

A computational account of bilingual aphasia rehabilitation

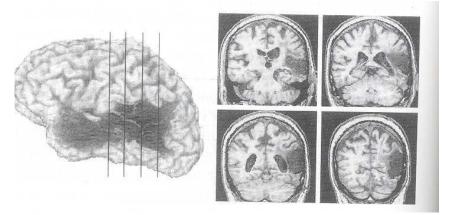
Swathi Kiran

Speech & Hearing Sciences; Boston University Department of Neurology; Massachusetts General Hospital Communication Sciences & Disorders; University of Texas at Austin

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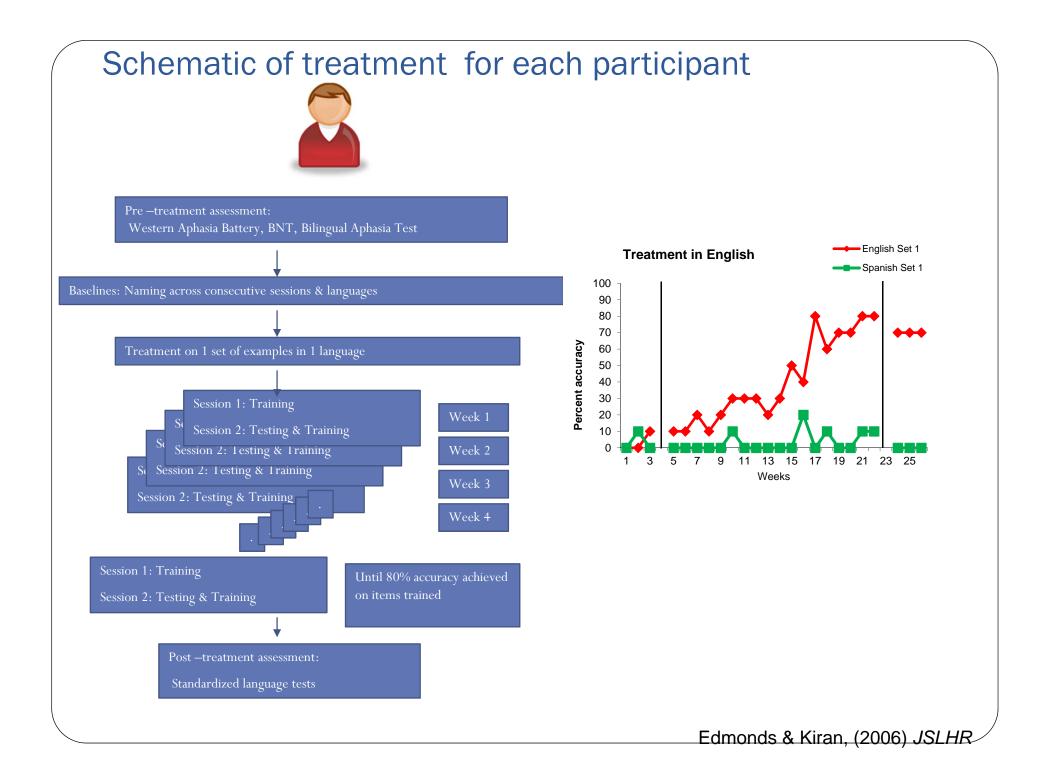
What is aphasia?

- Aphasia is characterized by language deficits such as problems speaking, understanding people, reading and writing
- Approximately 80,000 people incur aphasia each year
- It is estimated that 60% of the world is bi/multi-lingual



