

**ON THE WRONG TRACK?
BIAS OF PUBLIC SECTOR WORKER
PERFORMANCE MONITORING:
THEORY AND EMPIRICAL EVIDENCE FROM
MIDDLE SCHOOL TEACHERS**

Douglas N. Harris

Associate Professor of Economics

University Endowed Chair in Public Education

Tulane University

Andrew A. Anderson, UW-Madison

MSU Value-Added Conference, Oct. 10-11, 2013

Motivation

- Obvious concern about selection bias in using value-added for high-stakes decisions
- But, almost everything we know about selection bias comes from evidence of elementary schools
 - Rothstein
 - Kane and Staiger
 - Guarino, Reckase, and Wooldridge
- The problem is structurally different in middle and high schools where “tracking” is nearly universal

Tracking & 2 “Misalignment” Problems

- Metric Misalignment
 - ▣ Prime example: Ceiling effect (Koedel and Betts, 2009)
 - ▣ More generally, refers to anything we would normally define as measurement error for a given outcome that is systematically different across tracks (i.e., non-classical)
- Multidimensionality Misalignment
 - ▣ Well known that test scores capture only certain outcomes, i.e., educational output is “multidimensional”
 - ▣ This is a problem if only one output dimension is measured, and if the optimal level of the other output varies by track
- If the metric and multidimensionality problems may vary by track, then we have a new problem
- Both of these problems are much more likely in middle and high school compared with elementary

Optimal Monitoring and Tracking

- Decomposing the track effects is important to understanding how policymakers should use value-added
 - ▣ If metric misalignment, then get a better metric
 - ▣ If multidimensionality, then get more metrics
 - ▣ If neither, and it's sorting, then we may have a problem that simulations don't account for

Questions

- How much does it matter if we ignore tracks (no track indicators)?
- What share of the “track effect” is due to selection and the two forms of misalignment?

Preview

- Ignoring tracks can lead to considerable bias: 30-70% are in the wrong performance category
- The bias is composed mainly of selection bias, but there are two forms of “misalignment” that largely cancel out
- Jackson (2011) says we can address this by simply adding track indicators, but given the complexity of the potential role of tracks, it’s not clear this works

Model (a)

- Suppose we have the following measurement error:

$$A_{it} = A_{it}^* + \beta_0 r_{0it} + \beta_2 r_{2it} + \varepsilon_{it}$$

Metric Misalign.

- Suppose we have the following simple EPF:

$$A_{it}^* = \lambda A_{i,t-1}^* + \nu_{j(i,t)} + \gamma_i + \eta_{it}$$

NO Multid. Misalign.

- By substitution:

$$A_{it} = \lambda A_{i,t-1} + \beta_0 (r_{0it} - \lambda r_{0i,t-1}) + \beta_2 (r_{2it} - \lambda r_{2i,t-1}) + \nu_{j(i,t)} + \gamma_i + \varepsilon_{it} - \lambda \varepsilon_{i,t-1} + \eta_{it}$$

Model (b)

- Suppose we have the following measurement error:

$$A_{it} = A_{it}^* + \beta_0 r_{0it} + \beta_2 r_{2it} + \varepsilon_{it}$$

Metric Misalign.

- Suppose we have the following simple EPF:

$$A_{it}^* = \lambda A_{i,t-1}^* + \nu_{j(i,t)} + \gamma_i + \alpha_0 r_{0it} + \alpha_2 r_{2it} + \eta_{it}$$

YES Multid. Misalign.

- By substitution:

$$A_{it} = \lambda A_{i,t-1} + (\alpha_0 + \beta_0) r_{0it} - \lambda \beta_0 r_{0i,t-1} +$$

$$(\alpha_2 + \beta_2) r_{2it} - \lambda \beta_2 r_{2i,t-1} + \nu_{j(i,t)} + \gamma_i + \varepsilon_{it} - \lambda \varepsilon_{i,t-1} + \eta_{it}$$

Identification Issues

- Student ability (γ) typically unobserved
 - Here, we try FD and FE
- But probably imperfect and:

$$\text{cov}(r_{it}, \gamma_i) \neq 0$$

$$\text{cov}(v_{jt}, \gamma_i) \neq 0$$

- Identify from within-teacher variation in some cases (doesn't seem to matter)
- Instrument for tracks (track availability)

