Local Versus Central School Control and the Distribution of Principal Effectiveness

BY

Jeffrey Charles Schiman
BA, Calvin College, 2009
MA, University of Illinois at Chicago, 2012

THESIS

Submitted as partial fulfillment of the requirements
for the degree of Doctor of Philosophy in Economics
in the Graduate College of
the University of Illinois at Chicago, 2016

Chicago, Illinois

Dissertation Committee:

Steven Rivkin, Chair and Advisor
Ben Feigenberg
Robert Kaestner
Ben Ost
Steve Tozer, Educational Policy Studies
I dedicate this thesis to my wife and to my family.
ACKNOWLEDGMENTS

I am grateful to my dissertation committee – Steven Rivkin, Ben Feigenberg, Robert Kaestner, Ben Ost, and Steve Tozer – for their many valuable comments and suggestions. I especially thank Steven Rivkin for his guidance during my study. I thank the Department of Economics and the Smith Richardson Foundation for their support.
CONTRIBUTION OF AUTHORS

An earlier version of Chapter 1 is also available as a National Bureau of Economic Research working paper (Laing, Derek, Steven G. Rivkin, Jeffrey C. Schiman, and Jason Ward. 2016. “Decentralized Governance and the Quality of School Leadership.” NBER Working Paper 22061. National Bureau of Economic Research, Inc.) and was also presented at a CES-ifo conference. Authors retain the copyright of NBER working papers -- see Appendix section 5.2 for copyright details. I developed the idea, wrote large sections of the text, developed the datasets, and ran all the empirical analyses. Dr. Steven Rivkin assisted in developing the idea and writing the manuscript. Dr. Derek Laing assisted in developing and writing the theory. Jason Ward collected the data on principal contracts. Chapter 2 is my own sole authored work.
# TABLE OF CONTENTS

1. Decentralized Governance and the Quality of School Leadership ................................................. 1
   1.1. Introduction .......................................................................................................................... 1
   1.2. Data ..................................................................................................................................... 4
1.3. The Variance in Principal Quality .......................................................................................... 6
   1.3.1. Fixed Effect Estimates of Principal Effectiveness ......................................................... 8
   1.3.2. Principal Turnover-Based Estimates ........................................................................... 10
1.4. School Reforms and the Structure of School Governance in Chicago .................................. 15
1.5. Model of LSC Behavior ........................................................................................................ 16
   1.5.1. The Environment ........................................................................................................ 17
   1.5.2. Technology ................................................................................................................ 17
   1.5.3. Preferences ................................................................................................................ 20
   1.5.4. Monitoring and the Information Structure ................................................................ 21
   1.5.5. The Behavior of the LSC .......................................................................................... 23
1.6. Empirical Analysis of LSC Heterogeneity .......................................................................... 27
1.7. Conclusions .......................................................................................................................... 32
2. The Signaling Value of Merit Awards ....................................................................................... 33
   2.1 Introduction ........................................................................................................................ 33
   2.2. The Achievement Award Program ..................................................................................... 35
   2.3. How the Award Program Affects Transitions and Mobility ............................................. 37
   2.4. Measuring the Effects of Achievement Awards ............................................................... 40
   2.5. Regression Discontinuity Analysis .................................................................................... 42
   2.6. Results ............................................................................................................................... 43
      2.6.1. OLS Estimates ............................................................................................................ 44
      2.6.2. Regression Discontinuity Estimates ....................................................................... 45
      2.6.3. Movement ................................................................................................................ 47
      2.6.4. Merit Awards and Schools ....................................................................................... 48
   2.7. Discussion ........................................................................................................................ 50
3. Conclusions .............................................................................................................................. 50
4. Bibliography ............................................................................................................................. 52
5. Appendix ................................................................................................................................... 54
   5.1 Evidence of IRB Approval ................................................................................................. 54
   5.2 Evidence of Copyright ........................................................................................................ 66
6. Vita............................................................................................................................................. 88
LIST OF TABLES AND FIGURES

Table 1. Average Voting Intensity in LSC Election Cycles by Quartile of Socioeconomic Status........67
Table 2. The Distribution of Principal Transitions Experienced by Schools........................................68
Table 3. Estimates of the Standard Deviation of Principal Effectiveness Based on Regressions with Principal by Year Fixed Effects .................................................................69
Table 4. Average Principal Value Added, by Teacher Responses to Survey Questions on the Principal ..70
Table 5. Relationship between Teacher Ratings and Estimated Principal Value Added.....................71
Table 6. School Fixed Effect Estimates of the Variance in Principal Effectiveness ..........................72
Table 7. Distribution of Voting Intensity Conditional on Parental SES ...........................................73
Table 8. The Effects of Parental SES and Voting Intensity in LSC Elections on the Change in Principal Effectiveness ..................................................................................................................74
Table 9. Multinomial Logit Estimates of Principal Transitions.............................................................75
Table 10. Criteria For Principal Evaluation .......................................................................................76
Table 11. OLS Estimates of the Relationship Between Award Receipt and Turnover .......................77
Table 12. Reduced Form Regression Discontinuity Estimates of the Relationship between Turnover (t to t + 1) and Predicted Award Receipt .................................................................78
Table 13. 2SLS Estimates Weighted by the Probability of Treatment ..............................................79
Table 14. Change in Job Characteristics for Principals between t and t+1 for those Remaining in CPS ...80
Table 15. Distribution of performance ..............................................................................................81
Table 16. How the Turnover Generated by the Award Affects Schools .............................................82

Figure 1. Changes in the Probability of Award Receipt by Predicted Award Receipt .......................83
Figure 2. Probability of Overall Principal Transitions by Predicted Award Receipt .........................84
Figure 3. Pattern of Turnover Prior to the Award Program ..............................................................85

Appendix Figure 1. Density of the RD Running Variable .................................................................86
Appendix Table A1. Discontinuities in Observable Covariates ...........................................................87
SUMMARY

Elevating the quality of schooling in large urban districts is a challenge, and districts have tried many policies. In Chicago, after years of widespread dissatisfaction with the school system, legislators decentralized school governance in 1988 by forming elected local school councils (LSCs) responsible for principal hiring, evaluation, and contract renewal as well as other management functions. Although decentralization has the potential to strengthen local knowledge and incentives, there may exist a divergence between personal and school-quality maximizing incentives or differences throughout the city in the capacity to make such managerial decisions. Consequently, subsequent legislation in 1995 outlined circumstances in which the district could reclaim authority from the LSC, thereby limiting local control. More recently, the mayor introduced a district-level incentive policy aimed at improving schooling by recognizing and rewarding principals who were successful at raising test scores.

In my dissertation I study the effects of decentralization and merit incentives on the distribution of principal quality. To separate the contribution of the principal from other changes in the school, I use both a variety of approaches and supporting evidence. After having demonstrated that principals are important determinants of student success, I next explore differences in the quality of LSC contract decisions. I conclude by exploring the effect of winning a merit award on principal job mobility.

In Chapter 1 of my dissertation, we first establish the presence of significant variation in principal effectiveness based on both an analysis of variance approach and the estimation of principal fixed effects. Teacher survey responses support the findings based on the principal fixed effects, though the much smaller magnitude of the analysis of variance estimates suggest that unobserved shocks inflate many existing estimates of the variance in principal effectiveness.
We next consider potential differences in LSC behavior that contribute to the variation. Following Aghion and Tirole (1997) we develop a model that highlights the tensions between formal and real authority and incorporates potential differences in LSC capacity and incentives to maximize school quality. Using proxies for managerial capacity and incentives we find evidence largely consistent with the theory, showing that LSCs with higher management capacity and stronger incentives to raise school quality experience larger gains in principal effectiveness following the end of contracts.

In Chapter 2 of my dissertation, I study the recent principal merit award program in the CPS. The award program intended to recognize and reward school leaders who were successful at raising test scores. Although visible awards may increase worker effort and productivity, they may also produce unintended consequences. By acting as a signal of quality, awards may increase outside options and increase worker turnover. In this chapter, I use a regression discontinuity design exploiting the fact that Chicago provides principals with merit awards based solely on achieving a particular threshold of performance. Given the similarity between awardees and non-awardees at the threshold, any differences in productivity, motivation, or ability are unable to explain subsequent differences in job mobility. I find that principals who just exceeded the threshold for a merit award are more likely to exit their school the year after winning compared to principals who fell slightly short of the award threshold. Those who transition within the district move toward higher paying and higher test performance settings than principals who just lose. The findings are consistent with the predictions of a search model where awards that signal high-quality performance shift right the distribution of outside job offers, leading to turnover for award winners.
1. Decentralized Governance and the Quality of School Leadership

1.1. Introduction

In response to a persistently high dropout rate and widespread dissatisfaction with the schools, the 1988 Chicago School Reform Act decentralized school governance by forming Local School Councils (LSCs). LSCs are elected groups of parents, community members, and school personnel who are responsible for the hiring, evaluation, and contract renewal decisions of principals, the school budget, and other management functions. Supporters of the reform expected the greater knowledge and interest of local decision-makers to elevate the quality of school leadership and management. In addition, the reform had the potential to amplify the benefits of competition among attendance zones, although the low rate of homeownership for many poor families with children in the schools would likely dampen any such effect.

Yet there are reasons to question the benefits of decentralization. First, the interests of LSC members may not align with those of school children and community members; such misalignment may be more likely in neighborhoods with less participation in the political process. Second, development of budgets, supervision of the principal (herself a member of the council), and the completion of other tasks in a manner that improves principal and school quality likely require a range of skills which many LSC members, particularly those without a post-secondary degree or experience in a supervisory or management role, may lack.

Consequently, it is likely that decentralization will produce heterogeneous effects that benefit some schools and not others and that the reform may be less beneficial in areas with a lower capacity to manage and weaker incentives to improve the quality of schooling.
In an influential study, the Consortium on Chicago School Research surveyed LSC members about a number of issues including capacity to govern and the principal evaluation process (Ryan et al. 1997). The responses suggested that although highly productive LSCs did exist, inequality in governance practices led the decentralization reform to “produce highly varied outcomes across Chicago’s school communities.” Differences in reported principal evaluation practices were particularly striking. Responses suggested that just over half of LSCs had a formal evaluation process with explicit criteria the principal had to meet during her contract, while many of the remaining councils had minimal or even no evaluation procedures in place (Ryan et al. 1997).

Nonetheless, the survey responses do not provide evidence on the extent to which LSCs improve the quality of schools and school leaders. In fact, ongoing concerns that LSCs have not substantially raised the quality of schools and school leadership, especially in high poverty areas, likely contributed to subsequent legislation that restored some authority to the district. Over time, the district has assumed control over principal evaluation, hiring, and renewal decisions in a growing number of schools, although the majority of principals remain under the supervision of a local school council.

In this paper, we use Chicago Public School administrative data and information on principal contracts to investigate the distribution of principal effectiveness in a system of decentralized governance with the potential for central authority intervention. We pay particular attention to heterogeneity in managerial capacity and the incentives to maximize school quality in order to gain a better understanding of the distributional consequences of decentralization.

In the absence of differences in principal value added to achievement, neither LSC nor district actions will influence the quality of school leadership in that dimension. Therefore, we
first estimate the variance of principal value added using two different approaches and use teacher survey responses to provide supporting evidence. The analysis of variance estimates account most extensively for potential confounding factors and indicate educationally and statistically meaningful differences that are roughly one third the size of the estimated variance in teacher effectiveness. Importantly, the pattern of estimates suggests that time-varying factors will overstate the variance when using principal fixed effects.

We turn next to LSC decision-making and its relation to the distribution of principal quality. Following Aghion and Tirole (1997) we develop a model of LSC behavior that highlights the tension between formal and real authority in the district, incorporating potential variation in LSC managerial capacity and incentives and the possibility the District CEO intervenes. We estimate the extent to which LSC capacity and incentives relate to the change in principal effectiveness from one contract to the next as well as the determinants of principal transitions in LSC managed schools including those where the CEO intervenes. Our analysis reveals that LSCs with higher management capacity and stronger incentives appear to experience larger average gains in principal effectiveness following the completion of contracts.

The absence of information on contract offers and firings inhibits the identification of differences in LSC personnel practices with respect to principal effectiveness. Although such information would be valuable, the ultimate effect of the LSC on principal quality involves both the renewal decision and the desirability of working in a given school for current and prospective principals of varying levels of effectiveness. In contrast, CEO intervention represents a clear case of removal in response to poor performance or some other transgression, though other steps, including school restructuring, often accompany CEO interventions. Therefore, analyses of both voluntary and involuntary transitions must be interpreted with care.
Because our sample begins after the 1988 reform, we cannot identify the effect of decentralization on the distribution of principal effectiveness. Persistently poor performance in many schools precipitated the reform, so the continued presence of low-performing schools and questionable personnel decisions does not constitute evidence the reform harmed the district. Rather, our aim is to provide a better understanding of the nature of principal transitions under decentralization and the efficacy of CEO interventions, enhancing the capacity to improve school governance and potentially illuminating areas of weakness.

1.2. Data

To describe the distribution of principal effectiveness and transition patterns we combine CPS administrative data for the period 1993-4 to 2013-4 with US Census and American Community Survey data, teacher survey responses, LSC election data, and information we collected on principal transitions and contracts using public online documents from the proceedings of CPS Board of Education (BOE) and LSC meetings. The administrative data contain extensive information on students including test scores, attendance, demographic characteristics, special education status, eligibility for a subsidized or free lunch, school attended, grade, and school characteristics. School switchers can be followed if they move within the district. We measure the socioeconomic status of the block group in which each student lives using Census information on education levels and share of employed adults. Prior to 2010, the measure is based on the 2000 Census while from 2010 forward it is based on the 2010 ACS.

We also use the administrative data to construct a panel data set of principals that we merge by school and year to the student data. Principals must be linked by name, as there is no unique ID number that spans the period. Therefore, we take great care to account for name
changes following marriage or divorce, other name changes, and spelling or punctuation inconsistencies. The principal panel contains almost 1,500 principals in over 700 schools.

Principal contract approval records show BOE approval of principal selections by LSCs and BOE ratification of principal employment contracts. The records include contract start and end dates and school served. We also record any disciplinary action against a principal including remove and replace resolutions initiated by the CEO as well as school designations including under transformation or turnaround. In the case of remove and replace resolutions the LSC selected principal is removed and replaced with a principal appointed by the CEO, and until stated otherwise the CEO assumes authority over the hiring, evaluation, and contract renewal decisions of principals. Note that the data do not distinguish between a decision not to offer another contract and a decision not to accept a new contract.

The LSC election data contain the number of parent and community candidates and vote totals for each biennial election from 2002 to 2012. We focus on parental candidates and develop a measure of election voting intensity equal to the sum of votes cast for parent representatives divided by school enrollment in the election year. Table 1 reports mean values of our intensity measure by quartiles of parental socioeconomic status, revealing a strong, positive association between school average parental SES and election voting intensity.

Teacher responses to questions about their principals form the final component of the data. These are available for a number of years, though the content of the survey changes over time. Teachers are asked whether they agree or disagree to a series of statements, and we focus on the three statements that seem particularly relevant to effects on value-added and appear in virtually all years of the data. The teacher is asked if the principal (1) “makes clear to the staff his or her expectations for meeting instructional goals”; (2) “communicates a clear vision for our
school”; and (3) “carefully tracks student academic progress.”1 We use responses to these questions to create a scale of principal effectiveness from the teacher’s perspective that we relate back to our value-added estimates of principal effectiveness.

The analysis focuses on principal transitions, and Table 2 reports the distribution of schools by the number of principals during the sample period. Approximately 85 percent of schools experience at least one transition in our sample while 41 percent experience three or more. Because some schools enter and exit the sample either through new construction or through closure, we also present the distribution limited to schools observed in all twenty years. In the sample of schools observed in all years, over 99 percent of schools experience at least one transition and 46 percent experience three or more.

1.3. The Variance in Principal Quality

The measurement of principal value added and the variance of principal effectiveness share many similarities but also some important differences with the estimation of teacher value added. On the one hand, the residential location and school-choice decisions of families in combination with school assignment policies and practices introduce substantial variation in student composition among schools and classrooms that must be addressed in studies of both principals and teachers. On the other hand, the widely discussed problems for the estimation of teacher value added associated with purposeful allocation of students to classrooms and test measurement error are far less important in the case of principals given the focus on school-wide performance and much larger number of test-takers in schools than classrooms.2 However, the persistence of principal influences on the quality of instruction in the years after their departure

---

2 Kane et al. (2013), Chetty, Friedman, and Rockoff (2014), and Rothstein (2010) investigate the presence and magnitude of biases introduced by nonrandom assignment to classrooms.
and the absence of comparisons within a school at a single point in time present serious hurdles to the identification of principal effectiveness.

A comparison with the dynamics of teacher effects illuminates the problems introduced by the fact that many actions including teacher hiring, contract renewal decisions, mentoring and the establishment of a school climate will persist beyond a principal’s tenure. In the case of teachers, many of the longer-term effects are captured by lagged achievement measures for observations in later years, and the teacher in the previous grade generally has little or no involvement with instruction in the current year. Even if lagged test scores do not fully account for prior teacher effects due to the dynamics of learning, it is possible to account directly for prior teacher effects in the model.\(^3\) In the case of principals, however, it is clear that prior achievement does not account for effects of decisions such as teacher hiring that directly affect learning in periods.

