

# English Language Learners: Language growth within structured/sheltered English immersion programs

## Background

### The language growth of English Language Learners (ELLs; Rojas & Iglesias, under review)

Modeled the language growth of 1,723 (Spanish-English) ELLs

- Fall of kindergarten to spring of second grade
- 12,248 Narrative retell language samples:
  - 6,516 Spanish; 5,732 English
- Covariates: Gender; summer vacation
- Outcome measures: Mean length of utterance in words (MLUw); Number of different words (NDW)

#### Aims

- Differences with respect to prototypical language trajectories
- Intra- and inter-individual differences
- Systematic relationship between initial status and growth

## Purpose & Method

### What is the effect of program type on ELLs' language growth?

Structured/sheltered English immersion (SEI) programs involve all academic instruction in English

- Goal of SEI programs is to attain fluency in English

Language growth of ELLs in SEI programs was contrasted with the language growth of "overall" ELLs across a variety of programs

#### Participants

Subset of ELL children from overall sample used in Rojas & Iglesias (under review)

- ELLs enrolled in schools that offered SEI programs exclusively
- 419 ELLs: 198 girls; 221 boys
- 2,924 narrative retell language samples:
  - 1,427 Spanish; 1,497 English
- 65% of ELLs provided language samples in  $\geq 4$  semesters

