it is interesting that about two fifths of disruptive involuntary terminations occur for these teachers.

To examine whether extended work disruptions are more common for teachers who teach particular types of students, we calculate average student characteristics by type of disruption and compare them with averages for students whose teachers did not experience any extended disruption (Table 3). Overall, extended disruptions occur with similar frequency in schools serving different populations throughout New York City. However, there is noticeable variation in the characteristics of students when we look at individual disruption types. For example, black and Hispanic students comprise a relatively greater share of students in classrooms that experience a disruption due to resignation or retirement than white or Asian students, compared to their shares in classrooms with no disruption events.

Summary statistics on annual absences by type and by whether or not the teacher experienced an extended disruption are shown in Table 4. Teachers who experience an extended work disruption during the year have a similar number of total absences compared to other teachers (8.8 vs. 8.8) but more absences for poor health that are certified by a medical professional (3.3 vs. 2.0). Prior literature on teacher absences (e.g., Clotfelter et al. (2009)) finds that rates of absence are higher in schools that serve relatively disadvantaged populations of students. We do not find a similar pattern among schools within New York City. In our

matched student teacher data, students who receive free lunch have teachers who are absent 8.9 times per year on average, while the average for other students is 9.0 teacher absences per year.

3. Empirical Methods and Results

In our analysis, we examine the impacts of extended work disruptions separately from daily teacher absences. A typical disruptive event in our dataset covers roughly 60 instructional days, and should therefore have a much larger effect on student achievement than a day of absence. Whether these longer disruptions are more or less damaging to student achievement than isolated days of absence *on a per day basis* is an empirical question. For example, schools may find it difficult to find a single substitute to work full time for several weeks or months on short notice, and thus may cause additional disruption by filling this role with multiple individuals. On the other hand, the pool of substitutes available for long term assignments may be of higher quality, and schools may be willing to pay search costs needed to identify better candidates.

Our basic estimation specification takes the following form:

$$(1) Y_{it} = \beta_g X_{it} + \theta L_{it} + \delta A_{it} + \lambda W_{it} + \rho S_{it} + \pi_{gt} + \varepsilon_{it}$$

where Y_{it} is the math or reading score of student i in year t , L_{it} and A_{it} are, respectively, an indicator for an extended disruption and the number of daily absences taken by the teacher to whom the student is matched, X_{it} is a vector of student characteristics, W_{it} is a vector of teacher characteristics, S_{it} is a vector of school characteristics, and π_{gt} is a grade-year fixed effect.⁹

⁹ Student characteristics include a cubic polynomial in prior year math score, a cubic polynomial in prior year reading score, gender, race and ethnicity, free/reduced price lunch, special education, English Language Learner, and the number of absences and suspensions for the student in the previous year. We also interact all of these variables with the student's grade level. Teacher characteristics include teaching experience, type of certification, and possession of a graduate degree. School characteristics include school level averages of student characteristics.

Estimates of Equation 1 for math and English exams are shown in Columns 1 and 4 of Table 5. Students whose teacher has a disruptive event have 0.062 and 0.271 standard deviations lower achievement in math and English, respectively, on average. Additionally, student achievement is estimated to fall by 0.0025 and 0.0015 standard deviations in math and English, respectively, for each additional day of absence taken by their teacher. Of course, one potential explanation for these results is that disruptions and absences are more likely to occur with teachers that provide lower quality instruction, and that the disruptions themselves are not significantly detrimental. If this selection is related to time-invariant dimensions of teacher quality, then the addition of teacher-school-grade fixed effects will remove this potential source of bias. Estimates with teacher-school-grade fixed effects are smaller but still statistically significant. Students' math and English achievement are 0.055 and 0.0235 standard deviations lower, respectively, in years when an extended work disruption occurs, relative to other years for the same teacher in the same school and grade. When teachers take more (than their average) absences, math and English achievement are estimated to be 0.0017 and 0.0008 standard deviations lower, respectively, for each additional absence taken.¹⁰

The effect of a work disruption is likely to vary depending on its timing. As stated above, disruptions that occur after exams have been taken cannot have a direct causal impact on student test scores. In Columns 3 and 6, we present specifications that allow the coefficients on extended work disruptions to differ depending on whether the disruption began: before the school year started, during the school year but before the students' exam, and during the school

¹⁰ It is useful to note that the coefficient estimates on extended disruptions—which, on average, last 60 days—are less than 60 times the estimated coefficients on daily absences. This is in line with the view that unexpected absences are more detrimental than those known in advance. If a worker calls in sick one hour before the start of the work day, there may be little time for employers to find a highly qualified substitute. In contrast, if an employer knows that a worker will be going on leave for two months, more attention can be placed on the selection and training of a replacement.

year but after the students' exam. We also add an indicator for extended leaves and departures from teaching that began in the summer following the school year and are therefore non-disruptive. In both subjects, the coefficients on extended disruptions after the exams and after the school year are both small and statistically insignificant, while the estimated effects of disruptions starting before the year and starting within the year and before the exam are negative and statistically significant. Again, we find larger effects in math (-0.06) than in English (-0.04).