The presence of a single principal in each school at any point in time rules out within school and year comparisons. In contrast, teacher effects estimated on the basis of within school-year variation account fully for any school-wide shocks regardless of whether they persist over time or affect learning in a single year. Because the estimation of principal effects essentially involves the estimation of school value added during a principal’s tenure, with perhaps controls for school resources and other variables outside of the principal’s control, care must be taken in the interpretation of such estimates.

Similar to Branch, Hanushek, and Rivkin (2012), we estimate the variance in principal effectiveness using two distinct approaches. In our first approach, we estimate a value-added specification that includes principal-by-year fixed effects, which we use to calculate the variance

---

\(^3\)Rothstein (2010) shows a relationship between previous teacher quality and achievement even in a value-added specification.
in principal effectiveness both overall and within schools. We then use teacher survey responses to provide evidence on the validity of the fixed-effect estimates, given the ambiguity introduced by the aforementioned equivalence of principal-by-year and school-by-year effects.

In our second approach, we take additional steps to mitigate bias introduced by unobserved time-varying factors and the persistence of principal influences. This method identifies the variance directly on the basis of year-to-year fluctuations in achievement growth around principal transitions. Essentially the year-to-year fluctuations within principal spells capture shocks that affect achievement, and larger fluctuations in value added around transitions would identify the effects of differences in principal effectiveness.

Importantly, recent evidence in Miller (2013) reveals a systematic decrease in value added in the year prior to the arrival of a new principal. This may reflect a reduction in principal health, effort, or authority over the school or the impacts of other factors associated with the decision to leave a principal position. Therefore, we estimate specifications that exclude the years immediately surrounding transitions in both our approaches. Because of the possibility that value added in the first year might be inflated by a recovery from the achievement dip in the final year, we must exclude both the last and first years of spells. Note that by focusing on years away from transitions we can account better for persistence, unobserved school trends, and the confounding effects of potentially large shocks that coincide with principal changes.

1.3.1. Fixed Effect Estimates of Principal Effectiveness

In our first approach we estimate the following specification,

\[ A_{igst} = f(A_{igst,t-1}) + X_{igst} + S_{sgt} + \delta_{gt} + \theta_{pt} + \varepsilon_{igst} \]

\(^{4}\)The fixed-effect approach follows Bertrand and Schoar (2003), Grissom et al. (2012), Cannon, Figlio, and Sass (2013), and Branch, Hanushek, and Rivkin (2012).
where current achievement for student $i$ in grade $g$ in school $s$ in year $t$ ($A_{igst}$) equals a cubic function of prior-year achievement $f(A_{igst-1})$, a vector of student controls ($X_{igst}$) including race, sex, special education status, and whether or not the student is new to a school, a vector of school-level controls ($S_{sgt}$) including grade averages of the student controls as well as enrollment and parental SES, a principal-by-year fixed effect ($\theta_{pt}$), and a random error term ($e_{igst}$). Regressions also include year-by-grade effects ($\delta_{gt}$) to account for changes over time in test structure or other policies.

Year-to-year changes in value-added that occur during a principal’s tenure are captured by the principal-by-year fixed effect ($\theta_{pt}$). Of course, estimates of the principal-by-year fixed effects combine the true principal effect with any other fixed or time-varying influence not accounted for in the regression. Because of the likely presence of unobserved school influences not captured by prior achievement, we also compute variance estimates based on deviations from the school average of the principal-by-year fixed effects. This eliminates all variation in principal effectiveness between schools.

The top row of Table 3 reports estimates of the overall (left column) and within school (right column) standard deviation of principal quality produced by averaging the principal-year effects over a spell at a school. The more compelling within-school results suggest that a one standard deviation increase in principal effectiveness increases average test scores in a school year by 0.078 $\sigma$. To address the possibility of an Ashenfelter’s dip in performance, we drop the last and first year of all spells and recalculate the standard deviation. If a performance dip is present, we would expect the standard deviation to decrease after removing the last and first years because we are removing extra, within-principal variability around transitions. After
dropping these two years, we find that the standard deviation does in fact decline from 0.078 to 0.065.

The estimates reveal substantial variation in principal effectiveness, as a one standard deviation improvement in principal quality increases achievement by $0.065\sigma$ on average for all children in the school. Even if only half of the improvement persists in the long run, after 9 years (i.e. Kindergarten through 8th grade) in an elementary school such an improvement would increase average achievement by roughly $0.3\sigma$.

We turn now to the teacher survey responses and examine whether teachers rate higher-value added principals more favorably. This would provide confirmatory evidence that the fixed effects do in fact capture differences in principal contributions to achievement. Table 4 reports average value added by the response to each statement, and the estimates reveal strongly positive and monotonic relationships between estimated value added and teacher ratings for all three questions.

In order to ensure that these results are not driven by other differences among schools, we use factor analysis to compute a teacher rating index and then regress value added on the index in specifications with and without school fixed effects. The coefficients in Table 5 reveal a strong relationship between value added and teacher ratings both overall and within schools. This supports the belief that the estimates of principal effectiveness capture differences in the contributions of principals to school quality, even if the fixed effects also capture time-varying shocks that inflate estimates of variance.

1.3.2. Principal Turnover-Based Estimates

Our second approach to estimate the variation in principal effectiveness uses a method similar to Rivkin, Hanushek, and Kain (2005) and extended to principals by Branch, Hanushek,
and Rivkin (2012) and Coelli and Green (2012). Importantly, we take additional steps to account for persistent and not persistent time-varying shocks, performance dips in the transition years, and the persistence of principal effects on school quality. To illustrate the approach we draw heavily on Branch, Hanushek, and Rivkin (2012):

Equation (2) relates the average gain in achievement (current score minus prior year score) in school s, in year y as an additive function of principal quality (θ), the quality of other school and community factors including student composition not under the control of the principal (γ), a school fixed effect (δ), and the school by year average error that includes unobserved student influences:

\[ \Delta A_y^s = \theta_i^y + \gamma_i^y + \delta_i^y + \nu_i^y \]

Consider the difference between successive years y and y’ in average gain in achievement. This eliminates all school effects that do not vary across the two years, leaving only year-to-year differences in principal quality, other school influences and other unobserved time-varying factors as determinants of the difference in achievement gain. Squaring this difference yields a natural characterization of the observed achievement differences between years as a series of terms that reflect the variances and covariances of the principal and school effects plus a catchall component e that includes all random error and cross product terms between principal and other year specific effects.

\[ \left( \Delta A_y^s - \Delta A_y'^s \right)^2 = \theta_i^y + \gamma_i^y + \delta_i^y + \nu_i^y \]

Taking the expectation of Equation 3 and assuming principals are drawn from common distributions at each school over the restricted time period of the observations yields:

\[ E \left( \Delta A_y^s - \Delta A_y'^s \right)^2 = 2 \sigma_\theta^2 + 2 \sigma_\gamma^2 + 2 \gamma_i^y - 2 \theta_i^y \gamma_i^y + 2 \gamma_i^y \gamma_i'^y + E(e_{ij}) \]

where \( \sigma_\theta^2 \) (\( \sigma_\gamma^2 \)) is the variance of principal quality (other school influences) in school s and \( \sigma_{\theta_i,\gamma_i}^2 \) (\( \sigma_{\gamma_i,\gamma_i}^2 \)) is the covariance in principal quality (other school influences) across years.(Branch, Hanushek, and Rivkin 2012)
As Branch, Hanushek, and Rivkin (2012) describe, the three primary assumptions for this approach are “1) the effect of a principal is fixed (no change over time); 2) principals are drawn from a common distribution during this time period; and 3) principal turnover is orthogonal to other school changes that affect achievement gain.” If satisfied, the within-school variance in principal effectiveness can be uncovered from comparisons between annual fluctuations in achievement gains around transitions and fluctuations within the tenure of a single principal. Absent a transition, the covariance in principal quality between years $y$ and $y'$ equals the variance given the assumption that principal quality remains constant. By comparison, if the principal changes between years $y$ and $y'$, the covariance equals zero within schools given the assumption that principals are drawn from a common distribution. Based on these assumptions, “the within school variance in principal quality can be identified from a regression of the squared difference in cohort average gains on a dummy variable indicating that the two cohorts had different principals.” (Branch, Hanushek, and Rivkin 2012)

As noted above, the systematic value-added decline in the final year of a spell means that variance estimates that include this year do not strictly capture fixed differences in principal skill or effectiveness. Therefore, some specifications exclude the final year of a departing principal’s tenure in a school and also the first year of the incoming principal’s tenure, since the average gain in the first year may be inflated by the dip in the prior year. Exclusion of these years also mitigates bias introduced by additional turbulence associated with a transition.

Excluding these years increases the gap around transitions from one to three years, and this has two potentially offsetting effects on the estimates. One the one hand, measurement error that increases achievement and therefore achievement gain in year $y$ will tend to decrease achievement gain in year $y + 1$, because the positive error in year $y$ will tend to decrease the
difference between achievement in years $y + 1$ and $y$. This, in turn, will tend to amplify the squared difference in gains based on adjacent-year comparisons. Therefore, in specifications that use adjacent years to compute squared differences in gains within spells but nonadjacent years to compute squared differences in gains across transitions, such measurement error will tend to attenuate the estimates.

On the other hand, underlying achievement trends would tend to amplify differences computed over longer periods. This would likely increase the squared difference calculations around transitions based on gains three years apart relative to those within spells based on adjacent years and therefore introduce upward bias into the estimate of the variance. Consequently the direction of bias cannot be signed a priori, and computations across transitions and within spells should be comparable in terms of distance between years. Note that this restriction substantially decreases the sample size and therefore reduces the precision of the estimates.

Table 6 reports coefficients on the transition indicator for a series of specifications that differ according to whether or not the first and last years of spells are included and the number of years between calculations within a principal spell. The final column reports the estimate for the preferred specification in which squared differences both across transitions and within spells come from gains that are three years apart, and the estimate is slightly smaller than that from the specification that includes all observations, though it just misses significance at the 10 percent level.

The three estimates from specifications that exclude the first and last years are consistent with both measurement error induced attenuation bias in specifications that use adjacent-year differences within spells and unobserved trends that introduce upward bias in specifications that
calculate squared differences within spells across fewer than three years. First, the insignificant estimate of 0.0015 in Column 2 based on adjacent year calculations within principal spells that likely amplify the effects of measurement error is much smaller than the estimate of 0.0037 in Column 3 based on differences two years apart. Second, the Column 4 estimate of 0.0029 based on within spell differences calculated across the same three years as the differences across transitions is smaller than 0.0037, consistent with the notion that widening the gap tends to increase the counterfactual variance. Taken as a whole the results highlight the importance of using non-adjacent years and the same size gaps both across transitions and within spells.

Not surprisingly and similar to the pattern found in Branch et al. (2012), the Column 4 estimate of the standard deviation in principal effectiveness of 0.0383 is roughly half as large as those produced by the direct estimates of principal fixed effects. It is likely that the true magnitude lies somewhere between the two estimates.

Although the estimates of the standard deviation based on the principal fixed effects may be inflated by changes over time in other factors, the turnover estimate makes the quite strong and unrealistic assumption of fixed principal effects within spells. Not only is the principal’s influence likely to grow over time as more of the teachers hired, school practices adopted, and climate reflect the work of the incoming principal, principals may improve as they gain a better understanding of how to lead a school. Regardless, even the smaller estimates are educationally significant in magnitude, as they suggest an improvement of almost $0.2 \sigma$ in average achievement from 9 years of attendance at an elementary school with a one $\sigma$ more effective principal, assuming that half of the annual effect persists.
1.4. School Reforms and the Structure of School Governance in Chicago

Two institutions are charged with the task of overseeing Chicago's public schools: the Board of Education (BOE), which is headed by a chief-education officer (CEO), and Local School Councils (LSCs). These LSCs—one for each public school in the city—were introduced as part of the landmark 1988 Chicago School Reform Act. This Act shifted much authority including the hiring and evaluation of the principal away from the BOE to the LSC.

Each LSC includes the principal of the school together with a periodically elected membership that includes teachers, parents, and other community members with each constituency holding a fixed number of seats on the council. Importantly, the self-interests of LSC members may diverge from the welfare of the students. For example, a parent member could push for reallocation toward her child; a member of the Council could attempt to influence the hiring process to favor a friend or relative; and, of course, the school principal—a key player in each LSC—may use her position on the council to further her career.

With the potential hazards of decentralization in mind, the 1995 Chicago School Reform Amendatory Act gave the mayor the right to appoint a Public Schools CEO and restored elements of the decision-making authority to the BOE. In particular, the 1995 amendment gave the CEO the authority to remove and replace a principal under certain circumstances including persistent low achievement.

Using the terminology developed in Aghion and Tirole (1997), the 1995 Act endowed the BOE with all of the formal authority concerning the administration of each the city's public schools. It gave the CEO the right to place a public school on probation. Once on probation, the

---

5 In fact, there are three institutions, if one includes the Office of the Mayor. The BOE is under the direct formal control of the Mayor. Nevertheless, in practice, the role of the Office of the Mayor is limited to setting the annual school budget, appointing the CEO, and appointing members to the BOE. In particular, it has delegated all of its real authority in the practical business of running the CPS to the BOE and by extension the CEO.

principal has one year to remedy performance issues. If the principal fails to do so, the CEO has the power to assume control over decisions regarding the school principal—authority that otherwise rests with an LSC. Furthermore, the CEO can, in effect, dissolve the current LSC by mandating that new elections be held. In addition, she has the power to close educational programs, and she even has the power to shut down a school and then reopen it by hiring new staff altogether or by selectively rehiring some of those who were dismissed from the school in question. In each of these cases the BOE assumes real as well as formal authority.

Despite the considerable formal power possessed by the Board, the real authority concerning the practical day-to-day business of running the vast majority of the city's public schools rests with local school councils. Here, the economic substance of real authority is that the LSC is responsible for selecting the principal, renewing the principal's contract, determining the allocation of the budget, and crafting a School Improvement Plan. Although the CEO effectively rubber stamps the majority of LSC principal contract decisions, the process of monitoring and the possibility of intervention constrain LSC and principal behavior.

1.5. Model of LSC Behavior

Next, we construct a model that draws on Aghion and Tirole (1997) and is designed to capture the salient features of the Chicago public school system (CPS) as they pertain to principal effectiveness and transitions. We begin with a description of the environment and then describe the behavior of LSCs and potential variation by capacity and election pressure.

---

1.5.1. The Environment

We first describe the technology that governs school quality, the objectives of each LSC, the means whereby the CEO monitors the schools and the determinants of its decision of whether to overrule the LSC. Subsequently, we consider the relationship between the CEO and the LSCs.

1.5.2. Technology

We envisage a school as a production function that takes the following as its primary inputs: students, the labor of the principal, teachers, and staff, together with other resources, such as the quality of the library, buildings, and other educational capital. A school uses these inputs to produce educational value added, as measured by school quality, \( q \).

For the moment, put to one side issues pertaining to the quality of the administrative decisions made by an LSC or the principal. Consider a school with a resource endowment \( k \). We assume potential value added, \( \hat{Q} \), is given by

\[
\hat{Q} = F(x; k) = x \cdot k.
\]

The term \( x \in [0,1] \) is an index that captures the (endogenous) efficacy with which the resource endowment available to the school, \( k \), is allocated to the production of educational value added. The case in which \( x = 1 \) corresponds to the highest possible potential value added, which we denote \( \hat{Q}_m \equiv F(1; k) = k \).

As will be clear later, the allocation of resources, \( x \), plays an important role in our analysis. In anticipation, we assume each LSC also has an independent interest over the way

---

8 In practice, each school draws the large part of its student body from the geographic community in which it is located. This, in conjunction with the widespread socioeconomic disparities across Chicago’s neighborhoods, induces considerable variation in the composition of students who attend it. Therefore, certain neighborhoods draw students primarily from high income households and others draw them from economically challenged ones, especially those with high recent immigrant populations. To provide a veracious measure of value added by a school in general and assay the effectiveness of the principal in particular, it is necessary to control for these factors, which we do in the subsequent empirical analysis.
resources are allocated within the school and, as a result, its preferences over $x$ may be not fully aligned with the goal of maximizing value added, $Q$. The main economic substance of this structure is that it generates a potential conflict between the interests of the CEO and the LSC. In settings with less voting intensity/participation or low capacity to govern, the principal likely plays a more dominate role in resource allocation which may or may not ameliorate the potential conflict. Next, we consider the issue of authority and control.