## Method (continued)

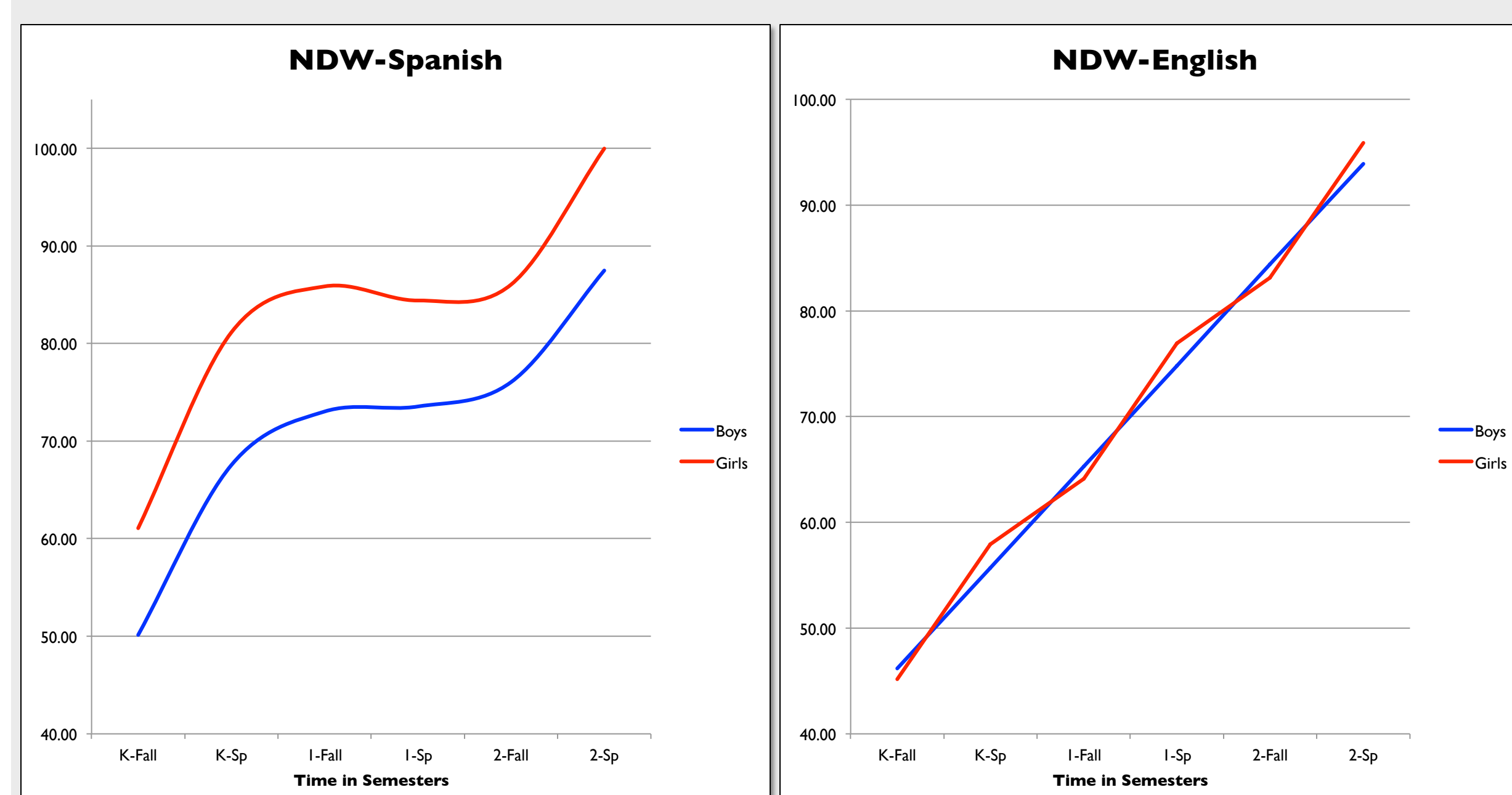
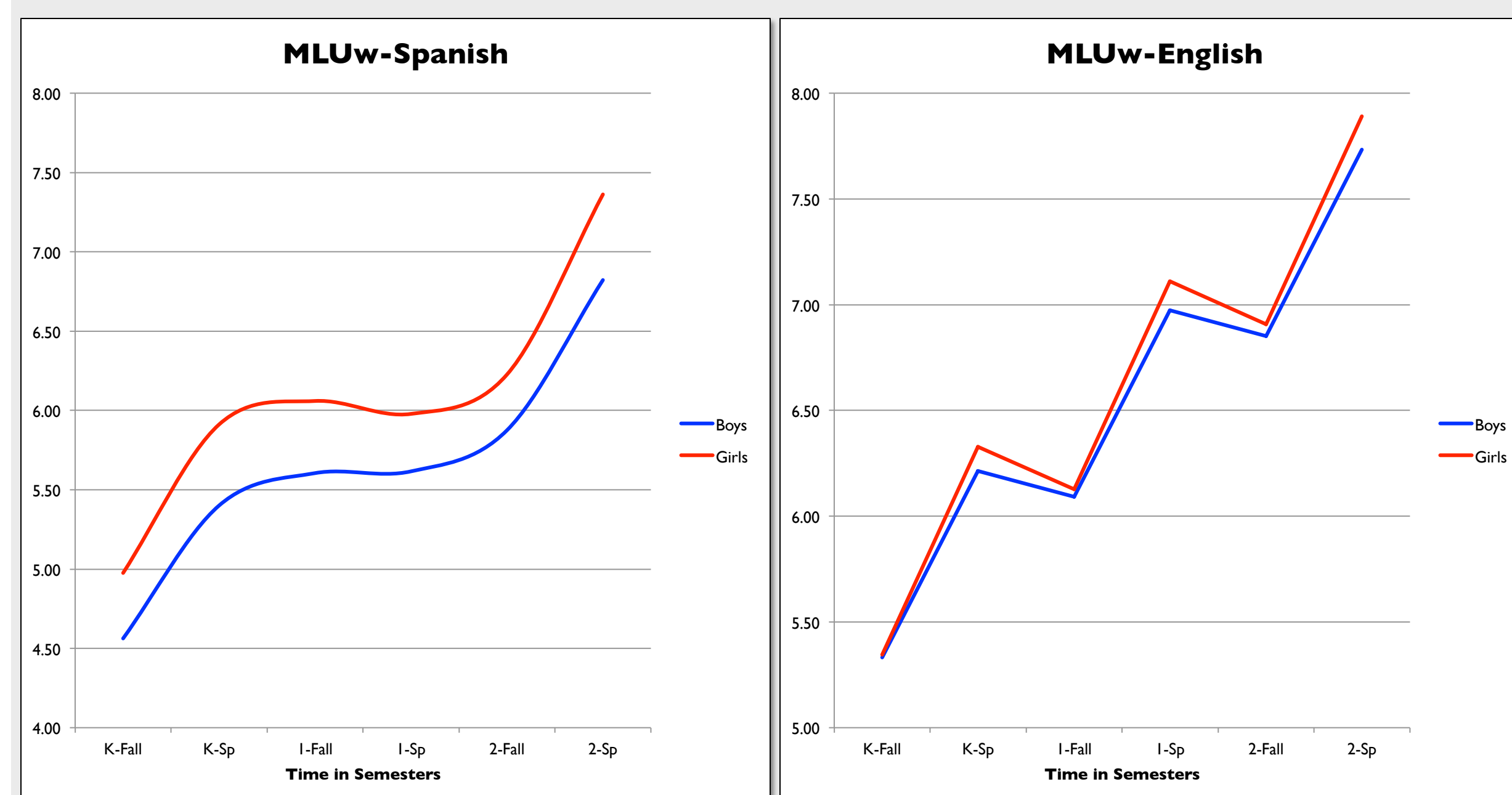
### Growth curve modeling (GCM)