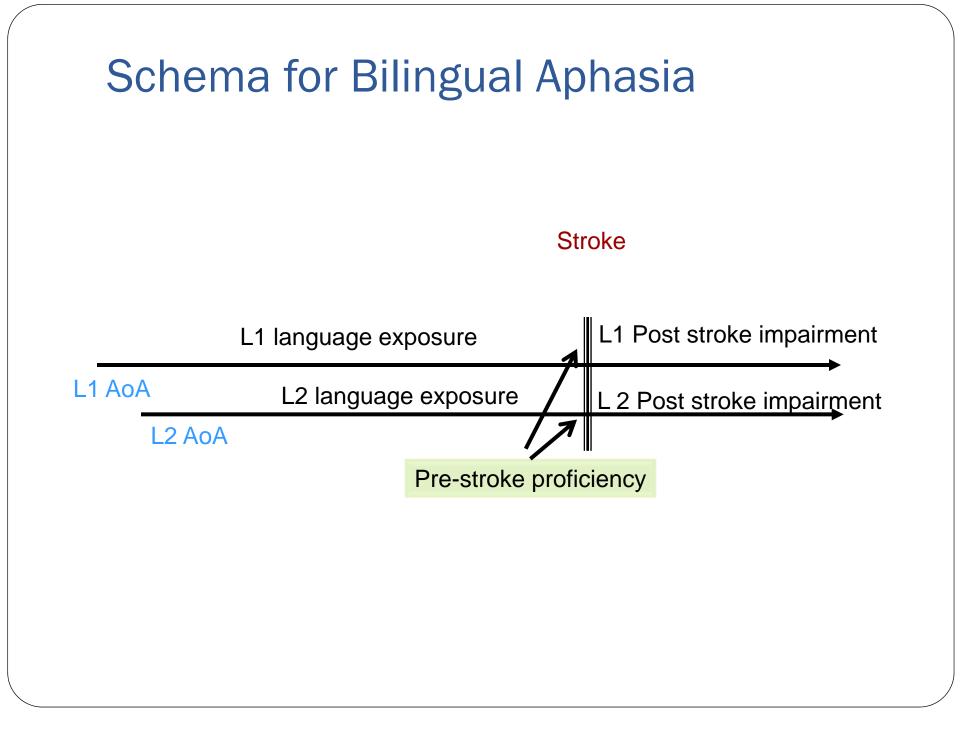
Bilingual Aphasia

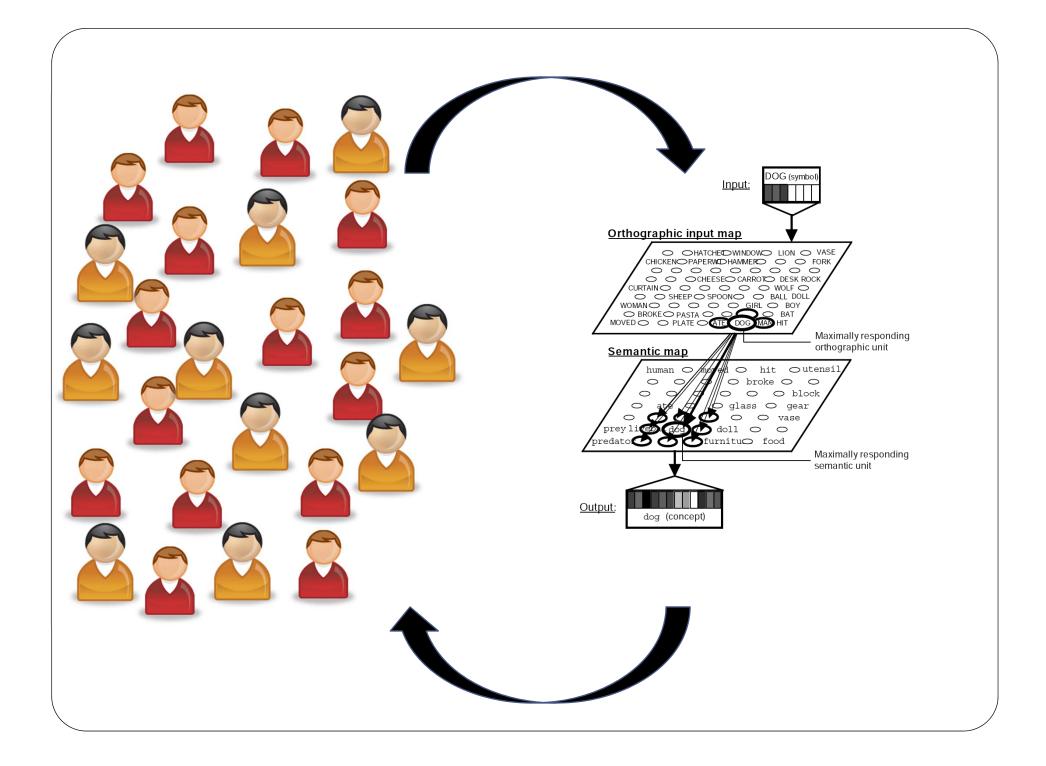




Bilingual Aphasia Rehabilitation

- Obviously, this translates to an increase in clinical need to address bilingual aphasia rehabilitation, but no clear guidelines on how to do so...
- No consistent results on rehabilitation of bilingual aphasia (Lorenzen & Murray, 2008; Faroqi-Shah et al., 2010)
- Few systematic studies that have examined and observed the extent of cross language transfer but results vary
 - (Croft et al., 2011; Edmonds & Kiran, 2006; Miertsch et al., 2009, Kiran & Roberts, 2009)





Is there another way to understand the nature of bilingual aphasia rehabilitation?

- Develop a computational simulation of bilingual aphasic naming deficits and rehabilitation of bilingual aphasia.
- Similar to predicting rehabilitation of naming deficits (Plaut, 1996) which has led to the complexity account of treatment deficits for naming deficits (Kiran, 2007)

Computational Modeling: SOM

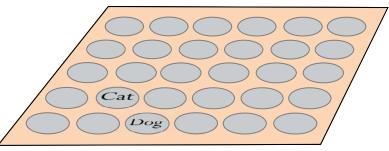
- Self Organizing Maps (Kohonen, 1995) operate in two modes
 - Training -builds the map using input examples
 - Mapping- classifies a new input vector
- SOMs have been used to understand bilingual language learning (Li, Zhao & McWhinney, 2007) and biological/psychiatric conditions (Hamalainen, 1994; Hoffman, Grasemann, & Miikkulainen, 2011)

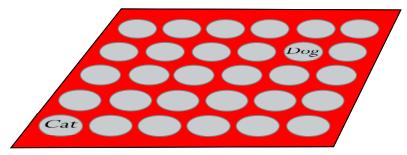
The Bilingual DISLEX Model

Semantic representations 260 hand-coded binary features

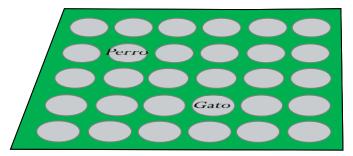
Semantic map

Phonetic representations
Based on English and
Spanish IPA transcriptions
Numerical
representations of
phonemes



