

# The limitations of using school league tables to inform school choice

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# Introduction

- Each year the government publishes schools' GCSE results and contextual value-added (CVA) performances in school league tables
  - They estimate value-added scores using multilevel models
- A principal justification for this is to inform parental choice of secondary schools
- A crucial limitation of these tables is that the most recent published information is based on a cohort of pupils who are 7 years ahead of the cohort of interest
- For choosing a school, it is the future performance of schools that is of interest
- The government make no adjustment for the statistical uncertainty that arises from making predictions into the future

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http://www.dcsf.gov.uk/cgi-bin/performance/tables/group\_07.pl?Mode=Z&Tyj

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**Achievement and attainment tables**

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### Secondary School achievement and attainment tables 2007

LA : Oxfordshire

Key Stage 2 to 4 Contextual Value Added

| GCSE and equivalent                                       | Key Stage 2 to 4 CVA  | Key Stage 3 to 4 CVA                               |       | Year on year comparisons  | Absence  | Background |
|---|---|--|-------|---|--|------------|
|   | Measure centred on 1000   | Limit of Key Stage 2 to 4 CVA Confidence Intervals |       | Coverage  | Number of qualifications   |            |
|   | CVA measure based on progress between Key Stage 2 and Key Stage 4 | Upper  | Lower | % of students at the end of Key Stage 4 included in CVA calculation | Average number of qualifications taken by students in Key Stage 2 to 4 CVA calculation |            |
| <a href="#">Banbury School</a>                            | 997.7   | 1005.8   | 989.7 | 97%   | 9.7  |            |
| <a href="#">Bartholomew School</a>                        | 1009.1  | 1018.4   | 999.8 | 97%   | 10.0   |            |
| <a href="#">Bicester Community College</a>                | 994.3   | 1002.8   | 985.7 | 92%   | 10.8   |            |
| <a href="#">Blessed George Napier Catholic School and</a> | 1002.1  | 1011.8   | 992.4 | 96%   | 10.2   |            |

Done

# Seven years out of date

- During October 2008 parents will choose which secondary schools to send their children to
- These pupils will start secondary schooling in September 2009 and will take their GCSE examinations in 2014
- When choosing their secondary schools, the most recent published information will be for the cohort of pupils who take their GCSEs in 2007
- These two cohorts are seven years apart

# Stability of school effects

- Previous literature has shown that whilst simple school averages are strongly correlated over time, value-added estimates of school effects are only moderately correlated
- Correlations of 0.5 - 0.6 for value-added estimates five years apart
- This limits the extent to which current school performance can be used as a guide to future performance

# Data

- National Pupil Database (NPD)
  - Census of all state school pupils in England
  - Pupils test scores data at ages 11 and 16
  - Same data as is used to produce government school league tables
- Pupil Level Annual School Census (PLASC)
  - Provides data on pupil background characteristics
  - These are included in the CVA model specification
- We use data on the cohort of pupils that took their GCSEs in 2007
- We analyse a 10% random sample of all English secondary schools
  - 274 schools, approximately 190 pupils per school

# Two-level multilevel model

- The traditional school effectiveness model is

$$y_{ij} = \beta_0 + \beta_1 x_{ij} + u_j + e_{ij}$$
$$u_j \sim N(0, \sigma_u^2), \quad e_{ij} \sim N(0, \sigma_e^2)$$

- $y_{ij}$  is the total GCSE score for pupil  $i$  in secondary school  $j$
- $x_{ij}$  is their achievement at age 11 intake
- $u_j$  is the value-added school effect for secondary school  $j$
- $e_{ij}$  is the pupil level random effect

# Predictor variables

- At the pupil level (level 1) we adjust for
  - Achievement at age 11
  - Month of birth
  - Gender
  - Free school meals
  - Special educational needs
  - English as an additional language
  - Ethnicity
  - Local neighbourhood deprivation
- We do not adjust for any school level (level 2) variables

# School effects for the 2007 cohort

- Posterior estimates of the school effects and their associated variances are given by