Let $\theta \in [0, 1]$ represent the given managerial or organizational capacity/quality of the institution that exercises real authority over the running of a specific public school. If the school is administered by an LSC, $\theta$ is a reduced-form measure of the LSC’s overall managerial capacity. In practice, it depends jointly on the abilities of its members and, most important, the ability of the current school principal. This is because the organizational structure is one in which the principal is a member—often the most influential member—of the LSC. This renders challenging the task of explicitly modeling the link between abilities of the committee members and the efficacy of the committee itself: one of the primary functions of the LSC is assaying the performance of the principal, and the principal herself has considerable indirect influence over the assessment of her performance by virtue of her membership of the LSC.$^9$

We assume that actual value added by the school, denoted $Q$, depends on the managerial quality parameter, $\theta$, and potential value added, $\hat{Q}$, according to

$$Q = \theta \cdot \hat{Q}.$$ 

$^9$Indeed, these concerns are the primary motivation for our use of a stochastic representation of the educational value-added production technology, described shortly. This approach allows us to adopt an agnostic approach concerning the explicit link between the abilities of the LSC committee members and the overall managerial capacity of the LSC. More specifically, it is inconsequential (to our model) whether the LSC’s capacity, $\theta$, is governed by the strength of the weakest link (the ability of the least able member), the sum of the parts (the simple sum of ability members), or some more complicated sub- or super-modular function of these abilities.
We can think of $\theta$ as an $X$-efficiency parameter a la Leibenstein: The effectiveness with which a given set of inputs are used to produce outputs. Here, $\theta = 1$ corresponds to the frictionless ideal that represents the perfect management of resources. For intermediate cases of $\theta$, the term $(1 - \theta) \times \hat{q}$ represents the degree of $X$-inefficiency—the loss of value added resulting from poor administrative decisions.

Next, we consider the determinants of $\theta$ for each of the city’s schools. To this end, consider a public school that is administered by a LSC. We assume the managerial quality of those schools controlled by LSCs, $\theta$, is distributed on the support $[\theta_0, 1]$ according to the distribution function $G(\theta)$, with mean $\bar{\theta} \in [\theta_0, 1]$, where $\theta_0 > 0$ represents some minimal managerial capacity.\(^{10}\)

To capture the potential benefits of decentralized local control, we assume that those public schools over which the Board of Education exercises real authority each have an organizational capacity $\theta_B \in (0, M]$, where $M < 1$. Though simple, this formulation is rich enough to capture, on the one hand, the potential benefits of decentralization, $\theta_B \leq M < 1$, and, on the other, the possibility that some schools would benefit by having important decision making rights taken away from the LSC and assigned to the CEO, $\theta_0 < \theta_B$.

In what follows, it is convenient to work with the intensive form of educational value added. That is, value added per dollar of resource endowment, $k$. Therefore, let $\hat{q} \equiv \hat{Q}/k$ denote potential educational value added per dollar and let $q \equiv Q/k$ denote actual educational value added per dollar. Therefore equation (5) becomes

\[
\hat{q} = \frac{F(x; k)}{k} = x,
\]

and (6) becomes

\[\hspace{10cm}\]

\(^{10}\)We can think of $\theta$ as a stochastic representation of technology, similar to that proposed in Eaton and Kortum (2002). The lower bound $\theta_0$ ensures that for all values of $x$ and $k, Q \geq \theta_0 \cdot x \cdot k > 0$, thereby ensuring no school is so bad as to generate zero value added.
\( q = \theta \cdot \hat{q} \leq 1. \)

In the interest of brevity, in what follows we refer to the intensive measure, \( q \), as educational value added and to \( Q \) as aggregate educational value added by the school, as we do for \( \hat{q} \) and \( \hat{Q} \) – their potential value added counterparts.

The principal advantage of using the intensive forms of the value-added measures is that they allow for meaningful comparisons of relative performance, in view of the realistic and practical heterogeneity in resource endowments that are available to different schools in the city. More specifically, consider two schools, \( i \in \{0,1\} \), with endowments \( k_0 < k_1 \). It may well be the case that total value added is greater in the latter school than in the former, \( Q_0 < Q_1 \).

Nevertheless, if, for example, \( q_0 = \frac{Q_0}{k_0} = 1 > q \), the school \( i = 0 \) has attained the greatest possible output with the resources available to it, whereas school \( i = 1 \) has not.

1.5.3. Preferences

Consider first, the various LSC’s that control Chicago’s public schools. We assume the objectives of each LSC are described by the following utility function of the representative committee member:

\[
V = v(q, c),
\]

where \( q \) is the value added by the school under its control, and \( c \equiv 1 - x \) represents the consumption-equivalent value arising from the distortion in the allocation of resources, as measured by \( 1 - x \). We assume \( v_q > 0, \lim_{q \to 0} v_q = \infty, v_c \geq 0 \), and that \( q \) and \( c \) are substitutes.

With the inclusion of \( c \) in the utility function, each LSC has a direct interest over the manner in which resources are allocated. For instance, as one of its members, the principal may lobby the LSC to renew his or her contract, despite the fact that another candidate would be
better suited for the position and so generate greater value added. Alternatively, a committee member may lobby in favor of hiring a friend or relative to fill an open position.

Despite its simplicity, this framework is rich enough to capture the cases of greatest interest to us. If $v_c = 0$, then the LSC purely is interested in maximizing school quality. Alternatively, if $v_c > 0$, there is an imperfect alignment of the preferences of the LSC with its core (stated) mission of maximizing the quality of the school. Most important, this misalignment of incentives potentially creates a conflict of interest between the objectives of the LSC and those of the Board, an issue we consider shortly.

Finally, if the CEO assumes authority over school governance, we normalize the LSC's utility to zero and, for simplicity, assume the Board eliminates any transparent allocative distortions by setting $x = 1$.

1.5.4. Monitoring and the Information Structure

Monitoring depends upon both the timing of information and actions and the structure of information. Assume events within a given period are timed as follows: (1) At the beginning of a period, each LSC chooses the values of $c$ and $q$; (2) these choices generate a noisy signal of school performance, $s$; (3) the CEO monitors each LSC with probability $m \in (0,1)$, and observes the performance signal, $s$, for those schools that are monitored; (4) on the basis of signal, $s$, the CEO decides whether to intervene in the affairs of the LSC; (5) the election for LSC seats is held; (6) the events just described occur in negligible time. During the remainder of the period, the LSC members derive utility $v(q, c)$ provided that neither the LSC members lose the election nor the CEO intervenes in LSC affairs. Otherwise, the LSC derives (normalized) zero utility, $v_0 = 0$. 
Assumption 5.1 describes the information structure.\(^\text{11}\)

**Assumption 1.5.1 (Information)**

(a) The performance signal, \(s\), is observed by members of the local community

(b) The signal, \(s\), depends on actual educational value added, \(q\), and a random disturbance \(z \geq 0\) according to \(s = q \ast z\). The value of \(z\) is governed by an \(i.i.d\) draw from the exponential distribution: \(z \sim 1 - \exp[-z/\lambda]\).

(c) Each LSC knows its own managerial capacity, \(\theta\). Furthermore, the distributions, signal generating process, and technological relationships are common knowledge

Part (a) is the root of the inherent informational advantage of decentralized versus centralized control: the community is aware of substantive issues pertinent to its own well-being, whereas the CEO only becomes privy to this information through an imperfect monitoring process.

The posited exponential distribution implies the disturbance \(z\) has a mean \(\lambda\) and variance \(\lambda^2\). The parameter \(\lambda\) governs the *informativeness* of the signal, \(s\), and lower values of \(\lambda\) reflect greater informational content. The expected value of the signal, \(\bar{s}\), generated by a school with quality \(q\) is \(\bar{s}(q) = q \ast \lambda\). Each LSC recognizes the expected value of \(s\) depends positively on actual value added, \(q\), of the school under its control. The significance of this fact is that the LSC—through its choice of \(c\)—can influence \(s\) and so the likelihood that the CEO will intervene. Hence each LSC faces a non-trivial decision concerning the choice of \(c\). On the one hand, it derives a direct benefit from \(c\) and, on the other, it has an incentive to temper its choice of \(c\) because it can reduce the likelihood the CEO will abrogate its power.

\(^{11}\)Our results would hold for any distribution function for the disturbance, \(z\), that satisfies the monotone-likelihood ratio property; the exponential distribution considered here, obviously is a particularly tractable member of this family.
If the actual quality of the school in question is $q$, Assumption 5.1 implies that, if monitored, the probability the LSC will lose its authority is $1 - \exp[-\hat{s}/(q * \lambda)]$, for this is the probability that $s$ falls below the critical threshold, $\hat{s}$. As a corollary, the probability that the LSC retains its real authority is $\exp[-\hat{s}/(q * \lambda)]$. Those schools that are not monitored and those that are monitored—but for which the observed performance signal, $s$, exceeds the threshold, $\hat{s}$—are deemed satisfactory, and the CEO delegates the real authority over decision making to the LSC.\(^{12}\)

Even if the incumbent LSC survives the CEO’s monitoring process with its authority intact, the next hurdle it face is the mandated electoral process. The probability it survives the re-election process is $\hat{\mu}(s, \rho)$, where $\rho$ indexes the intensity of local political involvement. We assume this is increasing and concave in the observed performance of the school under its stewardship, $s$, with $\hat{\mu}(0, \rho) = 0$ and $\hat{\mu}(1, \rho) = 1$. In addition, we suppose that $\hat{\mu}(\cdot)$ exhibits a single crossing property, in which for each $\rho$ there is a unique $s(\rho)$ such that $\hat{\mu}_\rho > 0$ and $s > s(\rho)$; $\hat{\mu}_\rho < 0$ if $s < s(\rho)$; and $s(\rho)_\rho > 0$. This formulation captures in the simplest way that a community with a high level of involvement demands high performance for their LSC (i.e. $s(\rho)_\rho > 0$) and the notion that the electoral process runs both ways. Specifically, an already outstanding incumbent LSC ($s > s(\rho)$) is more likely to be re-elected the greater the degree of involvement, $\rho$, whereas exactly the opposite is true for a substandard one ($s < s(\rho)$).

\textit{1.5.5. The Behavior of the LSC}

We assume that at the beginning of the period, the Committee chooses feasible values of $x$ and $c$ to maximize its expected utility. The structure of the model offers a parsimonious

\textsuperscript{12}In practice, there are significant costs incurred by the CEO whenever she exercises her formal authority by overruling the decisions made by a given LSC that temper her office’s behavior. First, if the LSC’s managerial capacity exceeds the Board’s, $>,$ then (all else equal) centralized control results in a direct efficiency loss because the LSC has the greater of the two managerial capacities. Second, she must devote scarce resources to managing newly acquired schools—in particular, costs in evaluating the school principal and deciding whether to terminate his or her contract. Finally, because each LSC is an elected body, there is a potential political cost incurred should her behavior run counter to the community’s wishes.
framework for studying the behavior of LSCs. Although formally a static problem, the sequence of events is such that the Committee anticipates that its choices of \( c \) and \( q \) have subsequent consequences for its well-being.

Given that \( s = q \times z \) and \( z \sim 1 - \exp[-z/\lambda] \), \( s \sim (1/q) \times \exp[-s/\lambda q] \). Therefore the ex ante probability the LSC anticipates surviving the re-election contest is \( u = \int \hat{\mu}(s, \rho) \exp[-s/(\lambda q)] \, ds \equiv \mu(q; \rho) \). Since the LSC’s normalized payoff in the event of CEO intervention is zero, its expected payoff, denoted \( \hat{V}(q, c; \hat{s}, m) \), is

\[
\hat{V}(q, c; \hat{s}, m) = \mu(q; \rho) \times \sigma(q; \hat{s}, m) \times v(q, c),
\]

where \( \sigma(q; \hat{s}, m) \equiv (1 - m) + m \times (\exp[-\hat{s}/(q \times \lambda)]) \) is the probability that the LSC survives the CEO’s monitoring process, with its control rights intact. The term \( (1 - m) \) is the probability it is not monitored, and \( m \times \exp(-\hat{s}/(q \times \lambda)) \) is the probability that it is monitored but retains its control rights because it generates a performance measure, \( s \), that exceeds the given threshold \( \hat{s} \).

By using the constraints \( q = \theta \times x \) and \( c = 1 - x \), each LSC’s problem takes the simple form

\[
V(\hat{s}, m, \theta) \equiv \max_{c, q \in [0, 1]} \hat{V}(q, c; \hat{s}, m, \rho)
\]

s.t., \( \frac{1}{\theta} \times q + c \leq 1 \)

By expressing the problem in this manner, it is immediate from inequality (9) that \( 1/\theta \) is the (implicit) price of school quality \( q \) in terms of foregone \( c \). It follows from this that, well-managed schools—those with a high \( \theta \)—face a lower effective price of providing school quality than do poorly managed ones. Since both \( c \) and \( q \) are valued by the LSC, the resource constraint binds at the optimum. Therefore we can use this constraint to write the LSC’s problem in terms of school quality alone.
Assuming an interior solution, the optimal choice of $q$, denoted $q^* = q(\hat{s}, m, \theta, \rho)$, is governed by the following first-order condition:

\begin{equation}
0 \equiv [\pi * v_q + \pi_q * v] - (\pi * v_c) / \theta,
\end{equation}

where subscripts represent partial derivatives and the FOC is evaluated at the maximum.

The first term in square brackets in equation (13), is the marginal benefit of an increase in school quality $q$. More specifically, provided it survives with its authority intact, which occurs with probability $\pi$, the LSC derives a direct benefit from the increase in school quality valued at $v_q$ at the margin. Furthermore, an increase in $q$ raises the probability that the LSC does indeed survive to see another day by $\pi_q$, which in turn allows it to derive utility, $v$.

The term $-(\pi * v_c) / \theta$ represents the marginal cost of raising school quality, $q$, in terms of forgone $c$. Again, the committee survives (CEO intervention and the electoral process) with probability $\pi$ and (provided it does) suffers a marginal loss in utility of $v_c / \theta$ from the increase in $q$. For the reasons just described, those schools that are most effectively managed—their managerial ability, $\theta$, is high—bear lower cost, in terms of forgone $c$, than do poorly managed schools. Finally, it follows immediately from equation (11) that if $v_c = 0$, then $c^* = 0$: If an LSC values only school quality, it will not distort the resource allocation of the school under its stewardship.

In what follows, let $q^* = q(\hat{s}, 0, \theta_0) > 0$ denote the value added by the school confronted with the least auspicious of circumstances: the given LSC is not monitored (and so has no fear the CEO will intervene based on its performance), and it possesses the lowest possible managerial capacity, $\theta_0$. (Since $m = 0$, $q_0^*$ is independent of $\hat{s}$.) The expected signal generated by such a school is $s_0 = q_0^* / \lambda$. Consider Condition 1 where we assume that the performance
threshold, \( \hat{s} \), satisfies \( m/(1 - m) > [\hat{s} - \bar{s}_0] \ast \exp((\hat{s}/\bar{s}_0)) \). Given our technological assumptions, the maximum expected performance signal is \( \hat{s}(1) = 1/\lambda \). In essence, Condition 1 ensures that the threshold \( \hat{s} \) is not set so high that every school—including the best of the best—is expected that it would be taken over in the event it were monitored.

**Proposition 1.5.1 (The Optimal Behavior of LSC’s)**

Letting \( q^* \) and \( c^* = 1 - q^*/\theta \) denote the Committee's optimal choices, we show

(a) \( q_m^* > 0, \ q_s^* > 0, \ q_{\lambda}^* > 0, \ q_{\rho}^* \leq 0, \) and (b) \( q_\theta^* > 0, c_\theta^* < 0 \)

Proof: All proofs are presented in an Appendix available from the authors.

Part (a), describes the impact of the effect of the joint threats on the behavior of the LSC that, on the one hand, the CEO will intervene and exercise her real and formal authority, and on the other the LSC may not win re-election. A greater audit probability, \( m \), or a tougher performance standard, \( \hat{s} \), induces the LSC to raise the school’s performance level, \( q^* \), to reduce the likelihood the CEO will intervene in its affairs. The effect of an increase in community participation, \( \rho \), is theoretically ambiguous. In particular, a high performing LSC that anticipates it will draw \( s > s(\rho) \) (the single crossing point threshold) anticipates that a marginal increase in \( \rho \) only will increase the probability of its re-election. As a result, it anticipates it can marginally increase \( c \), thereby marginally reducing performance, \( q \), and still survive re-election. An increase in \( \lambda \) corresponds to an increase in the informativeness of the signal, \( s \); it results in an increase in quality, since it is easier for the CEO to distinguish good from bad performance.

One implication of the threat effect just described is that it is insufficient to evaluate the overall effectiveness of the CEO by examining just the change in performance of those schools in which she intervenes directly (for example, via a before-and-after comparison of each school’s
performance). More specifically, even were this set of schools to experience little discernable improvement in performance, the prospect of intervention by the CEO and the LSC's fear of the loss of its control rights may already have increased substantially the performance of each school in the city. Another implication of the threat effect is that it reduces the extent of any room for the ex post observed improvement of any school under the management of a new school principal after being taken over and reorganized by the CEO.

Part (b) of the Proposition describes the effect of the LSC's own managerial capacity $\theta$ on the value added by the school under its control. The fact that better managed schools have higher value added, $q^*_\theta$ perhaps is not too surprising. The interesting part of the Proposition is that $c^*_\theta < 0$—or equivalently $x^*_\theta > 0$. Therefore, a greater managerial capacity, $\theta$, raises school quality directly (because $q = \theta \ast x$) and indirectly because the LSC responds to its own intrinsic higher ability by choosing a better resource allocation. Of course, this means that poorly managed schools generate a low educational value added because they are poorly managed and because they respond in part to their own ineffectiveness by pursuing goals other than maximizing quality as captured by the term $c^*$. In other words, they respond to their inherent low managerial capacity, $\theta$, by substituting away from the relatively high cost provision of quality, $q$, toward $c$.

