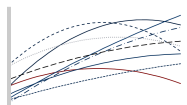

A Constellation of Findings From NCAASE

Drs. Stevens and Schulte on Across Year Growth and Gaps

Drs. Elliott, Tindal, and Nese on Within Year Growth

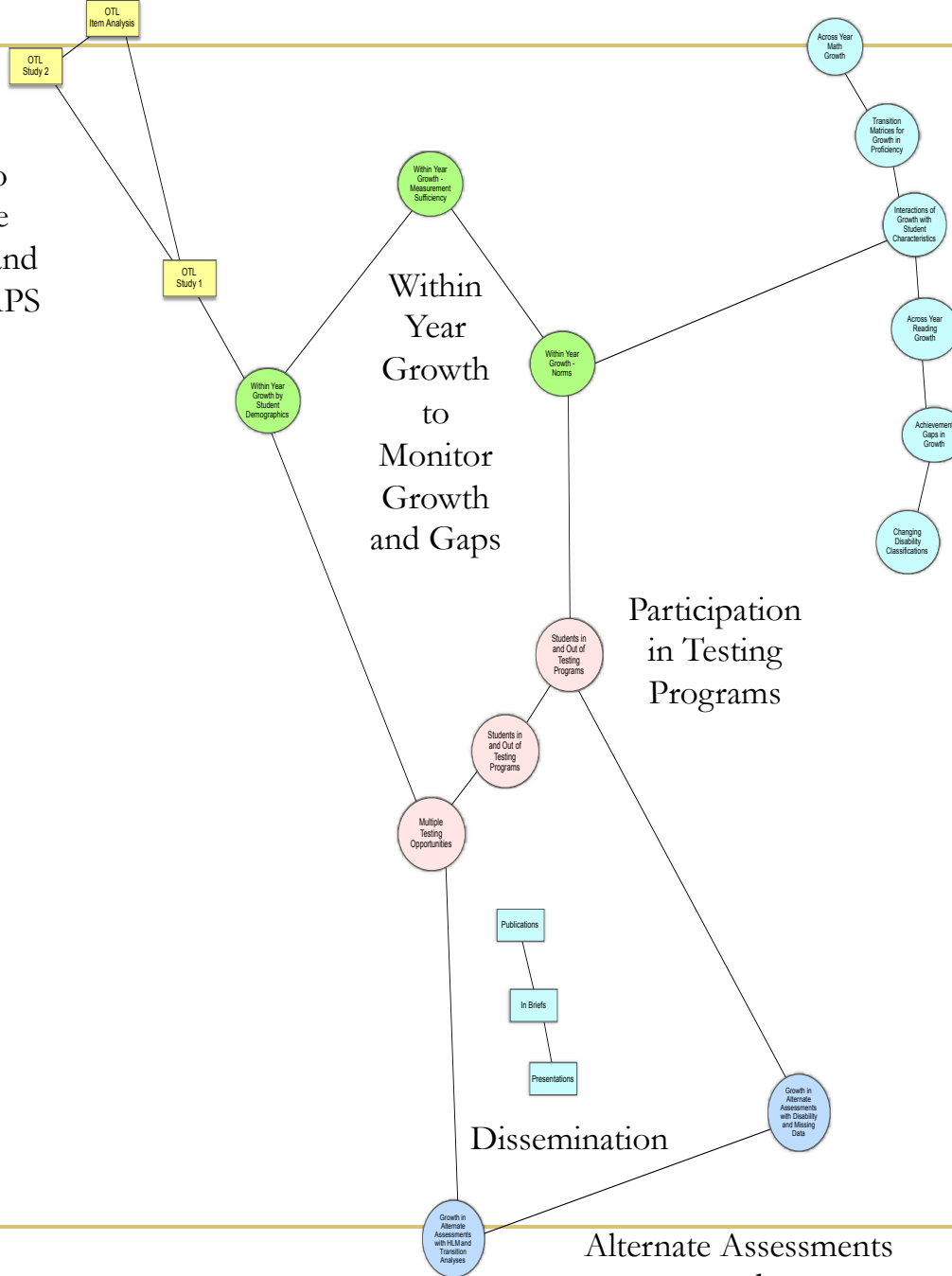
Dr. Stevens on Comparison of Growth Models



Six Questions turned in Numerous Studies

- 1. What is the **natural developmental progress in achievement** for students with disabilities?
- 2. What **models best characterize achievement growth for students with disabilities** who are participating in general achievement tests, as well as those taking alternate assessments?
- 3. How do various **growth models represent school effects** for students with and without disabilities, and how do results compare to those derived from status models now in use?
- 4. What are the **reliability and validity of the estimates of school effectiveness** for students with disabilities produced by alternative growth models and how are these estimates influenced by contextual differences among schools and students?
- 5. How do **results from different types of interim assessments of students' achievement** meaningfully contribute to a model of academic growth for students with disabilities?
- 6. How can **information about opportunity to learn and achievement growth** be used to enhance academic outcomes for students with disabilities?

OTL to
increase
Growth and
Close GAPS



Across Year Growth
(and Gaps) for
Students with and
without Disabilities

Participation
in Testing
Programs

Dissemination

Alternate Assessments
and
Growth for SWSCD





Achievement Growth for Students With and Without Disabilities

Ann Schulte
Arizona State University

Investigating Achievement Growth for SWD

- Widespread agreement that achievement growth is an important outcome to be considered in educational accountability models
 - Little information on achievement growth for SWD or how it compares to students without disabilities (SWOD) on which to base alternative models
-

Investigating Achievement Growth for SWD

- NCLB treats SWD as a unitary subgroup, but specific disabilities are likely to have different achievement trajectories—how different are the exceptionality groups in terms of intercept, growth, and gaps?
- What are some implications for charting growth of the changing nature of the special education population?

Schulte, A. C., Stevens, J. J., Elliott, S. N., Tindal, G., & Nese, J. F. T. (in press). Achievement gaps for students with disabilities: Stable, Widening, or Narrowing on a State-wide Reading Comprehension Test? *Journal of Educational Psychology*.

Stevens, J. J., Schulte, A. C., Elliott, S. N., Nese, J. F. T., & Tindal, G. (2015). Mathematics achievement growth of students with and without disabilities on a statewide achievement test. *Journal of School Psychology, 53*, 45-62.

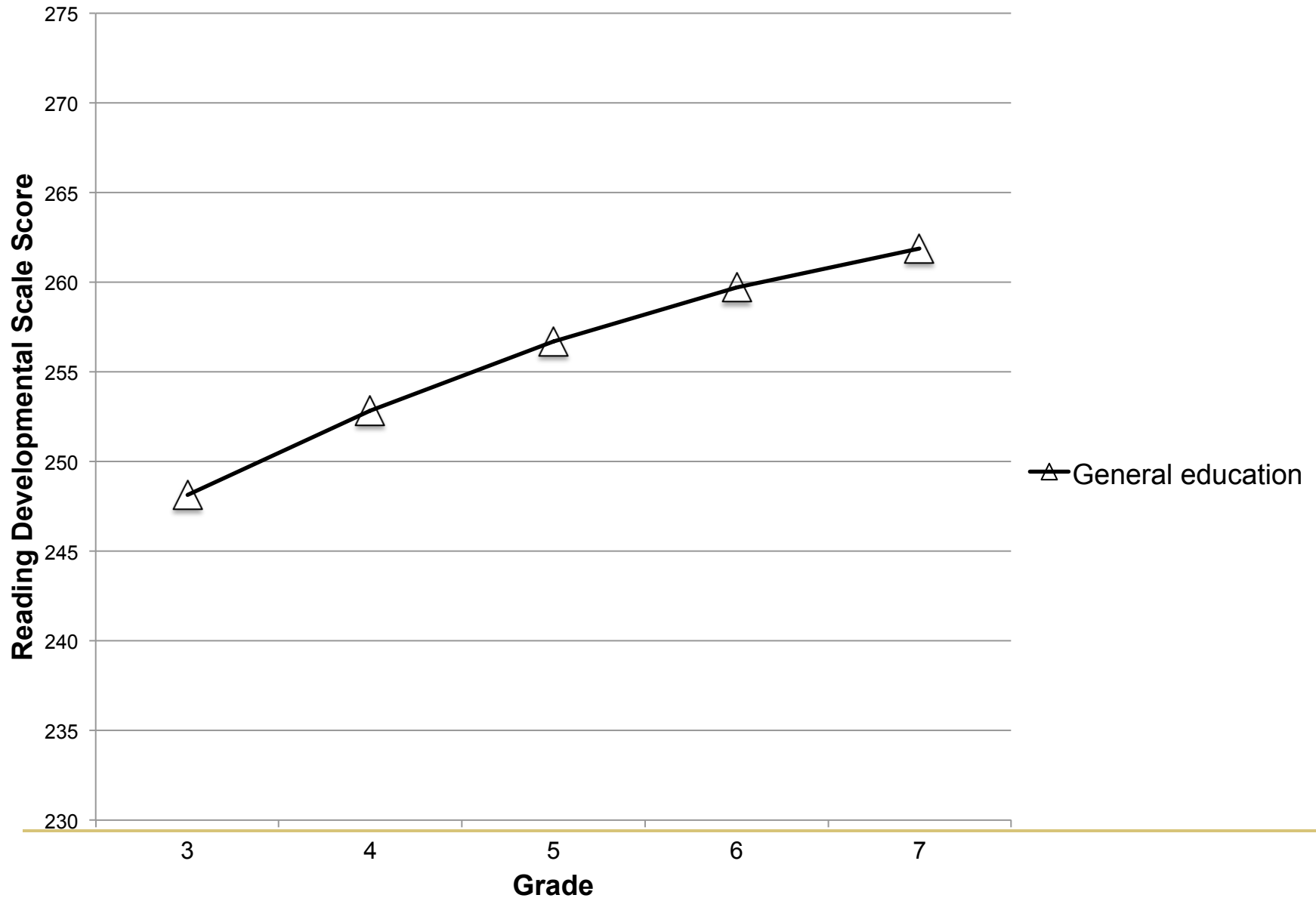
Cornerstone Growth Studies

- Followed two entire cohorts ($n > 95,000$ each) in one state across grades 3-7, single editions of state reading and mathematics tests
- Exceptionality based on classification at Grade 3 (students who entered system after Grade 3 not included in sample)
- Two level HLMs (time and students), controlling for demographic characteristics at Level 2
- Intercept at Grade 3, both linear and quadratic functions included

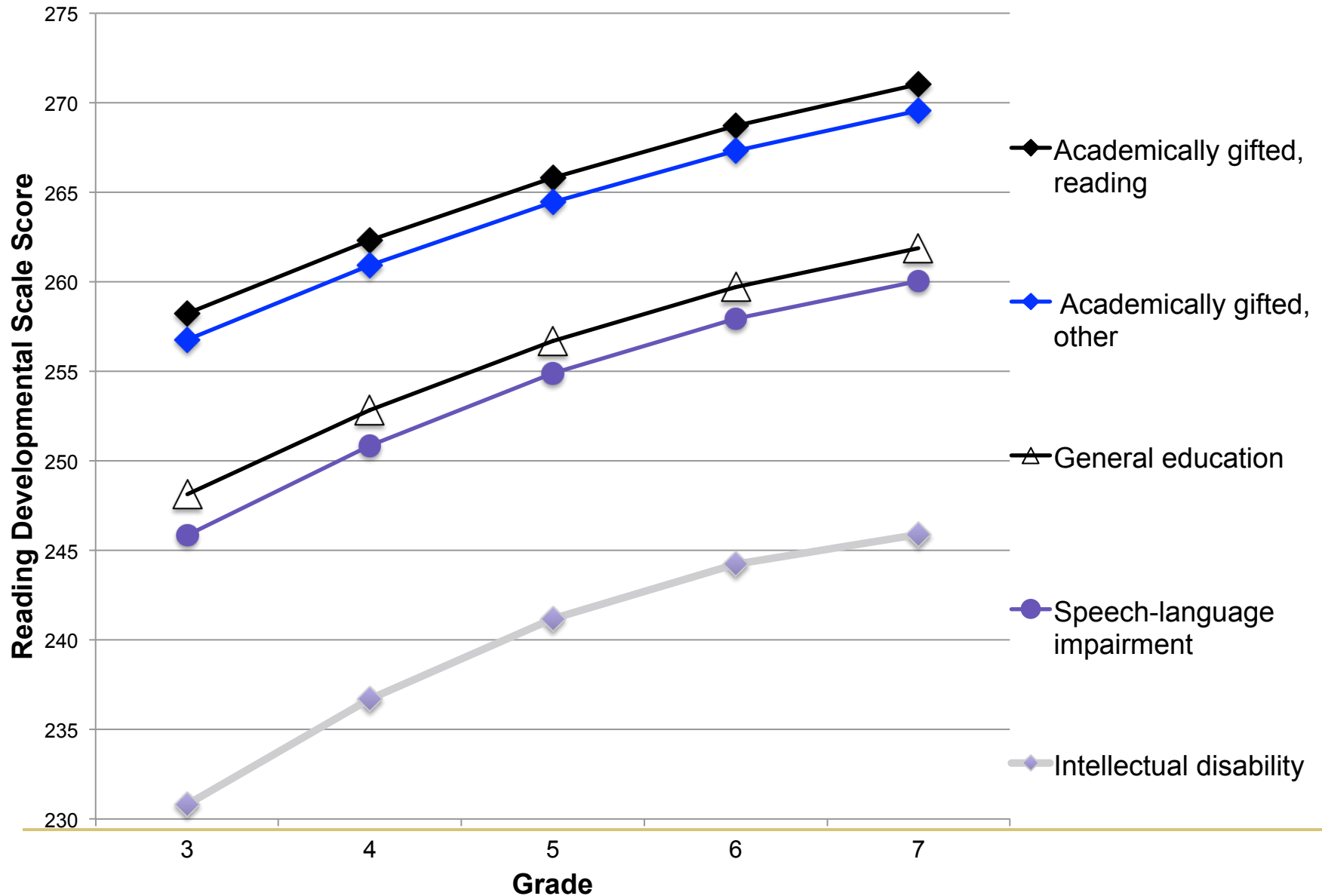
North Carolina End of Grade (EOG) Tests

- EOG-R (2nd ed.) Reading passages followed by multiple-choice items testing comprehension of passage content
 - EOG-M (2nd ed.) Mathematics stressed application and problem solving
 - Administered annually, grades 3-8
 - Developmental scale across grades
-

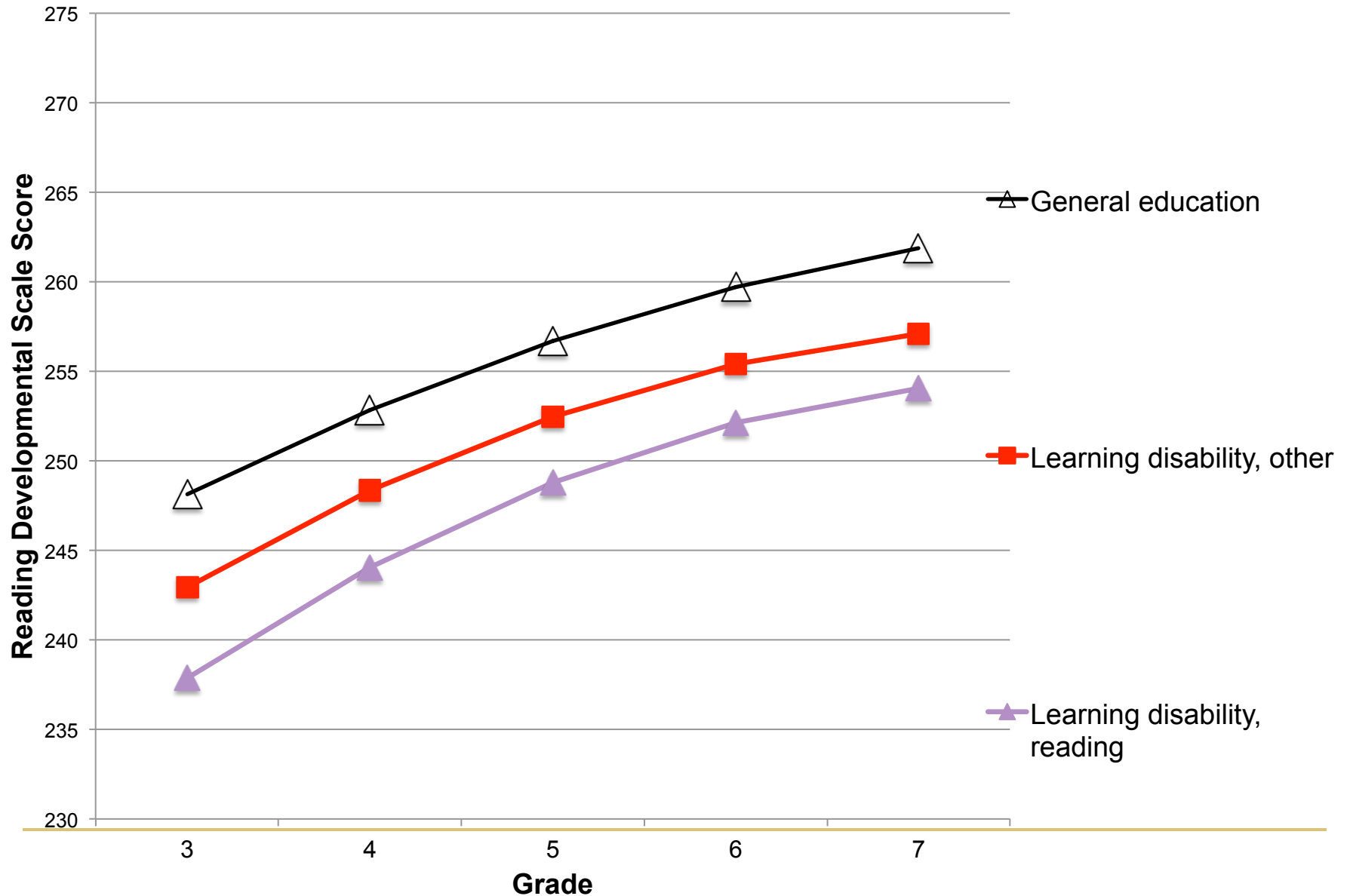
Reading Comprehension Growth by Exceptionality