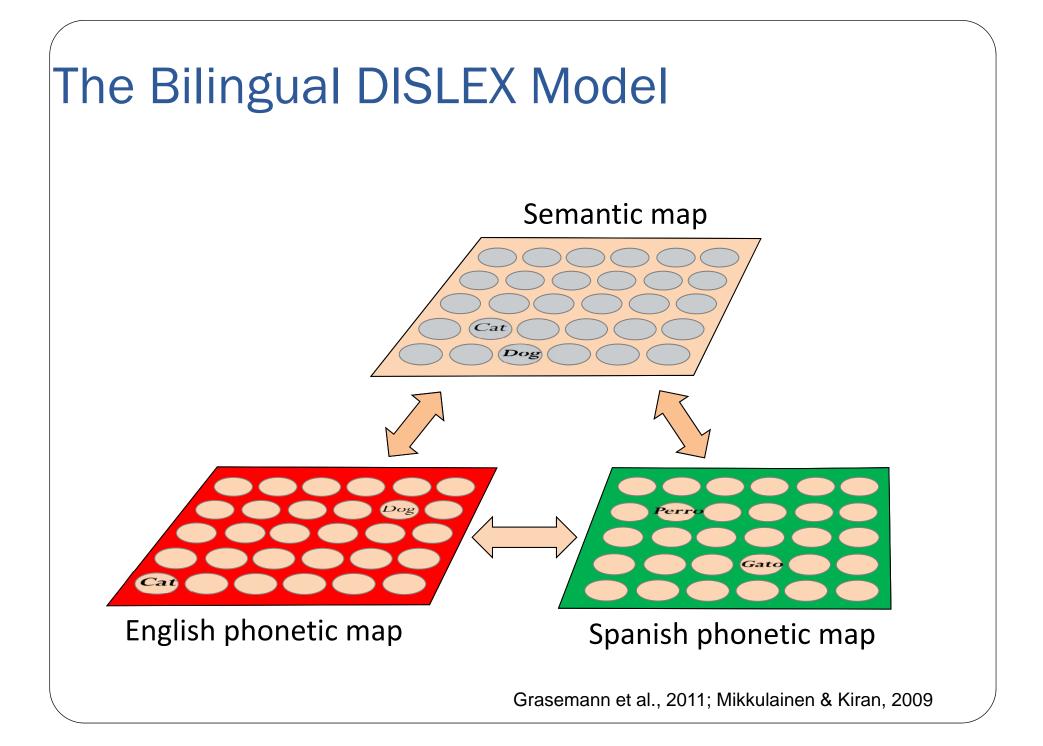


English phonetic map

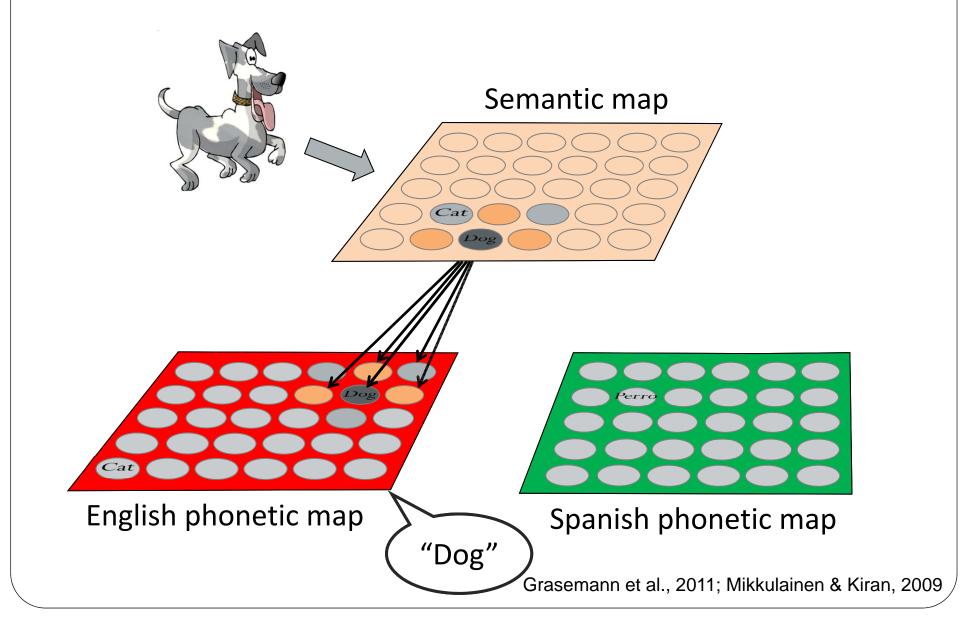


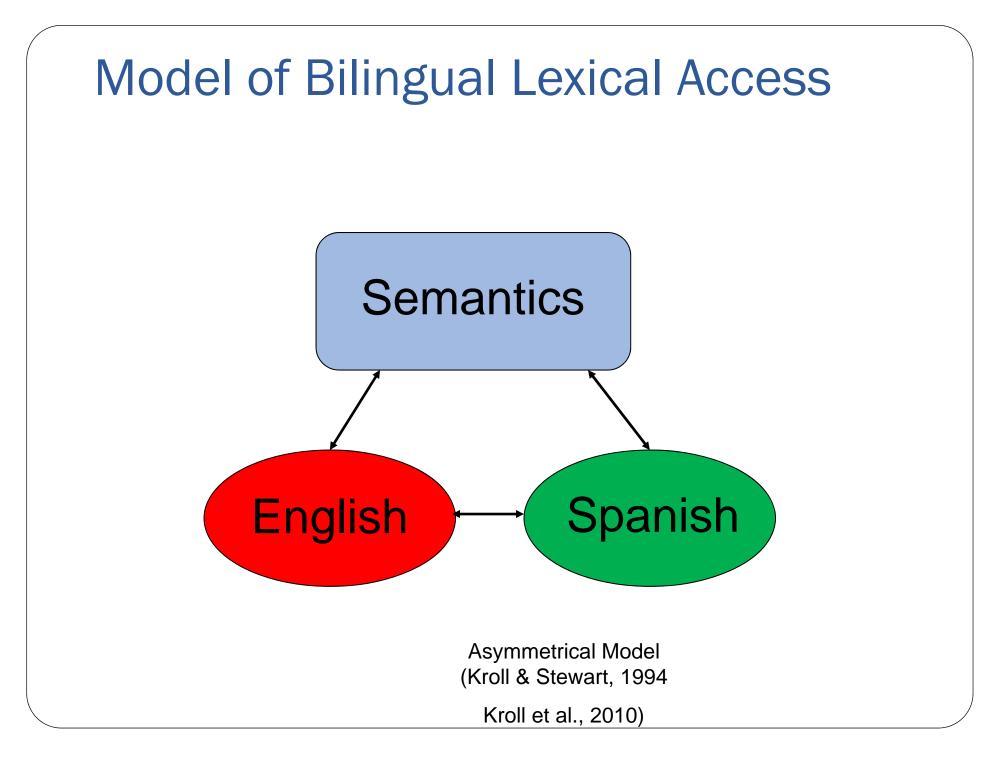
Spanish phonetic map

Grasemann et al., 2011; Mikkulainen & Kiran, 2009



Naming Task in Bilingual DISLEX Model





Develop a computational simulation of bilingual aphasic naming deficits and rehabilitation of bilingual aphasia.

