

Evaluating an Agent-Based Model: Evaluation Practice Meets Research on Evaluation

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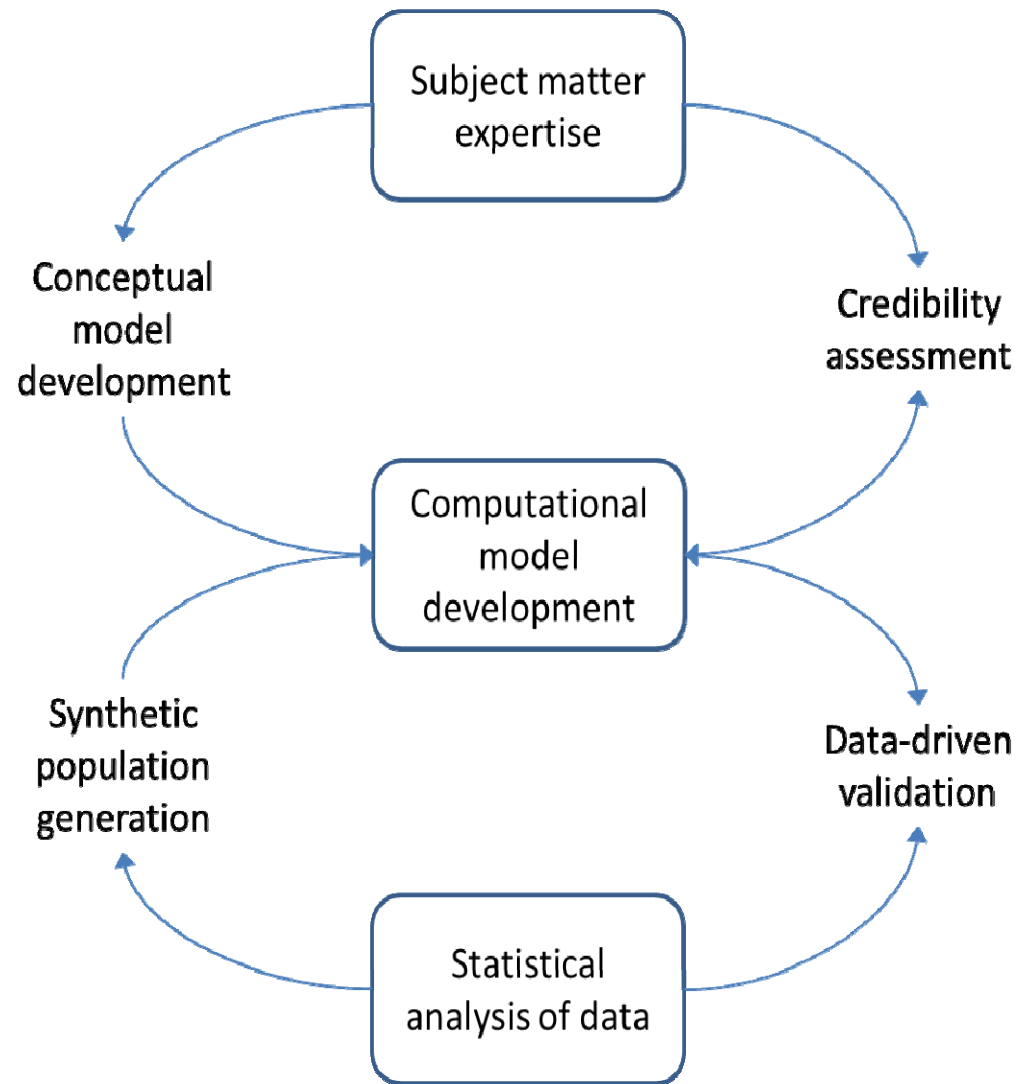
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Large-Scale Agent-Based Simulation of Complex Educational Systems

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Stephan Eidenbenz
Los Alamos National Laboratory

What are
we up to?