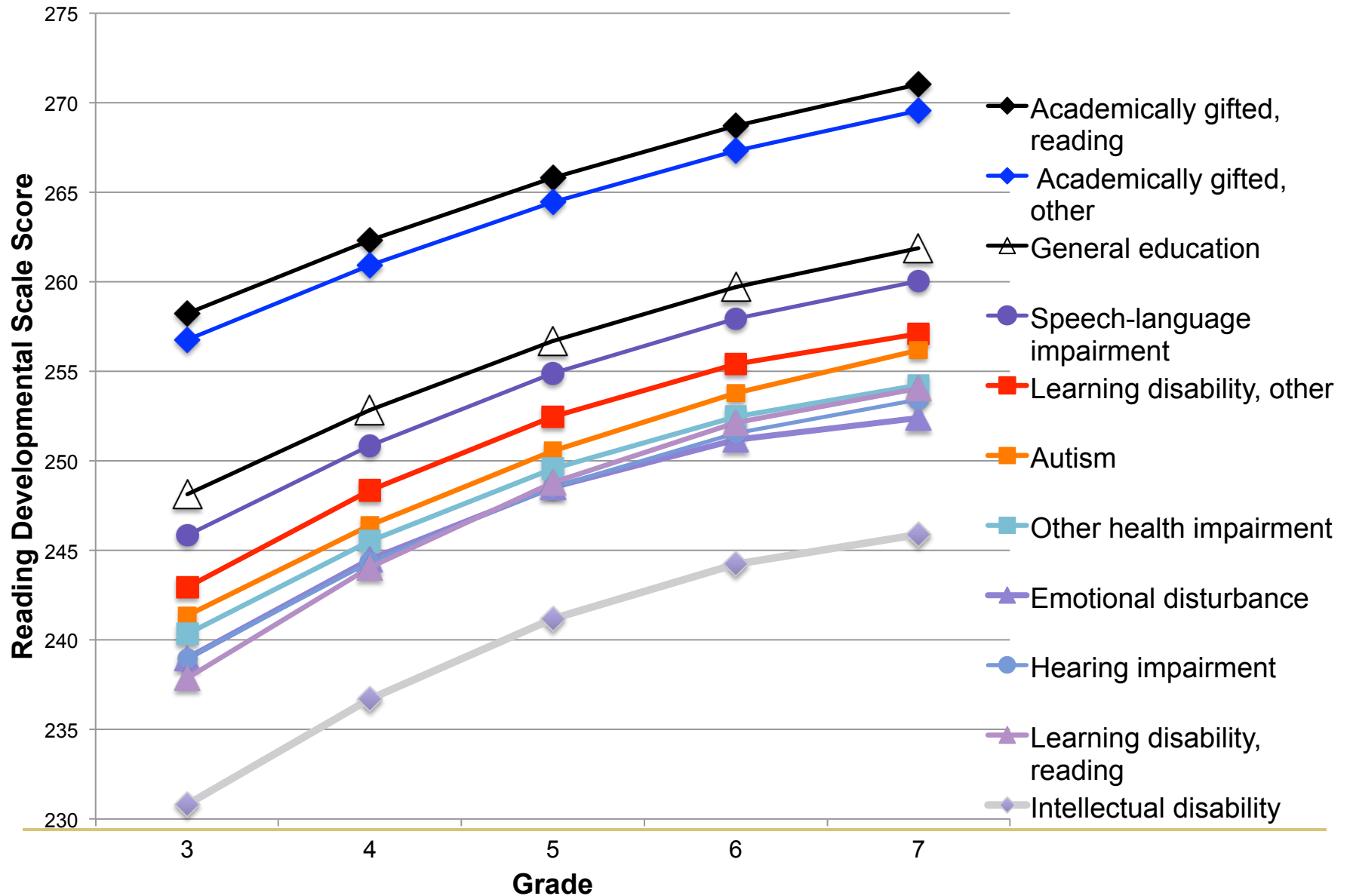
Reading Comprehension Growth by Exceptionality



Learning Disability in Reading vs. LD in Other Area



Reading Comprehension Growth by Exceptionality



Reading Comprehension Two-level HLM (Time and Students)

Predictor	Intercept	Linear	Quadratic
Grand Mean	250.82 (.04)	5.03 (.03)	-.42 (.01)
Sex	1.02(.05)	-0.17 (.03)	.07 (.01)
Free lunch	-3.61 (.06)	0.14 (.03)	-.05 (.01)
Limited English	-5.57(.16)	0.70 (.10)	-.05 (.02)
Asian	.31† (.17)	0.18† (.10)	.05 (.02)
American Indian	-2.74 (.21)	-0.62 (.13)	.12 (.03)
Black	-4.34 (.06)	0.25 (.04)	-.05 (.01)
Hispanic	-1.54 (.13)	0.40 (.08)	-.05 (.02)
Multiracial	-1.05 (.16)	0.19† (.10)	-.03† (.02)

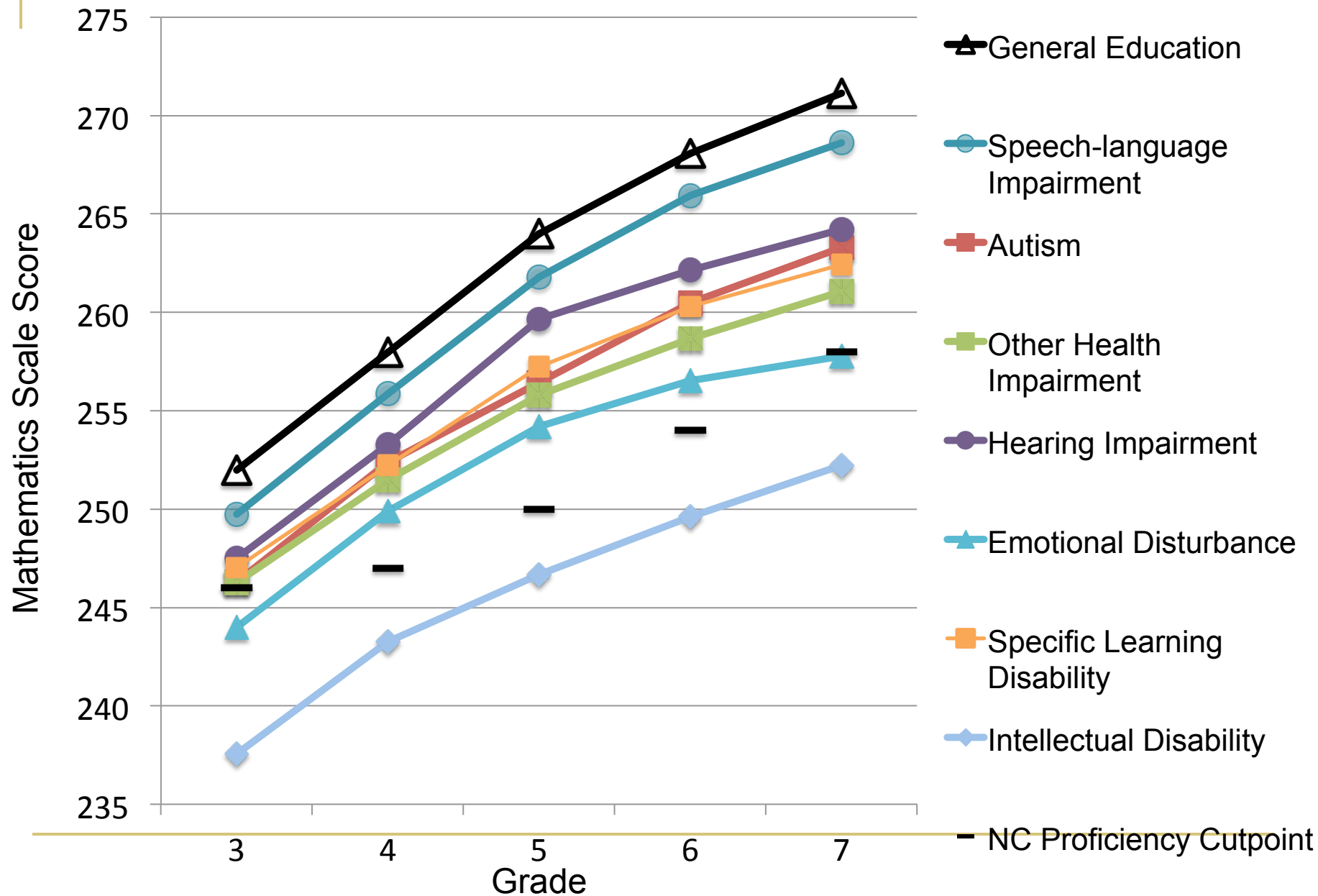
†Not significant, $p > .05$

Reading Two-level HLM (con't.)

Predictor	Intercept	Linear	Quadratic
Gifted-Rdg	7.82 (.08)	-0.59 (.06)	.09 (.01)
Gifted-Other	6.84 (0.15)	-0.54 (.12)	.08 (.03)
Autism	-7.80 (.74)	.38† (.41)	-.01† (.10)
Emotional Disturbance	-7.17 (.35)	1.00 (.23)	-.25 (.06)
Hearing Impairment	-8.89 (.65)	0.77† (.40)	-.15† (.10)
Intellectual Disability	-14.82 (.22)	1.38 (.18)	-.25 (.04)
Other Health Impairment	-7.86 (.22)	0.62 (.13)	-.13 (.03)
Specific Learning Disability-Rdg	-9.78 (.13)	1.72 (.08)	-.27 (.02)
Specific Learning Disability-Other	-4.96 (.25)	0.85 (.15)	-.18 (.03)
Speech-language Impairment	-2.64 (.15)	0.38 (.09)	-.06 (.02)
Pseudo R ² (as %)	39.44	8.45	8.26

†Not significant, $p > .05$

Mathematics Growth by Exceptionality



Special Education Subgroup Membership Changes Across Time

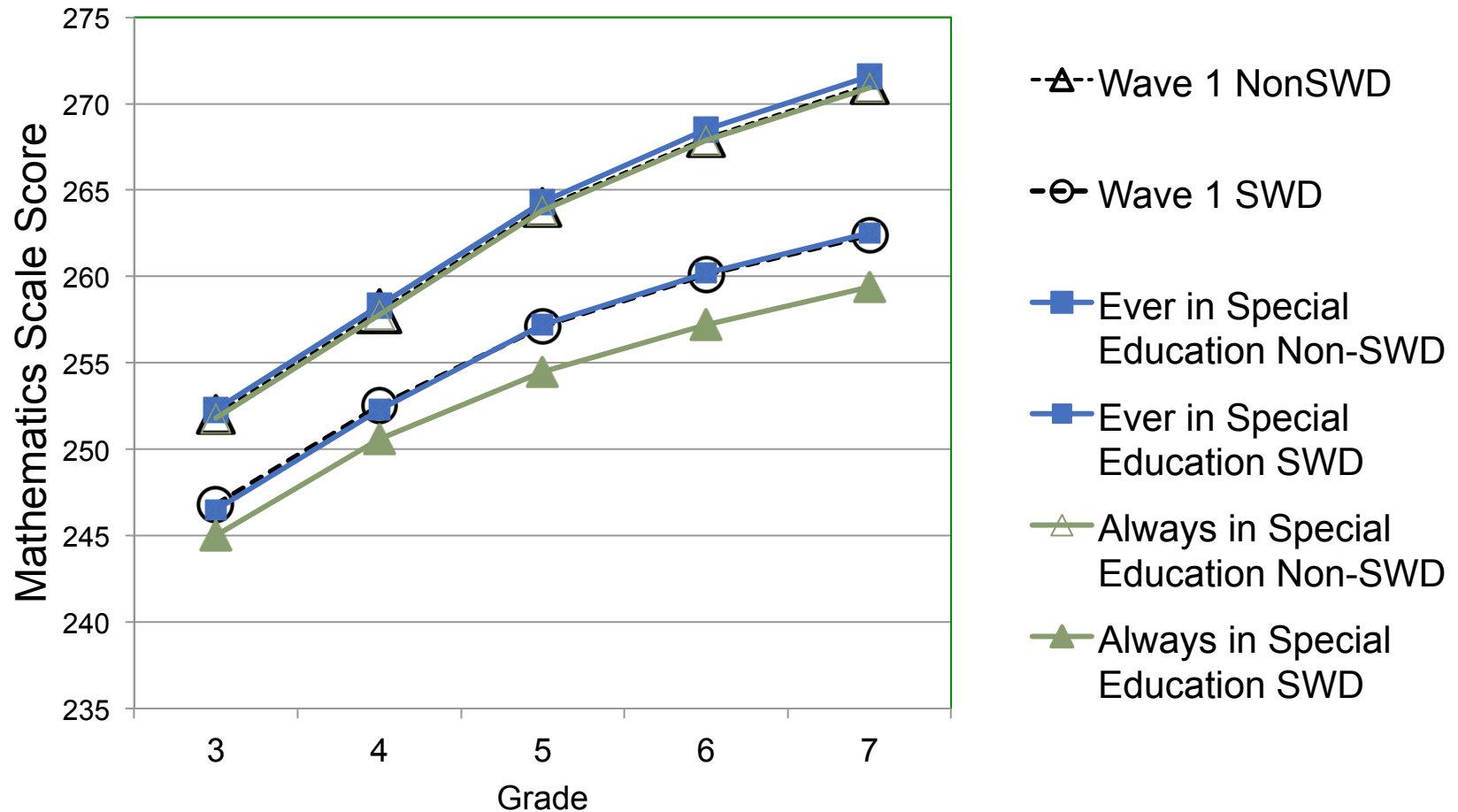
- Students exit and enter special education and changes are related to student achievement status.
- What are the implications of the changing membership?

Schulte, A. C., & Stevens, J. J. (2015). Once, sometimes, or always in special education: Mathematics growth and achievement gaps. *Exceptional Children*, 81, 370-387. doi: 10.1177/0014402914563695

Special Education Membership Grades 3-7

SWD Subgroup Identification Method	Percent
Current Year	11.1 to 12.4
Wave 1	11.8
Ever in Special Education	16.1
Always in Special Education	6.0

Observed Means by SWD Identification Method



Implications

- SWD subgroup is comprised of heterogeneous group of students, who vary greatly in achievement in grade 3
 - Most exceptionality groups made somewhat greater growth than general education students in reading and somewhat less growth in mathematics—although overall characterization is one of stable gaps
 - “One size may fit all” for growth, but only if differing starting points for SWD are recognized. Growth-to-standard expectations require much greater growth for most SWD groups than is typically observed
-

Individual Differences and Gaps for Students With Disabilities

Joseph Stevens
University Oregon

Individual Differences and Achievement Gaps in Math and Reading for SWD

- Summarize a number of our study results that focus on individual differences in academic performance
- Central NCLB and RTTT goal is universal proficiency and the reduction of achievement gaps between SWoD students and protected subgroups including SWD
- Previous research on achievement gaps has limitations:
 - ❑ Often gaps are not evaluated empirically; visual inspection rather than statistical testing; no common, empirical metric (effect size) to describe differences
 - ❑ Interactions not tested

SWD Growth Achievement Gaps

- What is the size of the achievement gap in mathematics and reading for students in specific exceptionality categories?
- Does the gap increase, decrease or stay the same over time?

Mathematics Achievement Gaps, see:

Stevens, J. J., Schulte, A. C., Elliott, S. N., Nese, J. F. T., & Tindal, G. (2015). Mathematics achievement growth of students with and without disabilities on a statewide achievement test. *Journal of School Psychology, 53*, 45-62.

Reading Achievement Gaps, see:

Schulte, A. C., Stevens, J. J., Elliott, S. N., Tindal, G., & Nese, J. F. T. (in press). Achievement Gaps for Students with Disabilities: Stable, Widening, or Narrowing on a State-wide Reading Comprehension Test? *Journal of Educational Psychology*.

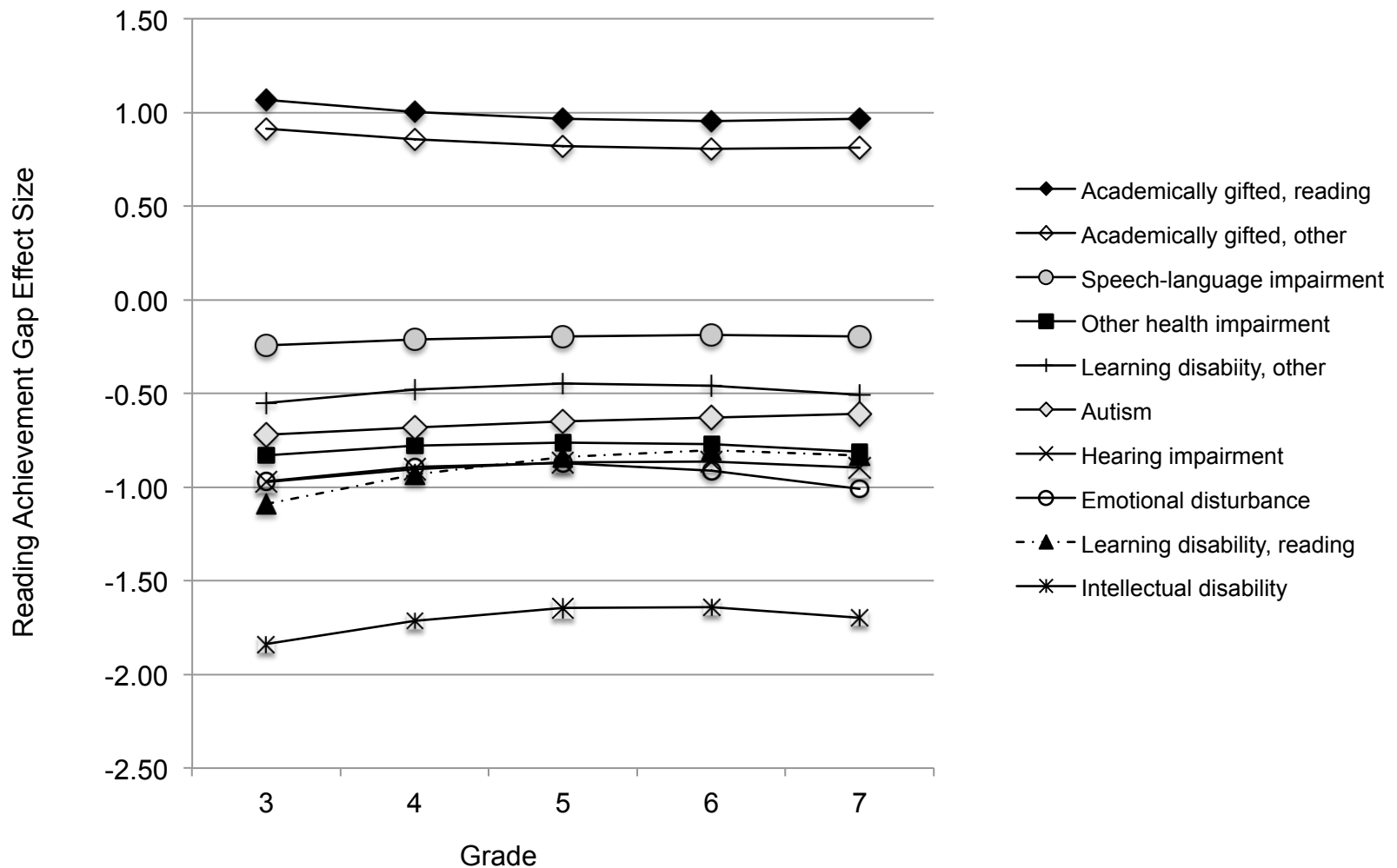
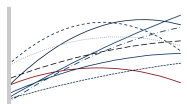


Figure. Reading achievement gap effect sizes based on differences in empirical Bayes estimated means across grades for students in different exceptionality categories compared to students in general education (from Schulte et al., in press).