- Model pre-stroke/normal bilingual language performance
 - Use AoA and exposure as training parameters
 - DISLEX should be able to match pre-stroke English and Spanish performance
 - Simulate damage to the lexicon

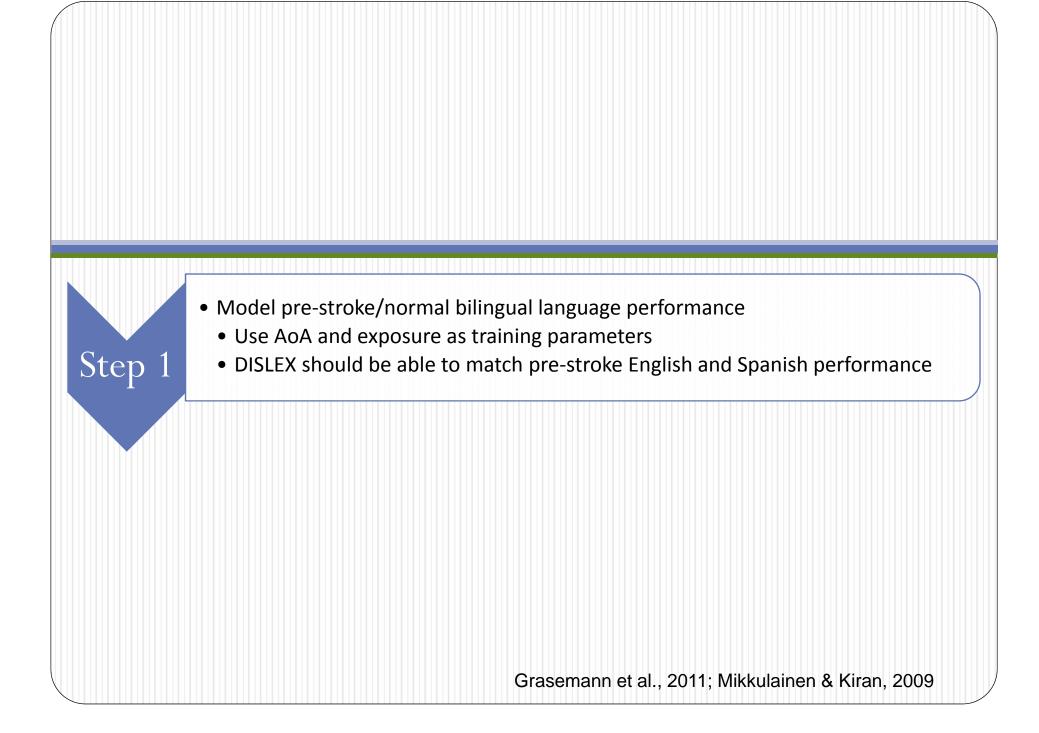
Step 1

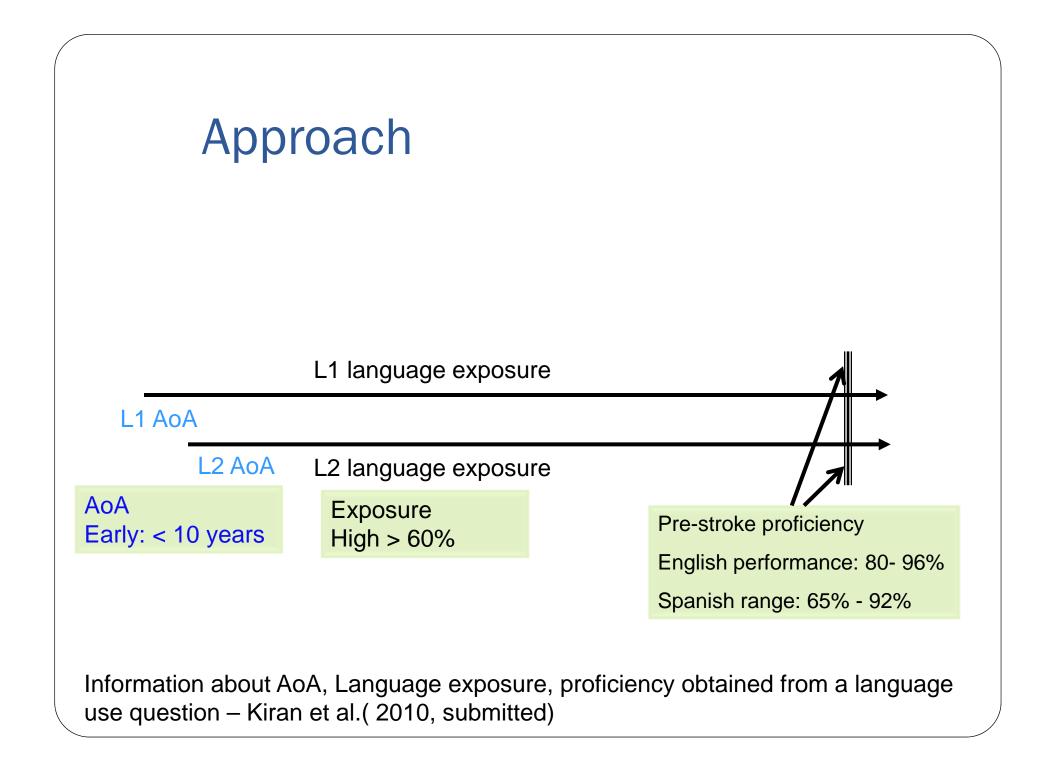
Step 2

Step 3

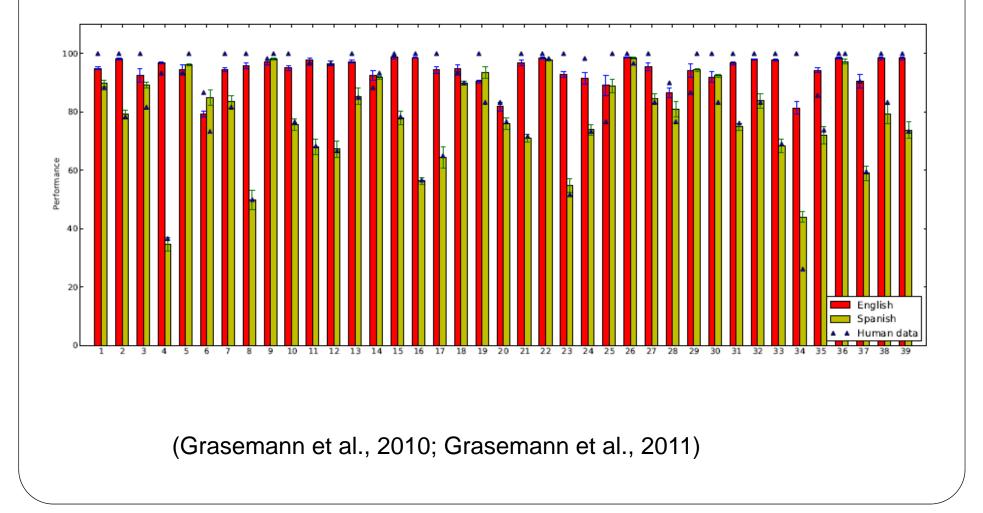
- Distort associative connections with noise
- DISLEX should be able to model impairment in patients

- Use the model to predict treatment outcomes
 - Examine improvements in trained language and cross language transfer





Results of simulation of normal bilingual individuals