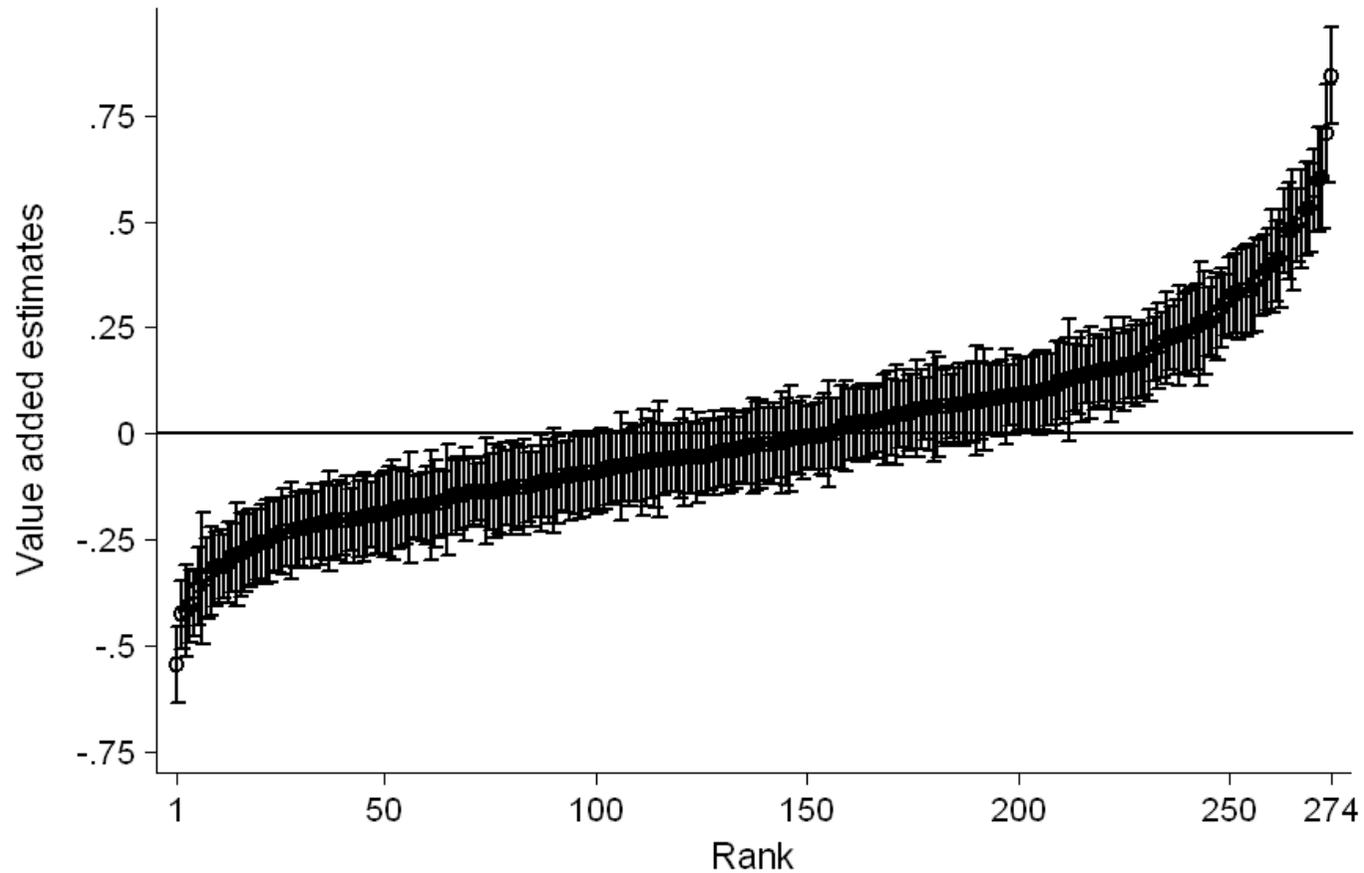
$$\hat{u}_j = \frac{n_j \sigma_u^2}{n_j \sigma_u^2 + \sigma_e^2} \tilde{y}_j, \quad \text{var}(\hat{u}_j - u_j) = \frac{\sigma_u^2 \sigma_e^2}{n_j \sigma_u^2 + \sigma_e^2}$$