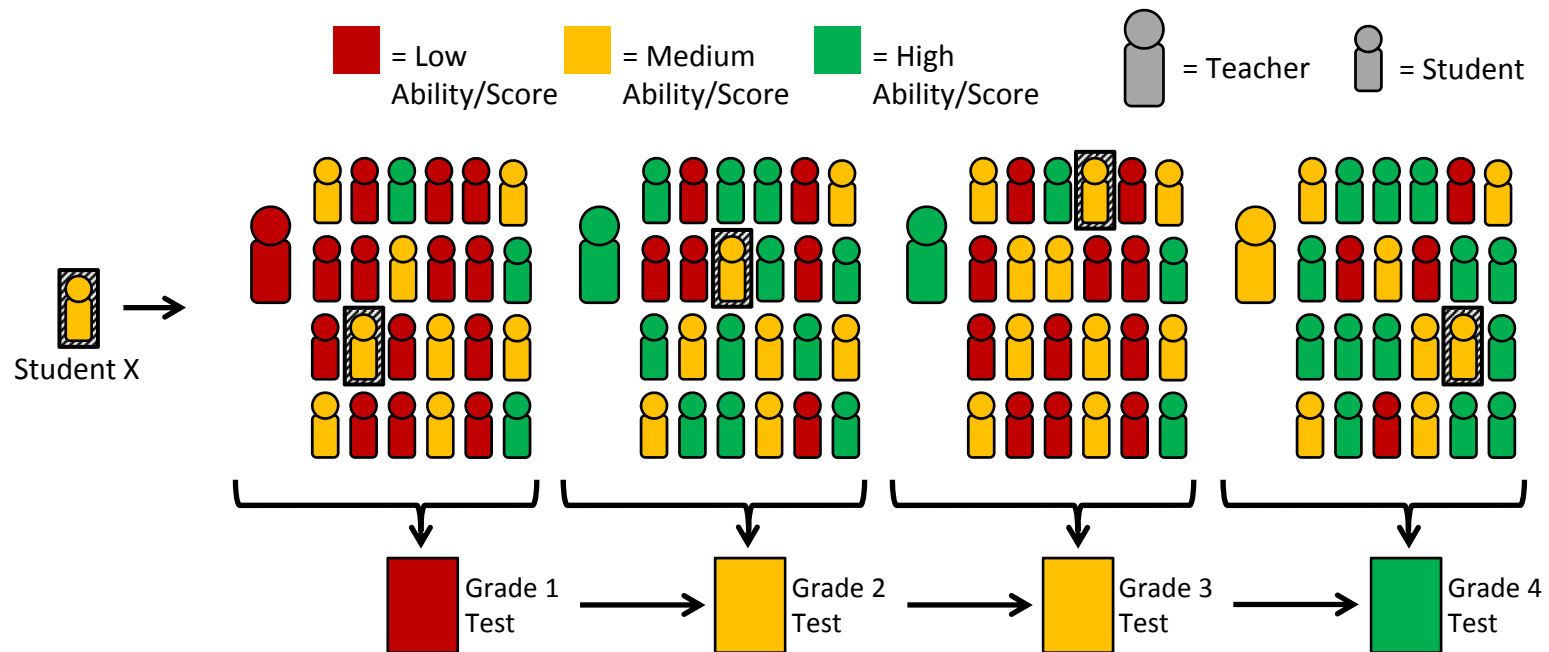
model development process



Model

1

Visual representation of score assignment



Student score function

■ Score Assignment Notation

- X_k^g = Score for student k at grade level g
- A_k = Baseline ability for student k to learn the material
- T = Baseline ability for the teacher to teach the material
- N = Number of students in the classroom
- w_i = Calibration Weights

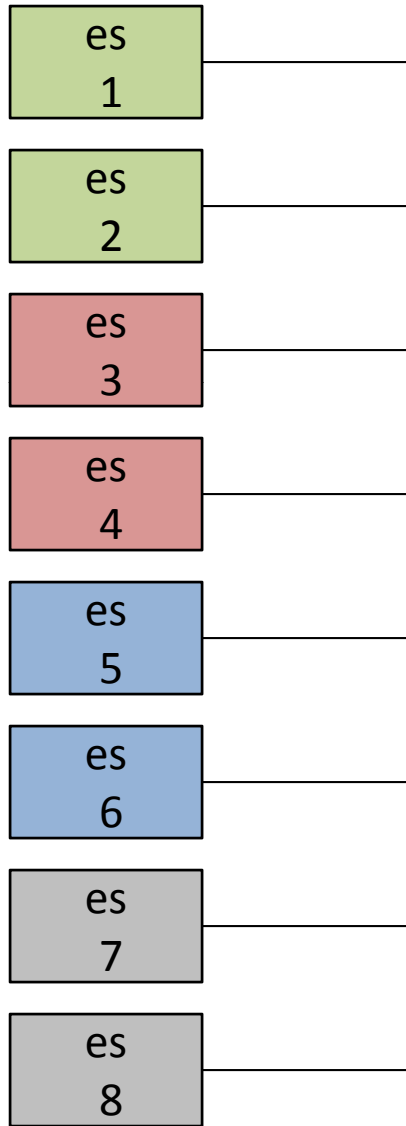
$$X_k^g = \frac{1}{4} \left[w_1 A_k + w_2 T + w_3 \frac{1}{N} \sum_{j=1}^N A_j + w_4 X_k^{g-1} \right]$$

For the first iteration, we only average over first three terms.

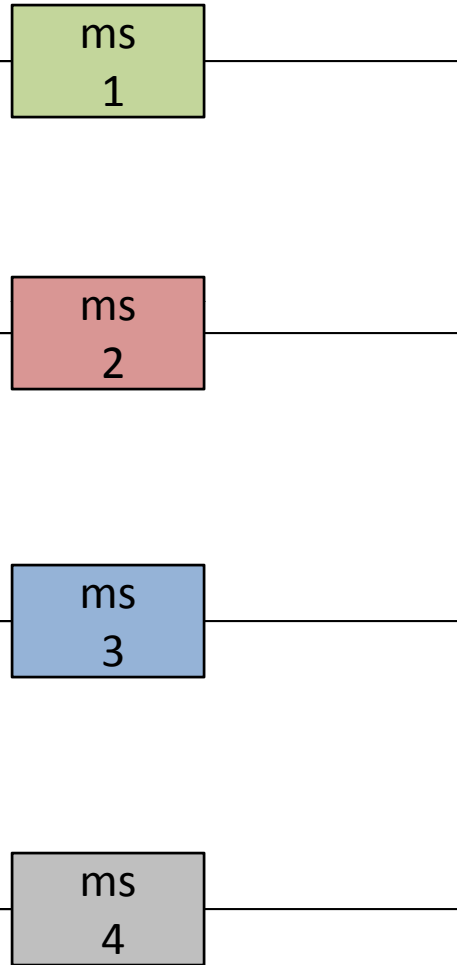
Model

2

8 schools
2 Teachers/School
1 Class/Teacher



4 schools
2 Teachers/School
2 Classes/Teacher



2 schools
2 Teachers/School
4 Classes/Teacher



Abilities
(True Scores)

versus

Observed
Test Score

Student's math ability at the end of this school year

Student's math ability at the end of last school year

Student's math ability at in the starting year

Mean class ability

Teacher's value added

School's value added

$$\theta_{ijkt} = w_1 \theta_{ijk(t-1)} + w_2 \theta_{ijk(t=0)} + w_3 \bar{\theta}_{jkt} + w_4 \alpha_{jkt} + w_5 \beta_{kt}$$