Achievement Gaps as Differences in Proficiency Rates

- We also examined SWoD-SWD achievement gaps in other ways
- Difference in percent proficient ($P-P$)
 - Most common method in public dissemination (e.g., report cards); district, state, and federal reports
 - Easy to interpret?
 - $P-P$ and Cohen's h reported below for North Carolina
- Problems with these metrics, however:
 - Size of gap depends on test used, score scale, and location of cutscore
 - Size of gap depends on shape of score distributions for the two groups
 - Proportions are ordinal, units may be different at different locations on the scale (i.e., not an interval scale)



North Carolina	Mathematics							Reading					
Grade	3	4	5	6	7	8		3	4	5	6	7	8
Student Group													
General Education <i>N</i>	78.3 (97680)	79.3 (94162)	77.2 (92973)	74.4 (91406)	75.4 (90642)	76.8 (91668)		61.7 (97625)	66.1 (94111)	64.1 (92935)	67.7 (91370)	60.0 (90607)	62.3 (91627)
All SWD <i>N</i>	53.2 (11208)	49.9 (11046)	45.3 (9934)	39.7 (9310)	38.5 (8728)	39.9 (8613)		30.8 (10759)	32.1 (10475)	27.8 (9466)	28.5 (9059)	23.0 (8510)	24.1 (8459)
<i>h</i>	.54	.63	.67	.72	.76	.77		.63	.69	.75	.81	.77	.79
Autism <i>N</i>	62.4 (351)	64.9 (365)	59.7 (365)	62.1 (330)	64.8 (244)	57.4 (284)		41.9 (346)	48.3 (360)	50.6 (360)	47.2 (335)	51.6 (252)	47.4 (289)
<i>h</i>	.35	.32	.38	.27	.23	.42		.40	.36	.27	.42	.17	.30
Communication <i>N</i>	67.3 (3842)	66.3 (2501)	59.7 (1354)	51.8 (651)	51.2 (369)	48.4 (223)		46.6 (3838)	50.8 (2496)	43.2 (1353)	42.8 (649)	32.2 (370)	29.1 (223)
<i>h</i>	.25	.29	.38	.47	.51	.60		.30	.31	.42	.51	.57	.68
Emotional <i>N</i>	44.4 (331)	43.9 (394)	35.1 (453)	29.2 (510)	26.5 (475)	22.6 (562)		28.8 (323)	44.6 (166)	26.3 (453)	22.4 (510)	22.7 (476)	17.9 (570)
<i>h</i>	.71	.75	.88	.94	1.02	1.15		.67	.44	.78	.95	.78	.95
Hearing <i>N</i>	49.0 (143)	50.0 (130)	56.0 (134)	46.7 (105)	41.0 (100)	44.1 (111)		20.6 (136)	30.9 (123)	32.3 (133)	34.3 (102)	21.9 (96)	24.1 (112)
<i>h</i>	.62	.63	.45	.58	.71	.68		.86	.72	.65	.68	.80	.79

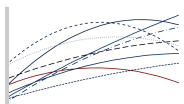
	Mathematics							Reading					
Grade	3	4	5	6	7	8		3	4	5	6	7	8
General Education	78.3 (97680)	79.3 (94162)	77.2 (92973)	74.4 (91406)	75.4 (90642)	76.8 (91668)		61.7 (97625)	66.1 (94111)	64.1 (92935)	67.7 (91370)	60.0 (90607)	62.3 (91627)
Intellectual	7.9	4.8	3.7	4.4	4.2	6.1		1.7	3.2	2.8	2.2	0.0	1.7
<i>N</i>	(252)	(229)	(215)	(225)	(240)	(296)		(239)	(218)	(216)	(230)	(250)	(295)
<i>h</i>	1.60	1.76	1.76	1.66	1.69	1.64		1.55	1.54	1.52	1.63	1.59	1.56
Orthopedic	55.6	40.5	62.5	47.7	58.8	54.7		42.1	45.9	50.0	42.6	58.8	40.0
<i>N</i>	(36)	(37)	(48)	(44)	(34)	(53)		(38)	(37)	(48)	(47)	(34)	(55)
<i>h</i>	.49	.82	.32	.56	.36	.47		.39	.41	.29	.51	.02	.45
Other	44.1	41.7	38.5	36.0	33.8	36.7		24.7	27.7	26.7	29.9	22.2	25.6
<i>N</i>	(1663)	(2085)	(2218)	(2347)	(2299)	(2172)		(1625)	(2043)	(2204)	(2371)	(2313)	(2189)
<i>h</i>	.72	.79	.81	.79	.86	.83		.77	.79	.77	.78	.79	.76
Language Disability	47.1	46.4	45.5	40.8	40.9	43.6		19.6	24.2	22.7	26.1	21.9	23.7
<i>N</i>	(4524)	(5246)	(5085)	(5058)	(4909)	(4865)		(4148)	(4761)	(4638)	(4776)	(4662)	(4680)
<i>h</i>	.66	.70	.67	.69	.72	.69		.89	.87	.86	.86	.80	.80
TBI	30.0	–	35.3	38.5	36.8	20.0		–	–	43.8	25.0	15.8	6.7
<i>N</i>	(10)	(8)	(17)	(13)	(19)	(15)		(9)	(7)	(16)	(12)	(19)	(15)
<i>h</i>	1.01	–	.87	.74	.80	1.21		–	–	.41	.89	.95	1.30
Visual Impairment	62.8	75.0	57.1	60.9	62.9	74.1		50.0	53.3	57.1	39.1	54.3	46.2
<i>N</i>	(43)	(48)	(42)	(23)	(35)	(27)		(44)	(45)	(42)	(23)	(35)	(26)
<i>h</i>	.34	.10	.43	.29	.27	.06		.24	.26	.14	.58	.12	.32

Achievement Gaps as Areas Between Score Distributions

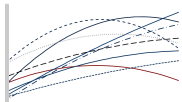
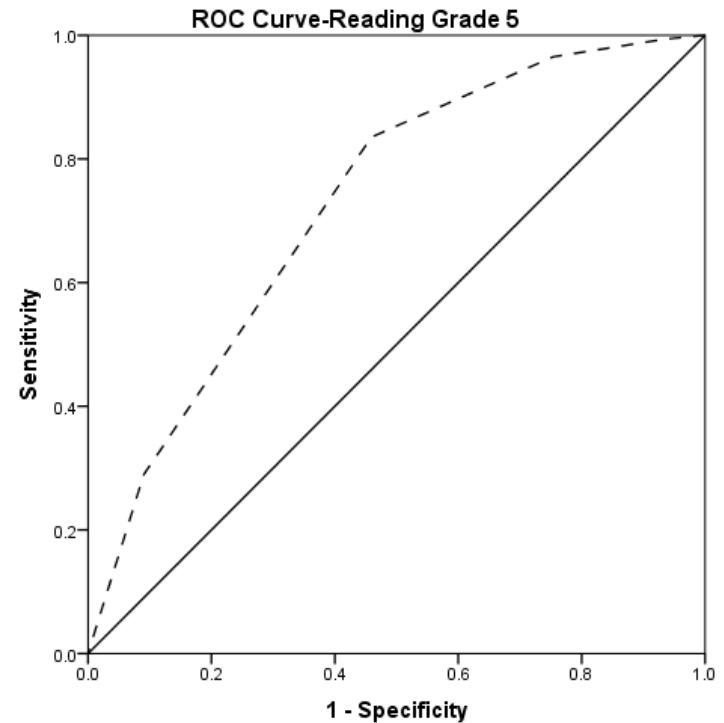
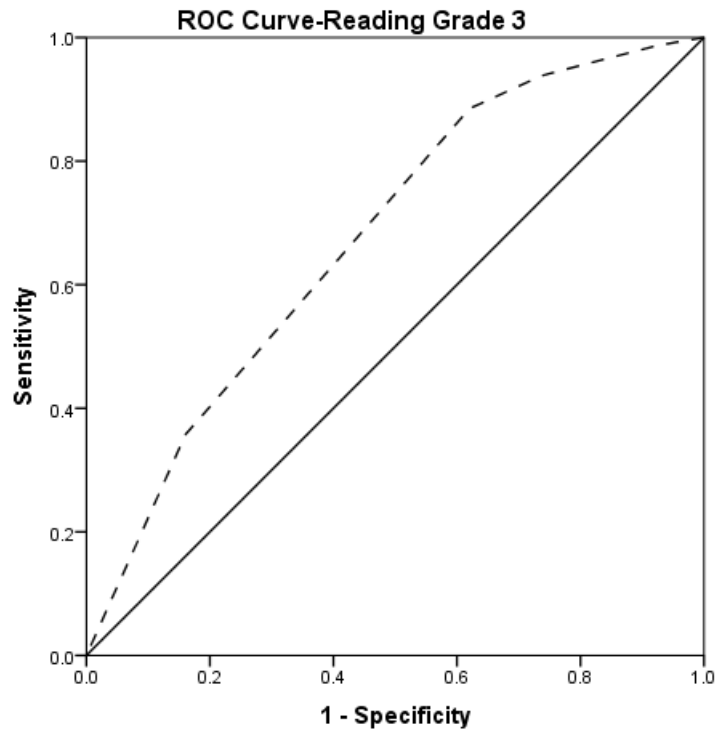
- A limitation of traditional ES measures is they only compare groups at the mean or at the proficiency cutpoint, possibly overlooking important group differences lower or higher on the score scale
- Alternatives are effect size measures based on nonparametric methods that examine group differences for all proficiency categories (see Ho & Reardon, 2012):
 - Area under the curve in Receiver Operating Curve (ROC) analysis
 - V' statistic
- Because of time constraints, we only report a few examples of ROC analysis

Whole distribution Achievement Gaps

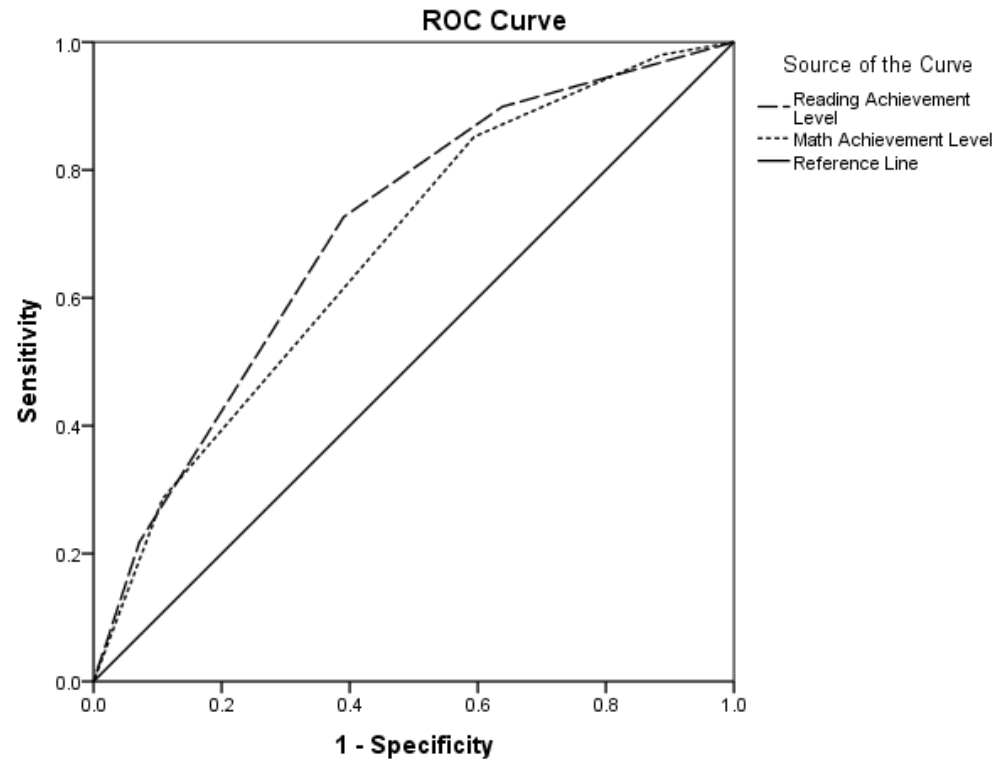
- ROC analysis (and V') uses nonparametric methods to address problems associated with characteristics of score distributions
- Advantage is estimation of gap across all proficiency levels
- ROC curve diagonal line represents no difference between reference group (SWoD) and focal group (SWD)
- Size of area between SWD group curve and diagonal is the area under the curve or the size of the difference between the two groups
- In following examples, note differences:
 - at different proficiency levels
 - for math vs. reading
 - by exceptionality subgroup



Achievement Gap for SWD vs. SWoD in Oregon Reading in Grade 3 (on left) and Grade 5 (on right)

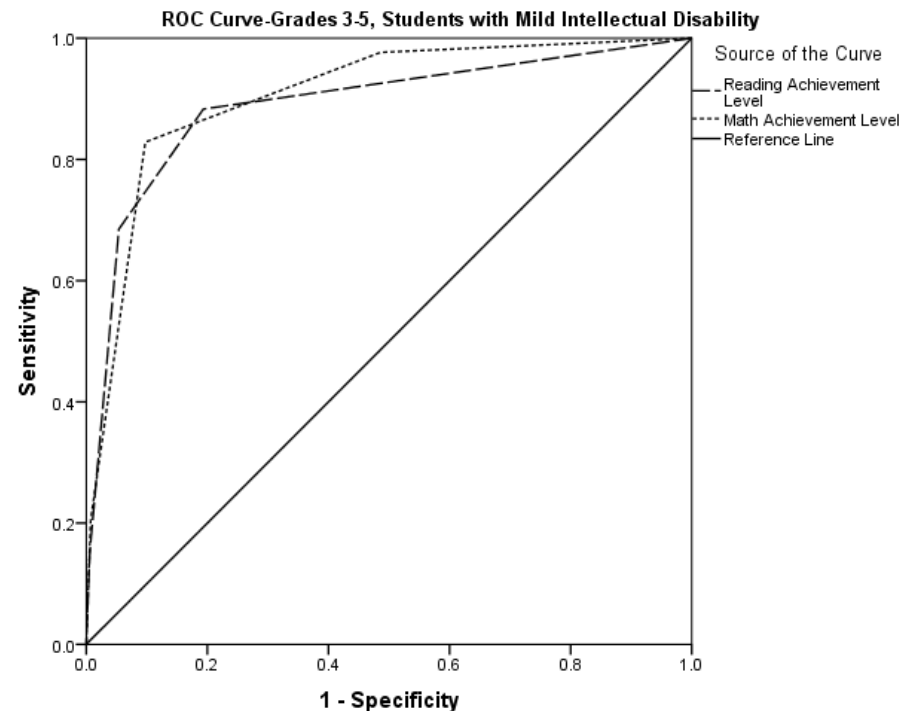
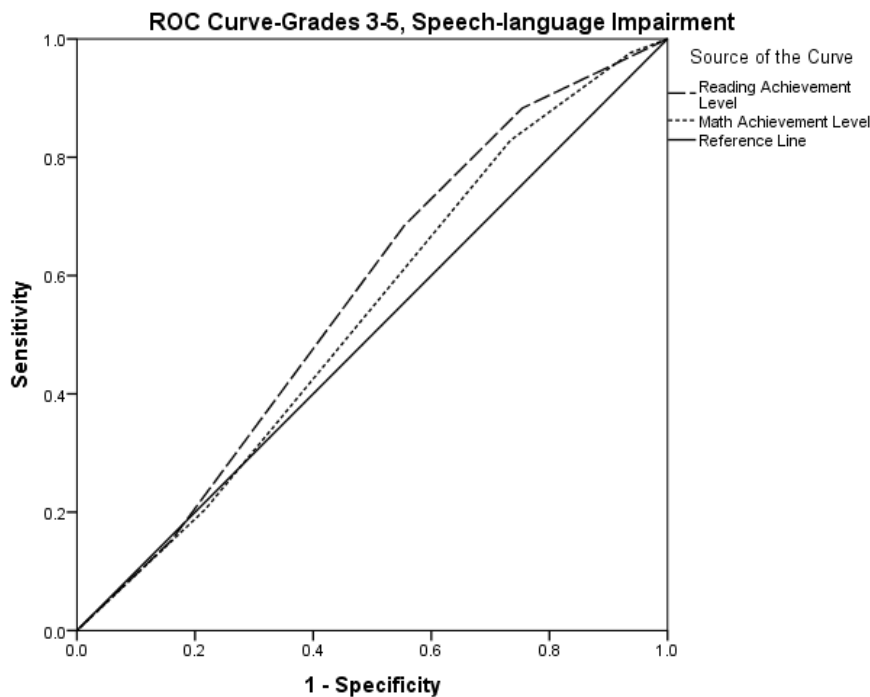


Achievement Gap for SWD vs. SWoD in NC Math and Reading Grades 3-5



iced by ties.