The coefficients on daily absences occurring before the exam (-0.0022 and -0.0008 in math and English) are also both statistically significant (Columns 3 and 6 of Table 5). However, while the estimates for absences occurring after the exam are considerably smaller (-0.0010 and -0.0002 for math and English) the coefficient for math scores is statistically significant. This suggests either a role for teacher health shocks or another form of selection bias that is not avoided by our use of teacher-school-grade fixed effects.

As stated above, one reason why an extended work disruption or absence occurring after an exam may be negatively associated with student test scores is that they are caused by worker health shocks that lowered productivity prior to the exams. In order to examine this explanation, we allow for heterogeneity across four types of extended disruptions: (1) maternity leave, (2) medical leave, (3) retirement or resignation and (4) all other types of events. While grouping all other events together ignores some substantial differences (e.g., sick family member leave vs. sabbatical), none of the remaining categories of extended disruption occur with sufficient frequency for us to reasonably expect to estimate a precise effect if we examine them separately.

For simplicity, we present results of specifications that limit attention to disruptions that start within the school year and include teacher-school-grade fixed effects.¹¹

The results of this analysis (shown in Table 6) provide little support for the worker health explanation. For both math and English, the estimated impacts of disruptions prior to the exam are always negative, though the effects for maternity and medical leaves (relative to other types of extended disruptions) are slightly smaller in math and slightly larger in English. In contrast, the estimated effects of extended disruptions that started after exams are all small, statistically insignificant and vary in sign. Thus, variation in student performance from extended disruptions does not present evidence of worker health shocks that affect productivity above and beyond the work disruption channel. However, this may simply be due to a weak relationship between teachers' health conditions prior to student examinations and their propensity to go on medical or maternity leave after exams.¹²

We further explore the worker health explanation by allowing the impact of daily teacher absences to depend on the cause for the absence. As mentioned above, we separate absences into three types: Self-Treated Sickness / Personal Days, Certified Sickness, and Other Absences. Self-Treated Sick and Personal Days are taken at the behest of the teacher, and may be due to sickness or other unexpected events but may also reflect teachers taking additional vacation time during the school year. Absences where the teacher has been certified as sick are highly likely to reflect real health problems, and, like the disruption estimates, we might reasonably expect instructional quality prior to the exam might be lower for teachers who are absent with health problems after the exam takes place. In math and English, all three types of absences occurring

¹¹ In other specifications—available upon request—we do not find any significant impacts of extended disruptions starting before or after the school year when we allow effects to differ by type. However, in line with the results in Table 5, the point estimates are all negative for disruption events that start before the school year.

¹² To investigate this possibility further, we estimated specifications that focus on extended disruptions occurring within 30 days of students' examinations. These regressions also produced small, insignificant coefficient estimates.

before the exam have (marginally) significant negative coefficients (Table 7).¹³ However, in math, there is also a statistically significant negative effect of absences that occur after the exam for Certified Sickness.

Overall, we believe these results strongly support the idea that work disruptions have an important causal impact on productivity in teaching, and the evidence from daily absences provides some support for the link between teacher health and teacher productivity on the job. While our use of within-teacher variation and the timing of student exams should do much to avoid most likely sources of bias, we cannot completely rule out all possible alternative explanations without an exogenous source of work disruption. To provide an additional check on the robustness of our results, we first take advantage of the fact that 8th, 7th, and some 6th graders typically have different teachers for math and English. If students perform worse for other reasons when their teachers take extended disruptions, then one might expect students to do worse in math when their English teacher goes on extended leave, and vice versa. The results of these estimates (Table 8) provide no support for this alternative explanation. We also test if the effects of teaching disruptions are driven by student absences. This relationship could be causal (i.e., teacher illness leads to student illness, which lowers achievement) or spurious (i.e., student illness leads to lower achievement and higher teacher illness), but either case would be important for the interpretation of our results. However, including current student absences as a control variable in our regression has almost no impact on our coefficient estimates, although these absences are negatively related to student achievement.

Finally, it is potentially important that all of these results presented above constrain the effect of disruptions and absences to be the same across different types of students and teachers.

¹³ We use the same specification as shown in Columns 2 and 4 of Table 6, except absences are broken out by both timing and type. We do not present the disruption coefficients, but they are nearly identical to those presented in Table 6.