Student i
Teacher j
School k
Time t

weights

$$\theta_{ijkt} = w_1 \theta_{ijk(t-1)} + w_2 \theta_{ijk(t=0)} + w_3 \bar{\theta}_{jkt} + w_4 \alpha_{jkt} + w_5 \beta_{kt}$$

$w_1 = [1 - (\text{grade level} / 12)] * \text{Constant}$

$w_2 = (\text{grade level} / 12) * \text{Constant}$

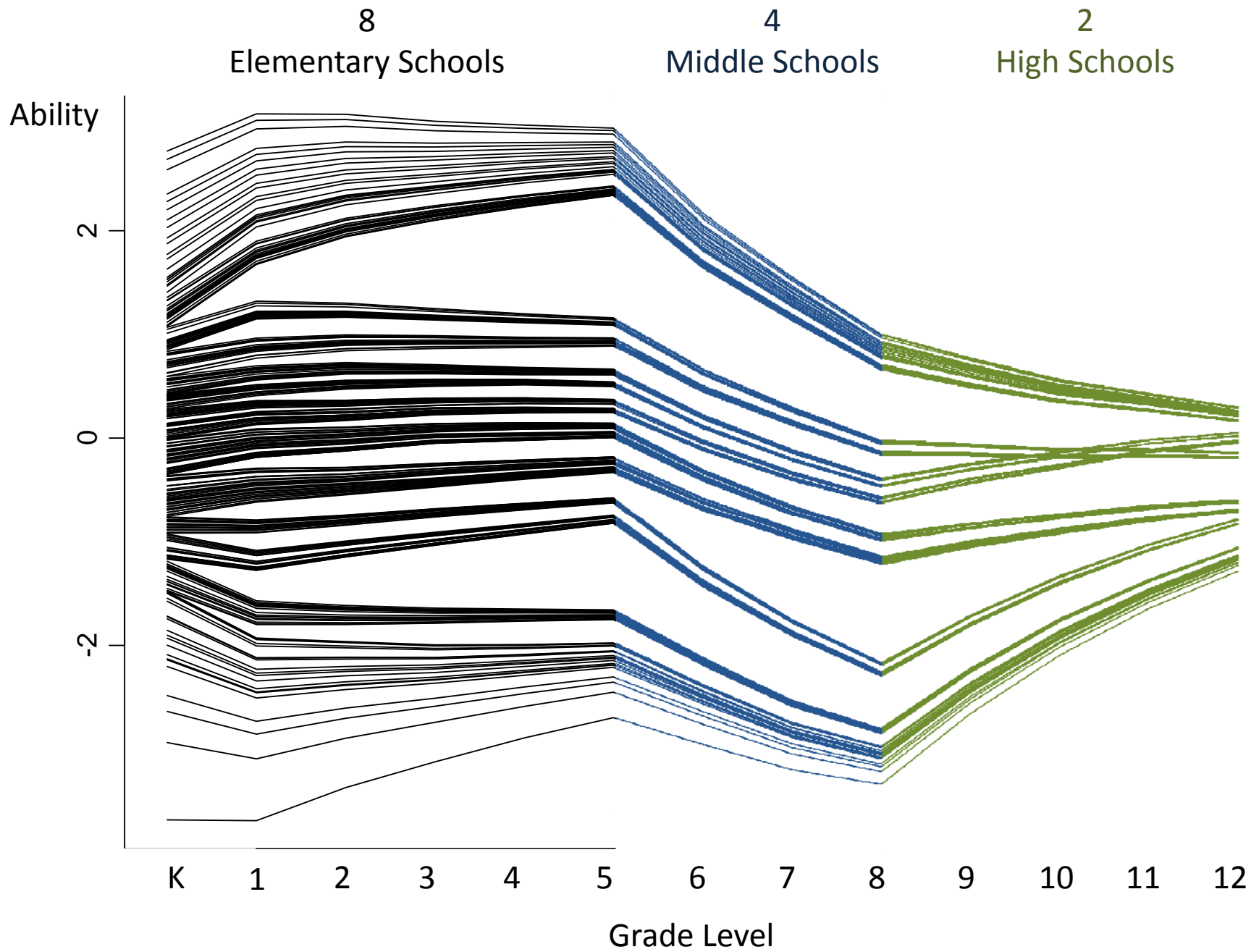
$w_3 = \text{Constant}$

$w_4 = \text{Constant}$

$w_5 = \text{Constant}$

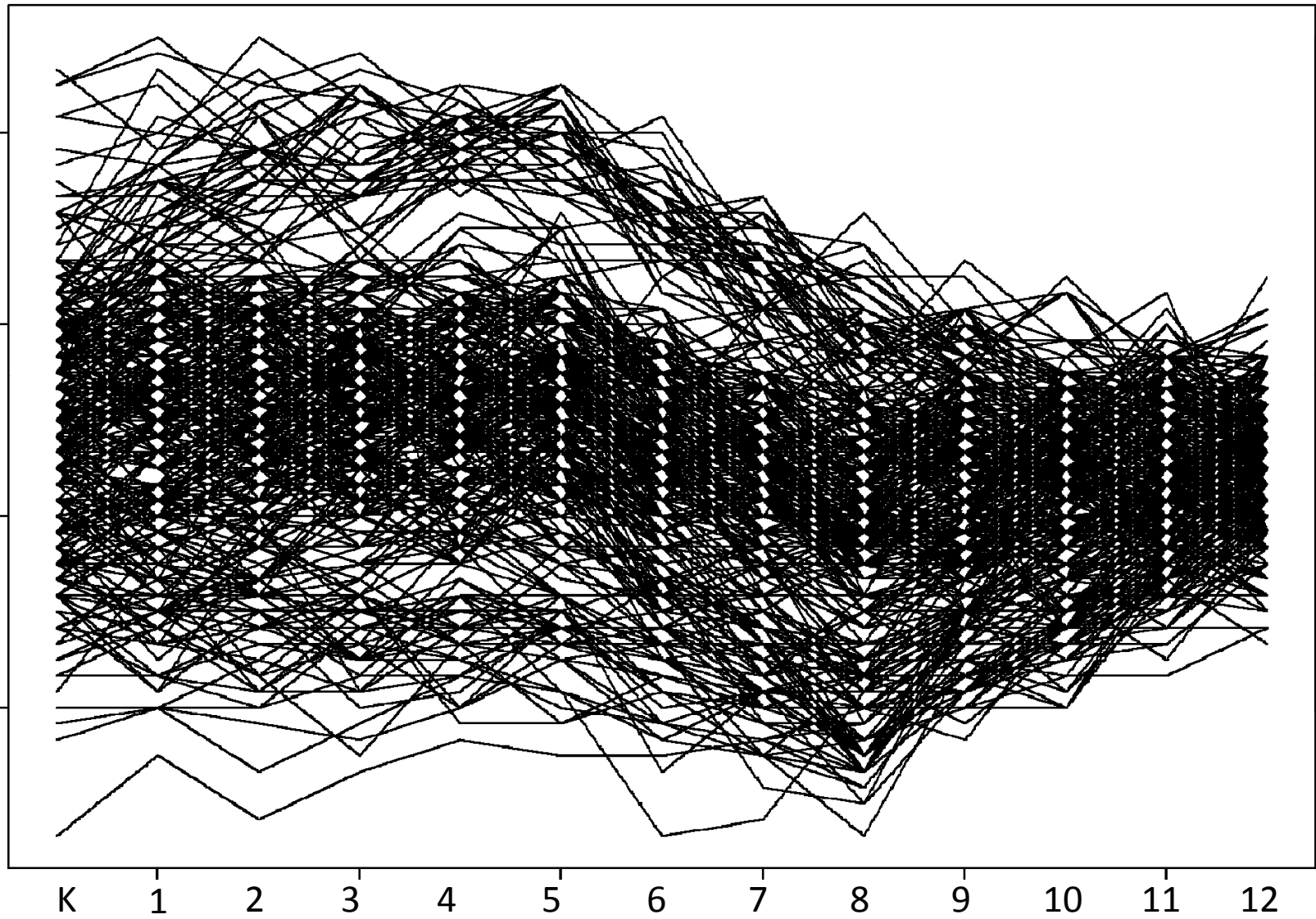
Exaggerated Model

Exaggerated Model (Abilities)



Exaggerated Model (Observed Test Scores)