- Maximum likelihood estimation method to handle missing data and estimate fixed effects and variance components
- Academic semester served as time metric
- Centering relative to fall of kindergarten as initial status
- GCM testing to determine final GCMs for each outcome measure:
  - Unconditional means model  $\rightarrow$  Unconditional growth models (linear, quadratic, and cubic; fixed and randomly varying slope configurations)  $\rightarrow$  Conditional growth models (gender and discontinuous time; gender x slope interactions)
  - Goodness of fit indices (-2LL for nested models; BIC for non-nested models) and Pseudo- $R^2$  statistics with  $\chi^2$  testing estimated and tested across models
  - Prototypical growth curve trajectories generated from final GCM parameter estimates

## Summary

	Spanish		English	
	MLUw	NDW	MLUw	NDW
Linearity	Curvilinear	Curvilinear	Linear	Linear
Direction	Non-monotonic	Non-monotonic	Non-monotonic	Non-monotonic
Continuity	Continuous	Continuous	Discontinuous	Discontinuous
Gender	Girls outpace boys	Girls outpace boys	Girls outpace boys (spring)	Girls outpace boys (spring)
Summer vacation	n/a	n/a	Negative growth	Slower growth
Initial status-growth covariance	Positive ( $\uparrow$ initial status = $\uparrow$ growth)	No systematic relationship	Negative ( $\downarrow$ initial status = $\uparrow$ growth)	Negative ( $\downarrow$ initial status = $\uparrow$ growth)