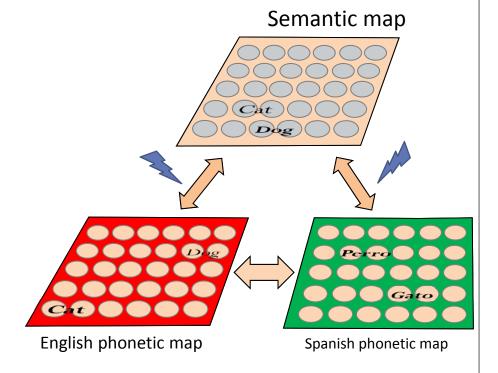
• Simulate damage to the lexicon

Step 2

- Distort associative connections with noise
- DISLEX should be able to model impairment in patients

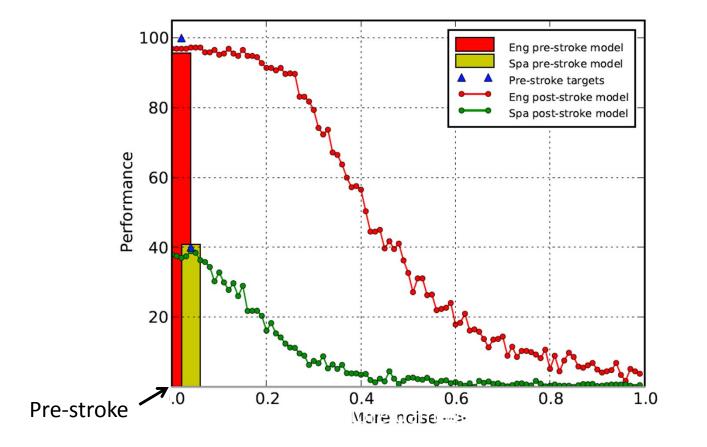
Simulation of bilingual aphasia-Lesion was applied to the

- Lesion was applied to the connections from the semantic map to the phonetic maps
- Adding Gaussian noise with µ = 0 to all these connections.
- The amount of damage (the "lesion strength") in each case was adjusted by changing the \sigma (σ) of the noise between 0 and 1.0 in steps of 0.01.



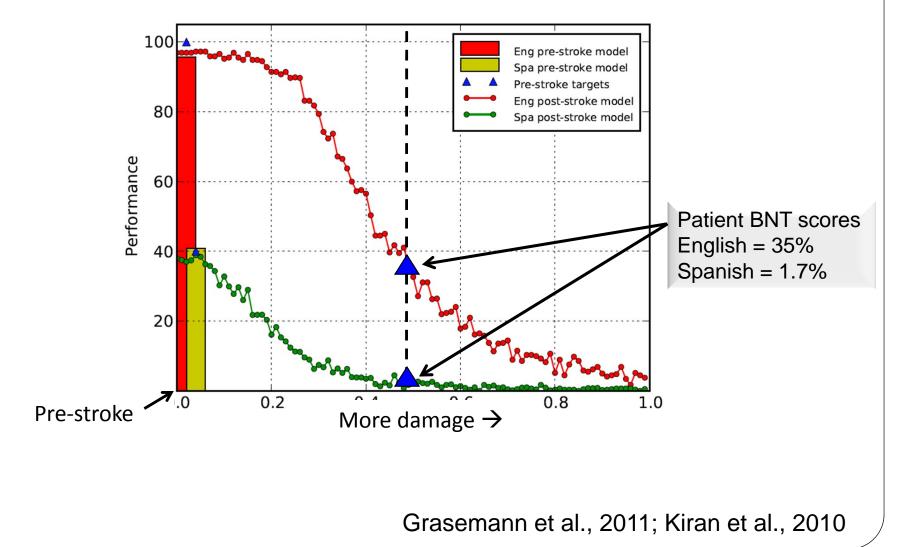
Grasemann et al., 2011; Kiran et al., 2010