Achievement Gap for SWoD vs. Speech-language Impairment (on left) or Mild Intellectual Disability (on right) on NC Math and Reading Grades 3-5



Interactions of SWD status and Other Student Characteristics

- Many studies do not directly test the interaction of SWD status and factors thought to be related to student performance (e.g., LD status and sex of student)
- When these factors are included in statistical models (especially regression and HLM models), only partial regression effects not the actual interactions are analyzed
- This can be very misleading and result in incorrect interpretations as well as incomplete understanding of achievement gaps

Stevens, J. J., & Schulte, A. C. (in press). The interaction of learning disability status and student demographic characteristics on mathematics growth. *Journal of Learning Disabilities*. DOI: 10.1177/0022219415618496

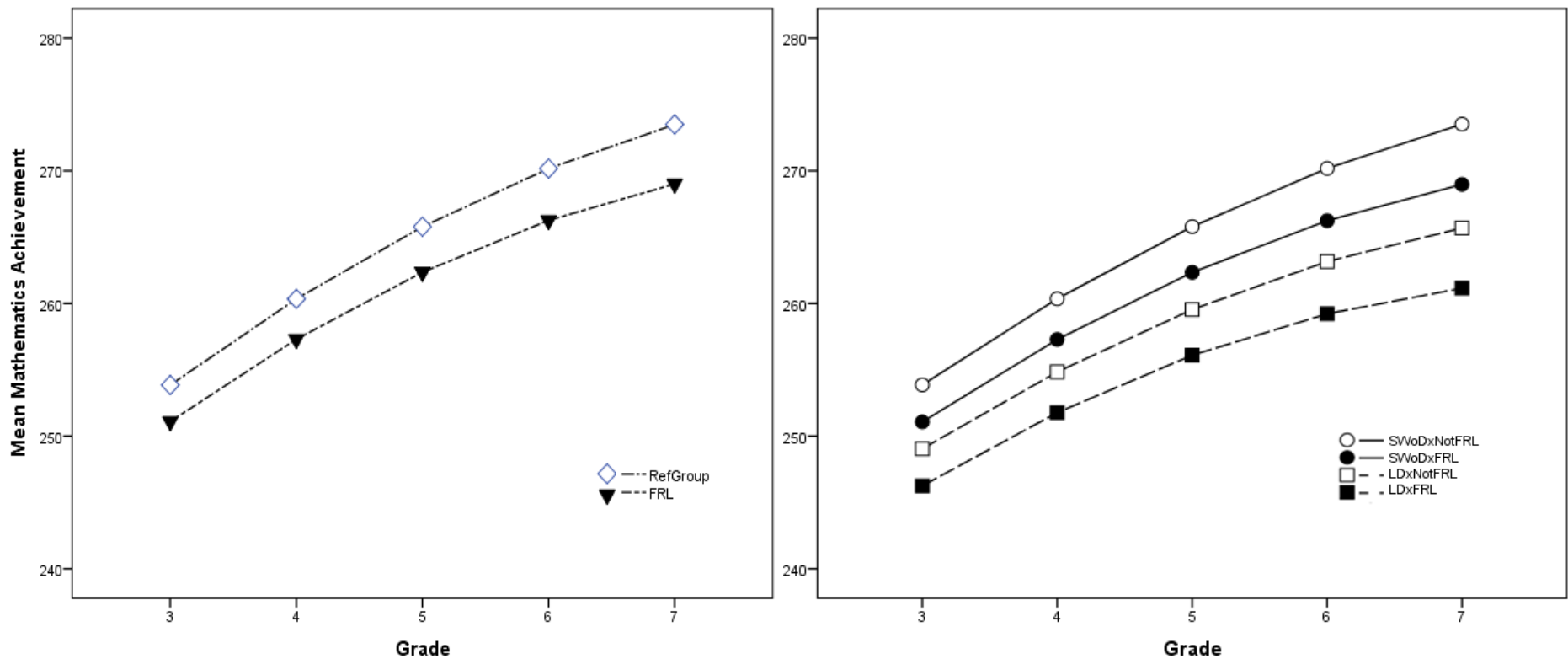
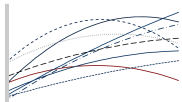


Figure. Partial regression of FRL compared to the reference group on left; three way interaction effect of LD x FRL x grade interaction on right.



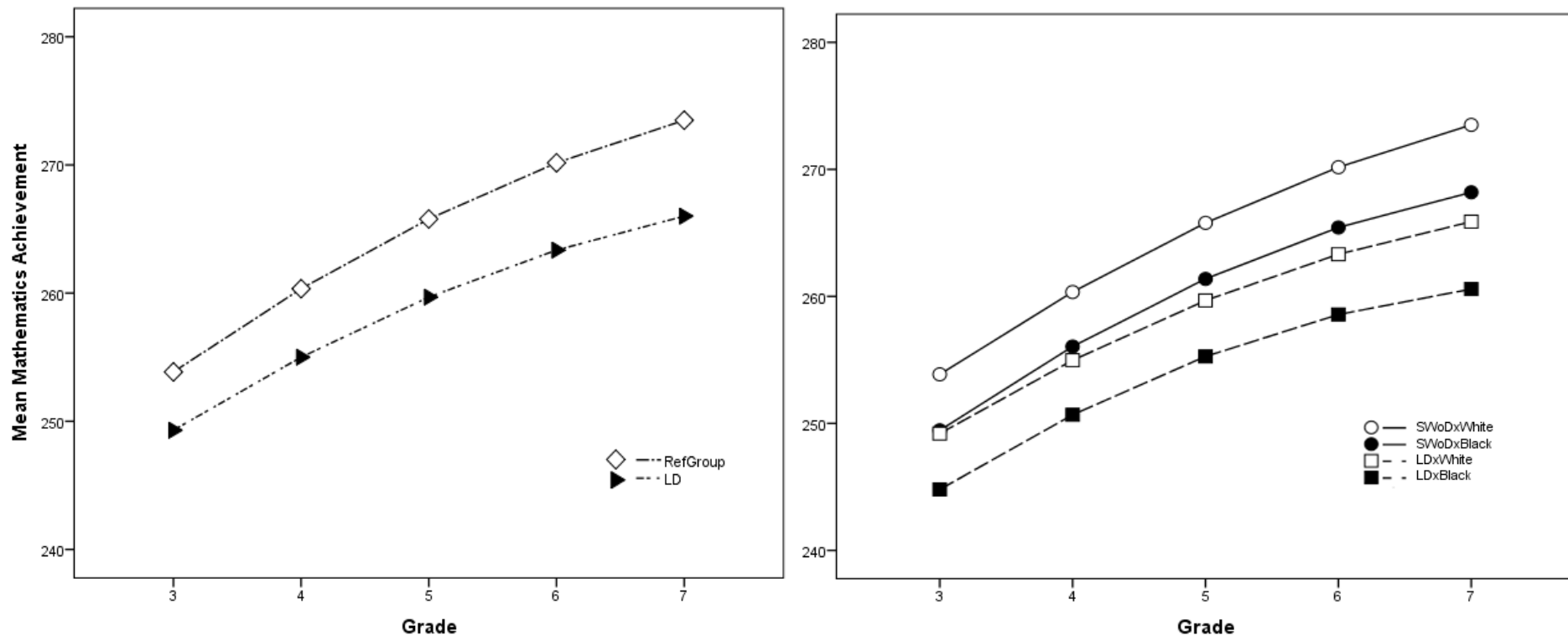


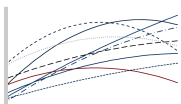
Figure. Partial regression of LD compared to the reference group on left; three way interaction effect of LD x Black race/ethnicity x grade interaction on right.

Mathematics Achievement Gaps for Elementary and Secondary Students:

The Influence of Opportunity to Learn and Special Education Status

Stephen N. Elliott

Arizona State University

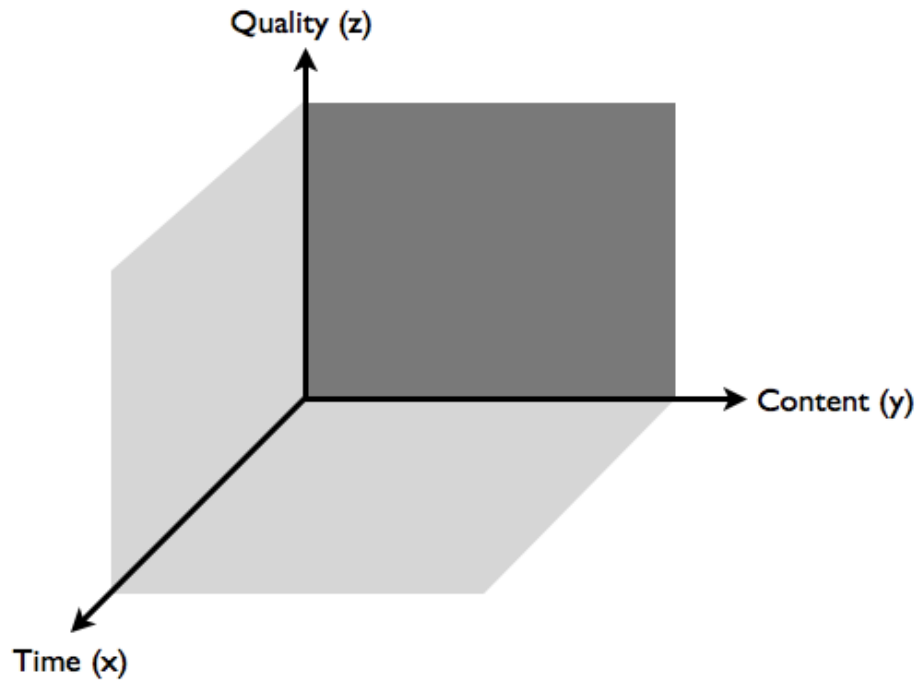


Research Questions

Specific research questions motivating the study were:

1. Do students with and without disabilities who received instruction in the same general education classrooms have an equal opportunity to learn mathematics?
2. What is the relationship among five instructional variables (characterized as OTL) and within year academic growth on an interim assessments?
3. What is the predictive relationship among five instructional OTL variables and students' end-of-year mathematics achievement?

Opportunity to Learn (OTL) the Intended Curriculum

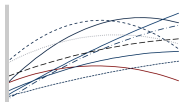


Definition: Opportunity to Learn

The degree to which a teacher dedicates instructional time and content coverage to the intended curriculum objectives emphasizing higher-order cognitive processes, evidence-based instructional practices, and alternative grouping formats.

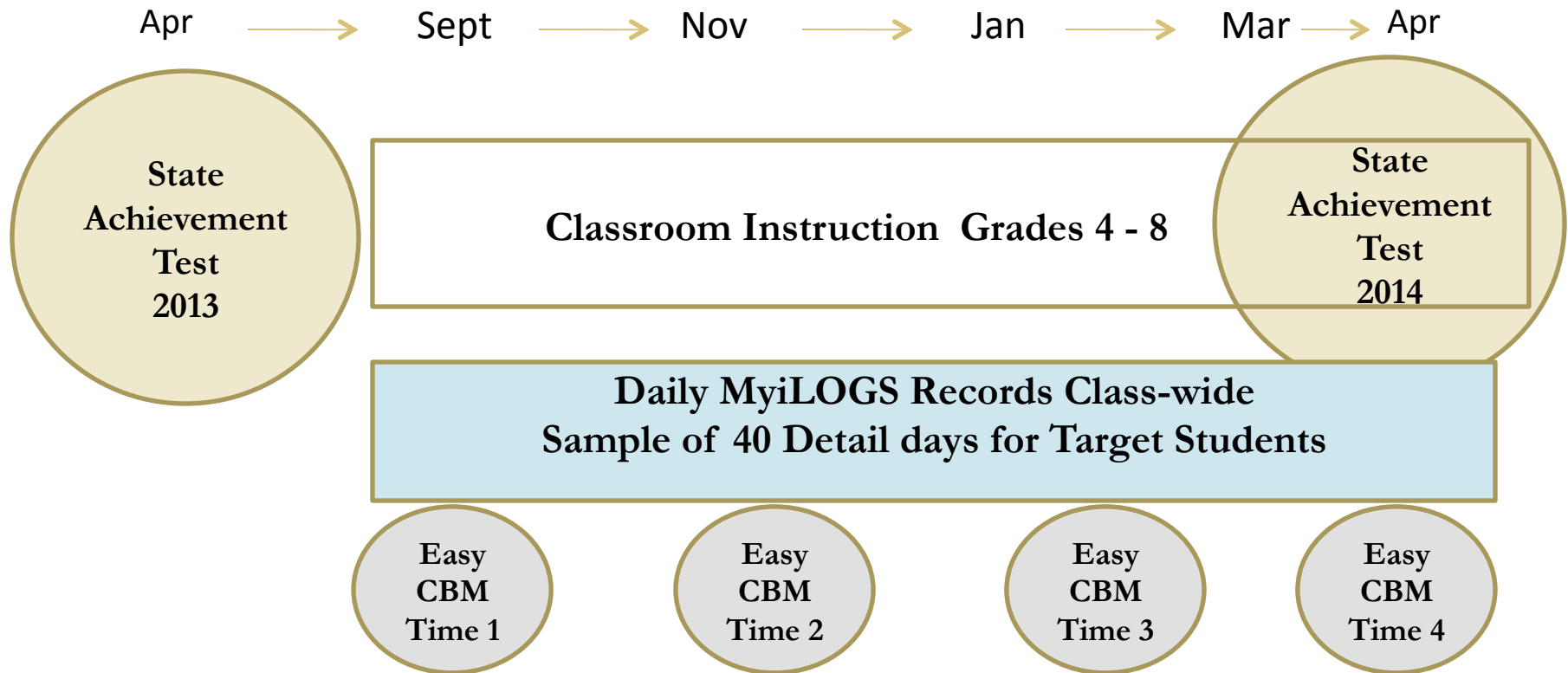
(Kurz, 2011)

**A unified conceptualization of OTL
based on 50+ years of empirical research.**

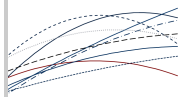


Multiple Measures Study Design*

Teachers (N = 78; AZ 49, OR 29) and students (N = 327; 162 SWD + 165 SWoD) from AZ & OR schools grades 4th-8th



*A 3-year study with longitudinal student cohorts

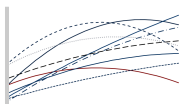


Summary of Year 1: Key Findings

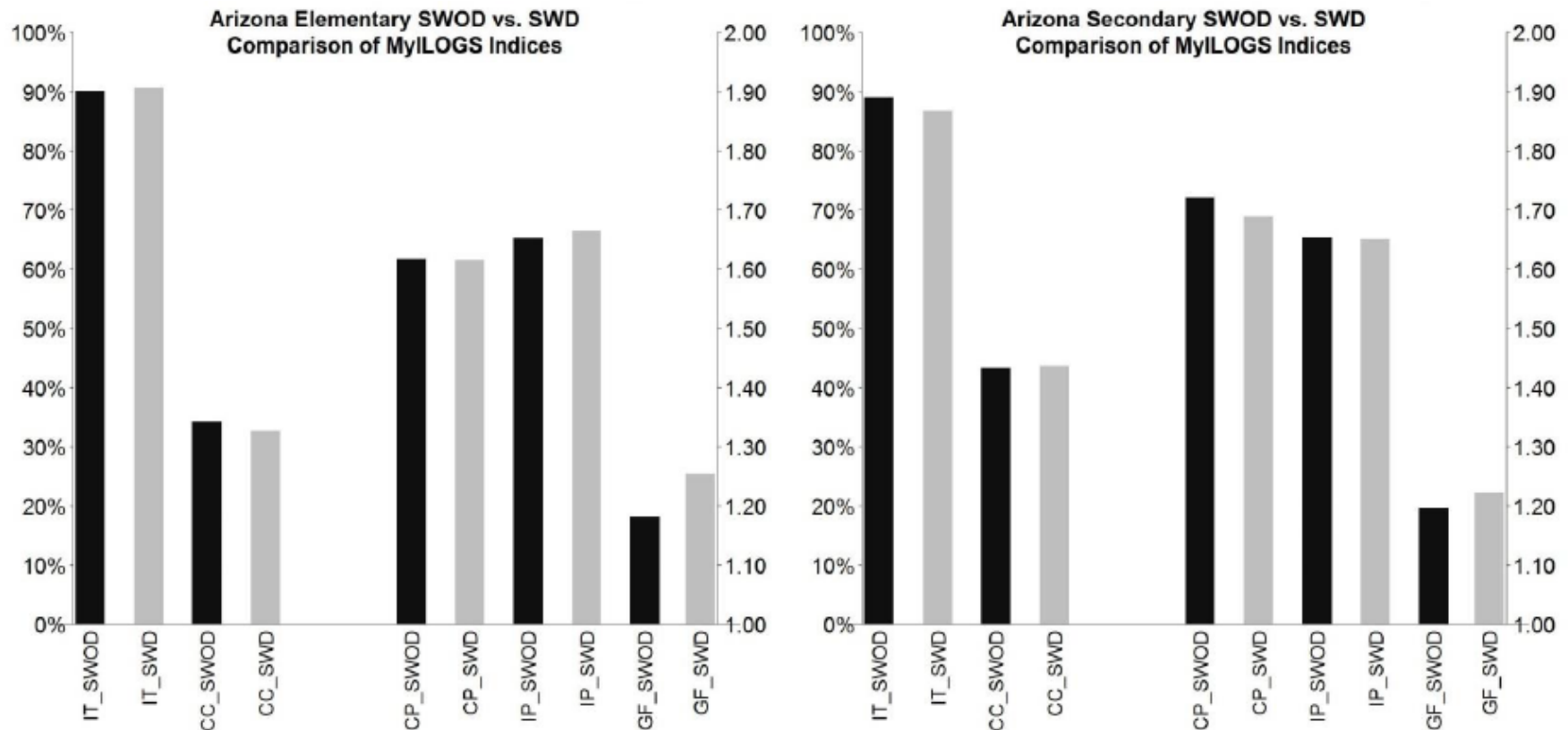
- We observed very similar instructional processes for students with and without disabilities learning mathematics in the same elementary or secondary classrooms in AZ and OR schools. Significant achievement gaps between these groups of students, however, existed on the four interim CBM assessments and the end-of-year achievement state test.
- We found that the collection of five MyiLOGS scores, along with grade level and special education status, accounted for a substantial amount (i.e., 43% to 44%) of the variance in student's end-of-year mathematics scores. A subset of OTL indices explained a statistically significant, although relatively small portion of unique variance in the end-of-year mathematics scores. The particular OTL scores found to be significant contributors varied across AZ and OR.