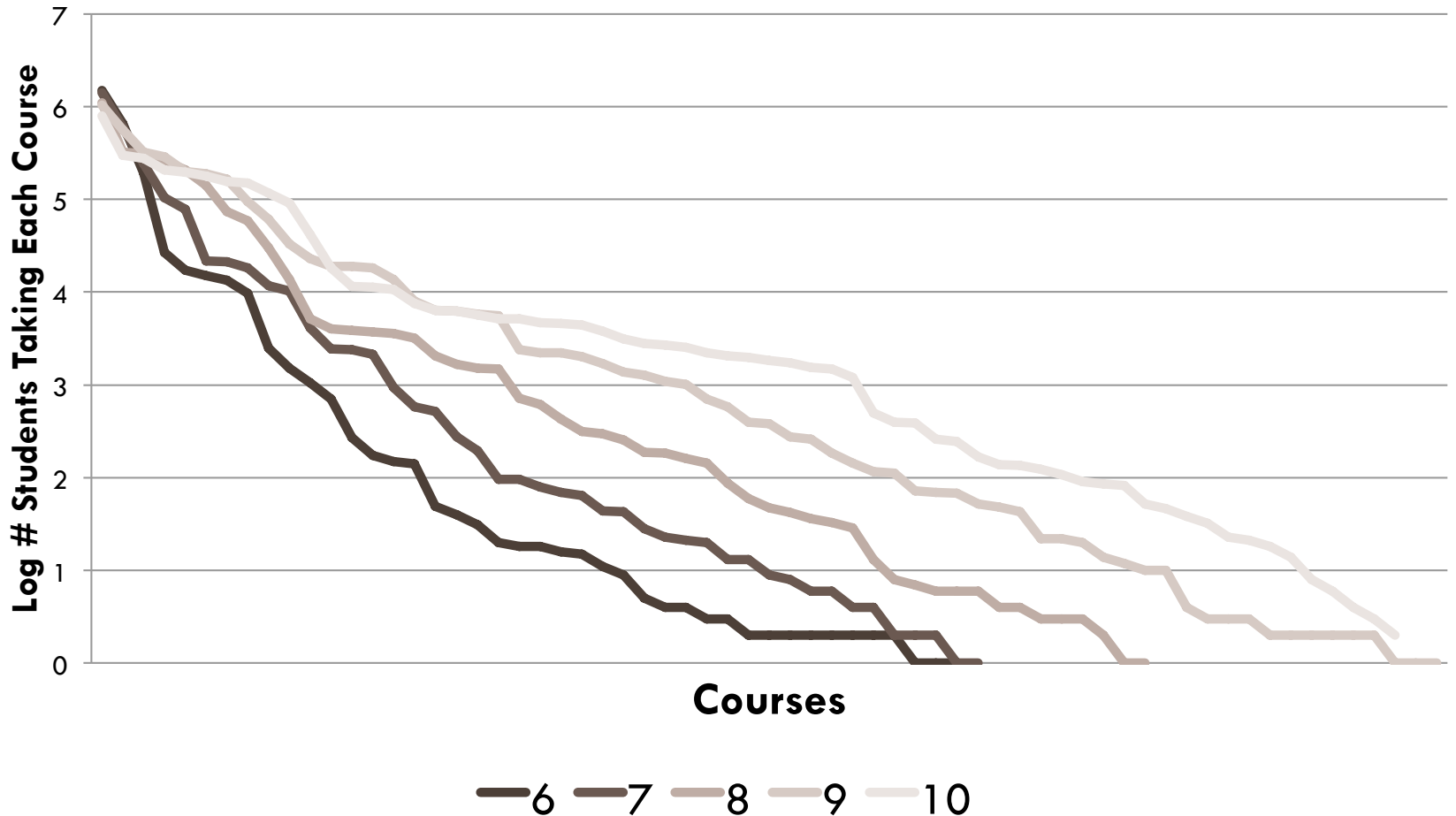
Florida Data

- Good matching of students to teachers
- Data on two tests: high-stakes (SSS) and low-stakes norm-referenced (NRT)
- Track data
 - ▣ Lower, middle, high (all by grade)
- Sample restrictions
 - ▣ Middle schools with grades 6-8
 - ▣ Only traditional public schools (no charters)
 - ▣ Classrooms with 10-40 students
 - ▣ Students with no more than 2 math teachers per year