- Assuming normality, standard 95% confidence intervals are calculated as

$$\hat{u}_j \pm 1.96 \sqrt{\text{var}(\hat{u}_j - u_j)}$$

- These school effects are published in the DCSF school league tables

# School effects for the 2007 cohort



~60% of schools are significantly different from the overall average

# School effects for the 2014 cohort

- The previous school effects allow us to make inferences about how schools performed for the cohort that took their GCSEs in 2007
- However, they do not allow us to make inferences about the likely performance of schools for future cohorts
- We want to know whether the same significant differences remain in 2014
- To do this, we need to adjust the estimates and standard errors of the 2007 school effects to reflect the additional uncertainty that arises from predicting into the future
- The bivariate response version of the school effectiveness model provides a way to do this

# Bivariate response model

- The traditional school effectiveness model for two cohorts of pupils is

$$y_{ij}^{(2007)} = \beta_0^{(2007)} + \beta_1^{(2007)} x_{ij}^{(2007)} + u_j^{(2007)} + e_{ij}^{(2007)}$$

$$y_{ij}^{(2014)} = \beta_0^{(2014)} + \beta_1^{(2014)} x_{ij}^{(2014)} + u_j^{(2014)} + e_{ij}^{(2014)}$$

$$\begin{bmatrix} u_j^{(2007)} \\ u_j^{(2014)} \end{bmatrix} \sim N \left( \begin{bmatrix} 0 \\ 0 \end{bmatrix}, \begin{bmatrix} \sigma_{u2007}^2 & \\ \sigma_{u2007,2014} & \sigma_{u2014}^2 \end{bmatrix} \right), \quad \begin{bmatrix} e_{ij}^{(2007)} \\ e_{ij}^{(2014)} \end{bmatrix} \sim N \left( \begin{bmatrix} 0 \\ 0 \end{bmatrix}, \begin{bmatrix} \sigma_{e2007}^2 & \\ 0 & \sigma_{e2014}^2 \end{bmatrix} \right)$$

- The level 2 residuals are allowed to be correlated. The correlation measures the stability of school effects between the two cohorts
- The level 1 residuals are modelled as independent as a pupil can only belong to one cohort

# School effects for the 2014 cohort

- It can be shown that the posterior estimates and variances of the school effects for the second cohort, given data only on the first cohort, are

$$\hat{u}_j^{(2014)} = \frac{\rho n_j^{(2007)} \sigma_u^2}{n_j^{(2007)} \sigma_u^2 + \sigma_{e2007}^2} \tilde{y}_j^{(2007)}, \quad \text{var}\left(\hat{u}_j^{(2014)} - u_j^{(2014)}\right) = \frac{n_j^{(2007)} \sigma_u^4 (1 - \rho^2) + \sigma_u^2 \sigma_{e2007}^2}{n_j^{(2007)} \sigma_u^2 + \sigma_{e2007}^2}$$

- Where, for simplicity, we have assumed that the school level variance is constant across cohorts