Year 2 Findings

- AZ teachers reported an average of 164 days and OR teachers reported 158 days of instruction; 25% of these days were Detail Days where instructional information on cognitive processes, practices, and grouping for SWD and SWOD was documented. Based on these Detail Days, we observed **very similar mathematics instructional processes for students with and without disabilities in the same elementary or secondary classrooms** in AZ and OR schools. Yet, there were **significant achievement gaps between these groups of students** on the four interim CBM assessments and the end-of-year achievement state test.
- We also found that Grade Level and Special Education Status, along with the collection of five MyiLOGS scores, accounted for a substantial amount (i.e., 30% OR, 39% AZ) of the variance in student's end-of-year mathematics scores. OTL indices explained a relatively small portion of unique variance in the end-of-year mathematics scores.
- ICCs (Teacher-Observer) for Observations on 6 random Detail Days each Year:
InstrTime = .80; CogProcess = .28; InstrPractice = .39; GroupFormat = .45

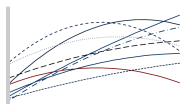


Comparison of OTL Indices for AZ Students



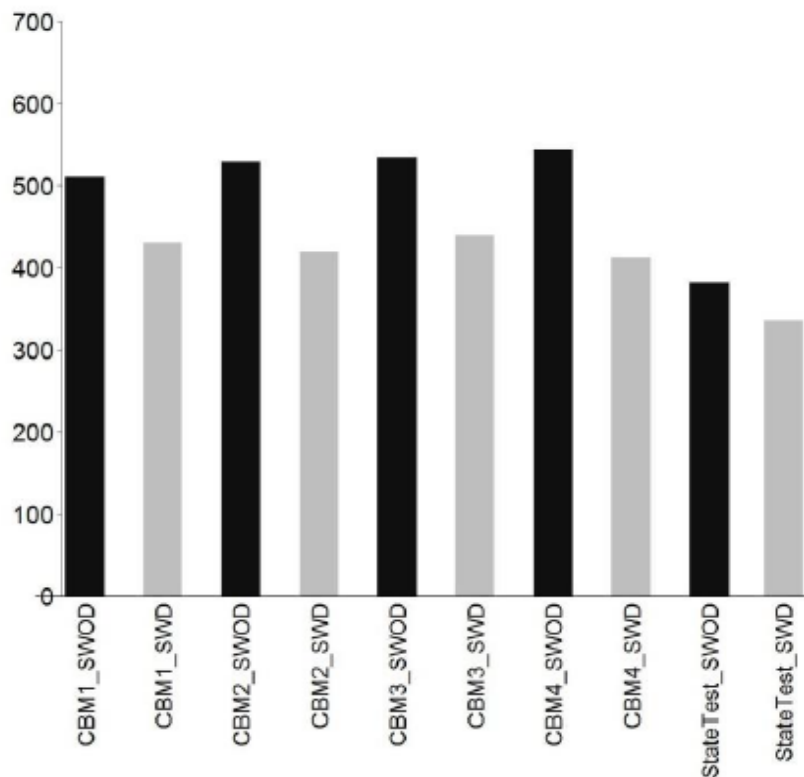
SWOD Black

SLD Gray

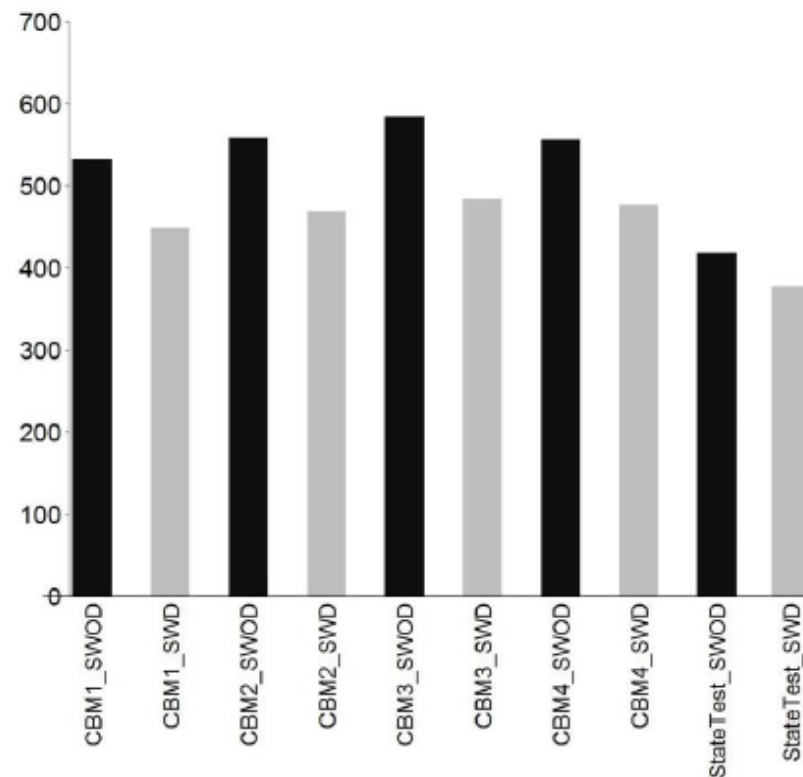


Comparison of Interim & End-of-Year Test Results for AZ Students

Arizona Elementary SWOD vs. SWD
Comparison of EasyCBM & State Test

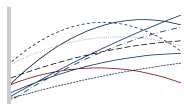


Arizona Secondary SWOD vs. SWD
Comparison of EasyCBM & State Test



SWOD Black

SWD Gray

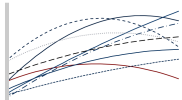
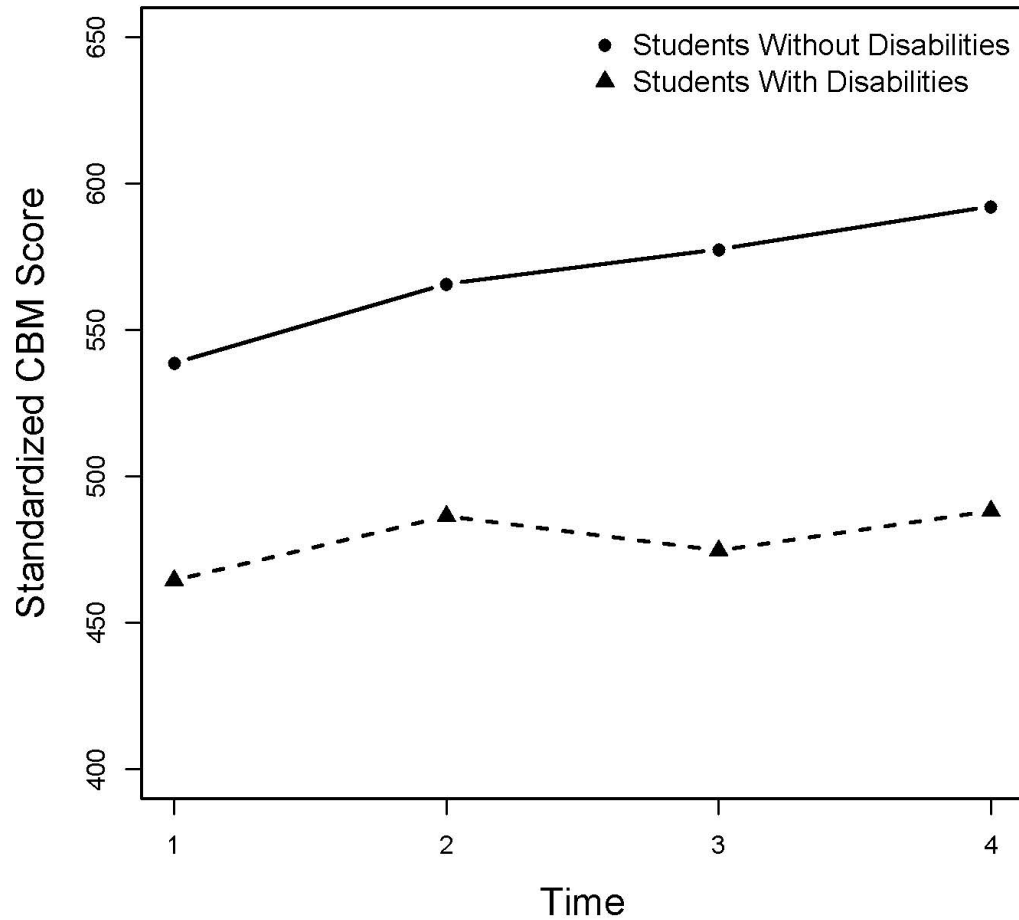


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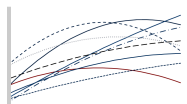
Within Year Standardized Mathematics CBM Growth



Conclusion

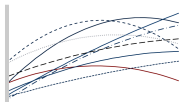
Offering students with disabilities the same amount of instruction on the same content standards in the same general education classrooms was found to offer **the same historic results—large and persistent gaps in achievement -- in comparison to students without disabilities.**

The findings in Year 2 replicated those from Year 1. Thus, it indicates that students with disabilities will need more instructional time on the intended curriculum, and perhaps more differentiated instruction to increase their rate of achievement enough to close gaps that currently exist between them and students without disabilities.



Within Year Growth

Drs. Tindal and Nese
University of Oregon



Measurement Sufficiency

- RTI within classrooms using CBMs to screen with tiers of support
- Grades 3-5 with oral reading fluency
- Benchmark and progress monitoring
- Teacher decision making on grade level of measures
- Multi-level model with
 - Time
 - Student characteristics
 - Measurement characteristics
- Major Findings

Measurement Sufficiency

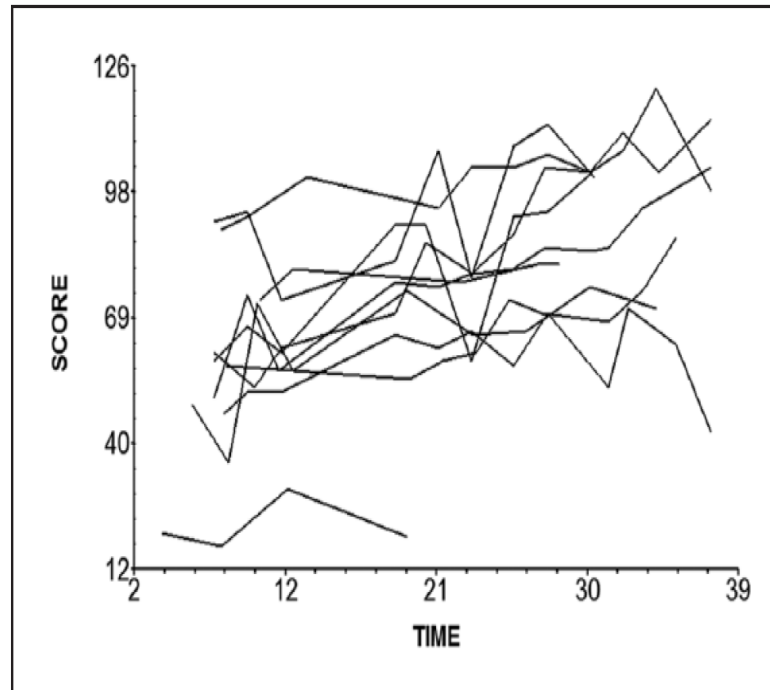
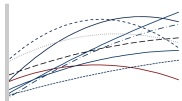


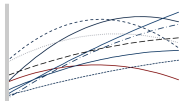
Figure 1. Plot of a random sample (.05%) of students' progress monitoring unconditional growth.



Measurement Sufficiency

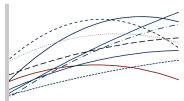
Table 4. Final Conditional Model With Special Education Status and Measurement Condition.

Fixed effect	Coefficient	SE	t ratio	df	p value
Grade 3					
Intercept, β_{00}	88.23	1.286	68.61	1278	<.001
Special education, β_{01}	-19.70	3.041	-6.48	1278	<.001
Sufficient measurement, β_{02}	-21.72	2.23	-9.74	1278	<.001
Slope, β_{10}	0.67	0.03	20.87	1278	<.001
Special education, β_{11}	0.08	0.12	0.67	1278	.505
Sufficient measurement, β_{12}	0.16	0.05	2.99	1278	.003
Grade 4					
Intercept, β_{00}	107.56	1.09	98.69	1235	<.001
Special education, β_{01}	-23.71	2.70	-8.79	1235	<.001
Sufficient measurement, β_{02}	-19.89	2.14	-9.31	1235	<.001
Slope, β_{10}	0.62	0.03	24.31	1235	<.001
Special education, β_{11}	0.01	0.05	0.16	1235	.872
Sufficient measurement, β_{12}	0.12	0.04	2.71	1235	.007
Grade 5					
Intercept, β_{00}	133.24	1.30	102.37	1094	<.001
Special education, β_{01}	-28.42	2.70	-10.53	1094	<.001
Sufficient measurement, β_{02}	-18.27	2.53	-7.22	1094	<.001
Slope, β_{10}	0.70	0.03	21.93	1094	<.001
Special education, β_{11}	-0.08	0.05	-1.53	1094	.126
Sufficient measurement, β_{12}	-0.10	0.05	-2.03	1094	.043



30 Years of Research on ORF

- Initial findings in 1984 – Pine County Norms
- Fuchs and Deno estimates of growth
- Deno and Shin estimation with a 'national sample'
- Recent studies with large data sets
 - Hasbrouck and Tindal publications
 - Florida studies
 - easyCBM



30 Years of Research on ORF

Table 1

Comparison of Studies on Research Variables Referenced in Studying Growth of Reading Fluency



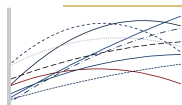
Authors (date)	Grades	Students	Measures	N-Measures	Slope Calculation	Averages and Growth – WCPM			
Marston, Lowry, <u>Deno</u> , & <u>Mirkin</u> (1981)	Grades (n):	Students from	Third grade	Three	Plot of raw scores and	Grades	Fall	Winter	Spring
	1 (13), 2 (9),	a small	basal reading	administrations:	percentage of change;	1	18.1	31.1	45.7
	3 (10), 4 (7),	Midwest city	series: <u>Allyn-</u>	Fall, winter, and	significance tests of	2	73.2	101.1	127.8
	5 (7), 6 (9)		<u>Bacon, Ginn</u>	spring	differences	3	108.3	123.6	136.2
			720, Houghton			4	125.4	131.7	155.3
			Mifflin			5	125.7	147.3	161.1
						6	142.9	176.7	182.8
Tindal, <u>Germann</u> , Marston, & <u>Deno</u> (1983)	Grades (n):	Students from	Curricula in use	Three	Change in raw score	Grades	Fall	Winter	Spring
	1 (5), 2 (13),	six districts	in the school	administrations:	and in discrepancy	1	6.8	14.3	9.6
	3 (17), 4 (22), 5	referred,	district	Fall, winter, and	from general education	2	5.5	16.2	23.6
	(18), 6 (21)	assessed, and		spring		3	20.5	36.6	41.3
		eligible for				4	31.0	50.3	52.9
		special				5	59.1	72.9	79.1
		education				6	59.5	65.9	66.8
Tindal, <u>Germann</u> , & <u>Deno</u> (1983)	Grades (n):	Students	Two passages	Three	Change in raw scores	Grades	Fall	Winter	Spring
	1 (276)	randomly	sampled from	administrations:		1	5	63	75
	2 (284)	sampled from	basal reading	Fall, winter, and		2	35	83	93
	3 (302)	six districts in	curriculum	spring		3	67	89	108
	4 (294)	Pine County				4	98	111	128
	5 (315)					5	121	120	138
	6 (328)					6	123	126	134
<u>Fuchs</u> , <u>Deno</u> , & <u>Mirkin</u> (1984)	Grades 3-5:	All students	3 rd grade	Pre-post (unknown:	Pre-post difference @	Condition*	Pre	Post	
	64 students in	were	passage reading	sometime between	28 weeks	Experimental*	41.6	70.2	
	(DBPM)	‘handicapped’	test from <u>Ginn</u>	Nov. and May)		Contrast	51.5	51.3	
	77 students in		720			<i>DBPM versus none</i>			
	no DBPM								

30 Years of Research on ORF

Authors (date)	Grades	Students	Measures	N-Measures	Slope Calculation	Averages and Growth – WCPM			
Al Otaiba, Petscher, Pappamihel, Williams, Dyrlund, & Connor (2009)	Grades 2- 3: 5,004 Latino students	(a) Proficient in English, (b) Not proficient and receiving English as a second language (ESL) services, and (c) Proficient enough to have exited from ESL	DIBELS (Good & Kaminski, 1996)	Four administrations: first 20–30 days of school (Sep); between the 65th and 75th days of school (Nov); between the 110th and 120th days of school (Feb); and between the 155th and 165th days of school (Apr).	Two level HLM with growth centered on the first testing time in third grade (September) and student characteristics (i.e., language group and special education subgroup) were entered at Level 2 designed to model both second- and third-grade growth trajectories	Grade 2 Weekly Growth:			
							Proficient	ESL	ESL-exit
						GE	1.2	.8	1.2
						LD	1.2	.8	1.1
						SL	1.3	.9	1.1
						Grade 3 Weekly Growth:			
							Proficient	ESL	ESL-exit
						GE	1.2	1.1	1.3
						LD	1.3	1.1	1.3
						SL	1.2	1.1	1.2
Ardoin & Christ (2009)	Grades 2-3: 28 and 40 respectively	Race and free or reduced-price lunch reported for students in each of two schools	Three passage sets: (a) FAIP-R, (b) AIMSweb, and (c) DIBELS	12 weeks (with approximately 2 administrations per week) with time of year unknown	Ordinary least square (OLS) regression used to calculate an intercept and slope for each passage set		Intercept	Slope*	
						FAIP-R	84.8	1.4	
						AIMSweb	94.3	1.8	
						DIBELS	100.4	.5	
						*Slope times 7 to estimate weekly gain			
Crowe, Connor, & Petscher (2009)	Grade 1 (n=9,993), Grade 2 (n=9,869), Grade 3 (n=10,141)	Lower SES (eligible for free or reduced price lunch) and non lower SES students	DIBELS (Good & Kaminski, 2002).	Two administrations in September and April	Hierarchical Linear Modeling (HLM) to estimate mean growth trajectories for curriculum interacting with SES over the school year (7 months)		Sept.	April	
						Grade 1	18	50	
						• Higher SES	25	63	
						• Lower SES	16	46	
						Grade 2	53	89	
						• Higher SES	63	77	
						• Lower SES	50	62	
						Grade 3	73	101	
						• Higher SES	85	113	
						• Lower SES	69	97	
						*Curriculum differences reported that interacted with students SES.			