Test Score



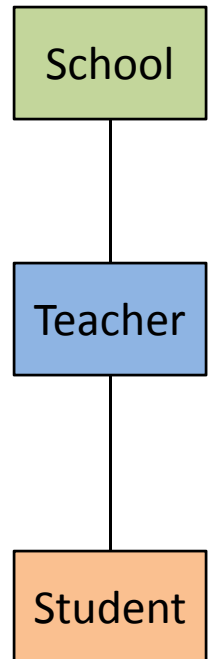
Grade Level

Systems
Models
Informed
by
Data

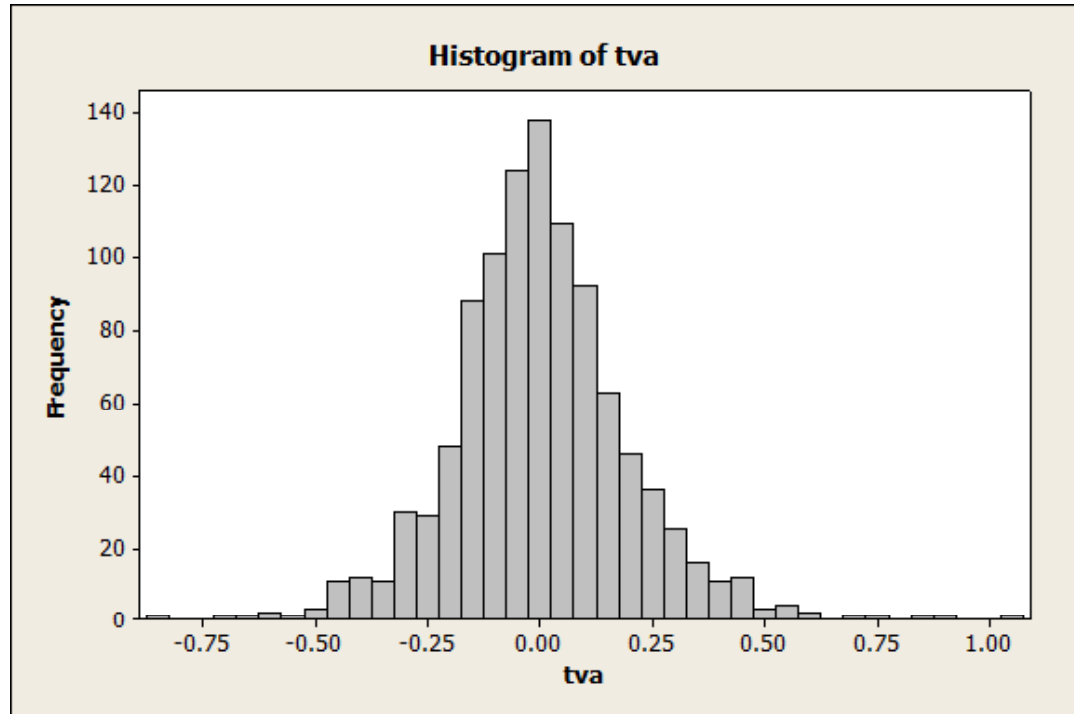
Used a variety of
multi-level models (HLMs)
to estimate a
variety of parameters

Proportion of Variance in Test Scores at Each Level

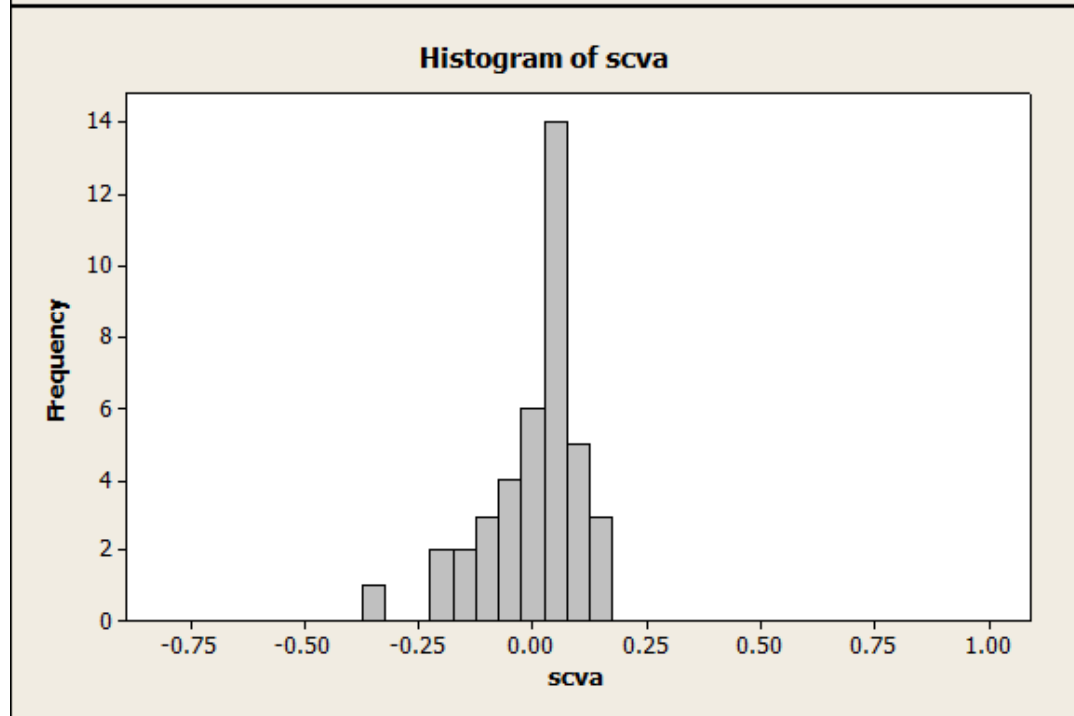
	"Start" of School Year	"End" of School Year
School	0.27	0.27
Teacher	0.06	0.09
Student	0.67	0.64