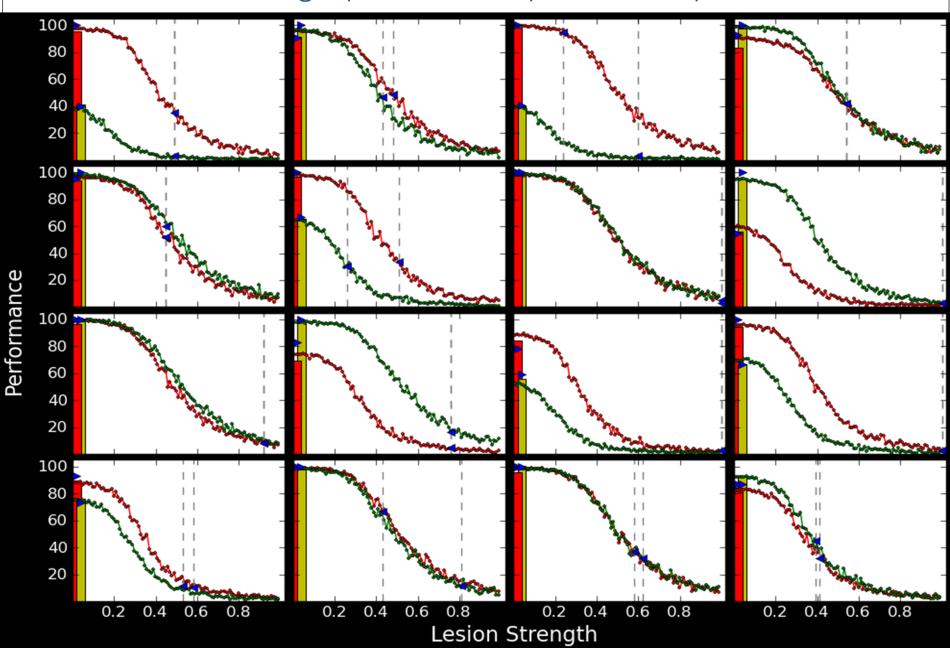
Results from DISLEX Model – Modeling Impairment in one patient



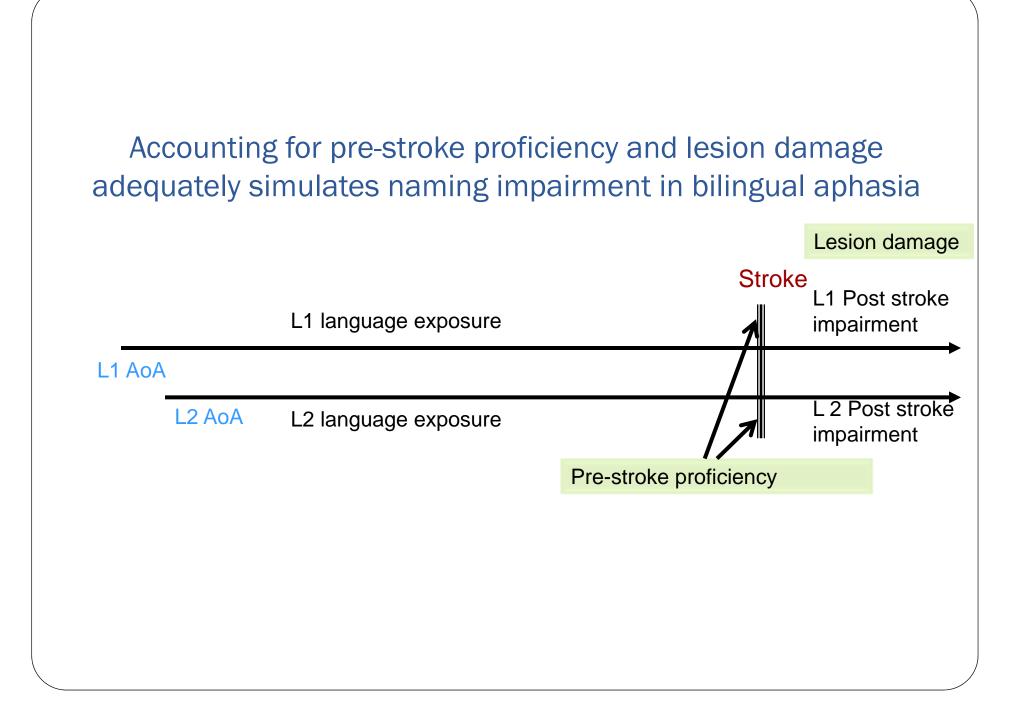
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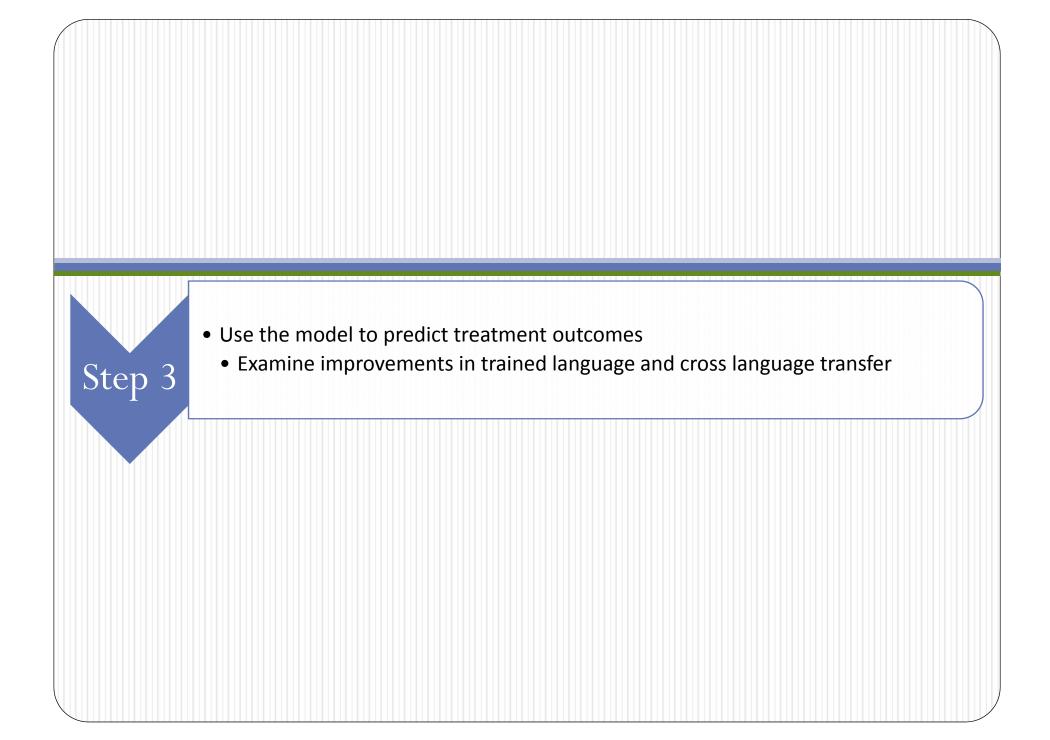
Results DISLEX Model – Modeling Impairment in one patient