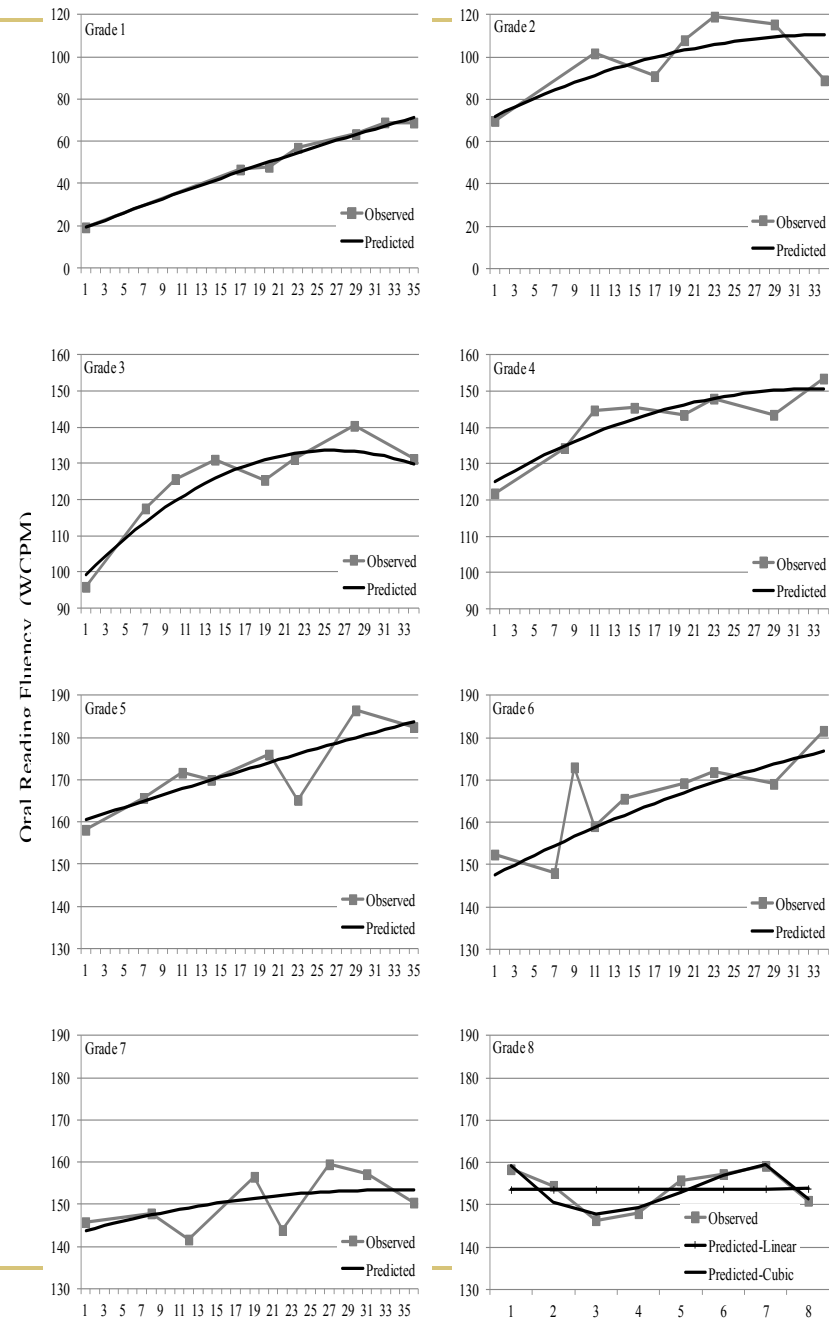
PRF WITHIN all Grades

- Rates of growth in research and aim lines in practice are used to characterize student growth; in either case, growth is generally defined as linear, increasing at a constant rate over time.
- Linearity assumption may be inaccurate.
- We examined ORF growth within-year for students in Grades 1-8.
 - Other research limited by using only 3 testing occasions.
 - Our sample included Grade 1 to 8 students, drawn from the full range of abilities within each grade level and assessed up to 8 times per year.

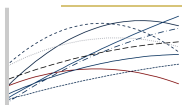


- Comparing the trajectories across grades, we found that a decelerating growth curve best described ORF data.
- On average, across grades, students exhibit a decrease in growth across the year.

Nese, Biancarosa, Cummings, Kennedy, Alonzo, Tindal. In search of average growth: describing within-year oral reading fluency growth for grades 1-8.
Journal of School Psychology.

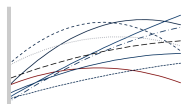


Time (grades 1-7 average weeks, grade 8 months)



PRF within GRADE 4

- The purpose of this study was to demonstrate ways to model nonlinear growth using three testing occasions: fall, winter, and spring passage reading fluency benchmark assessments.
- 2,100 Grade 4 students.
- Passage reading fluency (PRF)



PRF Within GRADE 4

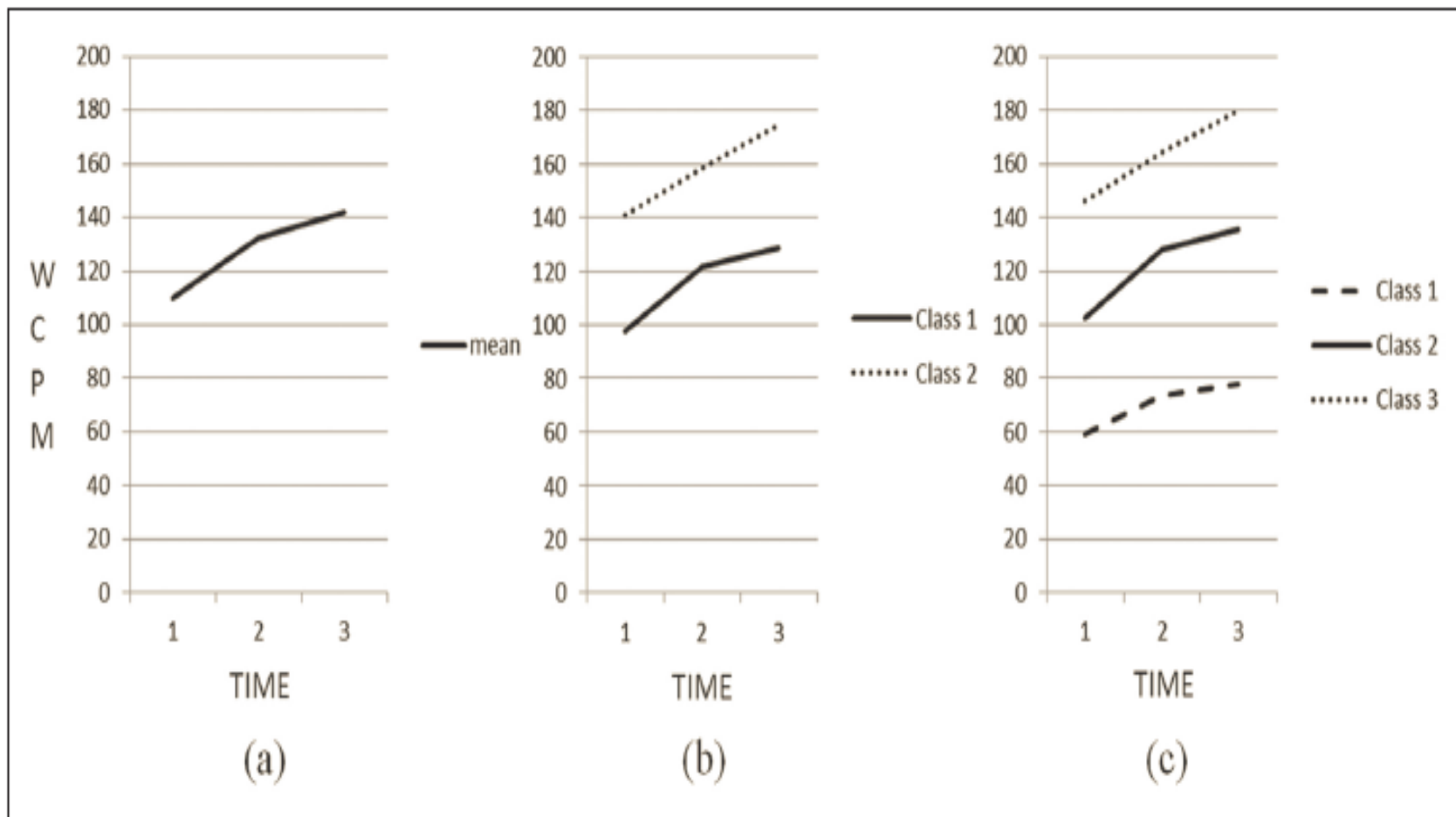
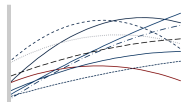


Figure 3. Estimated growth patterns for 1-class PGM and 2-class and 3-class piecewise growth mixture models (PGMM).



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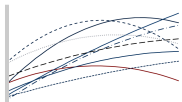
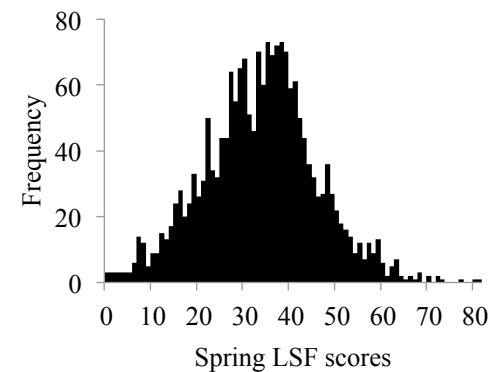
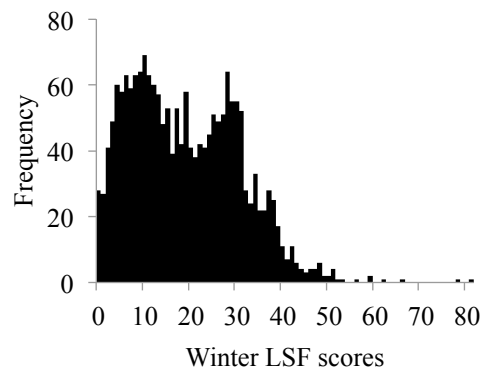
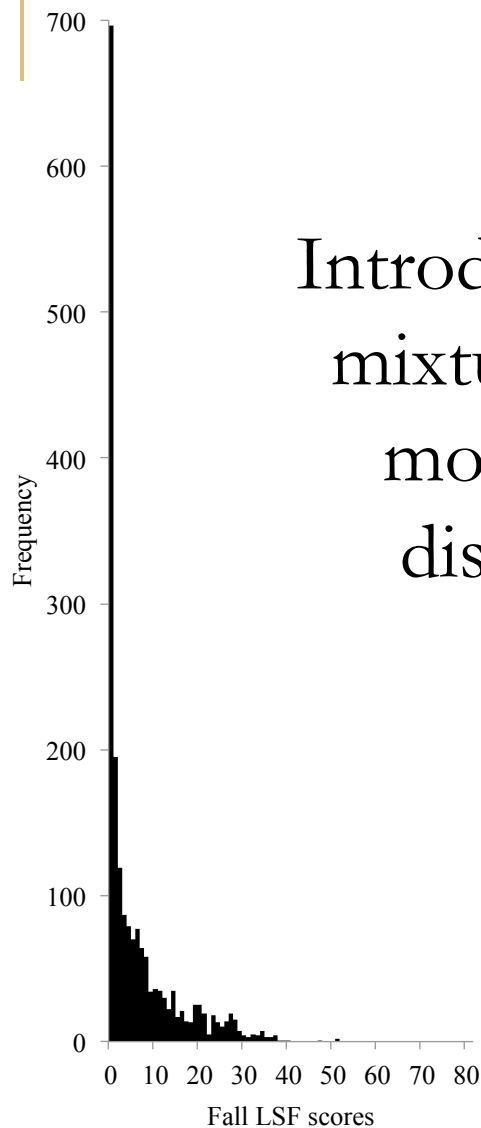
National Center on Assessment and
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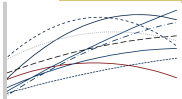
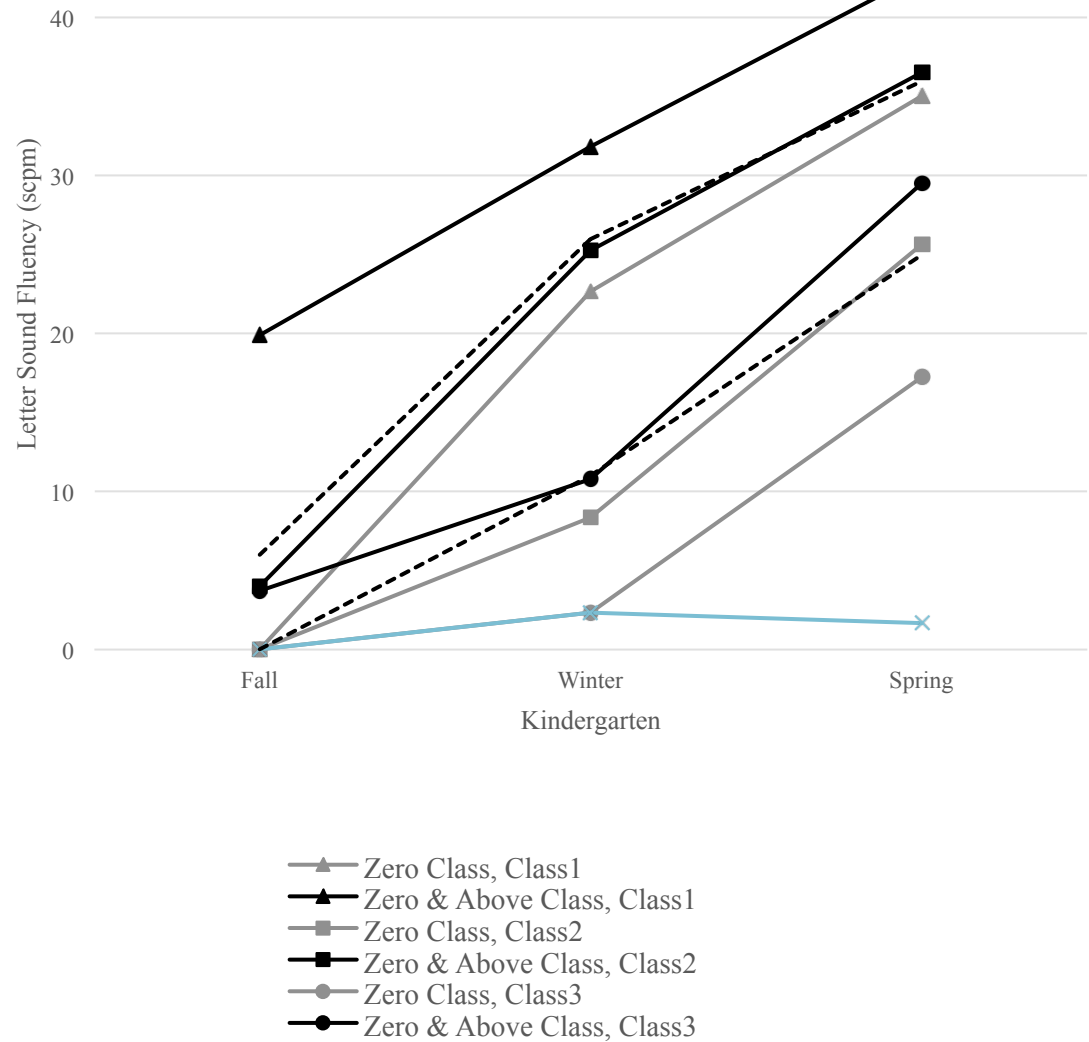
Kamata, Nese, Patarapichayatham, Lai. Modeling
Nonlinear Growth With Three Data Points:
Illustration With Benchmarking Data. *Assessment
for Effective Intervention*.

Purpose

Introduce and apply a two-step growth mixture model (GMM) approach for modeling repeated measures with distributions changing over time.