1.6. Empirical Analysis of LSC Heterogeneity

The formal model provides the structure under which we evaluate how LSC and CEO decisions affect principal quality as measured by value added. The initial empirical analysis examines the relationship between principal quality following a contract event and proxies for LSC capacity and pressure to focus on school quality. Given the uncertainty in quality at the time of hiring, more effective LSCs would be expected to make better contract renewal decisions with
respect to realized principal performance and the observed trajectory of school quality. In addition, a change in political participation would be expected to induce a larger change in school quality for a more effective LSC; as effectiveness approaches zero behavioral responses lead to little perceptible effect. Finally, the model highlights the uncertainty in the effect of political participation. Note that these regressions control for principal effectiveness prior to the end of the contract.

Estimates of the relationships between the changes in effectiveness and LSC characteristics combine the effects of principal transitions initiated by the principal, transitions initiated by the LSC, CEO interventions and contract renewals. Therefore the subsequent component of the analysis examines the relationship between the probability of a specific transition and principal effectiveness prior to the contract event using multinomial logit. We classify transitions into three types of contract events: contract renewal (48.8%), principal departure during or at the end of the contract (48.2%), and principal removal by the CEO (3%).

The absence of information on contract offers limits the analysis, as LSC decisions not to renew and principal decisions not to return are indistinguishable. Nevertheless, the estimates provide information on the relationship between the LSC characteristics and the net effect of contract offer and acceptance decisions. Moreover, CEO removals of a principal do not suffer from the same ambiguity as LSC decisions. However, when the CEO intervenes and removes the principal they may also change other aspects of the school. Therefore, comparisons of principal effectiveness prior to a CEO intervention to effectiveness prior to a transition at a school managed by an LSC must be interpreted with care.

The analysis requires proxies for both capacity and pressure, and we use census block group average parent SES computed over students in the school as the proxy for LSC capacity.
and election participation intensity as the proxy for awareness of school performance. Based on information from the US Census and American Community Survey, the SES index is a function of adult education and adult employment in managerial positions. In the case of the SES index, we assume that formal education elevates the capacity to comprehend information on school performance, and white collar occupations enhance the capacity to manage and supervise. In the case of the voting intensity measure, we assume that higher intensity reflects higher awareness of and interest in school performance.

SES and voting intensity are likely related to factors that affect achievement, raising the possibility that they influence principal value added directly. Yet because we focus on the change in principal effectiveness following a contract event and the principal transitions conditional on performance, we believe that they provide valid proxies for factors that affect LSC capacity and behavior. Moreover, the inclusion of variables that capture changes in demographic characteristics has virtually no effect on the SES or voting intensity coefficients.

Table 7 illustrates the joint distribution of SES and voting intensity. Although the table shows that the fraction in the top voting intensity quartile is much higher for the schools in the top SES quartile than for the others, there is substantial variation in voting intensity in all SES quartiles.

Table 8 reports estimates of the relationship between the change in principal effectiveness following a contract event and both SES and voting intensity using the following specification:

\[
\Delta E_s^{1,2} = \alpha + \gamma LSC_s^{1'} + \beta SES_s^{1'} + \Delta X_s^{1,2} + \delta y + \varepsilon_s^2
\]

where the change in principal effectiveness (effectiveness under contract 2 minus effectiveness under contract 1) is a function of LSC voter intensity \(LSC_s^{1'}\) and parental SES \(SES_s^{1'}\), each measured in the nearest year prior to the end of the contract, a vector measuring the change in
student demographics and school characteristics between the first and second contract ($\Delta x_{s}^{1,2}$), year fixed effects ($\delta_{y}$), and a random error. The change in principal effectiveness following the contract event equals VA measured during the 2nd year following the contract event minus VA measured during the second year of the initial contract period prior to the event. Note that the second specification includes the interaction between SES and voting intensity and the third specification also adds the interaction between voting intensity and effectiveness in order to examine potential heterogeneity in the responsiveness to voter participation by initial principal effectiveness as suggested by the theory.

The estimates in Table 8 support the predictions of a positive relationship between the change in principal effectiveness and both capacity and incentives. Both the SES and voter intensity coefficients in the left column are positive and significant at the 10 percent level despite the small sample size. Moreover, the interaction term coefficient in the middle column is consistent with the notion that higher capacity amplifies effects of voting intensity. However, the prediction that the effect of voter intensity should weaken with initial principal effectiveness is not supported by the estimates reported in Column 3. Rather, the interaction term is positive although not significant.

Principal productivity is likely to change through two primary channels: hiring and contract renewal. Because most new principals were not previously principals in CPS, it is difficult to evaluate the hiring decision. Therefore we focus on principal transitions out of a school and investigate whether systematic differences between the probability of exiting and principal effectiveness emerge by both SES and LSC election participation. To do so, we estimate the following multinomial specification,

$$\text{CONTRACT}_{pst} = \alpha + f\left(E_{pst}\right) + f\left(E_{pst}\right) \ast SES_{st} + f\left(E_{pst}\right) \ast LSC_{pst} + \epsilon_{pst},$$

(15)
where $CONTRACT_{pst}$ measures how the contract ends for principal $p$ in school $s$ at time $t'$ and takes on three values: renewal (omitted category), exit, and removal by CEO. The basic specification reported in Column 1 models the contract event as a function of principal effectiveness during the contract $f(E_{pst})$, the interaction of effectiveness and socioeconomic status of the parents $f(E_{pst}) \times SES_{st}$, the interaction of effectiveness and the voting intensity in the election nearest the end of the contract $f(E_{pst}) \times LSC_{pst}$, and an error $\epsilon_{pst}$. Column 2 adds the interaction between voting intensity and parental SES and the full three-way interaction between SES, voting intensity and effectiveness.

The coefficients in the top panel of Table 9 reveal a strong, negative relationship between principal value added prior to the contract event and the probability of exiting that appears to strengthen as parental SES increases but weaken as voting intensity increases. Inclusion of the triple interaction in the third column strengthens the negative effect of SES on the relationship between effectiveness and the probability of exit at lower but not higher levels of participation.

The bottom panel of Table 9 reveals a negative though only marginally significant relationship between value added and the probability of leaving by CEO removal relative to continuation. CEO removals comprise only three percent of outcomes, and given the inclusion of interaction terms the imprecision of the estimates is not a surprise. Nonetheless, the magnitude of the main effect suggests that moving from the 90th to the 10th percentile of the effectiveness distribution roughly doubles the probability if the values of voting intensity and SES are set to zero; the difference is much larger for schools near the middle of the SES and intensity distributions.
1.7. Conclusions

The devolution of authority over principals to local school councils constituted a major change in the structure of school governance. Whether this had a substantial effect on the distribution of principal effectiveness depends on both the underlying variance and the extent to which LSC heterogeneity amplified or contracted the existing distribution at the time of the reform.

The results reveal meaningful differences in principal effectiveness, confirming the importance of the LSC responsibility over principal hiring and contract renewal. Not only is there significant variation in principal effects that is strongly correlated with teacher survey responses regarding principal performance, but an analysis of variance approach based on principal turnover that goes to great extent to account for time-varying school differences and the dynamics of principal effects also shows substantial variation in principal effectiveness.

The analysis of LSC effects on the distribution of principal quality provides evidence consistent with the notion that higher LSC management capacity improves decision-making and principal effectiveness. There is also evidence of larger improvement in principal effectiveness following the end of a contract in schools with higher voter participation and SES.

The results provide reason to be concerned that decentralization may generate smaller improvements and may even harm schools serving the least advantaged students in terms of parental SES and election participation. Given the strong negative associations between school poverty and both SES and voter intensity, this suggests that decentralization may be least beneficial for some of the highest poverty schools.
2. The Signaling Value of Merit Awards

2.1 Introduction

Employers concerned with worker output often incentivize effort through merit awards. A large theoretical and empirical literature documents that merit awards may raise worker productivity and output per worker (Lazear 2000; Lavy 2009; Glazerman and Seifullah 2010; Woessmann 2011; Muralidharan and Sundararaman 2011; Gius 2013; Goodman and Turner 2013; Neckermann, Cueni, and Frey 2014; Brehm, Imberman, and Lovenheim 2015). Although visible awards may increase worker effort and productivity, they may also produce unintended consequences. By acting as a signal of quality awards may increase outside options and increase worker turnover. Researchers have noted the implications of merit awards systems for average job mobility (MacLeod and Malcomson 1998; Jones 2013). However, theoretical and empirical evidence on the effect of awards themselves is lacking because of the difficulty isolating the effect of an award from unmeasured differences in productivity. Even in cases where micro-level worker productivity data are available, endogeneity due to unobservable differences in productivity and ability often renders estimates of the effect of award receipt on job mobility unreliable.

In this paper, I present new evidence on the signaling effects of merit awards using a unique policy that rewarded school principals in Chicago for reaching goals related to test-score growth. Specifically, I examine how winning a merit award affects principal turnover. I estimate using a regression discontinuity design exploiting the fact that Chicago provides principals with merit awards based solely on achieving a particular threshold of performance. The threshold is not publicly known ahead of time and principal performance is measured based on student test scores. Both of these features make it extremely unlikely that principals have the ability to finely manipulate the metric that determines awards.
Using a model of job search that incorporates a signaling component, I show that award receipt is predicted to increase turnover for winners. The intuition for this result is that the award signals quality and increases the wage-offer distribution of award winners such that winners are predicted to receive and depart toward higher-quality offers. Consistent with the predictions of the theory, I find that being awarded a merit bonus increases principal turnover. Principals who just exceeded the threshold for a merit award are more likely to exit their school the year after winning compared to principals who fell slightly short of the award threshold, driven almost entirely by exits from the district. Award winning principals who transition within district are disproportionately more likely to move toward higher performing and higher-paying settings, a finding likely driven by movement toward higher enrollment schools.

In addition to providing direct evidence on the impact of merit award receipt on turnover, this paper informs the literature on principal evaluation. A growing literature finds that school management matters for student success (Branch, Hanushek, and Rivkin 2012; Coelli and Green 2012; Dhuey and Smith 2014; Bloom et al. 2015; Laing et al. 2016). Accordingly, education policy makers now increasingly focus on ways to incentivize principal effort. Many recent initiatives have focused on rewarding principals with both public recognition and financial incentives in hopes of improving principal performance and the skill distribution of the principal applicant pool. I conclude by discussing whether the award program is able to identify historically high performing principals and how turnover generated by the program affects schools.

I find that 75 percent of principals in the top quartile of principal test-score value added in the award year also performed in the top half of the distribution in the prior four years. Similarly, 67 percent of principals in the bottom quartile of principal value added in the award
year were in the bottom half of the distribution in the four prior year performance. Principals in the middle of the distribution are more difficult to pick out. These findings suggest that the award program does in fact reward historically high performing principals. Focusing on changes in performance, enrollment, and student poverty around the award threshold, I find that places just to the right of the threshold experienced test score declines although enrollments increased and student poverty declined.

The findings in this paper have important implications for the implementation of principal incentive programs and, more generally, employee incentive programs. Rewards have the potential to induce effort and increase the talent pool of principals. Despite this, one potential unintended consequence of providing high-quality workers a verifiable credential is that it may alter decisions over job mobility.\textsuperscript{13} Past research has documented that principal turnover is associated with school performance, so a policy that increases turnover is likely to significantly affect student success.

\subsection{2.2. The Achievement Award Program}

Beginning in the fall of 2011, the Chicago Public Schools (CPS) announced a principal achievement award that provided both public recognition of success and a one-time financial bonus ranging from $5,000 to $20,000. The program, backed by the mayor and funded by wealthy donors, aimed to raise principal accountability over several test-based dimensions.

In Table 10, I report the exact criteria a principal had to achieve in order to win. In both years of the program, two of the four criteria were related to gains in value added to math and

\footnote{\textsuperscript{13} In a recent dissertation, Wills (2016) considers if principal credentials (e.g. level of education received) influence student achievement and principal turnover in South Africa. Using a fixed effect framework, she finds that more credentialed principals compared to less credentialed principals are less likely on average to leave their school yet are no better at raising test performance. The principal award credential in Chicago differs in that it is received by principals who have demonstrated an ability to raise achievement.}
reading achievement. The remaining criteria captured growth in college readiness (only in 1st year) and test score growth for minority groups. In both years of the program, a principal earned $5,000 by reaching any two of criteria, $10,000 by reaching any three, and $20,000 by reaching all four. Although the payout varies, 71% of winning principals earned the $5,000 bonus and 24% earned the $10,000 bonus. An average principal in the CPS in 2011-12 earned $129,450 so the majority of winners received a one-time bonus of between 4 to 8 percent of salary. In total, approximately 24% of the 710 principals in the sample received an award.

Although the payout is identical and the criteria are largely the same between the first and second year, the way in which a principal met the criteria changed. In the first year, a principal reached targets if they ranked at or above a certain threshold based on the distribution of CPS principals (i.e. a rank-order tournament). A principal’s ranking was therefore not only determined by their own effort, but also by the effort of all other principals in the district. In the second year, a principal reached targets by ranking at or above a threshold based on rankings relative to a national distribution. In both years, principals had virtually no ability to precisely manipulate their standing.

In each year, awards were announced during the fall of the subsequent school year. To be eligible to win, principals had to still serve at a school within the CPS during the fall of the next school year (not necessarily the same school). For example, during the first year of the program, performance was measured during the 2011-12 school year and award winners were announced during the fall of the 2012-13 school year. If the principal exited the district between 2011-12 and 2012-13 they were not eligible to win. Therefore, the sample is limited to principals who still served in the CPS the years the awards were announced which leaves 710 principal-year observations.
2.3. How the Award Program Affects Transitions and Mobility

I use a model of signaling embedded into simple model of job search to generate predictions about the effect of job awards on employee turnover. The model follows the literature on signaling such as Aigner and Cain (1977) and the search framework of McCall (1970) and Zaretzky and Coughlin (1995).

Principals are currently employed at wage, $w$, and draw new wage offers from the distribution $F(w')$ that is centered on perceived principal productivity. Here wages include both pecuniary aspects of the job like salary and also non-pecuniary aspects of the job like working conditions. Each principal has true underlying productivity, $p_i$, with variance $\sigma_p$, and a publicly observable signal of productivity, $s_i$, with variance $\sigma_e$. The signal is a noisy measure of true productivity according to $s = p + e$ where $e$ is an independent and normally distributed error term.

Potential employers seek to pay principals their true productivity yet they only observe the overall average productivity of principals, $\bar{p}$, and the signal, $s_i$. As such, potential employers pay according to their best guess of true productivity which equals a weighted average of overall productivity and the signal. In particular, a potential employer will pay according to $w'_i = \bar{p}(1 - \alpha) + s\alpha$ where $\alpha = \frac{\text{cov}(p,s)}{\text{var}(s)} = \frac{\text{cov}(p,p+e)}{\text{var}(p+e)} = \frac{\text{cov}(p,p)+\text{cov}(p,e)}{\text{var}(p)+\text{var}(e)+2\text{cov}(p,e)}$. Assuming that $p$ and $e$ are orthogonal, $\alpha = \frac{\text{var}(p)}{\text{var}(p)+\text{var}(e)} = \frac{\sigma_p}{\sigma_p+\sigma_e}$. Therefore, a potential employer is willing to pay optimal wages according to $\mu_i = \bar{p} \left( \frac{\sigma_e}{\sigma_p+\sigma_e} \right) + s_i \left( \frac{\sigma_p}{\sigma_p+\sigma_e} \right)$.

A principal’s signal of productivity is a function of a portfolio of items including their resume, recommendations, and award receipt. Winning a merit award augments the signal which
will shift rightward the mean of the wage-offer distribution \( \mu_i \). In addition, the first order condition \( \frac{\partial \mu_i}{\partial s_i} \) also suggests that the rightward shift in \( w'_i \) is larger the smaller is the variance of the signal and the larger the variance in principals’ true productivity.\(^{14}\)

Having demonstrated that the award will shift right the distribution of wage offers from which a principal draws, I now develop a search model using a basic McCall framework and incorporate the predictions from the signalling model. Principals receive new wage offers, \( w' \), with probability \( \delta \) (offer arrival rate), according to the wage offer distribution \( F(w') \) which is centered around \( w' \) with mean \( \mu \). If a principal declines a new wage offer, she receives her current wage offer and searches again. Also assume that principals discount the future with discount factor \( \beta \).

The value of accepting a new wage offer is \( V_A = \frac{w'_i}{1-\beta} \) and the value of rejecting the wage offer is \( V_R = w_i + \beta(1-\delta)E[V(0)] + \beta \delta E[V(w'_i)] \), where \( E[V(0)] \) is the expectation of receiving no wage offer. To determine optimal job search behavior, principals chose the maximum of \( V_A \) or \( V_R \),

\[ V(w) = \max_{A,R} \{ V_A, V_R \} = \max_{A,R} \{ \frac{w'_i}{1-\beta}, w_i + \beta(1-\delta)E[V(0)] + \beta \delta E[V(w'_i)] \} \]

The reservation wage, \( W_R \), makes a principal indifferent between accepting an offer now and rejecting and continuing to search. To learn how a principal makes decisions between \( V_A \) and \( V_R \), one must set \( V_A = V_R \) and solve for the reservation wage,

\[ \frac{W_R}{1-\beta} = w_i + \beta(1-\delta)E[V(0)] + \beta \delta E[V(w'_i)] \]

\[ \vdots \]

\[ \frac{\partial \mu_i}{\partial s_i} = \frac{\sigma_p}{\sigma_p + \sigma_u} = \frac{\sigma_p}{1 + \frac{\sigma_p}{\sigma_u}} > 0 \]

\(^{14}\)}
(18) \( w_R - w_i = \frac{\beta \delta}{1 - \beta} \int_{w_R}^{\infty} (w' - w_R) dF(w') \)

The left hand side of equation (18) is the marginal cost of rejecting a new offer and the right hand side is the marginal benefit of continuing the job search. As the award signal increases the mean of the wage-offer distribution, \( \mu \), the marginal benefit of search increases relative to the marginal cost which will tend to raise the reservation wage, \( w_R \). In other words, as the mean of wage-offers rises, principals will tend to become more selective over which offers they accept and the reservation wage will rise. To see this, consider the first order condition of the reservation wage with respect to the mean of the wage-offer distribution

(19) \( \frac{dw_R}{d\mu} = \frac{1}{1 + \frac{1 - \beta}{\beta \delta [1 - F(w')]} } \in (0,1) \).