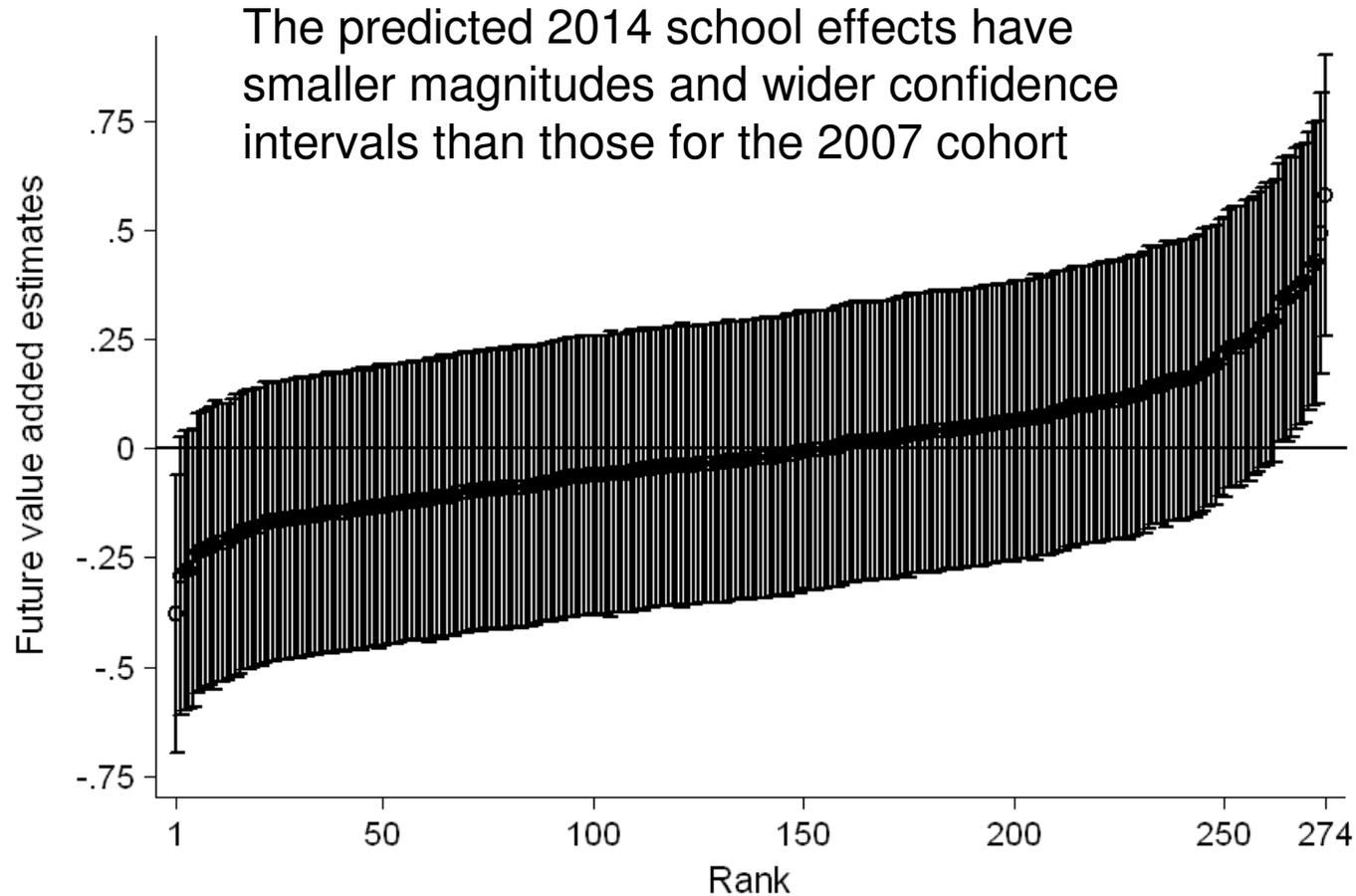
$$\sigma_{u2007}^2 = \sigma_{u2014}^2 = \sigma_u^2$$

- The two equations are the same as before, except for the addition of the terms in red
- The only term we don't know is  $\rho$  the correlation between the two sets of school effects