## Growth curve trajectories: Overall sample



## Final growth curve models

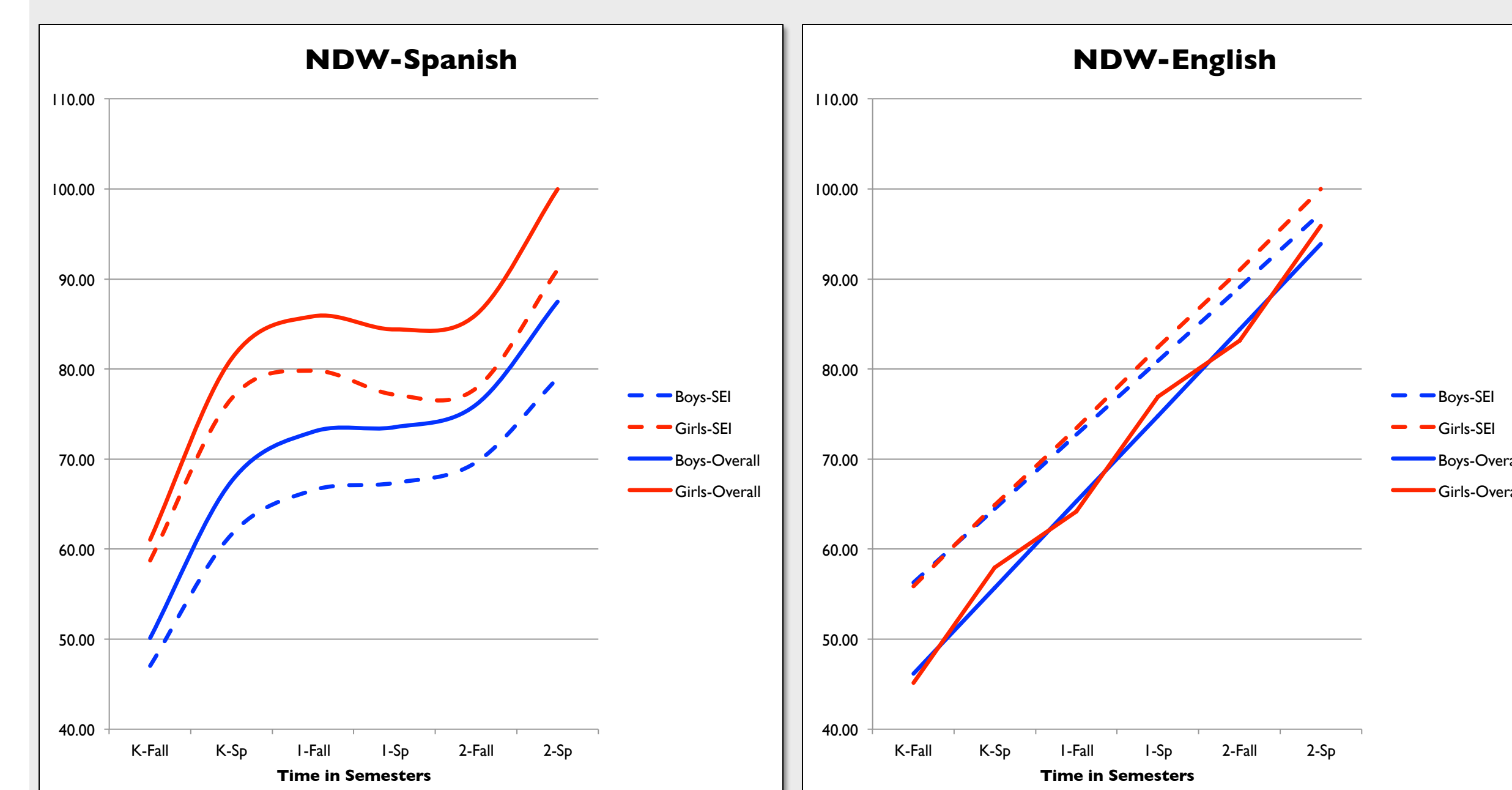
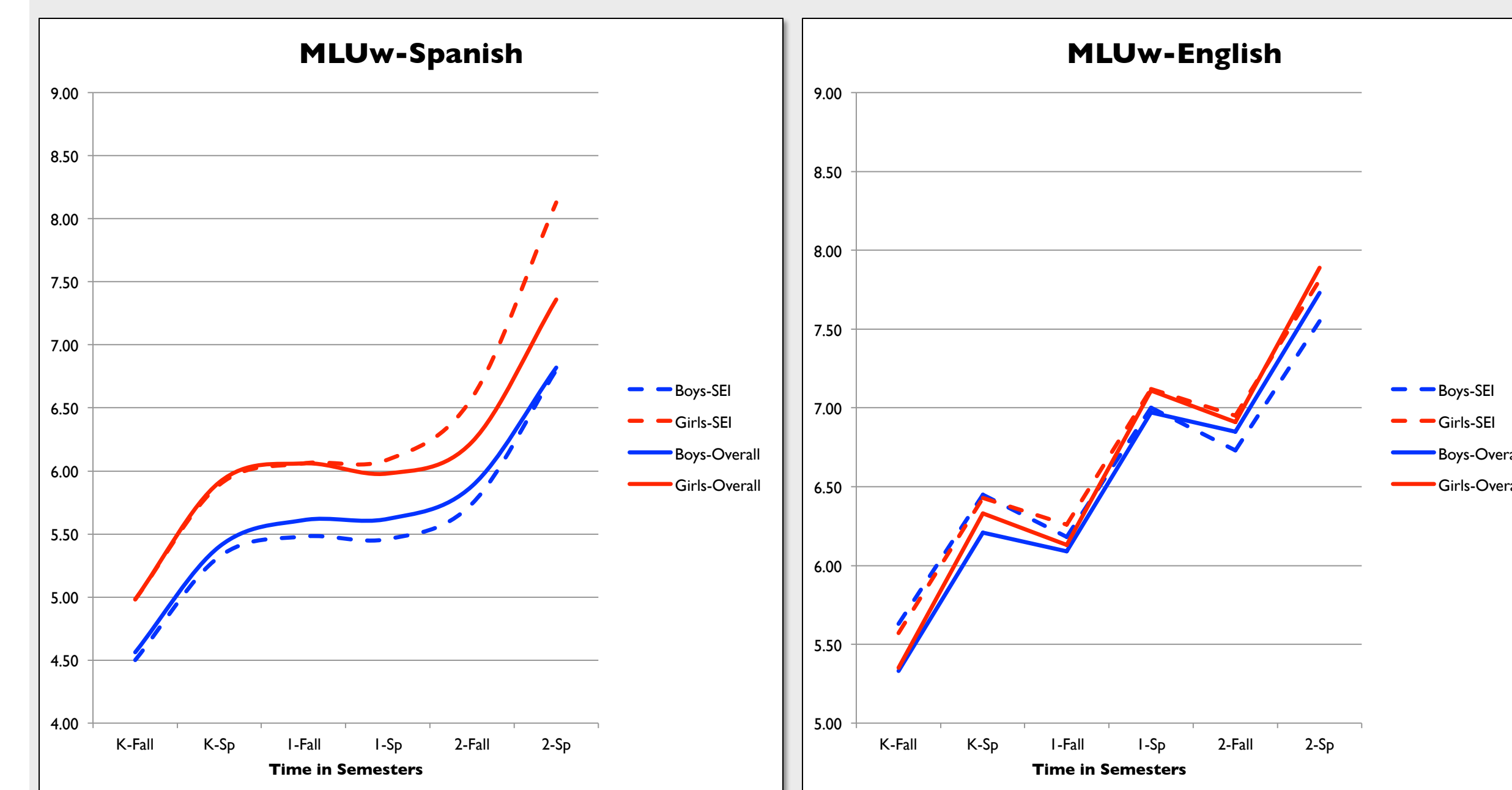
Table 1. Comparisons of continuous and discontinuous growth curve model parameter estimates for Mean Length of Utterance in words in Spanish and English (MLUw-S/E), and for Number of Different Words in Spanish and English (NDW-S/E)