To learn how principals search behaviour changes as a result of the rightward shift in the wage-offer distribution generated by award receipt, consider the probability of acceptance. The probability of accepting a job (i.e., hazard rate of leaving current job) equals the product of the probability of job offer and the probability of accepting the offer conditional on the presence of an offer. So the hazard rate of leaving a current job for a new one, \( \gamma \), is linked to the reservation wage by,

(20) \( \gamma = \delta \phi(w_R, \mu) \)

where \( \phi \) represents the probability of acceptance conditional on an offer and is a function of both \( w_R \) and \( \mu \).

The award signal ultimately increases the hazard rate of leaving because

(21) \( \frac{d\gamma}{d\mu} = \delta \left[ \frac{\partial \phi}{\partial w_R} \frac{d w_R}{d\mu} + \frac{\partial \phi}{\partial \mu} \right] > 0 \)

where \( \frac{\partial \phi}{\partial w_R} = -f(w_R) \). The channel driving the increase in the hazard rate of leaving is that award receipt raises the signal and optimal pay, which increases the mean of the wage-offer
distribution, $\mu$, and the reservation wage, $w_R$, but increases $w_R$ by less than the amount that the distribution shifted right ($\frac{\partial \phi}{\partial \mu} > \left| \frac{\partial \phi}{\partial w_R} \frac{\partial w_R}{\partial \mu} \right|$). In other words, as the signal shifts right the distribution of wage offers, principal turnover is expected to increase.

### 2.4. Measuring the Effects of Achievement Awards

I use administrative data on CPS students as well as publicly available data from the Chicago Public Schools Accountability Center to measure award receipt, principal turnover, school demographics, and subsequent test performance. The administrative data contain information on every student in grades 3 through 8 and include test scores for the Illinois Standard Achievement Test (ISAT) and Northwest Educational Assessment (NWEA) exams as well as student demographics and program characteristics. Information on principals includes employment records from which I learn salary and whether or not the principal switched schools or left the district. The records span seven years from 2009 to 2015. The main sample is limited to the award years 2012-13 and 2013-14 and consists of 710 school-principal level observations. Data prior to 2012-13 are incorporated in some analyses.

To define a transition, I focus on whether or not the principal exits their school following award receipt. In both years of the program, principals are given awards in period $t$ based on their performance in period $t-1$. After having an award in hand, a principal may continue in their current school or move in year $t+1$. The primary outcome then is principal turnover between $t$ and $t+1$. For example, in the first year of the program, I measure transitions between 2012-13 and 2013-14 and in the second year of the program between 2013-14 and 2014-15.

Although award receipt is endogenous, the unique structure of the merit award generates discontinuities in the probability of award receipt based on percentile rankings that allows one to
overcome these difficulties. A file from the Chicago Public Schools Accountability Center contains the exact ISAT value-added scores used to calculate the first two percentile rankings from the first year of the program. However, in the second year of the program I must reconstruct the percentile rankings using the administrative data on NWEA test scores following the procedure described in the School Quality Rating Policy Handbook produced by the CPS. Some aspects of the procedure involved imputing missing information and some elements of the calculation. Nonetheless, given the guidance of the handbook there was very little error in my prediction of who won.

In the main analyses, I focus on any award receipt and disregard differences in the amount of the award because there are simply too few winners at the highest payouts. To develop my RD framework, I focus on the award criteria related to growth in overall math and reading scores. In the first year these include ISAT math and reading scores and EXPLORE scores (3 of the 4 criteria) and in the second NWEA-MAP math and reading scores (2 of the 4 criteria). I focus on these criteria because I am best able to measure them and they are largely similar between both years of the program. Because my running variable does not incorporate all the criteria, some principals will be misclassified, making my research design a fuzzy regression discontinuity.

In this framework, the principal’s second highest percentile ranking becomes particularly important because as long as the second highest percentile ranking is at or above the award threshold (i.e. 75th percentile in the first year and the 90th percentile in the second year), then the principal will win an award. Consider an example from the first year of the program. A hypothetical principal who ranks in the 85th percentile of the ISAT math value-added distribution and the 80th percentile of the ISAT reading value-added distribution would win because their
second highest percentile ranking exceeds the 75th percentile. A second hypothetical principal who ranks in the 90th percentile of the ISAT math value-added distribution and the 70th percentile of the ISAT reading distribution would be unlikely to win because their second rating is below the 75th percentile.

I pool the two years of data together and treat them as independent from each other. Results are similar, but noisier when each year is analyzed separately. In both years I center the running variable around the award threshold (i.e. 75th percentile in the first year and 90th percentile in the second year) so that principals ranking zero or greater are classified as winners. The regression discontinuity analysis focuses on principals near the threshold, and I demonstrate that the results are robust to bandwidth choice.

2.5. Regression Discontinuity Analysis

In order to isolate the effects of the award signal from motivation, skill, or other unobserved differences I use a regression discontinuity approach. The RD design is implemented using the following regression model,

\[ y_{it,t+1} = f(r_{it-1}) + Above_{it-1} \delta + f(r_{it-1}) \cdot Above_i + \epsilon_{it,t+1} \]

where turnover for principal \( i \) between \( t \) and \( t+1 \) \( (y_{it,t+1}) \) is some function of their second largest percentile ranking \( (f(r_{it-1})) \), an indicator for passing the award threshold \( (Above_{it-1}) \), their interaction \( (f(r_{it-1}) \cdot Above_{it-1}) \), and a random error term \( (\epsilon_{it,t+1}) \). The interaction allows the turnover gradient to differ on either side of the award threshold.

The parameter \( \delta \) provides the causal effect of award receipt on turnover for principals at the margin of winning. The RD comparison is unbiased so long as treatment is as good as randomly assigned around the threshold. To support the primary identifying assumption, the
same cutoff must not be used for other policies and there must be no precise manipulation around the threshold. To my knowledge, there is no other policy in the CPS using precisely the same rule based on these four criteria. The possibility of precise manipulation around the threshold is \textit{a priori} unlikely given that where one principal falls in the distribution is a function not only of their own effort and performance, but also a function of the effort and performance of all other principals in the CPS (first year) or nationwide (second year).

Nevertheless, it is possible to test for the presence of precise manipulation by presenting density tests and the balance of covariates around the threshold. If baseline covariates are continuous through the threshold, then precise manipulation is unlikely. In Appendix Figure A1 I present the density of the RD running variable. The density of the running variable is smooth through the threshold with no evidence of bunching on other side suggesting that precise manipulation is unlikely. In Appendix Table A1 I present a series of balance tests to demonstrate that available baseline covariates (i.e. in year $t - 1$) including principal salary, student body race, share new students, share special education, share female students, share grade repeaters, enrollment, socioeconomic status, and concentration of poverty do not vary discontinuously around the threshold. None of the twelve baseline covariates presents a significant jump at the threshold suggesting that precise manipulation is unlikely.

\textbf{2.6. Results}

The first set of figures uses the discontinuity in award receipt generated by percentile rankings and presents the unconditional regression discontinuity estimates documenting a relationship between award receipt and turnover. After documenting the basic relationship, I present a full set of reduced-form and 2SLS regressions estimated over varying windows and
with or without controls. Subsequently, I consider the nature of transitions and focus on the types of schools to which principals move. Last, I explore the structure of the award program and whether or not it is able to pick out historically high performing principals.

2.6.1. OLS Estimates

Establishing the causal relationship of award receipt on turnover is challenging, because award receipt is correlated with underlying differences in productivity, motivation, and skill. Consider the following OLS specification,

\[ Y_{it,t+1} = \alpha + \text{WINNER}_{it} \beta + X_{it} \gamma + \epsilon_{it,t+1} \]

where turnover between \( t \) and \( t + 1 \) \((Y_{it,t+1})\) is a function of an indicator for award receipt \((\text{WINNER}_{it})\), a vector of observable baseline control variables \((X_{it})\) including principal salary, student demographics, and socioeconomic status, and a random error term \((\epsilon_{it,t+1})\). The estimate \( \beta \) describes the average difference in turnover for winners and losers, which will likely be biased due to unobservable differences in productivity, motivation, and skill.

In the panel A of Table 11, baseline OLS estimates for the entire sample reveal a negative relationship between award receipt and turnover, suggesting that principals who win are less likely to leave their school in the year following award receipt. After controlling for observable school characteristics, the estimate falls to. Moving to panels B and C, I narrow the sample to include those near the award threshold providing estimates more in the spirit of the regression discontinuity analysis. Narrowing the window over which I estimate begins to approximate the regression discontinuity approach, and the estimates become positive and significant. Although in some specifications the OLS estimates suggest that award receipt raises the probability of a
transition, the assumption of no omitted factors is a very strong assumption that may well be violated so I turn now to the RD analysis.

2.6.2. Regression Discontinuity Estimates

In Figure 1 I present the first-stage relationship between predicted and actual award receipt. I group the raw data into bins of size 5 based on predicted award receipt (i.e. the running variable). Each point on the figure represents average award receipt for the respective bin. Crossing the threshold is associated with a visible jump in the probability of actually winning. For those nearest the award threshold (±5 percentile rankings), the discontinuity is roughly 50 percentage points and is statistically significant at the 1% level. The discontinuity in treatment is both visually apparent and significant whether estimated with Local Linear Regression or a quadratic specification.

Using the discontinuity in award receipt depicted in Figure 1, I now explore differences in the probability of turnover. In Figure 2, I present the primary relationship of interest, the reduced form relationship between predicted award receipt and principal turnover. As expected, there is a downward gradient of turnover consistent with the notion that aggregate turnover falls as the level of school performance rises. This reflects the well-known fact that turnover is higher in schools with lower performance due primarily to less desirable working conditions (Béteille, Kalogrides, and Loeb 2011). 15 Although turnover generally falls with student performance, Figure 2 shows that principals at schools with the highest levels of performance have a similar level of turnover as principals at much lower performing schools. This appears to be driven by

---

15 The study also suggests the likelihood that the current district is less likely to remove a principal as performance increases and evidence that job satisfaction seems to be positively related to performance. These are reasons that turnover could fall as a result of award receipt.
the fact that there is a large and positive discontinuity right at the threshold such that those to the right are more likely to exit than those just to the left. Here there is a visible increase in the probability of turnover for those who just won compared to those who just lost. The discontinuity is large enough so that a principal 10 percentiles above the award threshold has approximately the same probability of turnover as a principal 20 percentiles below.

In the absence of the award program, one would expect turnover for high ranking principals to be near zero. To demonstrate the robustness of the results from Figure 2, in Figure 3 I hypothetically rank principals in the year before the award program as if the program existed. What is clear from the figure is that in the absence of the award program, highly ranked principals have very low turnover compared to more lowly ranked principals. The finding suggests that the turnover observed after the introduction of the award program may indeed be induced by the award signal.

Rather than rely on the figures alone, I now turn to the corresponding regression estimates with standard errors to understand if the discontinuity is plausible or explained by random noise. In Table 12 I present a series of reduced-form results estimated over different bandwidths in order to demonstrate my estimate are not driven by how I fit the data. Focusing on the estimates for those in a $\pm 10$ window, the point estimate suggests those who just won compared to those who just lost are approximately 16 percentage points more likely to exit their school the following year, a statistically significant increase of nearly 76% off the control-group mean. Moving from left to right I widen the bandwidth over which I estimate the regression and demonstrate that the estimates are largely insensitive to this choice. In Panel B, it is clear that the estimates are largely insensitive to baseline controls, as adding covariates changes the estimates little. Nearly 96% of the turnover is driven by principals exiting the district.
In Table 13, I present 2SLS estimates which provide the treatment effect for those who actually won the award. To do so, the 2SLS scale the reduced form by the corresponding first stage estimate which ranges from 0.5 to 0.6, so the 2SLS estimates will be roughly double in size. For those in the ±10 window, winners affected by treatment are 31 percentage points more likely to exit compared to those who just lost.

2.6.3. Movement

The results from section 2.6.2 demonstrate that winning principals are far more likely to exit their school in the subsequent years than are principals who just lost. The search model predicts, among other things, that winning principals who transition will move to places with greater pecuniary and non-pecuniary benefits like salary and better working conditions. Past work by Beteille et al. (2011) suggests that principals have a strong revealed preference for moving to higher SES and higher achieving schools. In this section I explore the change in these characteristics between origin and destination schools for those who remain principal in the Chicago Public Schools.

I now explore the change in average job characteristics between year $t$ and $t+1$ for principals who remain in the district using the following specification,

\begin{equation}
\Delta y_{it,t+1} = f(r_{it-1}) + Above_{it-1} \delta + f(r_{it-1}) \cdot Above_{it-1} + \Delta \epsilon_{it,t+1}
\end{equation}

where the change in job characteristics for principal $i$ between $t$ and $t+1$ ($\Delta y_{it,t+1}$) is some function of their percentile ranking ($f(r_{it-1})$), an indicator for passing the award threshold ($Above_{it-1}$), their interaction ($f(r_{it-1}) \cdot Above_{it-1}$), and an error term ($\Delta \epsilon_{it,t+1}$). Available job characteristics include salary, school enrollment, parental SES, school poverty, student
demographics and program characteristics. The regression also controls for program year effects to adjust for aggregate changes in principal pay or school characteristics.

Table 14 demonstrates no significant relationships. However, the search model predicts that the distribution of wage offers for award winners should shift to the right suggesting that winners are predicted to receive higher quality offers. Consistent with the predictions of the model, winning principals appear to transition to schools with greater pecuniary (salary) and non-pecuniary (test scores) job traits. Those just to the right of the threshold who stay within the district experience small increases in pay. Focusing on those within ±5 rankings around the award threshold, those just exceeding the threshold experience a $204 increase in salary relative to the control group. In the CPS, principal salary is largely determined by years of experience and school enrollment, so the primary way in which a principal may increase salary is by moving toward higher enrollment schools. I find that those just to the right of the threshold do indeed move to schools with roughly 17 more students. Although past research has demonstrated that principals have a revealed preference for higher SES settings, no clear pattern emerges when I focus on changes in parental SES or poverty. Finally, principals just exceeding the award threshold move toward schools with slightly higher test scores. The sample size is smaller because test scores from 2015 are not yet available so the estimates are identified only from the first year of the program and should be interpreted with care.

2.6.4. Merit Awards and Schools

Merit awards can affect principal effort, the desirability of working in particular schools due to perceived differences in the probability of winning an award, perceived differences in principal productivity and the availability of outside job offers, and over the longer-term the skill
distribution of the principal applicant pool. The structure of the award program allowed me to study how perceived differences in principal productivity alter job mobility. Although the remaining potential effects of the program are more difficult to identify because of endogeneity, in this section I explore descriptively whether or not awards actually go to better principals as opposed to those who got lucky and whether schools that lose award winners are hurt.

In Table 15, I describe the distribution of award year performance conditional on a principal’s performance over the prior four years in terms of value added to test scores and test score levels. If the award isolates actual performance, then one would expect principals in the top (bottom) of the award-year performance distribution to also be in the top (bottom) of the prior performance distribution. In panel A, I focus on principal value added to test scores and find that approximately 75% of principals who performed in the highest quartile in the award year also performed in the top half of the distribution over the prior four years. Similarly, approximately 67% of principals performing in the lowest quartile in the award year also performed in the bottom half of the distribution the prior four years. In panel B, a very similar pattern emerges when focusing on the level of performance. Table 15 suggests that the achievement award program does indeed reward persistently high performing principals.

Work by Beteille et al. (2011) and Miller (2013) suggests that principal turnover is associated with declines in school performance. Previous results demonstrated that schools to the right of the threshold lost their principals at higher rates than schools just to the left. Using the same discontinuity in award treatment, in Table 16 I explore the change in school characteristics between $t$ and $t + 1$ in schools around the award threshold. Consistent with past research, there appears to be a decline in performance, though estimates are largely insignificant. Enrollment
tends to rise and poverty of the student body tends to fall suggesting that schools where a principal won an award may attract more and wealthier students.

2.7. Discussion

Although much evidence exists demonstrating that merit awards likely raise employee effort, little is known about potential unintended consequences of such policies with respect to employment decisions. This paper provides new evidence with respect to these issues using administrative data on school principals and a unique policy that generated discontinuities in the probability of award receipt.

Consistent with the predictions of a search model with a signaling component, the empirical work provides strong evidence that award receipt increases the probability a principal will exit their school for opportunities outside of the CPS district. Moreover, award winning principals who transition within the CPS appear to move to higher paying and higher performing schools than the school they previously led.