# Stability of school effects

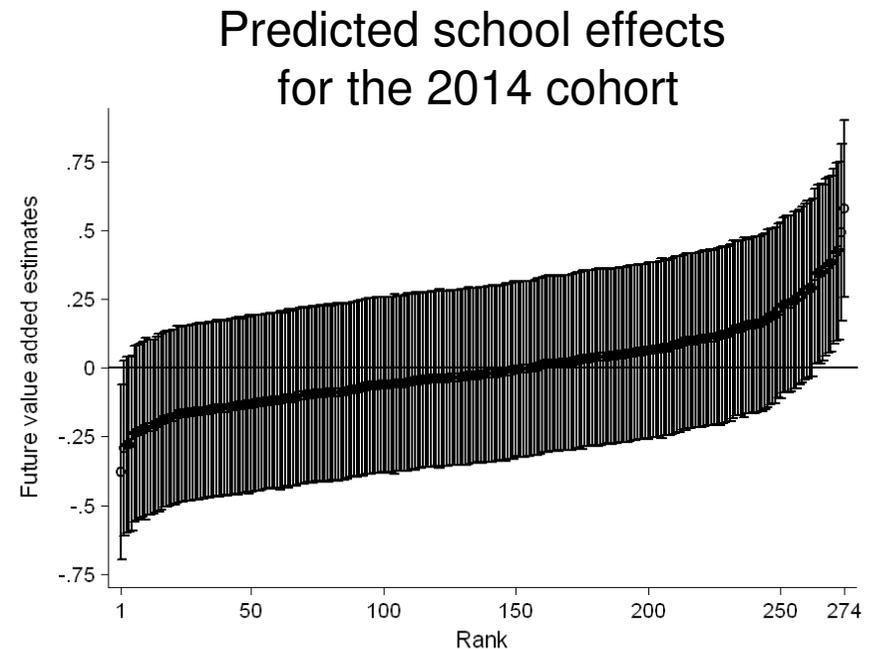
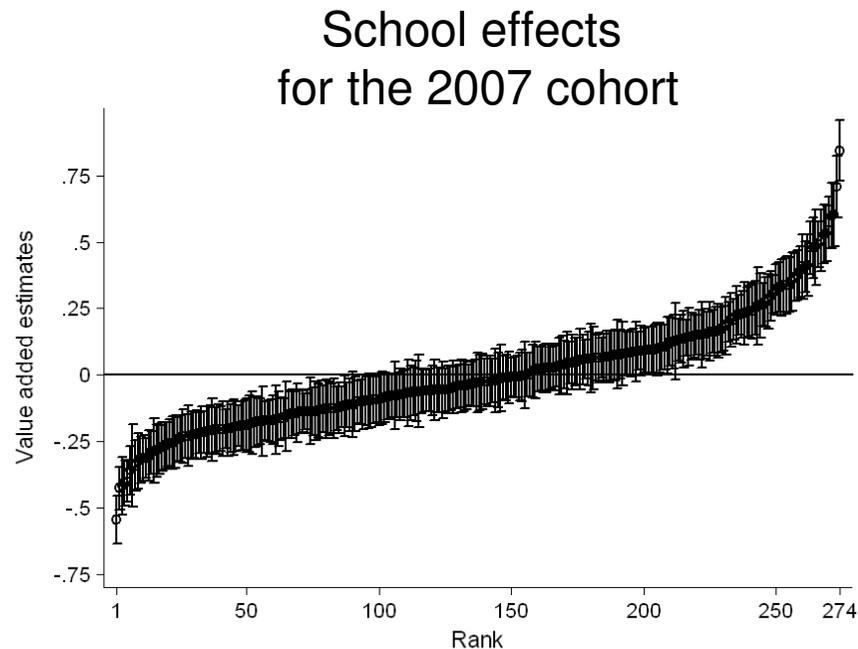
- We want to estimate  $\rho$  the 7 year apart correlation
- Ideally, we would estimate the bivariate response model based on two cohorts of pupils 7 years apart to obtain an estimate of  $\rho$ 
  - Note, we assume that  $\rho$  remains stable over time
- However, we only have data for cohorts five years apart (2002 and 2007)
  - The estimated correlation is 0.69
  - This is an overestimate of the 7 year apart correlation
- We can now adjust the estimates and standard errors of the 2007 school effects