Most Common Courses (& Tracks)

	Grade 6	Grade 7	Grade 8	Grade 9	Grade 10
Most Common	MATH 1	MATH 2	MATH 3	ALGEBRA 1	GEOMETRY
N	1,500,697	1,420,752	1,107,434	1,033,988	798,907
2nd Most	MATH 1, ADVANCED	MATH 2, ADVANCED	ALGEBRA 1	ALGEBRA 1-A	ALGEBRA 1-B
N	651,925	599,285	320,710	565,827	297,175
3rd Most	INTENSIVE MATH	INTENSIVE MATH	ALGEBRA 1 HONORS	GEOMETRY HONORS	ALGEBRA 2 HONORS
N	192,772	244,072	308,151	324,246	277,588
4th Most	GREAT EXPLOR. PRE-ALG.	PRE- ALGEBRA	MATH 3, ADVANCED	INTENSIVE MATH	GEOMETRY HONORS
N	26,799	103,879	263,034	290,304	209,644

Different Courses by Grade (log)



Results: Proof of Concept (MS Only)

Table 3: Track Effect Estimates from Standard Value-Added Specification (DOLS)

<i>Variable</i>	<i>SSS Test</i>	<i>NRT Test</i>
Lagged achievement	0.841** (0.004)	0.795** (0.004)
Lower track indicator	-0.138** (0.013)	-0.108** (0.013)
Advanced track indicator	0.241** (0.003)	0.357** (0.006)
Student / year observations:	253,730	209,563

s.d. in teacher effects is 0.25

What if we omitted the track indicators? (MS Only)

<i>Quartile of Teacher Value-Added with Track Coefficients Omitted</i>	<i>Quartile of Teacher Value-Added with Track Coefficients Included</i>			
	<i>1 (Low)</i>	<i>2</i>	<i>3</i>	<i>4 (High)</i>
<i>Lower Track Teachers</i>				
1	0.706	0.221	0.074	0.000
2	0.056	0.250	0.389	0.306
3	0.000	0.000	0.275	0.725
4	0.000	0.012	0.000	0.988
<i>Upper Track Teachers</i>				
1	0.917	0.083	0.000	0.000
2	0.422	0.477	0.099	0.002
3	0.092	0.383	0.429	0.096
4	0.000	0.090	0.342	0.568

Track Coefficients with FD-IV(α) (MS)

Table 4: Track Effects from First-Differenced Model: All Teachers in All Schools

<i>Model</i>	<i>Sunshine State Standards Test (SSS)</i>			<i>Norm-Referenced Test (NRT)</i>		
	$\widehat{\beta}_0$	$\widehat{\beta}_2$	$\hat{\lambda}$	$\widehat{\beta}_0$	$\widehat{\beta}_2$	$\hat{\lambda}$
FD-Most Exogenous IVs:						
λ unrestricted	0.121*** (0.022)	0.019** (0.009)	0.037 (0.078)	0.044* (0.025)	0.116*** (0.012)	0.195*** (0.060)
$\lambda=0.4$	0.129*** (0.023)	0.018* (0.010)	0.400 --	0.045* (0.025)	0.119*** (0.013)	0.400 --
FD-Stronger IVs (less exogenous):						
λ unrestricted	0.117*** (0.021)	0.019** (0.009)	-0.021 (0.017)	0.040* (0.023)	0.111*** (0.011)	-0.038** (0.015)
$\lambda \geq 0$	0.119*** (0.021)	0.019** (0.009)	0.000 (0.001)	0.042* (0.023)	0.112*** (0.011)	0.000 (0.001)
$\lambda = 0.4$	0.138*** (0.023)	0.016 (0.010)	0.400 --	0.056** (0.026)	0.115*** (0.013)	0.400 --

We also report results for FD-only (no IV) and the results are essentially the same

Track Coefficients with FE(b)