3. Conclusions

Both chapters of my dissertation illustrate the importance of considering the unequal or unintended consequences of policies. In the first chapter, we studied the devolution of authority over principal hiring, evaluation, and retention and how it influenced student achievement. We found that the benefits of decentralized decision making depend crucially on the ability of the local decision makers to make managerial decisions and on the strength of the incentives they face to make decisions in line with maximizing student success. Importantly, the findings demonstrate that decentralization may provide little benefit to schools serving the most
disadvantaged students. In the second chapter, I studied a recent program that provided principals public recognition and financial rewards if they produced large gains in student performance. Although the original intent was to induce principal effort and increase the talent pool of principals, I found that the program induced very high performing principals to exit their schools after winning an award.
4. Bibliography


5. Appendix

5.1 Evidence of IRB Approval

UNIVERSITY OF ILLINOIS
AT CHICAGO

Office for the Protection of Research Subjects (OPRS)
Office of the Vice Chancellor for Research (M/C 475)
190 Administrative Office Building
1737 West Polk Street
Chicago, Illinois 60612-7227

Approval Notice
Initial Review (Response To Modifications)

October 3, 2012

Steven Rivkin
Economics
601 S Morgan Street 729UH
M/C 144
Chicago, IL 60612
Phone: (312) 413-2368 / Fax: (312) 996-3344

RE:  Protocol # 2012-0744
“A Study of the Dynamics of the Market for Principals and the Distribution of Principal Effectiveness in the Chicago Public Schools”

Dear Dr. Rivkin:

Your Initial Review application (Response To Modifications) was reviewed and approved by the Expedited review process on September 27, 2012. You may now begin your research.

Please note the following information about your approved research protocol:

Please note that UIC will defer to the requirements of the organizations granting access to the data (Chicago Public Schools, University of Chicago) regarding the signatures required for release of the data. Kindly notify the UIC OPRS/IRB if the signature of the Vice Chancellor for Research and/or the Board of Trustees will be required.

Protocol Approval Period: September 27, 2012 - September 27, 2013
Approved Subject Enrollment #: 5,000,000 cases
Additional Determinations for Research Involving Minors: The Board determined that this research satisfies 45CFR46.404, research not involving greater than minimal risk.
Performance Sites: UIC, University of Chicago Consortium on Chicago School Research
Sponsor: None
Research Protocol:

a) A Study of the Dynamics of the Market for Principals and the Distribution of Principal Effectiveness in the Chicago Public Schools; Version 1; 09/04/2012

Recruitment Material:

a) No recruitment materials will be used - data will be accessed under an agreement with the University of Chicago Consortium on Chicago School Research

Phone: 312-996-1711 http://www.uic.edu/depts/ovcr/irb/ FAX: 312-413-2929
Informed Consent:

a) A waiver of consent/permission/assent has been granted under 45 CFR 46.116(d) for access to Chicago Public Schools data under an agreement with the University of Chicago Consortium on Research on Chicago School Research.

Your research meets the criteria for expedited review as defined in 45 CFR 46.110(b)(1) under the following specific categories:

(5) Research involving materials (data, documents, records, or specimens) that have been collected, or will be collected solely for non-research purposes (such as medical treatment or diagnosis).

(7) Research on individual or group characteristics or behavior (including but not limited to research on perception, cognition, motivation, identity, language, communication, cultural beliefs or practices and social behavior) or research employing survey, interview, oral history, focus group, program evaluation, human factors evaluation, or quality assurance methodologies.

Please note the Review History of this submission:

<table>
<thead>
<tr>
<th>Receipt Date</th>
<th>Submission Type</th>
<th>Review Process</th>
<th>Review Date</th>
<th>Review Action</th>
</tr>
</thead>
<tbody>
<tr>
<td>09/07/2012</td>
<td>Initial Review</td>
<td>Expedited</td>
<td>09/13/2012</td>
<td>Modifications</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>Required</td>
</tr>
<tr>
<td>09/21/2012</td>
<td>Response To Modifications</td>
<td>Expedited</td>
<td>09/27/2012</td>
<td>Approved</td>
</tr>
</tbody>
</table>

Please remember to:

⇒ Use your research protocol number (2012-0744) on any documents or correspondence with the IRB concerning your research protocol.

⇒ Review and comply with all requirements on the enclosure, "UIC Investigator Responsibilities, Protection of Human Research Subjects".

Please note that the UIC IRB has the prerogative and authority to ask further questions, seek additional information, require further modifications, or monitor the conduct of your research and the consent process.

Please be aware that if the scope of work in the grant/project changes, the protocol must be amended and approved by the UIC IRB before the initiation of the change.

We wish you the best as you conduct your research. If you have any questions or need further help, please contact OPRS at (312) 996-1711 or me at (312) 996-2014. Please send any correspondence about this protocol to OPRS at 203 AOB, M/C 672.

Sincerely,

Sandra Costello
Assistant Director, IRB # 2
Office for the Protection of Research Subjects

Enclosures:

1. UIC Investigator Responsibilities, Protection of Human Research Subjects
2. Data Security Enclosure

cc: John Fudacz, Assistant Dean LAS, M/C 228
Approval Notice
Continuing Review

September 27, 2013

Steven Rivkin
Economics
601 S Morgan Street 729UH
M/C 144
Chicago, IL 60612
Phone: (312) 413-2368 / Fax: (312) 996-3344

RE: Protocol # 2012-0744
"A Study of the Dynamics of the Market for Principals and the Distribution of Principal Effectiveness in the Chicago Public Schools"

Dear Dr. Rivkin:

Your Continuing Review was reviewed and approved by the Expedited review process on September 24, 2013. You may now continue your research.

Please note the following information about your approved research protocol:

Please note that at the next Continuing Review please provide the number of people in the data set on page 6, Section VIII (Subject Enrollment and Demographics) of the application.

Protocol Approval Period: September 27, 2013 - September 27, 2014
Approved Subject Enrollment #: 5000000
Additional Determinations for Research Involving Minors: The Board determined that this research satisfies 45CFR46.404, research not involving greater than minimal risk.
Performance Sites: UIC, University of Chicago Consortium on Chicago School Research
Sponsor: Smith Richardson Foundation
PAF#: 2013-03396
Grant/Contract No: 2013-9106
Grant/Contract Title: Not available
Research Protocol:

a) A Study of the Dynamics of the Market for Principals and the Distribution of Principal Effectiveness in the Chicago Public Schools; Version 1; 09/04/2012

Phone: 312-996-1711  http://www.uic.edu/depts/ovcr/ocra/  FAX: 312-413-2929
Recruitment Material:
  a) No recruitment materials will be used - data will be accessed under an agreement with the University of Chicago Consortium on Chicago School Research

Informed Consent:
  a) A waiver of consent/permission/assent has been granted under 45 CFR 46.116(d) for access to Chicago Public Schools data under an agreement with the University of Chicago Consortium on Research on Chicago School Research

Your research meets the criteria for expedited review as defined in 45 CFR 46.110(b)(1) under the following specific categories:

(5) Research involving materials (data, documents, records, or specimens) that have been collected, or will be collected solely for nonresearch purposes (such as medical treatment or diagnosis),
(7) Research on individual or group characteristics or behavior (including but not limited to research on perception, cognition, motivation, identity, language, communication, cultural beliefs or practices and social behavior) or research employing survey, interview, oral history, focus group, program evaluation, human factors evaluation, or quality assurance methodologies.

Please note the Review History of this submission:

<table>
<thead>
<tr>
<th>Receipt Date</th>
<th>Submission Type</th>
<th>Review Process</th>
<th>Review Date</th>
<th>Review Action</th>
</tr>
</thead>
<tbody>
<tr>
<td>09/18/2013</td>
<td>Continuing Review</td>
<td>Expedited</td>
<td>09/24/2013</td>
<td>Approved</td>
</tr>
</tbody>
</table>

Please remember to:

➔ Use your research protocol number (2012-0744) on any documents or correspondence with the IRB concerning your research protocol.

➔ Review and comply with all requirements on the enclosure,

"UIC Investigator Responsibilities, Protection of Human Research Subjects"
(http://uic.edu/depts/ovcr/research/protocolreview/irb/policies/0924.pdf)

Please note that the UIC IRB has the right to seek additional information, require further modifications, or monitor the conduct of your research and the consent process.

Please be aware that if the scope of work in the grant/project changes, the protocol must be amended and approved by the UIC IRB before the initiation of the change.

We wish you the best as you conduct your research. If you have any questions or need further help, please contact OPRS at (312) 996-1711 or me at (312) 355-2764. Please send any correspondence about this protocol to OPRS at 203 AOB, M/C 672.
Sincerely,

Betty Mayberry, B.S.
IRB Coordinator, IRB # 2
Office for the Protection of Research Subjects

cc: John Fudacz, Assistant Dean LAS, M/C 228
OVCN Administration, M/C 672
Office for the Protection of Research Subjects (OPRS)
Office of the Vice Chancellor for Research (MC 672)
R1 Administrative Office Building
1122 West Polk Street
Chicago, Illinois 60612-7227

Approval Notice
Amendment to Research Protocol and/or Consent Document – Expedited Review
UIC Amendment # 1

September 27, 2013

Steven Rivkin
Economics
601 S Morgan Street 729UH
M/C 144
Chicago, IL 60612
Phone: (312) 413-2368 / Fax: (312) 996-3344

RB: Protocol # 2012-0744
“A Study of the Dynamics of the Market for Principals and the Distribution of Principal Effectiveness in the Chicago Public Schools”

Dear Dr. Rivkin:

Members of Institutional Review Board (IRB) #2 have reviewed this amendment to your research under expedited procedures for minor changes to previously approved research allowed by Federal regulations [45 CFR 46.110(b)(2)]. The amendment to your research was determined to be acceptable and may now be implemented.

Please note the following information about your approved amendment:

Amendment Approval Date: September 24, 2013

Amendment:
Summary: UIC Amendment #1 dated and received September 18, 2013, is an investigator-initiated amendment to add Smith Richardson Foundation as a funding source. The Investigator’s research was approved on September 27, 2012 and the grant was applied for in December 2012. The approval was received March 8, 2013 and will expire March 31, 2015. An Appendix Z was included in the submission.

Approved Subject Enrollment #: 500000
Performance Sites: UIC, University of Chicago Consortium on Chicago
School Research
Sponsor: Smith Richardson Foundation
PA#: 2013-03396
Grant/Contract No: 2013-0106
Grant/Contract Title: Not available

Phone: 312-996-1711  http://www.uic.edu/cepa/ovce/opsf  FAX: 312-413-2929
Please note the Review History of this submission:

<table>
<thead>
<tr>
<th>Receipt Date</th>
<th>Submission Type</th>
<th>Review Process</th>
<th>Review Date</th>
<th>Review Action</th>
</tr>
</thead>
<tbody>
<tr>
<td>09/18/2013</td>
<td>Amendment</td>
<td>Expedited</td>
<td>09/24/2013</td>
<td>Approved</td>
</tr>
</tbody>
</table>

Please be sure to:

➔ Use your research protocol number (2012-0744) on any documents or correspondence with the IRB concerning your research protocol.

➔ Review and comply with all requirements on the enclosure,
   "UIC Investigator Responsibilities, Protection of Human Research Subjects"
   (http://dgser.uchicago.edu/depts/oovr/research/proojectreview/irb/policies/0924.pdf)

Please note that the UIC IRB #2 has the right to seek additional information, or monitor the conduct of your research and the consent process.

Please be aware that if the scope of work in the grant/project changes, the protocol must be amended and approved by the UIC IRB before the initiation of the change.

We wish you the best as you conduct your research. If you have any questions or need further help, please contact the OPRS at (312) 996-1711 or me at (312) 355-2764. Please send any correspondence about this protocol to OPRS at 203 AOB, M/C 672.

Sincerely,

[Signature]
Betty Mayberry, B.S.
IRB Coordinator, IRB #2
Office for the Protection of Research Subjects

cc: John Fudacz, Assistant Dean LAS, M/C 228
October 7, 2014

Steven Rivkin  
Economics  
601 S Morgan Street 729UH  
M/C 144  
Chicago, IL 60612  
Phone: (312) 413-2368 / Fax: (312) 996-3344

RE: Protocol # 2012-0744  
“A Study of the Dynamics of the Market for Principals and the Distribution of Principal Effectiveness in the Chicago Public Schools”

Dear Dr. Rivkin:

Your Continuing Review was reviewed and approved by the Expedited review process on October 7, 2014. You may now continue your research.

Please note that this research did not have Institutional Review Board (IRB) approval beginning on 9/27/2014; approval re-commenced on 10/7/2014. Any research activities conducted during this time were done without IRB approval and were not compliant with UIC’s human subject protection policies, The Belmont Report, UIC’s Assurance awarded by the Office for Human Research Protection (OHRP) at EHSS, and with the federal regulations for the protection of human research subjects, 45 CFR 46.

Please note the following information about your approved research protocol:

| Approved Subject Enrollment #: | 5,000,000 (data analyses from 2,100,000 cases) |

The Board determined that this research satisfies 45 CFR 46.404, research not involving greater than minimal risk.

| Performance Sites: | UIC, University of Chicago Consortium on Chicago School Research |
| Sponsor: | Smith Richardson Foundation |
| PAP#: | 2013-63396 |
| Grant/Contract No: | 2013-9106 |
| Grant/Contract Title: | Not available |

Phone: 312-996-1711  
http://www.uic.edu/depts/ovcr/irb/  
FAX: 312-413-2929
Research Protocol:
   a) A Study of the Dynamics of the Market for Principals and the Distribution of Principal Effectiveness in the Chicago Public Schools; Version 1; 09/04/2012

Recruitment Material:
   a) No recruitment materials will be used - data will be accessed under an agreement with the University of Chicago Consortium on Chicago School Research

Informed Consent:
   a) A waiver of consent/permission/assent has been granted under 45 CFR 46.116(d) for access to Chicago Public Schools data under an agreement with the University of Chicago Consortium on Research on Chicago School Research

Your research meets the criteria for expedited review as defined in 45 CFR 46.110(b)(1) under the following specific categories:

(5) Research involving materials (data, documents, records, or specimens) that have been collected, or will be collected solely for nonresearch purposes (such as medical treatment or diagnosis), (7) Research on individual or group characteristics or behavior (including but not limited to research on perception, cognition, motivation, identity, language, communication, cultural beliefs or practices and social behavior) or research employing survey, interview, oral history, focus group, program evaluation, human factors evaluation, or quality assurance methodologies.

Please note the Review History of this submission:

<table>
<thead>
<tr>
<th>Receipt Date</th>
<th>Submission Type</th>
<th>Review Process</th>
<th>Review Date</th>
<th>Review Action</th>
</tr>
</thead>
<tbody>
<tr>
<td>10/01/2014</td>
<td>Continuing Review</td>
<td>Expedited</td>
<td>10/07/2014</td>
<td>Approved</td>
</tr>
</tbody>
</table>

Please remember to:

→ Use your research protocol number (2012-0744) on any documents or correspondence with the IRB concerning your research protocol.

→ Review and comply with all requirements on the OPRS website under:
   "UIC Investigator Responsibilities, Protection of Human Research Subjects" (http://irbinfo.uic.edu/dept/over/research/protocolreview/irb/policies/0924.pdf)

Please note that the UIC IRB has the right to seek additional information, require further modifications, or monitor the conduct of your research and the consent process.

Please be aware that if the scope of work in the grant/project changes, the protocol must be amended and approved by the UIC IRB before the initiation of the change.

We wish you the best as you conduct your research. If you have any questions or need further help, please contact OPRS at (312) 996-1711 or me at (312) 996-9299. Please send any correspondence about this protocol to OPRS at 203 AOB, M/C 672.
Sincerely,

Anna Bernadska, M.A.
IRB Coordinator, IRB # 2
Office for the Protection of Research Subjects

Enclosure: None

cc: John Fudacz, Assistant Dean, LAS, M/C 228
OVCRA dministration, M/C 672
Approval Notice
Continuing Review

October 2, 2015

Steven Rivkin
Economics
601 S Morgan Street 729UH
M/C 144
Chicago, IL 60612
Phone: (312) 413-2368 / Fax: (312) 996-3344

RE: Protocol # 2012-0744
"A Study of the Dynamics of the Market for Principals and the Distribution of Principal Effectiveness in the Chicago Public Schools"

Dear Dr. Rivkin:

Your Continuing Review was reviewed and approved by the Expedited review process on September 24, 2015. You may now continue your research.

Please note the following information about your approved research protocol:

Protocol Approval Period: October 7, 2015 - October 6, 2016
Approved Subject Enrollment #: 5,000,000 (data analysis from 2,100,000 cases)
Additional Determinations for Research
Involving Minors: 
The Board determined that this research satisfies 45CFR46.404, research not involving greater than minimal risk.