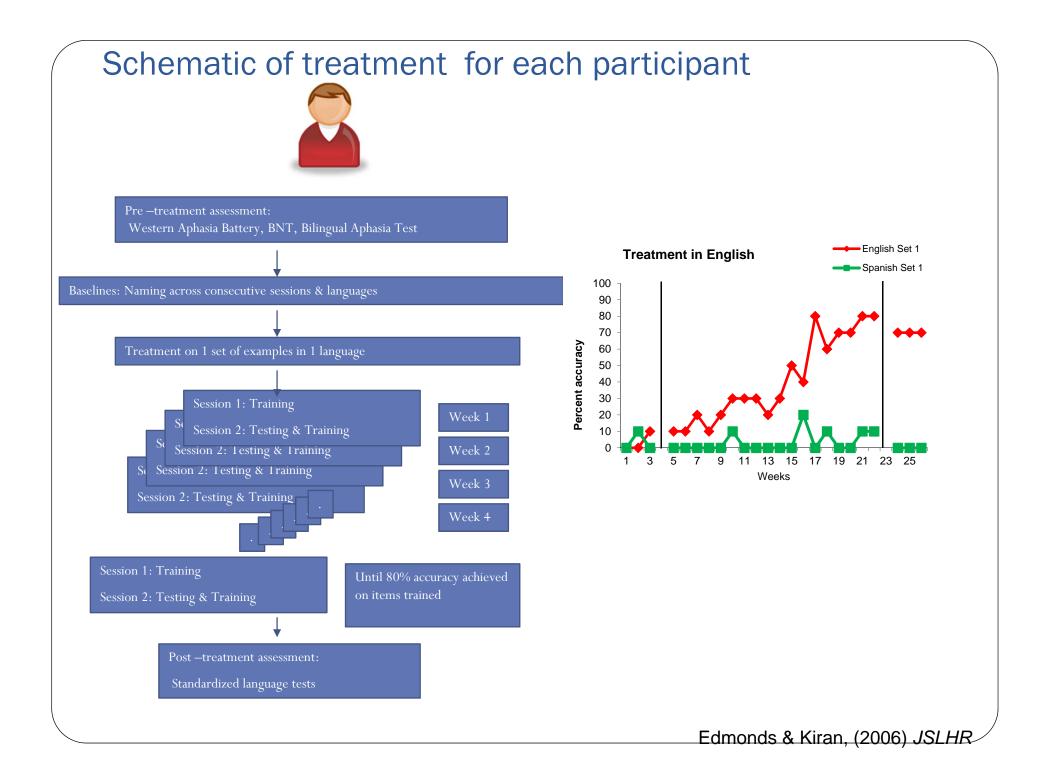
DISLEX Model: Modeling impairment for 16 patients with aphasia





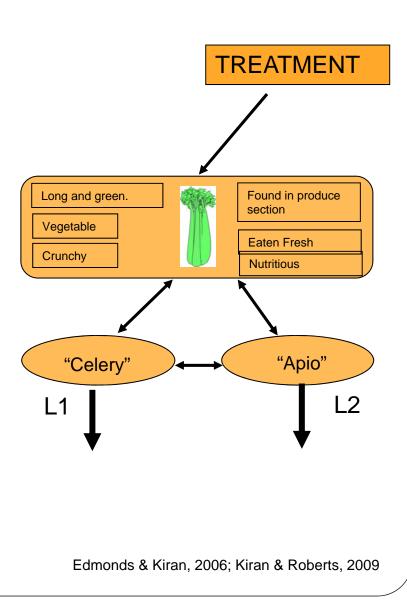
Patient Study 3: (N = 17)

	AOA		Lifetime Exposure		Treatment Effect Size	
	English	Spanish	English	Spanish	English	Spanish
UTBA01	Native	native	high	low	12.70	0.58
UTBA02	late	native	low	high	4.95	11.08
UTBA07	native	native	moderate	moderate	3.11	12.41
UTBA09	early	native	moderate	moderate	2.07	10.97
UTBA11	late	native	moderate	high	14.90	1.15
UTBA16	native	native	high	low	6.82	0.83
UTBA17	early	native	high	low	5.32	1.19
UTBA18	late	native	moderate	moderate	1.73	15.17
BUBA01	late	native	low	high	4.92	1.42
BUBA04	early	native	high	low	2.61	16.50
BUBA07	late	native	low	high	2.89	4.08
UTBA19	late	native	low	high	1.44	4.90
UTBA20	late	native	low	high	0	0
UTBA21	early	native			0	0
UTBA22	late	native	low	high	0.13	12.73
UTBA23	early	native	low	high	10.68	13.84
BUBA12	late	native	low	high	8.16	0