	7th		8th		9th		10th	
	SSS	NRT	SSS	NRT	SSS	NRT	SSS	NRT
α_0	-0.2829*** (0.0099)	-0.0970*** (0.0129)	-0.1639*** (0.0057)	-0.0661*** (0.0067)	-0.1142*** (0.0049)	-0.0559*** (0.0063)	-0.2456*** (0.0084)	-0.1474*** (0.0095)
α_2	0.0977*** (0.0074)	0.0421*** (0.0098)	0.0573*** (0.0045)	0.0440*** (0.0052)	0.0385*** (0.0038)	-0.0121* (0.0049)	0.0169* (0.0072)	-0.1364*** (0.0083)
β_0	0.3228*** (0.0102)	0.1290*** (0.0135)	0.2158*** (0.0060)	0.1016*** (0.0071)	0.1521*** (0.0053)	0.0738*** (0.0067)	0.2278*** (0.0085)	0.1585*** (0.0097)
β_2	-0.0718*** (0.0073)	-0.0174 (0.0098)	-0.0282*** (0.0048)	0.0029 (0.0055)	-0.0052 (0.0040)	0.0724*** (0.0052)	0.0600*** (0.0070)	0.2569*** (0.0082)

Track Coefficients with FE(b)

	7th		8th		9th		10th	
	SSS	NRT	SSS	NRT	SSS	NRT	SSS	NRT
α_0	-0.2829*** (0.0099)	-0.0970*** (0.0129)	-0.1639*** (0.0057)	-0.0661*** (0.0067)	-0.1142*** (0.0049)	-0.0559*** (0.0063)	-0.2456*** (0.0084)	-0.1474*** (0.0095)
α_2	0.0977*** (0.0074)	0.0421*** (0.0098)	0.0573*** (0.0045)	0.0440*** (0.0052)	0.0385*** (0.0038)	-0.0121* (0.0049)	0.0169* (0.0072)	-0.1364*** (0.0083)
β_0	0.3228*** (0.0102)	0.1290*** (0.0135)	0.2158*** (0.0060)	0.1016*** (0.0071)	0.1521*** (0.0053)	0.0738*** (0.0067)	0.2278*** (0.0085)	0.1585*** (0.0097)
β_2	-0.0718*** (0.0073)	-0.0174 (0.0098)	-0.0282*** (0.0048)	0.0029 (0.0055)	-0.0052 (0.0040)	0.0724*** (0.0052)	0.0600*** (0.0070)	0.2569*** (0.0082)
$\alpha_0 + \beta_0$	0.040	0.032	0.052	0.036	0.038	0.018	-0.018	0.011
$\alpha_2 + \beta_2$	0.026	0.025	0.029	0.047	0.033	0.060	0.077	0.121
Diff	-0.014	-0.007	-0.023	0.011	-0.005	0.042	0.095	0.109

A Problem in High School

- At the very least, the role of tracking is different at the high school level
- Also, many reasons not to trust the high school results
 - ▣ Challenge of placing courses into tracks
 - ▣ Other coefficients in model are surprising
 - Teacher experience is insignificant and wrong sign
 - Avg. class score is significant and wrong sign
- This may be why Jackson finds that current VA is a worse predictor of future VA in high school compared with elementary
- Two other related reasons . . .

More on High School Problem

- What if you're the teacher who teaches "Geometry, Honors" in both 9th and 10th grade
 - ▣ The test aligns poorly with Geometry in 9th grade, but well in 10th
 - ▣ Possible solution: Estimate separately by grade
 - And can just use course indicators rather than grouping into tracks
 - Jackson uses 9th grade (not 10th)
- As our model shows, the lagged track also matters (through its effect on lagged achievement)

Simulation Evidence

- Preliminary simulation evidence suggests that using only contemporaneous track is sufficient when students don't switch tracks
- Much more bias for teachers of students who switch tracks
- Another case of the “exception is the rule problem”
 - ▣ Other examples: inexperienced teachers, team teaching, teaching highly advanced students in elementary grades, etc.

Conclusions

- General finding is the same in models (a) and (b): track coefficients largely reflect sorting, not misalignment
- The complex structure of the tracking problem has not apparently been addressed in prior simulation evidence
- This may explain the Jackson result that VA, as currently estimated, works less well in high school