Performance Sites:
UIUC, University of Chicago Consortium on Chicago School Research

Sponsor:
Smith Richardson Foundation

PAF#: 2013-03396

Grant/Contract No: 2013-9106
Grant/Contract Title: Not available

Research Protocol:
a) A Study of the Dynamics of the Market for Principals and the Distribution of Principal Effectiveness in the Chicago Public Schools; Version 1.1, 09/16/2015

Recruitment Materials:
a) No recruitment materials will be used - data will be accessed under an agreement with the University of Chicago Consortium on Chicago School Research

Phone: 312-996-1711 http://www.uic.edu/depts/eovr/cprs/ FAX: 312-413-2929
Informed Consent:

a) A waiver of consent/permission/assent has been granted under 45 CFR 46.116(d) for access to Chicago Public Schools data under an agreement with the University of Chicago Consortium on Research on Chicago School Research.

Your research meets the criteria for expedited review as defined in 45 CFR 46.110(b)(1) under the following specific categories:

(5) Research involving materials (data, documents, records, or specimens) that have been collected, or will be collected solely for non-research purposes (such as medical treatment or diagnosis).

(7) Research on individual or group characteristics or behavior (including but not limited to research on perception, cognition, motivation, identity, language, communication, cultural beliefs or practices and social behavior) or research employing survey, interview, oral history, focus group, program evaluation, human factors evaluation, or quality assurance methodologies.

Please note the Review History of this submission:

<table>
<thead>
<tr>
<th>Receipt Date</th>
<th>Submission Type</th>
<th>Review Process</th>
<th>Review Date</th>
<th>Review Action</th>
</tr>
</thead>
<tbody>
<tr>
<td>09/18/2015</td>
<td>Continuing Review</td>
<td>Expedited</td>
<td>09/24/2015</td>
<td>Approved</td>
</tr>
</tbody>
</table>

Please remember to:

→ Use your research protocol number (2012-0744) on any documents or correspondence with the IRB concerning your research protocol.

→ Review and comply with all requirements on the OPRS website under: "UIC Investigator Responsibilities, Protection of Human Research Subjects" (http://higer.uic.edu/depts/over/research/protocolreview/irb/policies/0921.pdf)

Please note that the UIC IRB has the right to seek additional information, require further modifications, or monitor the conduct of your research and the consent process.

Please be aware that if the scope of work in the grant/project changes, the protocol must be amended and approved by the UIC IRB before the initiation of the change.

We wish you the best as you conduct your research. If you have any questions or need further help, please contact OPRS at (312) 996-1711 or me at (312) 996-9299. Please send any correspondence about this protocol to OPRS at 203 AOB, M/C 572.

Sincerely,

Anna Bernadaska, M.A.
IRB Coordinator, IRB #2
Office for the Protection of Research Subjects

Enclosure: None

cc: Steven G. Rivkin, Economics, M/C 144
5.2 Evidence of Copyright

The following is the title page of the NBER working paper which is an earlier version of chapter 1 of my dissertation. The page notes that authors retain the copyright of NBER working papers.

This research was conducted in collaboration with the University of Chicago Consortium on School Research. We thank Eva Halacheva for excellent research assistance and Abhijit Singh and seminar participants at Chicago Consortium for School Research, CES-Ifo and UIC for valuable comments. This research is supported by a grant from the Smith-Richardson foundation. The views expressed herein are those of the authors and do not necessarily reflect the views of the National Bureau of Economic Research.

NBER working papers are circulated for discussion and comment purposes. They have not been peer-reviewed or been subject to the review by the NBER Board of Directors that accompanies official NBER publications.

© 2016 by Derek Laing, Steven G. Rivkin, Jeffrey C. Schiman, and Jason Ward. All rights reserved. Short sections of text, not to exceed two paragraphs, may be quoted without explicit permission provided that full credit, including © notice, is given to the source.
Table 1. Average Voting Intensity in LSC Election Cycles by Quartile of Socioeconomic Status

<table>
<thead>
<tr>
<th>Quartile of Socioeconomic Status</th>
<th>Average Voting Intensity</th>
</tr>
</thead>
<tbody>
<tr>
<td>1\textsuperscript{st} (Lowest)</td>
<td>0.84</td>
</tr>
<tr>
<td>2\textsuperscript{nd}</td>
<td>0.82</td>
</tr>
<tr>
<td>3\textsuperscript{rd}</td>
<td>0.92</td>
</tr>
<tr>
<td>4\textsuperscript{th} (Highest)</td>
<td>1.60</td>
</tr>
</tbody>
</table>

Note. The quartile of socioeconomic status is based on Census measures of education levels and share of adults employed in managerial positions in the census block group where each student lives. Prior to 2010, the measure is based on the 2000 Census while from 2010 on it is based on the 2010 ACS. Higher quartiles of socioeconomic status indicate higher status. Voting intensity is calculated as the number of votes cast for parents running for an LSC seat divided by total school enrollment. Therefore, voting intensity may exceed 1.
Table 2. The Distribution of Principal Transitions Experienced by Schools

<table>
<thead>
<tr>
<th>Number of Principal Transitions</th>
<th>0</th>
<th>1</th>
<th>2</th>
<th>3</th>
<th>4+</th>
<th>Total</th>
</tr>
</thead>
<tbody>
<tr>
<td>Share of schools (observed in any years)</td>
<td>14.63</td>
<td>18.75</td>
<td>25.71</td>
<td>21.59</td>
<td>19.31</td>
<td>100%</td>
</tr>
<tr>
<td>Share of schools (observed in all years)</td>
<td>0.68</td>
<td>20.95</td>
<td>32.43</td>
<td>26.01</td>
<td>19.94</td>
<td>100%</td>
</tr>
</tbody>
</table>

Note. The estimates are based on a school-level sample of all available schools (704 in total) between 1993-94 and 2013-14 schools years.
Table 3. Estimates of the Standard Deviation of Principal Effectiveness Based on Regressions with Principal by Year Fixed Effects

<p>| | | |</p>
<table>
<thead>
<tr>
<th></th>
<th></th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>σ</strong> of average effectiveness during entire principal spell</td>
<td>0.110</td>
<td>0.078</td>
</tr>
<tr>
<td><strong>σ</strong> of average effectiveness dropping last and first years of spell</td>
<td>0.098</td>
<td>0.065</td>
</tr>
</tbody>
</table>

Removes school average N Y

Notes. All regressions control for student race, sex, special education status, and socioeconomic status of the student’s block group, whether the student is new to the school, school-grade averages of these characteristics, share of grade repeaters, school enrollment, and year-by-grade fixed effects.
Table 4. Average Principal Value Added, by Teacher Responses to Survey Questions on the Principal

<table>
<thead>
<tr>
<th>Teacher Response</th>
<th>Strongly Disagree</th>
<th>Disagree</th>
<th>Agree</th>
<th>Strongly Agree</th>
</tr>
</thead>
<tbody>
<tr>
<td>Makes clear expectations</td>
<td>-0.025</td>
<td>-0.017</td>
<td>-0.008</td>
<td>0.002</td>
</tr>
<tr>
<td>Communicates clear vision</td>
<td>-0.025</td>
<td>-0.018</td>
<td>-0.006</td>
<td>0.002</td>
</tr>
<tr>
<td>Tracks student progress</td>
<td>-0.023</td>
<td>-0.015</td>
<td>-0.007</td>
<td>0.003</td>
</tr>
</tbody>
</table>

Notes. Here we plot average value added by teacher’s responses to a series of questions about their principal, asked in 11 of the years between 1994 and 2013. Teachers are asked if the principal “makes clear to the staff his or her expectations for meeting instructional goals”; (2) “communicates a clear vision for our school”; and (3) “carefully tracks student academic progress.”
Table 5. Relationship between Teacher Ratings and Estimated Principal Value Added

<table>
<thead>
<tr>
<th></th>
<th>(1)</th>
<th>(2)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Teacher’s principal rating index</td>
<td>0.017***</td>
<td>0.013***</td>
</tr>
<tr>
<td></td>
<td>(0.002)</td>
<td>(0.002)</td>
</tr>
<tr>
<td>School fixed effect</td>
<td>N</td>
<td>Y</td>
</tr>
</tbody>
</table>

Notes. We regress estimated principal value added on an index of teacher’s rating of their principal. The index is based on responses to the three questions in Table 4. Each regression controls for school averages of student race, sex, special education, enrollment, share new students, parental SES, and year effects. Both regressions are based on 4,004 school-year level observations. Standard errors clustered by school are in parentheses. * p<0.10, ** p<0.05, *** p<0.001
Table 6. School Fixed Effect Estimates of the Variance in Principal Effectiveness

<table>
<thead>
<tr>
<th></th>
<th>(1)</th>
<th>(2)</th>
<th>(3)</th>
<th>(4)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Coefficient on principal transition indicator</td>
<td>0.0034**</td>
<td>0.0015</td>
<td>0.0037**</td>
<td>0.0029</td>
</tr>
<tr>
<td></td>
<td>(0.0015)</td>
<td>(0.0022)</td>
<td>(0.0019)</td>
<td>(0.0019)</td>
</tr>
<tr>
<td>Standard deviation</td>
<td>0.0414</td>
<td>0.0276</td>
<td>0.0429</td>
<td>0.0383</td>
</tr>
<tr>
<td>Exclude last and first years of spell</td>
<td>N</td>
<td>Y</td>
<td>Y</td>
<td>Y</td>
</tr>
<tr>
<td>Number of years between observations within spells</td>
<td>n.a.</td>
<td>1</td>
<td>2</td>
<td>3</td>
</tr>
<tr>
<td>Number of School-Year Observations</td>
<td>8,514</td>
<td>6,139</td>
<td>5,062</td>
<td>4,141</td>
</tr>
</tbody>
</table>

Notes. The regressions correspond to the approach described in section 1.3.2. Principal Turnover Based Estimates. The outcome is the squared difference in test scores between $t$ and $t-n$, where $n$ is the number of years between observations, and the principal transition variable is an indicator equal to one if there is a new principal in year $t$. The regressions also control for the squared difference in student race, sex, special education, and mobility, school enrollment, and parental SES. Standard errors clustered by school are in parentheses. * p<0.10, ** p<0.05, *** p<0.001
Table 7. Distribution of Voting Intensity Conditional on Parental SES

<table>
<thead>
<tr>
<th>Quartiles of Parental SES</th>
<th>Quartiles of Voting Intensity</th>
<th>Observations</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>1&lt;sup&gt;st&lt;/sup&gt; (Lowest)</td>
<td>2&lt;sup&gt;nd&lt;/sup&gt;</td>
</tr>
<tr>
<td>1&lt;sup&gt;st&lt;/sup&gt; (Lowest)</td>
<td>0.32</td>
<td>0.25</td>
</tr>
<tr>
<td>2&lt;sup&gt;nd&lt;/sup&gt;</td>
<td>0.23</td>
<td>0.29</td>
</tr>
<tr>
<td>3&lt;sup&gt;rd&lt;/sup&gt;</td>
<td>0.26</td>
<td>0.29</td>
</tr>
<tr>
<td>4&lt;sup&gt;th&lt;/sup&gt; (Highest)</td>
<td>0.18</td>
<td>0.17</td>
</tr>
<tr>
<td>Observations</td>
<td>314</td>
<td>317</td>
</tr>
</tbody>
</table>
Table 8. The Effects of Parental SES and Voting Intensity in LSC Elections on the Change in Principal Effectiveness

<table>
<thead>
<tr>
<th></th>
<th>(1)</th>
<th>(2)</th>
<th>(3)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Parental SES</td>
<td>0.0074*</td>
<td>0.0013</td>
<td>0.0016</td>
</tr>
<tr>
<td></td>
<td>(0.0043)</td>
<td>(0.0047)</td>
<td>(0.0048)</td>
</tr>
<tr>
<td>Voting intensity</td>
<td>0.0067*</td>
<td>0.0024</td>
<td>0.0017</td>
</tr>
<tr>
<td></td>
<td>(0.0038)</td>
<td>(0.0045)</td>
<td>(0.0045)</td>
</tr>
<tr>
<td>Prior contract VA</td>
<td>0.309***</td>
<td>0.308***</td>
<td>0.280***</td>
</tr>
<tr>
<td></td>
<td>(0.046)</td>
<td>(0.046)</td>
<td>(0.060)</td>
</tr>
<tr>
<td>Voting intensity*prior contract VA</td>
<td>-</td>
<td>-</td>
<td>0.0249</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td>(0.0341)</td>
</tr>
<tr>
<td>Parental SES*voting intensity</td>
<td>0.0051**</td>
<td>0.0047**</td>
<td></td>
</tr>
<tr>
<td></td>
<td>(0.0024)</td>
<td>(0.0024)</td>
<td></td>
</tr>
<tr>
<td>Sample size</td>
<td>767</td>
<td>767</td>
<td>767</td>
</tr>
</tbody>
</table>

Notes. We measure principal effectiveness as the change in value added from $t+1$ to $t+2$ from the first contract to the second contract. LSC managerial capacity is measured by parental SES, which captures the share of adults working managerial jobs in each student’s home census block. The strength of LSC incentives is measured by the voting intensity during the election (i.e. total number of votes cast divided by school enrollment). Both capacity and incentives are measured during the end of the previous contract. Each regression controls for the principal effectiveness at the same point during the previous contract, the change in school average demographics and program characteristics, and year fixed effects. Standard errors clustered by school are in parentheses. * $p<0.10$, ** $p<0.05$, *** $p<0.001$
<table>
<thead>
<tr>
<th></th>
<th>Relative to Continuation</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>(1)</td>
</tr>
<tr>
<td><strong>Left Mid or End of Contract</strong></td>
<td></td>
</tr>
<tr>
<td>Value Added</td>
<td>-3.619***</td>
</tr>
<tr>
<td></td>
<td>(0.795)</td>
</tr>
<tr>
<td>Voting Intensity</td>
<td>0.170***</td>
</tr>
<tr>
<td></td>
<td>(0.057)</td>
</tr>
<tr>
<td>Parental SES</td>
<td>0.023</td>
</tr>
<tr>
<td></td>
<td>(0.080)</td>
</tr>
<tr>
<td>Value Added*Voting Intensity</td>
<td>0.982**</td>
</tr>
<tr>
<td></td>
<td>(0.461)</td>
</tr>
<tr>
<td>Value Added*Parental SES</td>
<td>-1.021</td>
</tr>
<tr>
<td></td>
<td>(0.745)</td>
</tr>
<tr>
<td>Voting Intensity*Parental SES</td>
<td>-</td>
</tr>
<tr>
<td></td>
<td></td>
</tr>
<tr>
<td>Value Added<em>Voting Intensity</em>Parental SES</td>
<td>-</td>
</tr>
<tr>
<td></td>
<td></td>
</tr>
<tr>
<td><strong>Left by CEO Removal</strong></td>
<td></td>
</tr>
<tr>
<td>Value Added</td>
<td>-4.025*</td>
</tr>
<tr>
<td></td>
<td>(2.372)</td>
</tr>
<tr>
<td>Voting Intensity</td>
<td>-1.420</td>
</tr>
<tr>
<td></td>
<td>(1.182)</td>
</tr>
<tr>
<td>Parental SES</td>
<td>-0.295</td>
</tr>
<tr>
<td></td>
<td>(0.641)</td>
</tr>
<tr>
<td>Value Added*Voting Intensity</td>
<td>-4.792</td>
</tr>
<tr>
<td></td>
<td>(3.691)</td>
</tr>
<tr>
<td>Value Added*Parental SES</td>
<td>-3.100</td>
</tr>
<tr>
<td></td>
<td>(3.936)</td>
</tr>
<tr>
<td>Voting Intensity*Parental SES</td>
<td>-</td>
</tr>
<tr>
<td></td>
<td></td>
</tr>
<tr>
<td>Value Added<em>Voting Intensity</em>Parental SES</td>
<td>-</td>
</tr>
<tr>
<td></td>
<td></td>
</tr>
<tr>
<td><strong>Sample Size</strong></td>
<td>1,104</td>
</tr>
</tbody>
</table>

Notes. The outcome takes on three possible values related to how the principal’s contract ends. We compare principals who leave mid or end contract and those who leave by CEO removal to those who have their contract renewed (the base outcome). Each regression also controls for school averages of student race, sex, special education, enrollment, and share new students. Standard errors clustered by school are in parentheses. * p<0.10, ** p<0.05, *** p<0.001
Table 10. Criteria For Principal Evaluation

<table>
<thead>
<tr>
<th>Principal criteria</th>
<th>2011-12 School Year</th>
<th>2012-13 School Year</th>
</tr>
</thead>
<tbody>
<tr>
<td>“must rank in the”</td>
<td>75th percentile or above in reading growth (measured by value added on ISAT reading).”</td>
<td>90th percentile or above in the national distribution of the fall-to-spring growth on NWEA Reading.”</td>
</tr>
<tr>
<td></td>
<td>75th percentile or above in math growth (measured by value added on ISAT math).”</td>
<td>90th percentile or above in the national distribution of the fall-to-spring growth on NWEA Math.”</td>
</tr>
<tr>
<td></td>
<td>75th percentile or above in the growth of the percent of students meeting college readiness benchmarks on Grade 8 EXPLORE for both reading and math. Or, they must maintain 90% or more of students meeting college readiness benchmarks on Grade 8 EXPLORE for both reading and math.”</td>
<td>70th percentile or above in the national distribution of growth for priority groups (ELL, IEP, Black, or Hispanic).”</td>
</tr>
<tr>
<td></td>
<td>75th percentile or above in reading and math for more than 50% of the school’s subgroups in the percent of students improving by a performance level or staying at exceeds on ISAT reading and math.”</td>
<td>70% of students achieving individual growth targets.”</td>
</tr>
</tbody>
</table>