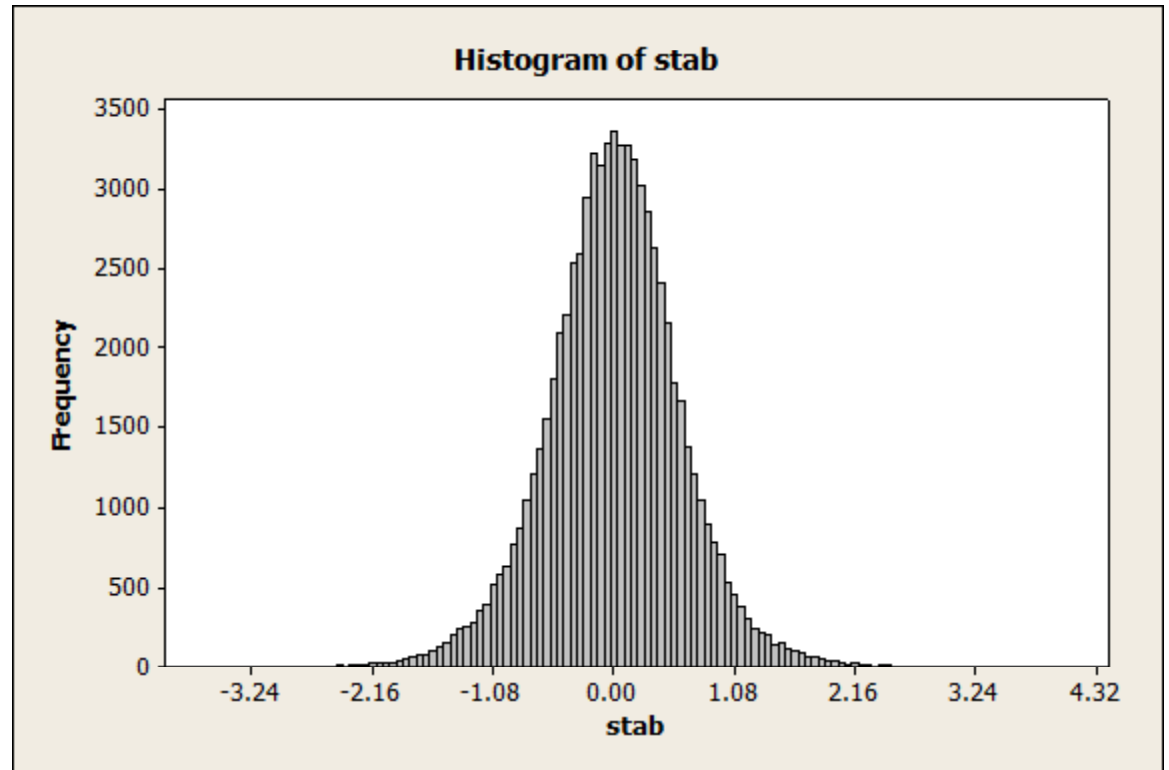
Teacher
VA



School
VA



Student VA

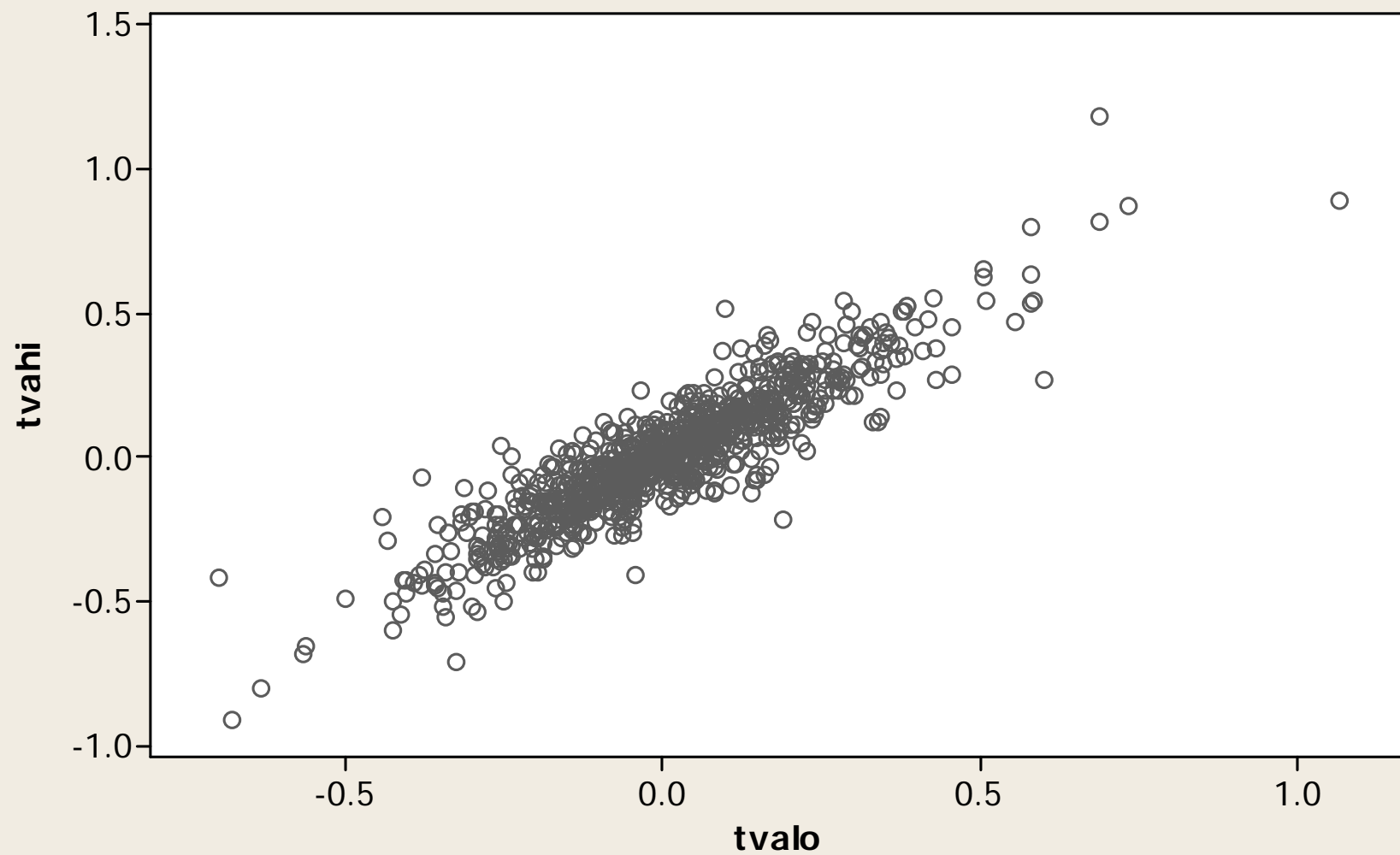


VA for ELA

High & Low Groups