In order to investigate the plausibility of this restriction, we have estimated two additional specifications, the results of which are available upon request. Motivated by the idea that high achieving students may have greater resources outside the classroom that insulate them from variation in teacher productivity, we allow work disruptions to have different effects on students with prior test scores below and above the citywide median. We find similar negative impacts of both extended disruptions and daily absences on both sets of students. We also test whether work disruptions are less harmful for students taught by experienced teachers. On one hand, work disruptions may be more harmful to students taught by less experienced teachers since, new teachers may have few lessons planned in advance and may provide little guidance to substitute teachers. On the other hand, the difference in teaching quality may be larger between experienced teachers and substitutes than between new teachers and substitutes. We find evidence of the latter; an extended disruption prior to the exam is expected to cause a greater reduction in student test scores when the teacher has with 3 or more years of experience than when the teacher has two or fewer years of experience (Table 9). The coefficient on the interaction of experience with a teaching disruption is statistically significant in math and marginally significant (p-value 0.11) in English. Absences prior to the exam also have a worse effect on the students of more experienced teachers in math.

4. Conclusion

Few existing studies credibly estimate the impact of work disruption or worker health on productivity. Using extremely detailed data from New York City on the timing and cause of work disruptions in teaching, we present considerable evidence that work disruption is an economically important determinant of teacher productivity. We also find some evidence supporting the role of worker health in determining productivity on the job. We take advantage

of the panel structure and detail of our data, as well as the timing of student exams, to support the notion that the effects we estimate are causal.

Our findings also have implications for the larger literature on estimating the impacts of teachers on student outcomes. First, there is some debate over whether teacher quality is an important determinant of student achievement, or whether estimates of teacher “value-added” are driven spuriously by non-random sorting of students to teachers (see Rothstein (forthcoming), Kane and Staiger (2008)). Our results provide additional support to the idea that variance in student learning across teachers is in large part due to variation in instruction. Second, researchers have documented considerable variance in student learning within teachers over time in addition to stable variance in quality across teachers. While part of this within-teacher variance is likely a result of test measurement error and other idiosyncratic “noise” (e.g., a dog barking outside on the day of the test), there may be a role for teacher health in explaining variation in teacher performance over time. Also, our results suggest that studies of the impact of teacher absences on student achievement that use teacher fixed effects as an identification strategy may produce biased results when researchers only observe teachers’ total number of absences for the year. Future work on this issue should focus on strategies to isolate exogenous variation in absences.

Extended work disruptions in teaching are rare—occurring in roughly 2.5 percent of classrooms per year in New York City—but our results suggest that they have appreciable negative effects on student achievement. Thus, there may be an important role for policies that help stabilize instructional quality when these disruptions arise. For example, individual schools may spend few resources in preparation for extended disruptions due to the rarity of these events. However, at more aggregate levels of administration, the frequency of extended disruptions is

more steady and predictable, perhaps making the maintenance of a staff of professional “replacement teachers” more cost effective. Another policy that may mitigate the effects of extended disruptions is the institution of a common curriculum across schools, making it easier for well trained substitutes to teach effectively.

Economists have done a considerable amount of research on the impact of financial incentives on worker absence, including several studies that focus on teachers.¹⁴ Our estimates suggest that teachers provide lower quality instruction when they work during periods of illness. Thus, the impact of absence reduction policies on teacher productivity will depend crucially on how the elasticity of absences with respect to price varies across absences with different causes. While it is reasonable to think that teachers’ benefits from absence (and their reservation price) would be high when they are in poor health, this is ultimately an empirical question.

¹⁴ See, for example, Winkler (1980), Jacobson (1989), Ehrenberg et al. (1991), Barmby et al. (1991), Brown and Sessions (1996), Lindeboom and Kerkhofs (2000), Duflo and Hanna (2005), and Clotfelter et al. (2009).