Notes.
### Table 11. OLS Estimates of the Relationship Between Award Receipt and Turnover

<table>
<thead>
<tr>
<th></th>
<th>A. Full sample</th>
<th>B. Bandwidth ± 10</th>
<th>C. Bandwidth ± 5</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Award winner</td>
<td>Award winner</td>
<td>Award winner</td>
</tr>
<tr>
<td></td>
<td>-0.103 -0.085 -0.063</td>
<td>0.036 0.047 0.055</td>
<td>0.101 0.116 0.123</td>
</tr>
<tr>
<td></td>
<td>(0.028) (0.030) (0.028)</td>
<td>(0.056) (0.057) (0.058)</td>
<td>(0.071) (0.799) (0.055)</td>
</tr>
<tr>
<td>Sample size</td>
<td>728 728 728</td>
<td>201 201 201</td>
<td>90 90 90</td>
</tr>
</tbody>
</table>

Notes. Each estimate is from a separate regression of the form \( y_i = \alpha + \text{winner}\beta + X\delta + \epsilon \). The outcome in each regression is an indicator for principal turnover between \( t \) and \( t + 1 \). Controls include baseline (i.e. year \( t \)) controls student racial composition, share female students, share new students, share of grade repeaters, student socioeconomic status, student poverty, enrollment, and principal salary. In Panel A, the sample includes all principals. In panel B and Panel C, the sample is narrowed to principals near the award threshold, similar to the RD analysis. Standard errors clustered by the running variable are in parentheses.
Table 12. Reduced Form Regression Discontinuity Estimates of the Relationship between Turnover (t to t + 1) and Predicted Award Receipt

<table>
<thead>
<tr>
<th></th>
<th>±5</th>
<th>±10</th>
<th>±15</th>
<th>±20</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>(n = 90)</td>
<td>(n = 201)</td>
<td>(n = 296)</td>
<td>(n = 341)</td>
</tr>
<tr>
<td><strong>A. Without Controls</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Passed award threshold</td>
<td>0.325***</td>
<td>0.158**</td>
<td>0.142**</td>
<td>0.094</td>
</tr>
<tr>
<td></td>
<td>(0.039)</td>
<td>(0.068)</td>
<td>(0.063)</td>
<td>(0.064)</td>
</tr>
<tr>
<td><strong>B. With Controls</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Passed award threshold</td>
<td>0.239***</td>
<td>0.126*</td>
<td>0.164**</td>
<td>0.142**</td>
</tr>
<tr>
<td></td>
<td>(0.079)</td>
<td>(0.068)</td>
<td>(0.065)</td>
<td>(0.063)</td>
</tr>
<tr>
<td>Control group mean</td>
<td>0.094</td>
<td>0.121</td>
<td>0.134</td>
<td>0.131</td>
</tr>
</tbody>
</table>

Notes. Each estimate is from a separate RD regression of the form \( y_i = f(\tau_i) + Above_i \delta + f(\tau_i) \cdot Above_i + \epsilon_i \). The outcome in each regression is an indicator for principal turnover between \( t \) and \( t + 1 \). Panel B includes baseline (i.e. year \( t \)) controls for program year, student racial composition, share female students, share new students, share of grade repeaters, student socioeconomic status, student poverty, enrollment, and principal salary. Standard errors clustered by the running variable are in parentheses.
Table 13. 2SLS Estimates Weighted by the Probability of Treatment

<table>
<thead>
<tr>
<th></th>
<th>±5</th>
<th>±10</th>
<th>±15</th>
<th>±20</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>(n = 90)</td>
<td>(n = 201)</td>
<td>(n = 296)</td>
<td>(n = 341)</td>
</tr>
<tr>
<td><strong>A. Without Controls</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Winner</td>
<td>0.679**</td>
<td>0.312**</td>
<td>0.262</td>
<td>0.166</td>
</tr>
<tr>
<td></td>
<td>(0.124)</td>
<td>(0.126)</td>
<td>(0.112)</td>
<td>(0.114)</td>
</tr>
<tr>
<td>First-stage estimate</td>
<td>0.478</td>
<td>0.507</td>
<td>0.542</td>
<td>0.564</td>
</tr>
<tr>
<td></td>
<td>(0.090)</td>
<td>(0.073)</td>
<td>(0.066)</td>
<td>(0.063)</td>
</tr>
<tr>
<td><strong>B. With Controls</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Winner</td>
<td>0.412**</td>
<td>0.203*</td>
<td>0.309**</td>
<td>0.267**</td>
</tr>
<tr>
<td></td>
<td>(0.146)</td>
<td>(0.111)</td>
<td>(0.128)</td>
<td>(0.126)</td>
</tr>
<tr>
<td>First-stage estimate</td>
<td>0.580</td>
<td>0.621</td>
<td>0.531</td>
<td>0.532</td>
</tr>
<tr>
<td></td>
<td>(0.071)</td>
<td>(0.058)</td>
<td>(0.068)</td>
<td>(0.066)</td>
</tr>
<tr>
<td>Control group mean</td>
<td>0.094</td>
<td>0.121</td>
<td>0.134</td>
<td>0.131</td>
</tr>
</tbody>
</table>

Notes. The 2SLS estimates equal the reduced for estimates (Table 12) divided by the corresponding first-stage estimates. Standard errors clustered by the running variable are in parentheses.
Table 14. Change in Job Characteristics for Principals between $t$ and $t+1$ for those Remaining in CPS

<table>
<thead>
<tr>
<th></th>
<th>Change ($t$ to $t + 1$) in</th>
<th></th>
<th></th>
<th>Test Scores</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Salary</td>
<td>Enrollment</td>
<td>Poverty</td>
<td>Scores</td>
</tr>
<tr>
<td>A. Bandwidth ±10</td>
<td>204</td>
<td>17</td>
<td>0.012</td>
<td>0.06</td>
</tr>
<tr>
<td>Passed award</td>
<td>(442)</td>
<td>(10.3)</td>
<td>(0.03)</td>
<td>(0.116)</td>
</tr>
</tbody>
</table>

Notes. Each estimate is from a separate regression. The outcome is the change in job characteristics between $t$ and $t + 1$ for those still serving as principal in the CPS. Standard errors clustered by the running variable are in parentheses.
Table 15. Distribution of performance

<table>
<thead>
<tr>
<th>Award year performance growth</th>
<th>Prior 4 years performance growth</th>
<th>1&lt;sup&gt;st&lt;/sup&gt; (Lowest)</th>
<th>2&lt;sup&gt;nd&lt;/sup&gt;</th>
<th>3&lt;sup&gt;rd&lt;/sup&gt;</th>
<th>4&lt;sup&gt;th&lt;/sup&gt; (Highest)</th>
<th>%</th>
<th>Size</th>
</tr>
</thead>
<tbody>
<tr>
<td>1&lt;sup&gt;st&lt;/sup&gt; (Lowest)</td>
<td></td>
<td>0.31</td>
<td>0.36</td>
<td>0.19</td>
<td>0.14</td>
<td>1.00</td>
<td>100</td>
</tr>
<tr>
<td>2&lt;sup&gt;nd&lt;/sup&gt;</td>
<td></td>
<td>0.30</td>
<td>0.22</td>
<td>0.29</td>
<td>0.18</td>
<td>1.00</td>
<td>92</td>
</tr>
<tr>
<td>3&lt;sup&gt;rd&lt;/sup&gt;</td>
<td></td>
<td>0.23</td>
<td>0.26</td>
<td>0.27</td>
<td>0.23</td>
<td>1.00</td>
<td>95</td>
</tr>
<tr>
<td>4&lt;sup&gt;th&lt;/sup&gt; (Highest)</td>
<td></td>
<td>0.12</td>
<td>0.13</td>
<td>0.29</td>
<td>0.46</td>
<td>1.00</td>
<td>92</td>
</tr>
<tr>
<td>Size</td>
<td></td>
<td>92</td>
<td>93</td>
<td>99</td>
<td>95</td>
<td>379</td>
<td></td>
</tr>
</tbody>
</table>

Notes. This table presents the distribution of award-year performance compared to performance during the prior 4 years. To be included in the sample, principals had to serve during the prior 4 years and have valid test scores or value added measures.
### Table 16. How the Turnover Generated by the Award Affects Schools

<table>
<thead>
<tr>
<th>Bandwidth ±</th>
<th>Passed award threshold</th>
<th>Value Added</th>
<th>Enrollment</th>
<th>Poverty</th>
</tr>
</thead>
<tbody>
<tr>
<td>±5</td>
<td></td>
<td>-0.225</td>
<td>25.8</td>
<td>-0.140</td>
</tr>
<tr>
<td></td>
<td></td>
<td>(0.138)</td>
<td>(9.6)</td>
<td>(0.043)</td>
</tr>
<tr>
<td></td>
<td>Control group mean</td>
<td>0.077</td>
<td>11.5</td>
<td>0.041</td>
</tr>
<tr>
<td></td>
<td>Sample size</td>
<td>21</td>
<td>63</td>
<td>63</td>
</tr>
<tr>
<td>±10</td>
<td></td>
<td>-0.165</td>
<td>7.7</td>
<td>-0.022</td>
</tr>
<tr>
<td></td>
<td></td>
<td>(0.072)</td>
<td>(13.2)</td>
<td>(0.050)</td>
</tr>
<tr>
<td></td>
<td>Control group mean</td>
<td>0.028</td>
<td>6.2</td>
<td>0.022</td>
</tr>
<tr>
<td></td>
<td>Sample size</td>
<td>50</td>
<td>163</td>
<td>163</td>
</tr>
<tr>
<td>±15</td>
<td></td>
<td>-0.089</td>
<td>7.9</td>
<td>-0.001</td>
</tr>
<tr>
<td></td>
<td></td>
<td>(0.063)</td>
<td>(12.5)</td>
<td>(0.047)</td>
</tr>
<tr>
<td></td>
<td>Control group mean</td>
<td>0.012</td>
<td>4.8</td>
<td>0.032</td>
</tr>
<tr>
<td></td>
<td>Sample size</td>
<td>71</td>
<td>233</td>
<td>233</td>
</tr>
<tr>
<td>±20</td>
<td></td>
<td>-0.064</td>
<td>6.9</td>
<td>0.026</td>
</tr>
<tr>
<td></td>
<td></td>
<td>(0.056)</td>
<td>(11.5)</td>
<td>(0.044)</td>
</tr>
<tr>
<td></td>
<td>Control group mean</td>
<td>0.005</td>
<td>2.1</td>
<td>0.047</td>
</tr>
<tr>
<td></td>
<td>Sample size</td>
<td>87</td>
<td>263</td>
<td>263</td>
</tr>
</tbody>
</table>

**Notes.** Each estimate is from a separate regression. The outcome equals the change in school characteristics between \( t \) and \( t + 1 \). Standard errors clustered by school are in parentheses.
Figure 1. Changes in the Probability of Award Receipt by Predicted Award Receipt

Notes. The figure presents the raw mean relationship between predicted and actual award receipt (i.e. the first-stage relationship). Each circle represents mean award receipt within bins of size 5. Those ranking at or above are predicted to win an award.
Figure 2. Probability of Overall Principal Transitions by Predicted Award Receipt

Notes. The figure presents the raw mean relationship between predicted award receipt and principal turnover (i.e. the reduced-form relationship). Each circle represents mean turnover within bins of size 5. The predicted award threshold is at 0.
Figure 3. Pattern of Turnover Prior to the Award Program

Notes. The figure presents the raw mean relationship between predicted rank and actual turnover. Each circle represents mean turnover within bins of size 5.
Appendix Figures and Tables

Appendix Figure 1. Density of the RD Running Variable

Notes.
## Appendix Table A1. Discontinuities in Observable Covariates

<table>
<thead>
<tr>
<th>Passed award threshold</th>
<th>Salary</th>
<th>SES</th>
<th>Poverty</th>
<th>Enrollment</th>
<th>% Black</th>
<th>% Latino</th>
<th>% Other race</th>
<th>% Special Ed.</th>
<th>% Grade repeaters</th>
<th>% Female</th>
<th>% new students</th>
<th>Max Rank</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>-0.044</td>
<td>-0.043</td>
<td>-0.049</td>
<td>-0.091</td>
<td>-0.007</td>
<td>0.031</td>
<td>-0.0003</td>
<td>-0.018</td>
<td>0.015</td>
<td>-2.717</td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>(3920)</td>
<td>(0.112)</td>
<td>(0.106)</td>
<td>(64)</td>
<td>(0.100)</td>
<td>(0.072)</td>
<td>(0.031)</td>
<td>(0.035)</td>
<td>(0.003)</td>
<td>(0.016)</td>
<td>(0.023)</td>
<td>(5.319)</td>
</tr>
<tr>
<td>Control group mean</td>
<td>131218</td>
<td>-0.074</td>
<td>0.109</td>
<td>486</td>
<td>0.424</td>
<td>0.363</td>
<td>0.071</td>
<td>0.132</td>
<td>0.012</td>
<td>0.493</td>
<td>0.115</td>
<td>56.2</td>
</tr>
</tbody>
</table>

*Notes.* Each estimate is from a separate RD regression where the indicated baseline covariate (i.e. in period $t - 1$) is the outcome. Each regression is estimated over a bandwidth of $±10$. Standard errors clustered by the running variable are in parentheses.
6. Vita

JEFFREY C. SCHIMAN
1 May 2016

EDUCATION
2016 Ph.D. in Economics, University of Illinois at Chicago
   Dissertation: “Local Versus Central School Control and the Distribution of
   Principal Effectiveness”
   Committee: Steven Rivkin (Chair), Ben Feigenberg, Robert Kaestner, Ben
   Ost, and Steve Tozer
2012 M.A. in Economics, University of Illinois at Chicago
2009 B.A. in Economics, Calvin College

FIELDS OF INTEREST
Economics of Education, Labor Economics, Public Economics, and Applied Econometrics

HONORS & AWARDS
2014–2015 The Dean’s Scholar Award, UIC, $56,748
2013, 2014 The Graduate Winifred Geldard Memorial Award, Dept. of Economics, UIC, $500
2014 Travel Award for ESTIMATE Econometrics Workshop at MSU, Dept. of
   Economics, UIC, $500

RESEARCH GRANTS
   of Principals to Student Outcomes and the Pattern of Principal Job
   Transitions.” Smith Richardson Foundation, $175,000

PEER-REVIEWED PUBLICATIONS
“Instruction Time, Classroom Quality, and Academic Achievement” (with Steven G. Rivkin),

“Grade-Specific Experience, Grade Reassignments, and Teacher Turnover” (with Ben Ost),
Economics of Education Review 46: 112-26, June 2015

WORKING PAPERS

“Decentralized Governance and the Quality of School Leadership” (with Derek Laing, Steven

“Dynamic Effects of Teacher Turnover on the Quality of Instruction in U.S. Urban Schools”
   (with Eric A. Hanushek and Steven G. Rivkin), Revise and Resubmit, Economics of Education
   Review, 12/2015

“Workload and Teacher Effort” (with Ben Ost), Revise and Resubmit, Economics of Education
   Review, 10/2015

“Comparing Standard Deviation Effects Across Contexts” (with Anuj Gangopadhyaya and
"Understanding the Divergent Trends in PISA Test Results for Poland and the Czech Republic" (with Mikolaj Herbst, Daniel Münich, and Steven G. Rivkin), 7/2012

RESEARCH IN PROGRESS
“Bilingual Education and the Short and Longer-Term Outcomes of English Language Learners”

“Academic and Longer-Term Outcomes for English Language Learners and Their Classmates in the Texas Public Schools” (with Eric A. Hanushek and Steven G. Rivkin)

“Principal Effectiveness and Longer-Term Outcomes for Students in the Chicago Public Schools” (with Steven G. Rivkin)

“Classroom Instruction Time, School-Year Length, and Academic Achievement”

“Neighborhood Violence and Academic Performance” (with Marcus Casey)

“Infant Health Shocks and Adult Wellbeing: Evidence from Wartime Britain” (with Robert Kaestner and Anthony Lo Sasso)

RESEARCH EXPERIENCE
2015 – Present Researcher, Texas Schools Project, UTD
2012 – 2015 Research Assistant to Professor Steven G. Rivkin, Dept. of Economics, UIC
2009 – 2010 Research Assistant, Center for Social Research, Calvin College

TEACHING EXPERIENCE
University of Illinois at Chicago
Summer 2014 Instructor, Principles of Microeconomics (Undergraduate)
Summer 2013 Instructor, Econometrics (Undergraduate)
Fall 2011 Teaching Assistant, Econometrics (Undergraduate)
Spring 2011 Teaching Assistant, Urban Economics (Undergraduate)

RECENT PROFESSIONAL PRESENTATIONS
2015 GVSU, Calvin College, APPAM Fall Research Conference, Hope College
2014 Calvin College, University of Chicago Consortium for Chicago School Research, APPAM Fall Research Conference, UIC Economics Research Lunch
2013 UIC Economics Research Lunch (x2), University of Chicago Consortium for Chicago School Research, UIC Economics Graduate Student Research Seminar

PROFESSIONAL ACTIVITIES
Referee: Journal of Policy Analysis and Management (2); Economics of Education Review (1)
Member: American Economic Association; Society of Labor Economists; Association of Public Policy Analysis and Management; Association of Education Finance & Policy