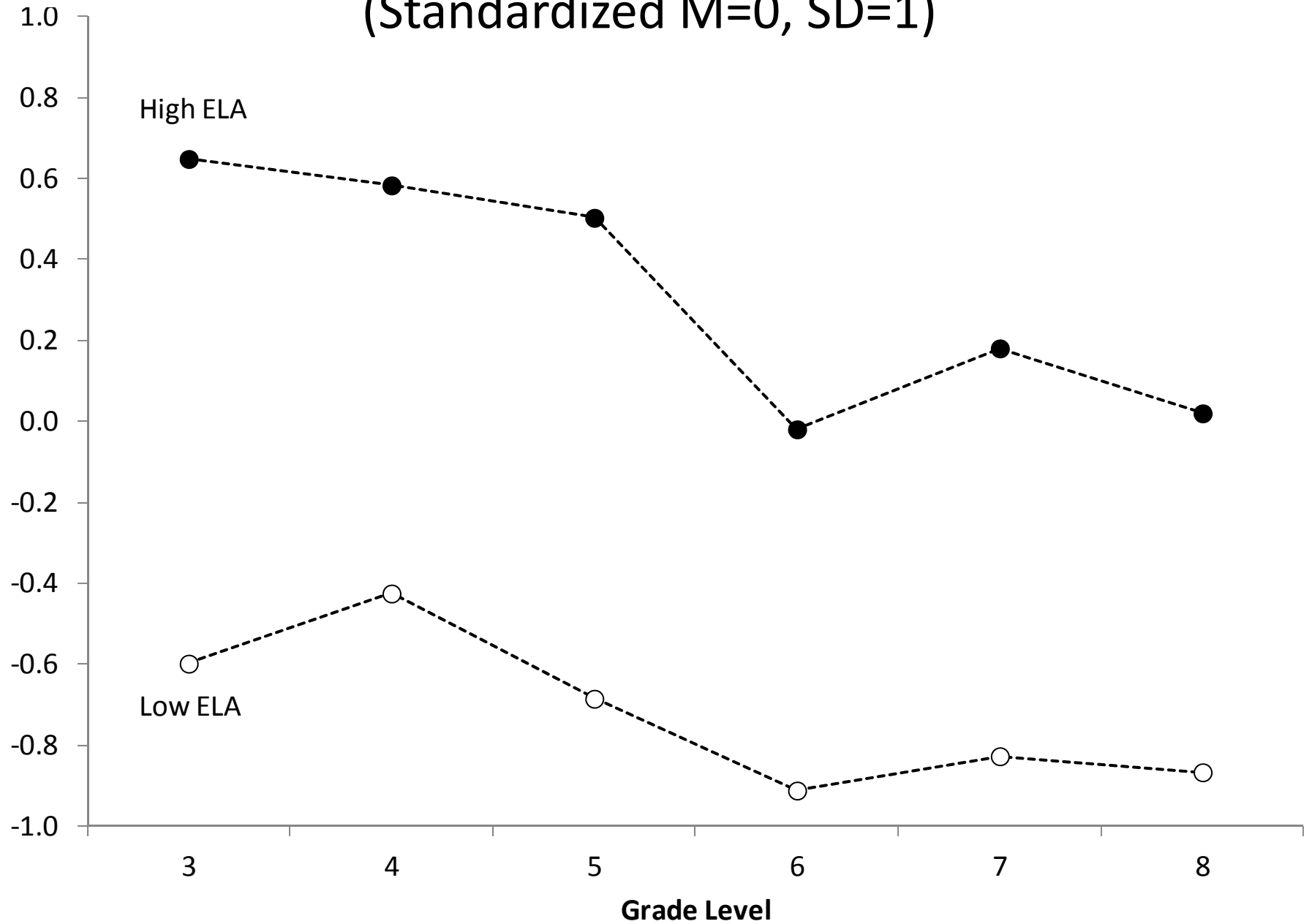
- averaged ELA scores over all years
- LOW ELA = bottom 1/3 of students
- HIGH ELA = top 2/3 of students