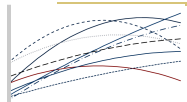


Results



Growth Models for School Effects

Joseph Stevens
University Oregon

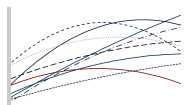


Models Evaluating School Level Growth

- Comparison of different models of estimating school performance using OR, AZ, NC, and PA state data
- Models include:
 - status; gain and residual scores
 - transition matrix models
 - value-added models
 - Student Growth Percentiles (SGP)
 - Hierarchical linear growth models (HLM)
- Model variations include: (a) three cohorts; (b) two grade levels, elementary (Grades 3-5) and middle (Grades 6-8), and (c) unconditional vs. conditional models (school size, student composition of school)

Some Preliminary Results

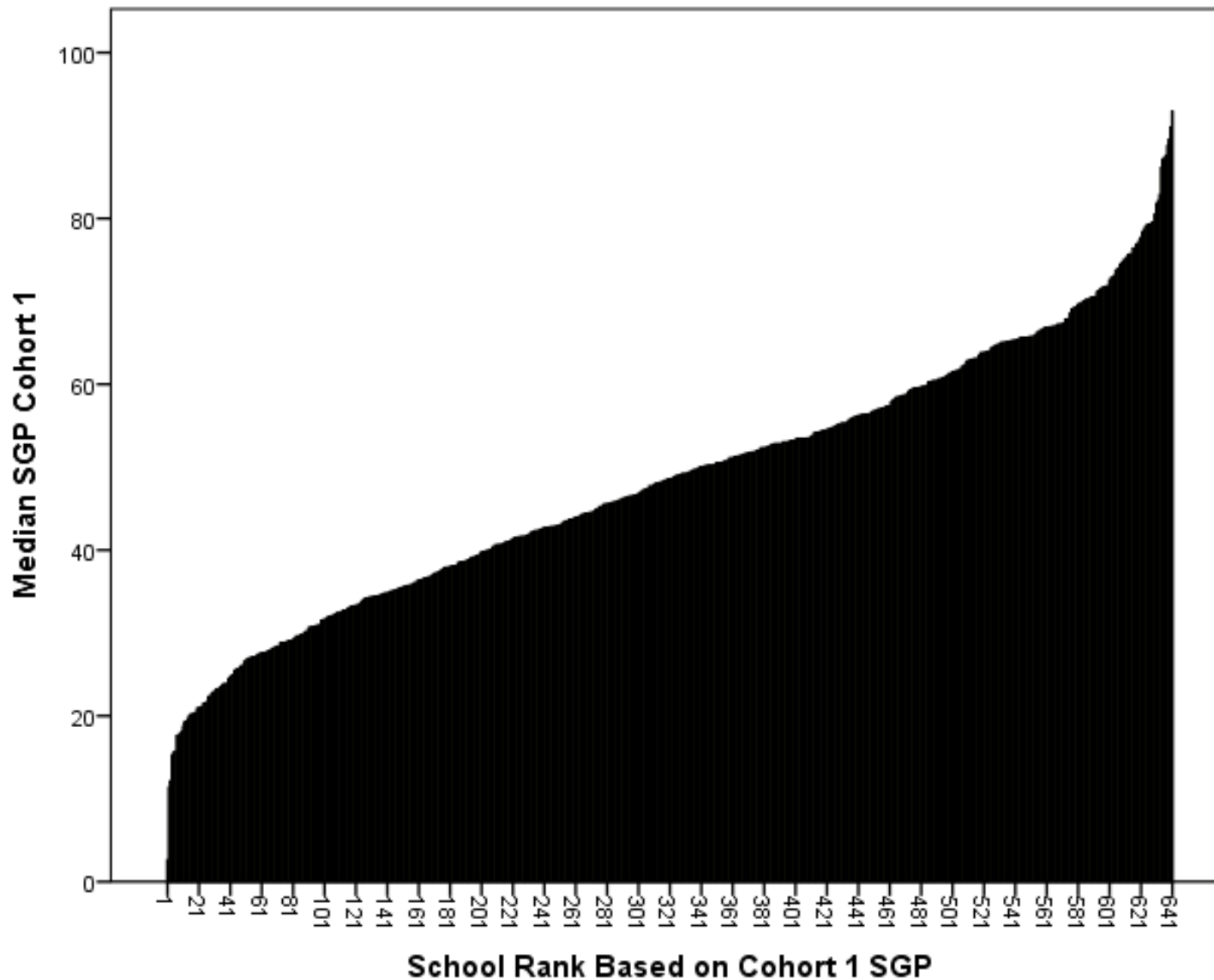
- Estimates of school effects across models sometimes very consistent, but other times very discrepant:
 - Most transition matrix models moderately to highly correlated with each other
 - SGPs almost perfectly correlated with conditional regression
 - Low to moderate correlation of SGPs with HLM growth
 - Low correlation of status models with HLM Growth
- Substantial cohort instability in the first state studied (Oregon)
- Several examples follow

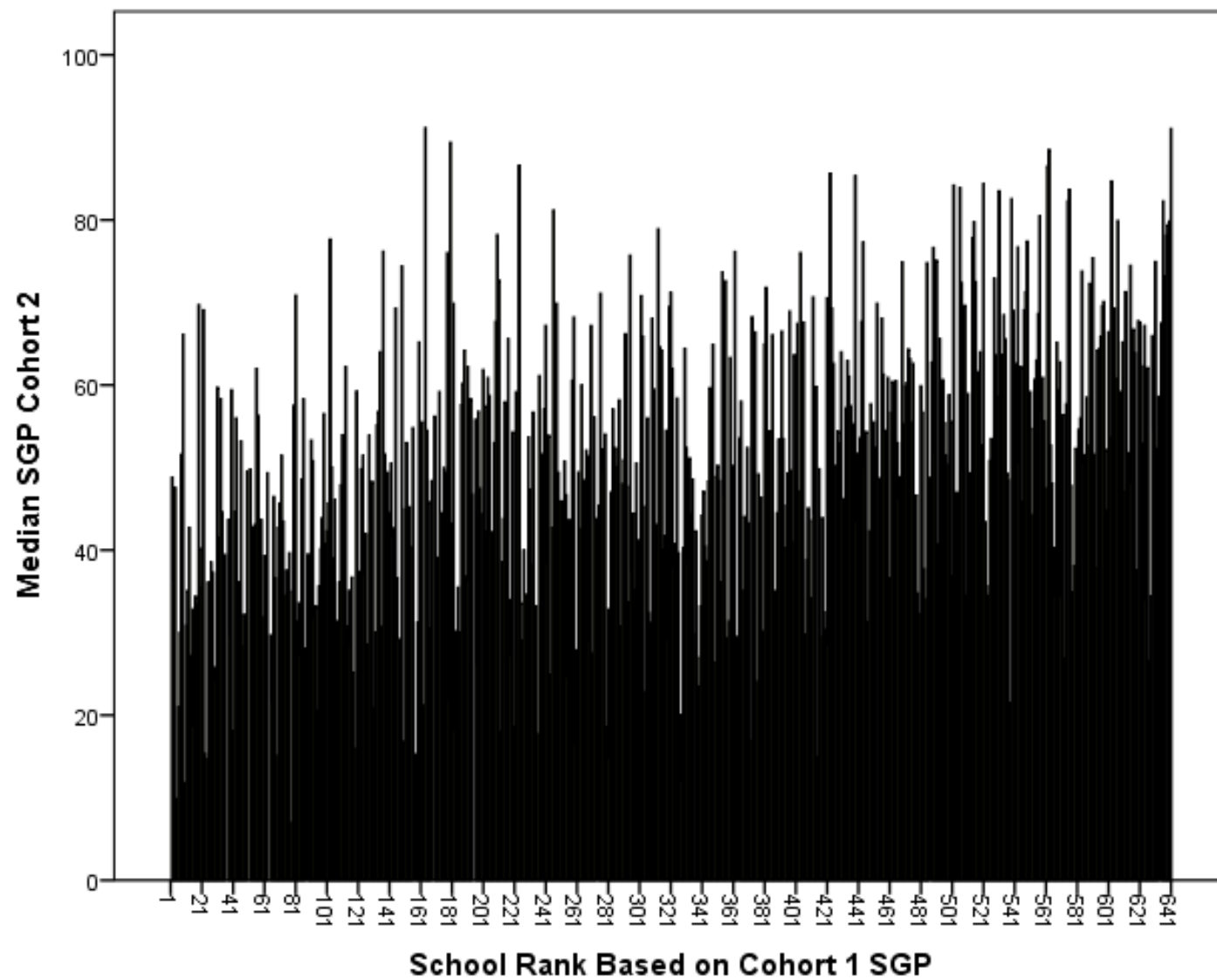


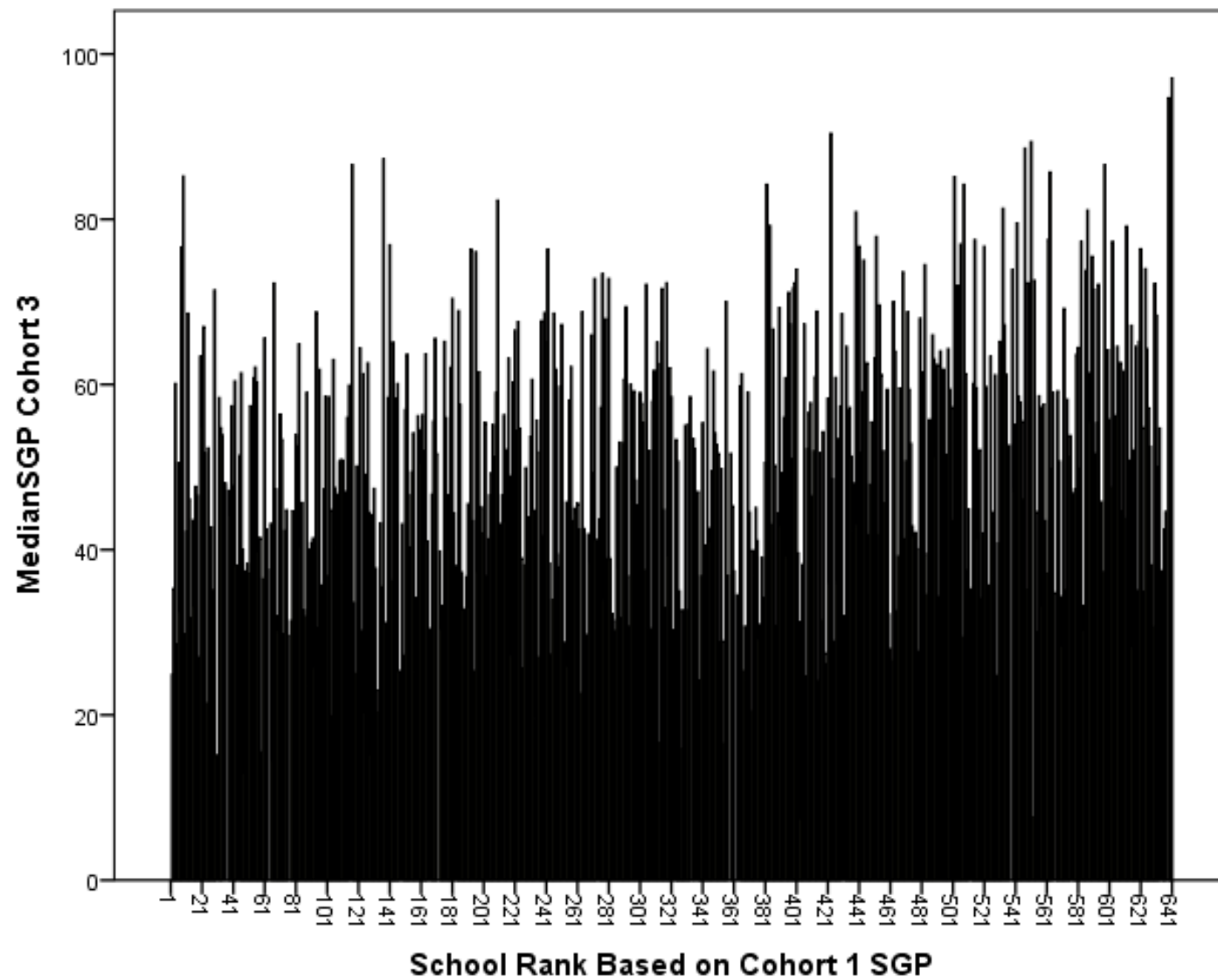
Student Growth Percentiles (SGP)

- Described as a Relative Growth Model
 - Current year performance conditioned on prior year(s) of performance using quantile regression
 - Relative rank in a distribution of those who had similar scores in previous years
- Oregon sample composed of all those who had a math or reading-language score in 2011 and at least one prior year score in years 2008-2010
- Currently popular approach in state accountability models used in dozens of states
- Result illustrated here is cohort instability using SGPs

Cohort Differences in SGP Models



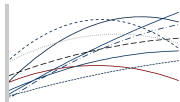
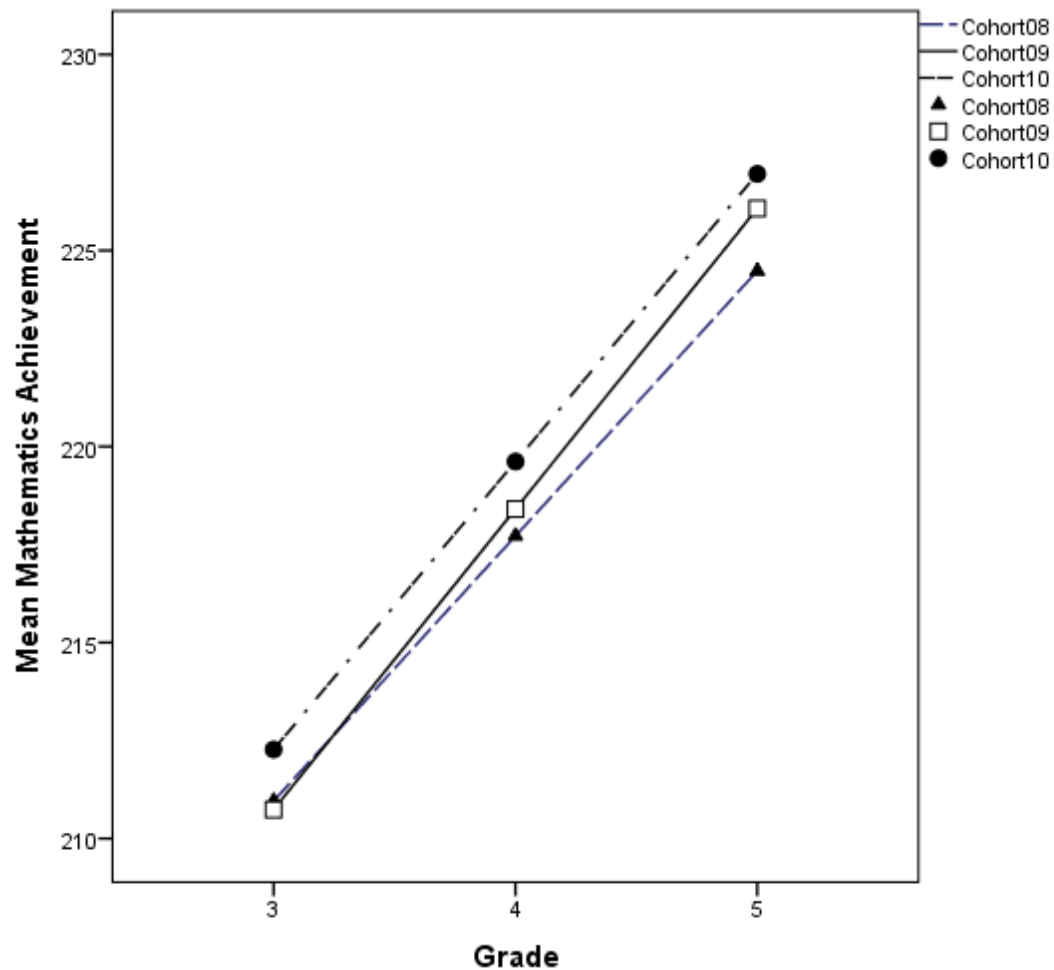




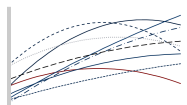
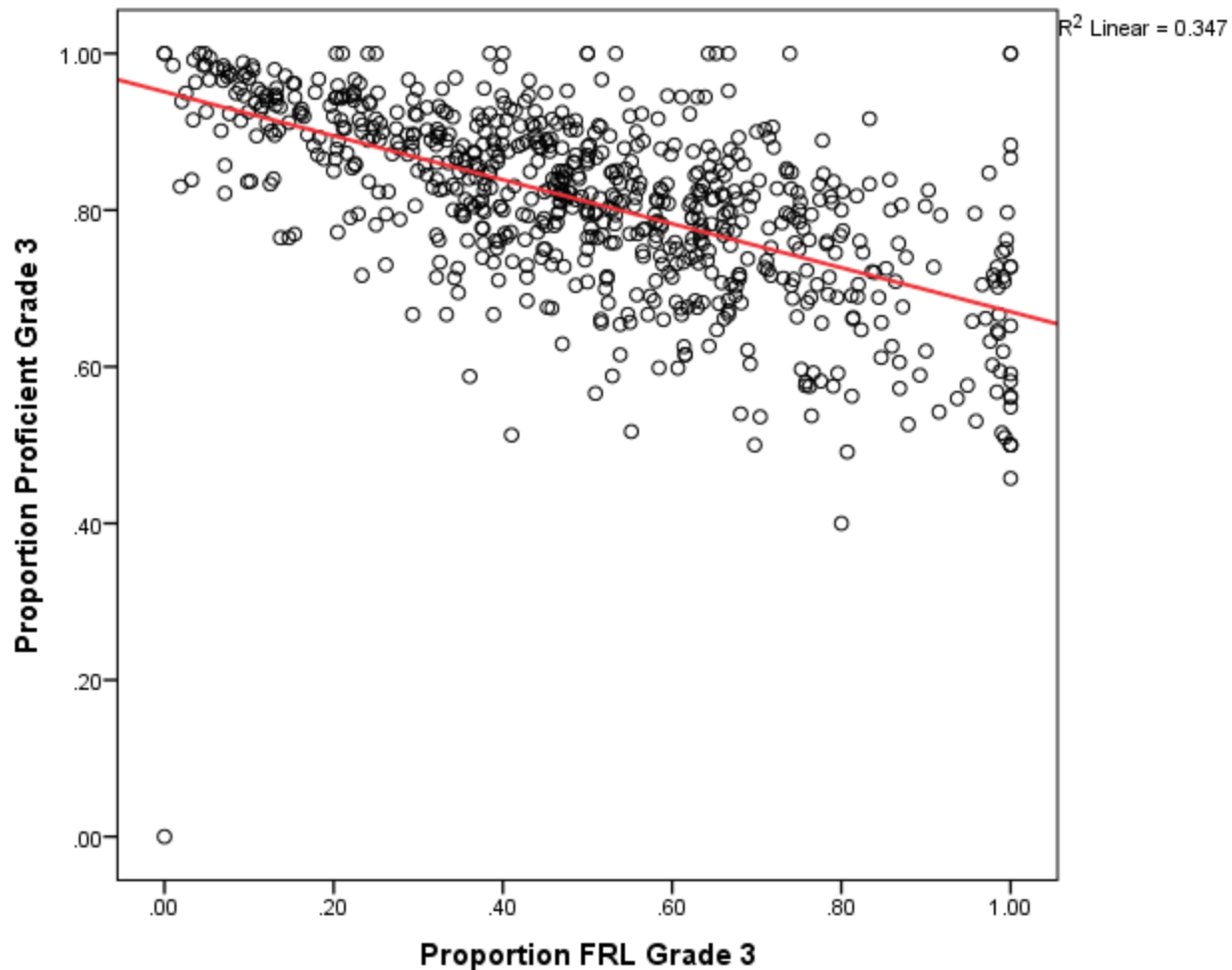
Hierarchical Linear Models