		Overall-MLUw-S	SEI-MLUw-S	Overall-MLUw-E	SEI-MLUw-E	Overall-NDW-S	SEI-NDW-S	Overall-NDW-E	SEI-NDW-E
		CG-Cb+G	CG-Cb+G	CG-Dc+G	CG-Dc+G	CG-Cb+G	CG-Cb+G	CG-Dc+G	CG-Dc+G
<b>Fixed effects</b>									
Intercept	$\gamma_{00}$	4.56 <sup>c</sup>	4.5 <sup>c</sup>	5.33 <sup>c</sup>	5.63 <sup>c</sup>	50.13 <sup>a</sup>	47.04 <sup>a</sup>	46.17 <sup>a</sup>	56.27 <sup>a</sup>
Linear Slope	$\gamma_{10}$	1.30 <sup>c</sup>	1.31 <sup>c</sup>	0.76 <sup>c</sup>	0.55 <sup>c</sup>	25.69 <sup>a</sup>	21.29 <sup>a</sup>	19.11 <sup>a</sup>	16.42 <sup>a</sup>
Quadratic Slope	$\gamma_{20}$	-0.54 <sup>c</sup>	-0.57 <sup>c</sup>	0.88 <sup>c</sup>	0.82 <sup>c</sup>	-9.44 <sup>a</sup>	-7.63 <sup>a</sup>	9.52 <sup>a</sup>	8.21 <sup>a</sup>
Cubic Slope	$\gamma_{30}$	0.07 <sup>c</sup>	0.08 <sup>c</sup>			1.16 <sup>c</sup>	0.93 <sup>c</sup>		
Gender (G)	$\gamma_{01}$	0.21 <sup>a</sup>	0.24 <sup>a</sup>	0.01	-0.06	5.46 <sup>a</sup>	5.85 <sup>a</sup>	-1.02	-0.4
G x Linear Slope	$\gamma_{11}$	0.21 <sup>a</sup>	0.17	0.02	0.14 <sup>a</sup>	5.17 <sup>a</sup>	7.17 <sup>a</sup>	-0.12	1.13
G x Quadratic Slope	$\gamma_{21}$	-0.14 <sup>b</sup>	-0.10	G x Summer	0.04	-2.87 <sup>b</sup>	-4.37 <sup>b</sup>	G x Summer	3.25 <sup>c</sup>
G x Cubic Slope	$\gamma_{31}$	0.02 <sup>b</sup>	0.02			0.38 <sup>b</sup>	0.59 <sup>b</sup>		0.79
<b>Variance components</b>									
L1: Within-person variance	$\sigma_1^2$	0.56 <sup>a</sup>	0.53 <sup>a</sup>	0.69 <sup>a</sup>	0.65 <sup>a</sup>	211.74 <sup>a</sup>	216.35 <sup>a</sup>	207.97 <sup>a</sup>	245.67 <sup>a</sup>
L2: B/w-person intercept	$\sigma_0^2$	0.30 <sup>a</sup>	0.28 <sup>a</sup>	0.99 <sup>a</sup>	0.67 <sup>a</sup>	200.64 <sup>a</sup>	180.74 <sup>a</sup>	484.94 <sup>a</sup>	314.73 <sup>a</sup>
L2: B/w-person linear slope	$\sigma_1^2$			0.15 <sup>a</sup>	0.05		1.18	18.25 <sup>a</sup>	
L2: B/w-person quadratic slope	$\sigma_2^2$	0.003				0.006		6.12	Summer
L2: B/w-person cubic slope	$\sigma_3^2$	0.0001	0.00001 <sup>a</sup>						
Covariance ( $\sigma_{01}, \sigma_1^2$ )	$\sigma_{01}$			-0.31 <sup>a</sup>	-0.15 <sup>a</sup>		-1.07	-46.17 <sup>a</sup>	
Covariance ( $\sigma_{02}, \sigma_2^2$ )	$\sigma_{02}$	-0.01				-0.85			-15.38
Covariance ( $\sigma_{03}, \sigma_3^2$ )	$\sigma_{03}$	0.003 <sup>a</sup>	0.0002						Summer
<b>Proportional variance reduction</b>									
L1: Within-person variance	$R_c^2$	48%	44%	50%	42%	38%	31%	57%	46%
L2: B/w-person intercept	$R_0^2$	6%	7%	<1%	<1%	5%	9%	<1%	<1%
L2: B/w-person linear slope	$R_1^2$			<1%	<1%		10%	<1%	
L2: B/w-person quadratic slope	$R_2^2$	<1%				<1%			<1%
L2: B/w-person cubic slope	$R_3^2$	50%	2%						Summer
<b>Goodness-of-fit</b>									
-2LL		16757.6 <sup>a</sup>	3607.0 <sup>a</sup>	15237.8 <sup>a</sup>	3648.5 <sup>a</sup>	55748.8 <sup>a</sup>	12253.5 <sup>a</sup>	46608.7 <sup>a</sup>	11713.2 <sup>a</sup>
BIC		16889.3	3694.1	15323.6	3720.4	55854.2	12340.6	46694.4	11785.1