Scatterplot of tvahi vs tvalo



Mean Math Scores Over Time

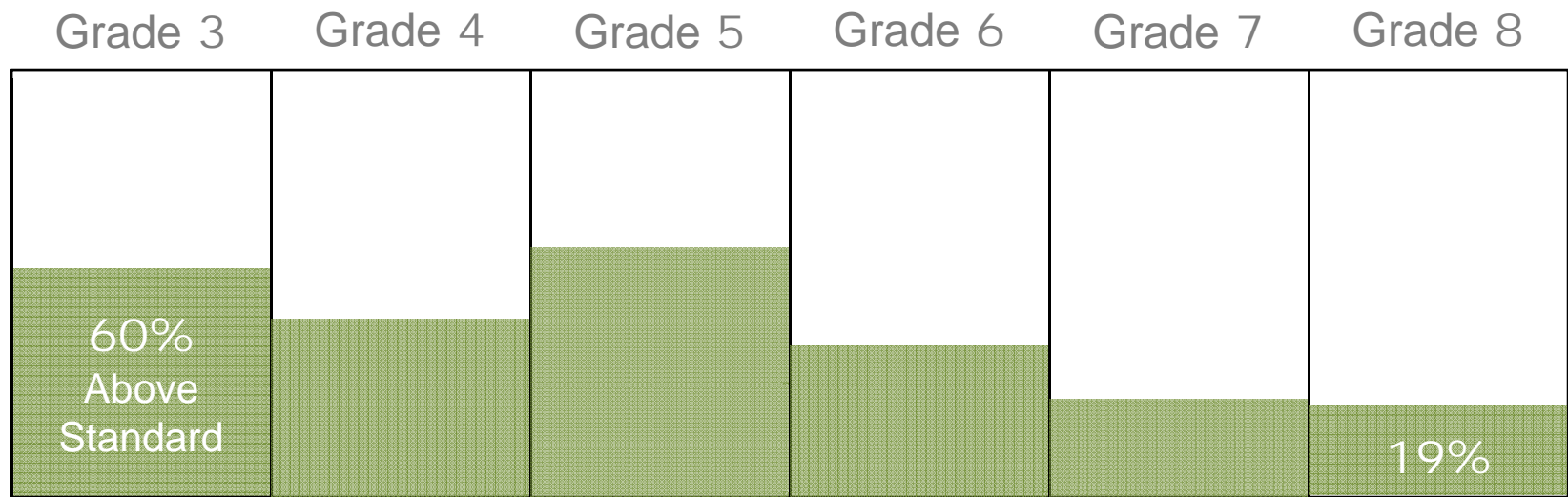
(Standardized M=0, SD=1)



Matching Students with Teachers

Fixed Effect	Coefficient	Standard Error	T-ratio	Approx. d.f.	P-value
For LOWELA slope, P1					
For INTRCPT2, B10					
INTRCPT3, G100	-0.118530	0.012738	-9.305	39	0.000
For VALO33, B11					
INTRCPT3, G110	0.355074	0.015701	22.615	1023	0.000
For HIELA slope, P2					
For INTRCPT2, B20					
INTRCPT3, G200	-0.171179	0.023128	-7.401	39	0.000
For VAHI33, B21					
INTRCPT3, G210	0.377650	0.016493	22.897	1023	0.000

Evaluating Utility



Professional Development for Teachers

Which teachers receives it?

weakest teachers

any teacher

Percent of all teachers who receive it?

30%

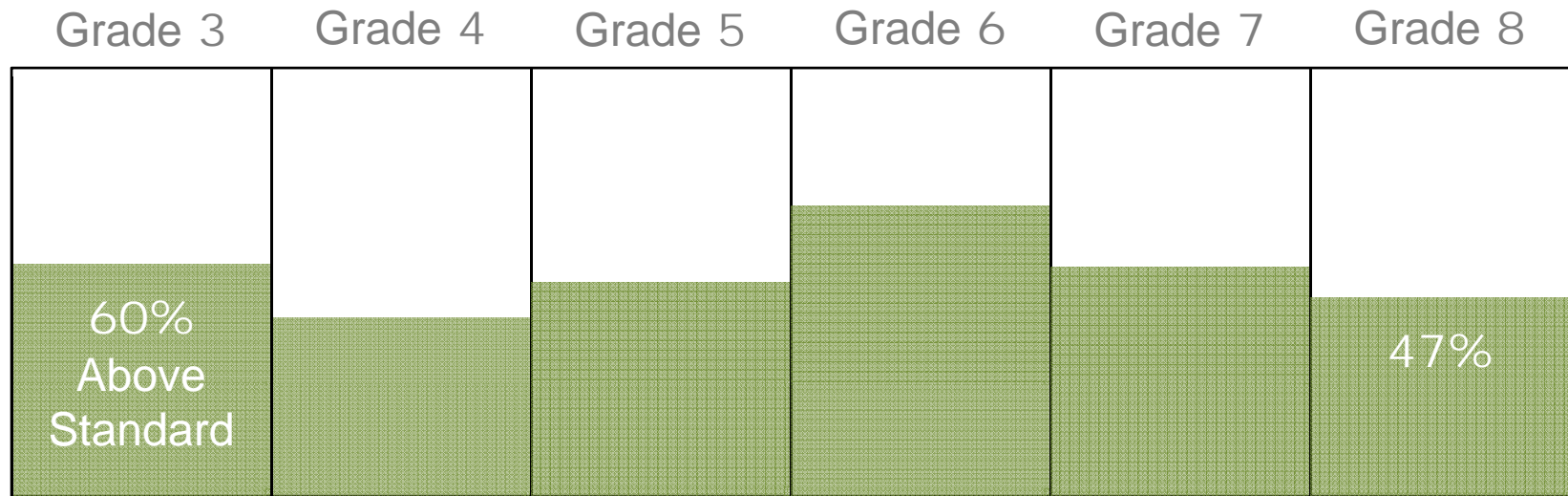
Results
Simulation 1: 19%

Matching

Percentage

Students matched to teachers: 90%

Students matched to students: 60%



Professional Development for Teachers

Which teachers receives it?

- weakest teachers
- any teacher

Percent of all teachers who receive it?

70%

Results

Simulation 1: 19%

Simulation 2: 47%

Matching

Percentage

Students matched to teachers: 90%

Students matched to students: 60%

Thank You