- Another alternative representation of student growth rests on the statistical modeling of change over time
- These models are absolute growth models in that they relate change to a time function and maintain the metric of the score scale
- Therefore a vertically linked score scale is necessary
- HLM models produce two primary outcomes, intercept (very similar to other status measures like school mean or PP) and growth estimate (slope)
- HLM results differ substantially from some other models especially status models (e.g., percent proficient)
- Illustration of cohort instability with HLM follows

Cohort Differences in HLM Growth Models

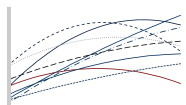


Covariate Relations with School Estimates

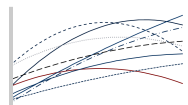
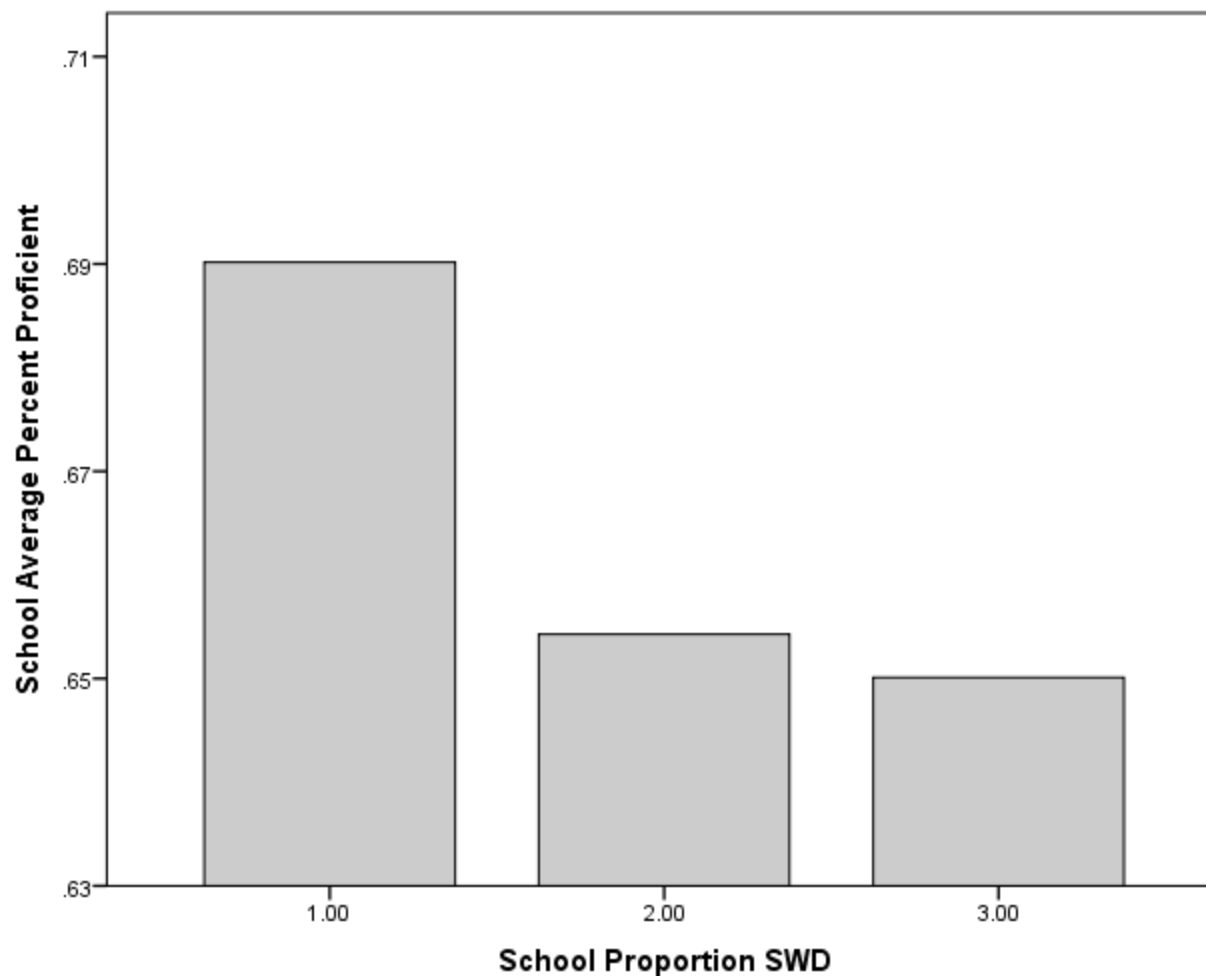


Relations of Model Estimates to School Proportion SWD

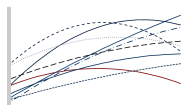
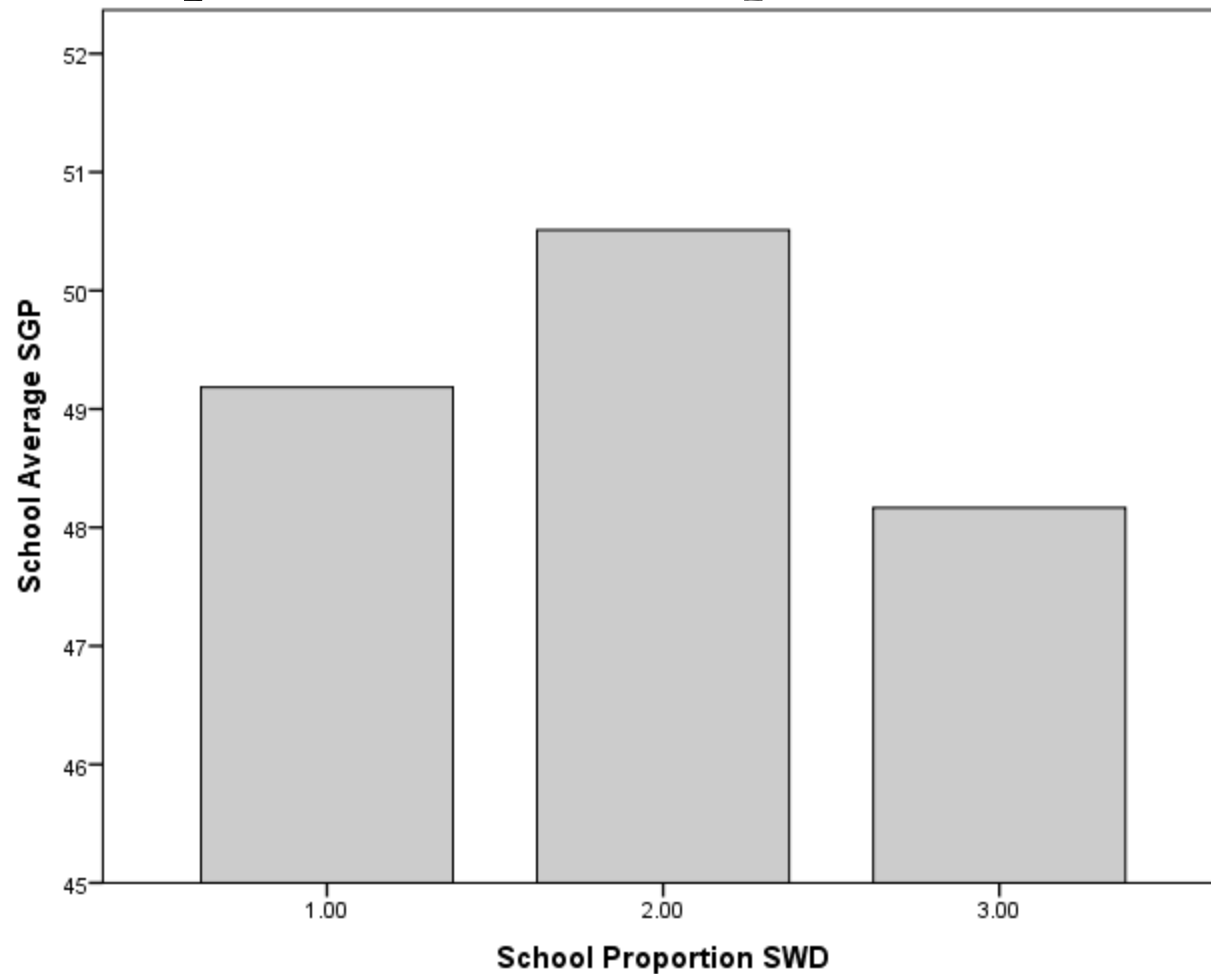
- School proportion SWD very positively skewed; variable was categorized for comparisons: Group 1, schools with 10% SWD or less; Group 2, 11-14% SWD; and Group 3: more than 14% SWD
- Significant differences by School SWD group for PP and HLM intercept; not statistically significant for SGP, HLM Slope



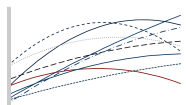
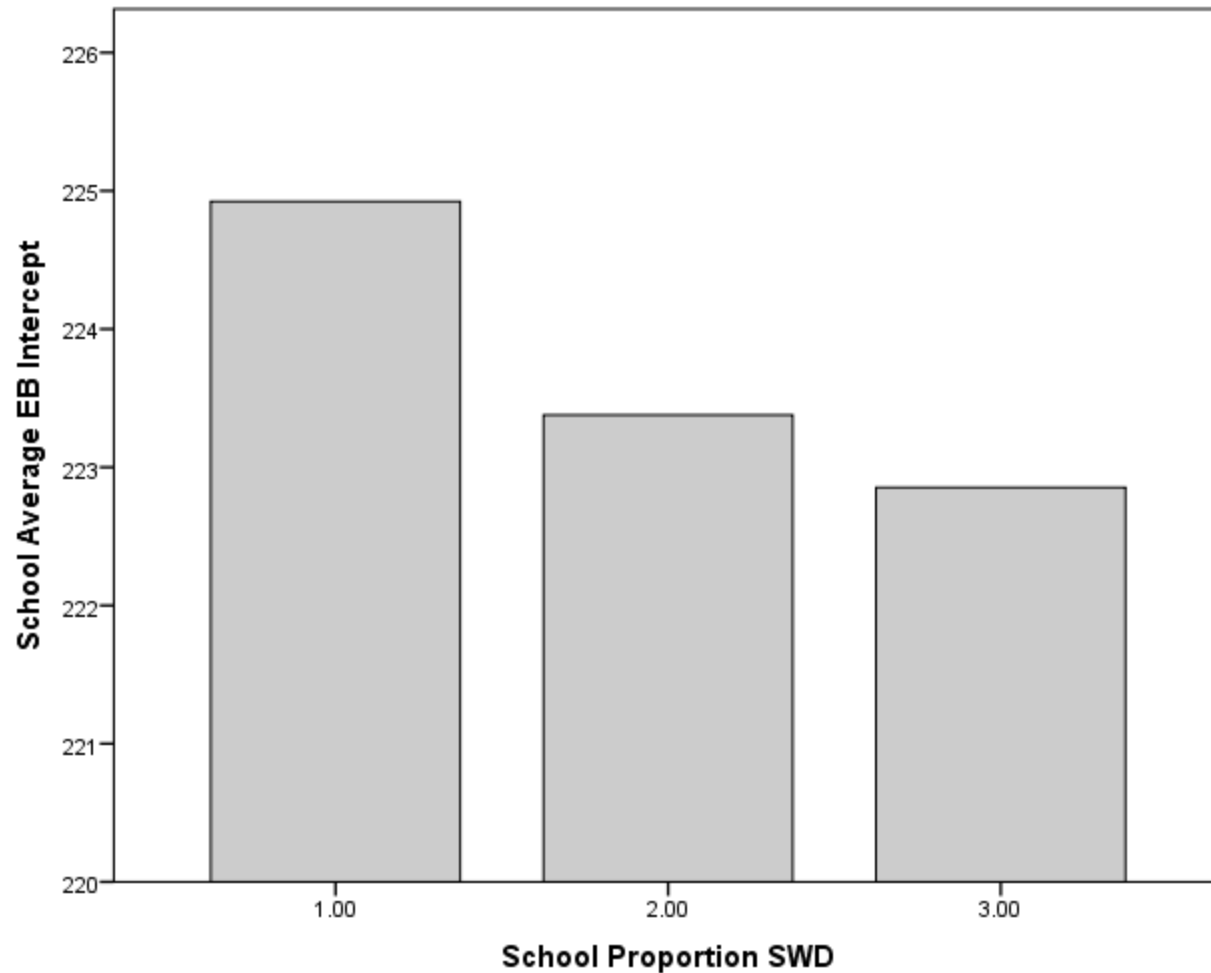
Relation of School Percent Proficient with School Proportion SWD ($p = .018$)



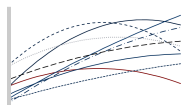
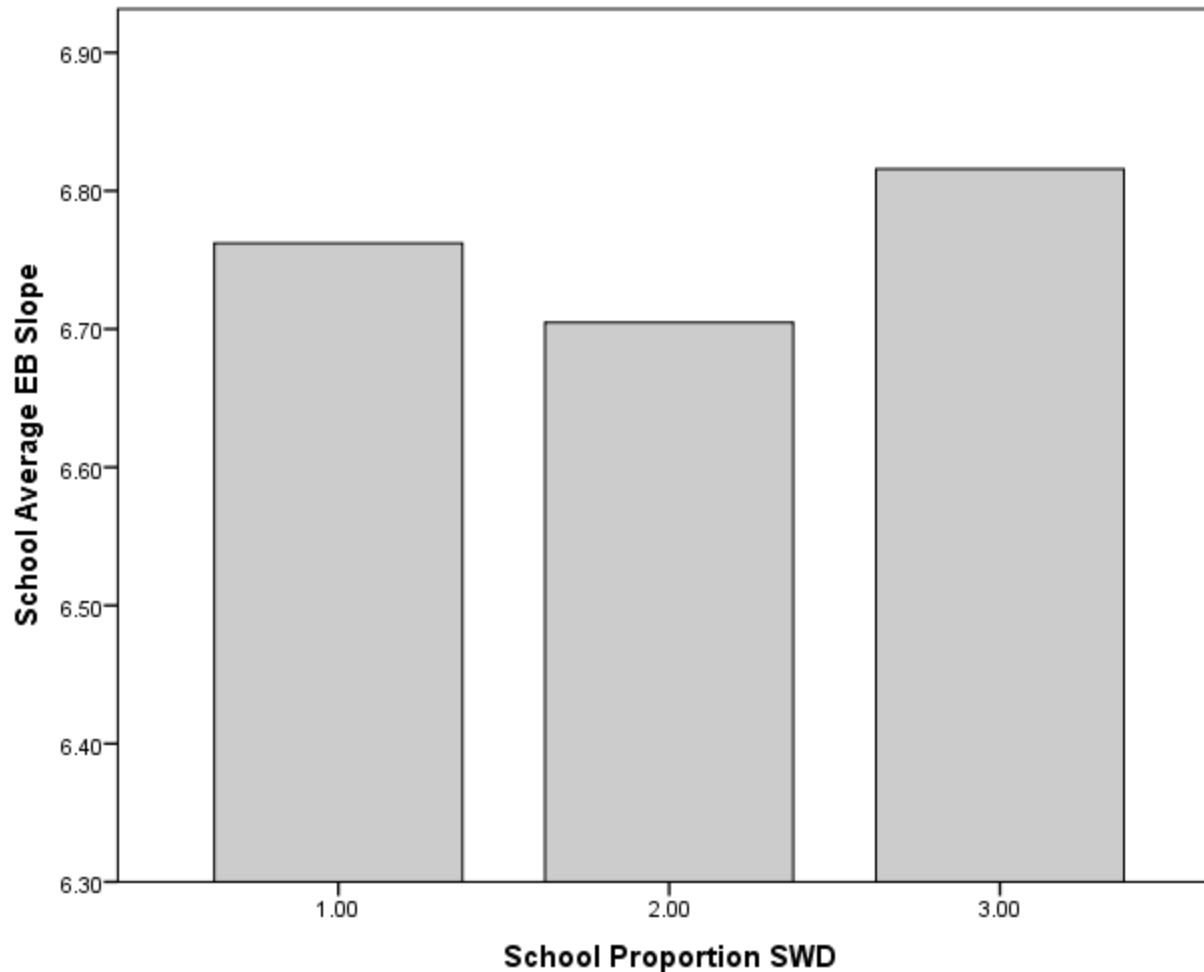
Relation of School SGP with School Proportion SWD (ns, $p = .654$)



Relation of HLM EB Intercept with School Proportion SWD ($p < .001$)



Relation of HLM EB Slope with School Proportion SWD (ns, $p = .796$)



Q and A

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