<sup>a</sup> $p < .10$ . <sup>b</sup> $p < .05$ . <sup>c</sup> $p < .01$ . <sup>d</sup> $p < .001$ .

Note. SEI: Structured/Sheltered English Immersion; CG-Cb+G: Conditional cubic growth model with gender; CG-Dc+G: Conditional discontinuous growth model with gender; L1: Level-1 submodel; L2: Level-2 submodel; -2LL: -2 log-likelihood deviance statistic; BIC: Schwarz's Bayesian information criterion.

## Growth curve trajectories: SEI programs vs. Overall sample



## Conclusions & Next steps

**ELLs in SEI programs differed from overall sample**  
 Based on systematic, academic instruction in English, some growth patterns were expected:

- Boys' MLUw- and NDW-Spanish slower than overall sample
  - Girls' NDW-Spanish slower than girls in overall sample
  - Girls' and boys' NDW-English faster than overall sample
- However, other growth patterns were unexpected:
- Girls' MLUw-Spanish faster than overall sample
  - Girls' MLUw-English with similar growth rates to overall sample
  - Boys' MLUw-English slower than overall sample

### Necessary to model bilingual programs

- Transitional bilingual education programs involve initial instruction in the native language, which gradually transitions to English
- Beyond "program type", could also consider the fidelity of language instruction by teacher in order to use actual language of instruction as a covariate of language growth