# School effects for the 2014 cohort



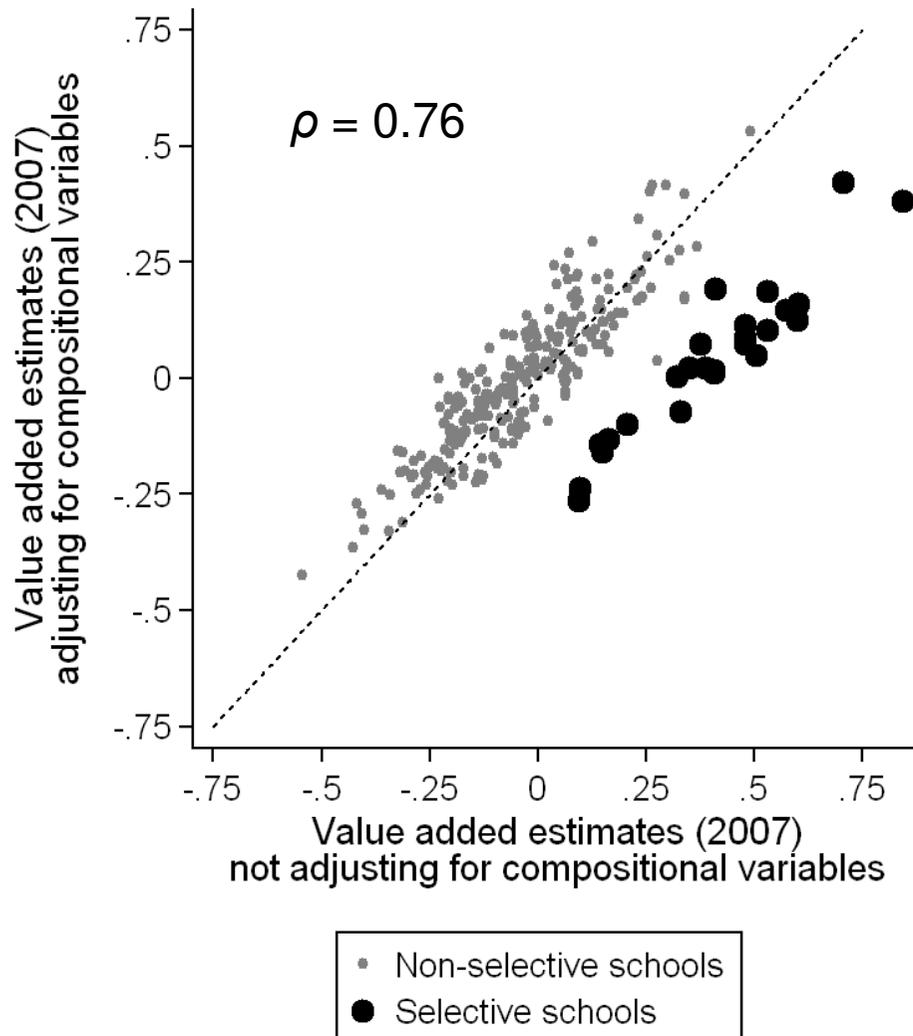
Only ~5% of schools are significantly different from the overall average

# Comparison of the school effects for the 2007 and 2014 cohorts



- Note, these caterpillar plots only allow schools to be compared at the 5% level to the *average* school
- Different confidence intervals are required for pairwise comparisons
  - These are the types of comparisons which parents are interested in

# Adjusting and not adjusting for school compositional variables



- The CVA model adjusts for two school level compositional variables
  - School mean of intake achievement
  - School spread of intake achievement
- This lowers the rankings of grammar schools
- Grammar schools admission policies lead them to have a high mean and narrow a spread of achievement at intake
- However, parents are interested in which schools will produce better subsequent achievement irrespective of whether this is due to school composition, policies or practices

# Conclusions

- School league tables make no adjustment for the statistical uncertainty that arises when current school performance is used to predict future school performance
- Our main result is that, when we adjust for this uncertainty, the number of schools that can be separated from the average school drops from 60% to almost none
- We also argue that, for the purpose of school choice, value-added measures should not adjust for school-level factors, since this is part of the very thing that parents are interested in
- We show that adjusting for the school-level intake composition substantially alters the rank order of school effects
  - Grammar schools drop down the rankings

# Conclusions (cont.)

- We do not propose our approach as a new means of producing league tables
- What we focus on is just one of a long list of statistical concerns that have been expressed about using results as indicators of school performance
  - Other concerns include the side effects and perverse incentives generated by the use of league tables
- However, we do feel that there is an accountability role for performance indicators as monitoring and screening devices to identify schools for further investigation
  - In which case, estimates for the 2007 cohort are the most appropriate
  - However, it is not clear whether to adjust for school compositional variables
  - Performance indicators will be of most use if combined with other sources of school information

# Conclusions (cont.)

- Whilst we have focussed on secondary school league tables, the issues we have discussed are relevant for other stages of schooling
- Indeed, for primary schools our main result will be even more dramatic, since the small size of primary schools makes their estimated schools effects particularly imprecise
- Scotland, Wales and Northern Ireland no longer publish school league tables, perhaps now is the time for England to stop