References

- Aaronson, D., Barrow, L. & Sander, W. (2007) "Teachers and Student Achievement in the Chicago Public High Schools," *Journal of Labor Economics*, 25(1): 95-135.
- Aronsson, G. Gustafsson, K. and Dallner, M. (2000) "Sick but Yet at Work: An Empirical Study of Sickness Presenteeism," *Journal of Epidemiology and Community Health* 54(7): 502-509.
- Barmby, T. A., Orme, C. D., Treble, J.G. (1991) "Worker Absenteeism: An Analysis Using Microdata" *The Economic Journal*, 101(405): 214-229.
- Brown, S. and Sessions, J.G. (1996) "The Economics of Absence: Theory and Evidence," *Journal of Economic Surveys* 10(1): 23-53.
- Bureau of Labor Statistics, U.S. Department of Labor (2008a) Current Population Survey, Annual Averages, Household Data, Table 46, <http://www.bls.gov/cps/tables.htm>.
- Bureau of Labor Statistics, U.S. Department of Labor (2008b) Current Population Survey, Annual Averages, Household Data, Table 47, <http://www.bls.gov/cps/tables.htm>.
- Chaudhury, N., Hammer, J., Kremer, M., Muralidharan, K., and Rogers, F.H. (2006) "Missing in Action: Teacher and Health Worker Absence in Developing Countries," *Journal of Economic Perspectives*, 20(1) pp. 91–116.
- Clotfelter, C., Ladd, H., and Vigdor, J. (2009) "Are Teacher Absences Worth Worrying About in the U.S.?" *Education Finance and Policy*, 4(2): 115-149.
- Currie, J. and Madrian, B.C. (1999) "Health, Health Insurance and the Labor Market," in: O. Ashenfelter & D. Card (ed.), *Handbook of Labor Economics*, pp. 3309-3416.
- Currie, J. (2009) "Healthy, Wealthy, and Wise: Socioeconomic Status, Poor Health in Childhood and Human Capital Development," *Journal of Economic Literature* 47(1): 87-122.
- Duflo, E. and Hanna, R. (2005) "Monitoring Works: Getting Teachers to Come to School," NBER Working Paper 11880.
- Ehrenberg, R., Ehrenberg, R., Rees, D., and Ehrenberg, E. (1991) "School District Leave Policies, Teacher Absenteeism, and Student Achievement," *The Journal of Human Resources*, 26(1) pp. 72–105.
- Goetzel, R.Z., Long, S.R., Ozminkowski, R.J., Hawkins, K., Wang, S., Lynch, W. (2004) "Health, Absence, Disability, and Presenteeism Cost Estimates of Certain Physical and Mental Health Conditions Affecting U.S. Employers," *Journal of Occupational and Environmental Medicine* 46(4): 398–412.
- Jacobson, S. (1989) "The Effects of Pay Incentives on Teacher Absenteeism," *The Journal of Human Resources*, 24 (2) pp. 280-286.

- Kane, T. J., Rockoff, J. and Staiger, D. O. (2008) "What Does Certification Tell Us About Teacher Effectiveness? Evidence from New York City" *Economics of Education Review* 27(6) 615-631.
- Kane, T. J., and Staiger, D. O. (2008) "Estimating Teacher Impacts on Student Achievement: An Experimental Evaluation," NBER Working Paper 14607.
- Kleiner, M.M., Leonard, J.S., and Pilarski, A.M. (2002) "How Industrial Relations Affects Plant Performance: The Case of Commercial Aircraft Manufacturing," *Industrial and Labor Relations Review*, 55(2): 195-218.
- Krueger A. and Mas, A. (2004) "Strikes, Scabs and Tread Separations: Labor Strife and the Production of Defective Bridgestone/Firestone Tires" *Journal of Political Economy*
- Lindeboom, M and Kerkhofs, M. (2000) "Multistate Models for Clustered Duration Data-An Application to Workplace Effects on Individual Sickness Absenteeism," *The Review of Economics and Statistics*, 82 (4) pp. 668-684.
- Mas, A. (2008) "Labor Unrest and the Quality of Production: Evidence from the Construction Equipment Resale Market," *Review of Economic Studies*, 75(1): 229-258.
- Miller, R., Murnane R., and Willett J. (2008) "Do Worker Absences Affect Productivity? The Case of Teachers," *International Labour Review* 147(1) pp. 71-89.
- Pauly, M.V., Nicholson, S., Polsky, D., Berger, M.L., Sharda, C. (2008) "Valuing Reductions in On-The-Job Illness: 'Presenteeism' from Managerial and Economic Perspectives," *Health Economics* 17(4): 469-485.
- Rivkin, S., Hanushek, E. A. & Kain, J. (2005) "Teachers, Schools, and Academic Achievement," *Econometrica*, 73(2): 417-458.
- Rockoff, J. E. (2004) "The Impact of Individual Teachers on Student Achievement: Evidence from Panel Data," *American Economic Review*, 94(2): 247-252.
- Rothstein J., (forthcoming) "Teacher Quality in Educational Production: Tracking, Decay, and Student Achievement," *Quarterly Journal of Economics* 125(1).
- Smith, J.P. (1999) "Healthy Bodies and Thick Wallets: The Dual Relation between Health and Economic Status," *Journal of Economic Perspectives* 13(2): 145-166.
- Strauss, J. and Thomas, D. (1998) "Health, Nutrition, and Economic Development" *Journal of Economic Literature*, Vol. 36(2): 766-817.
- Winkler, D. (1980) "The Effects of Sick-Leave Policy on Teacher Absenteeism," *Industrial and Labor Relations Review*, 33(2) pp. 232-240.