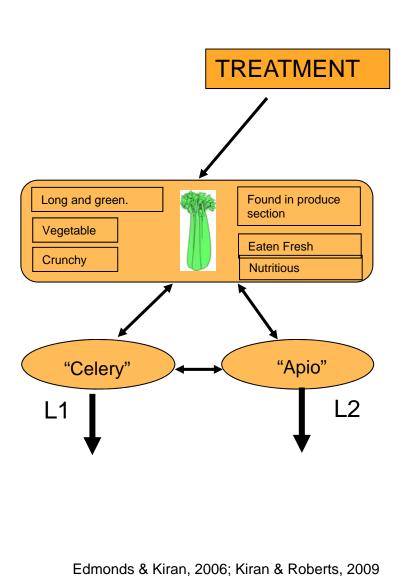
Treatment protocol in Behavioral studies

- 1. Name picture
- 2. If incorrect, told correct name
- 3. Choose 6 correct features from 12 cards
- 4. Answer 15 yes/no questions about the item
- 5. Named item again with feedback
- Treatment always provided only in one language (either English/Spanish) and amount of improvement examined
- Generalization (cross language transfer) examined to untrained language



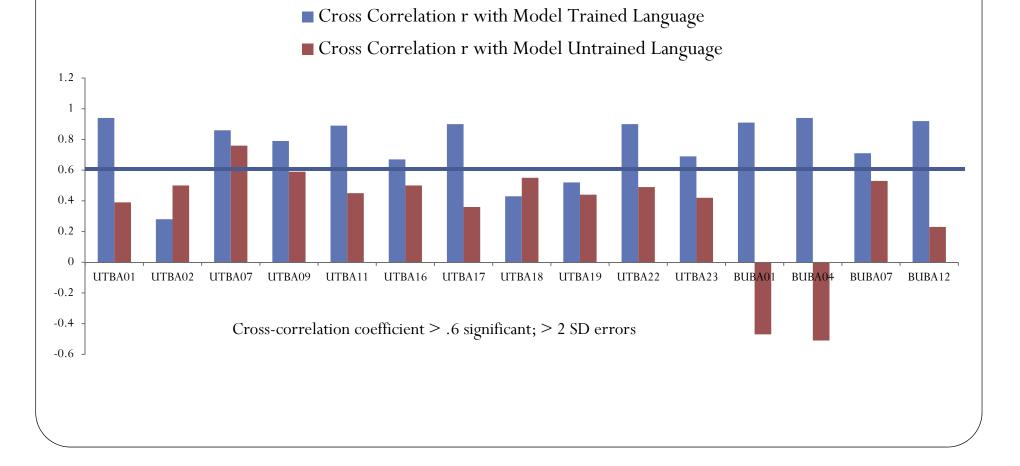
Rehabilitation in the DISLEX Model

- The starting point was set to either a severe impairment in naming (30% or less accuracy) or mild impairment (70% or high naming accuracy).
- Model retrained trained with different number and schedule of presentations of words in one language
- Treatment always provided only in one language (either English/Spanish) and amount of improvement examined
- Generalization (cross language transfer) examined to untrained language

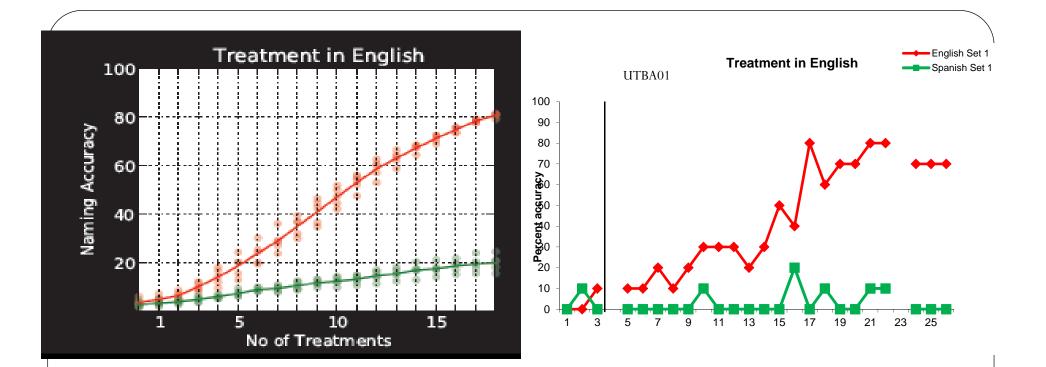


In order to evaluate the model

• Match the patient and model's parameters on AoA, exposure and damage parameters and see if the model's predictions match the actual data obtained.



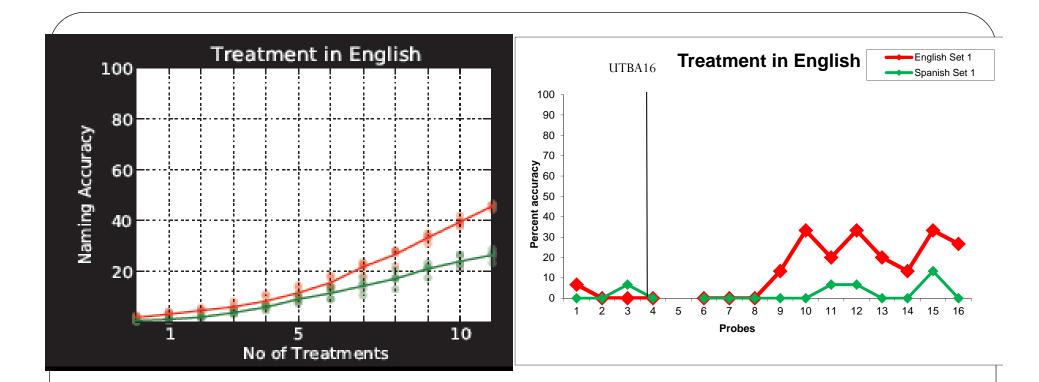
Patient and computational results: Both languages high damage



UTBA 01: English: Early Spanish: Native

English: High exposure Spanish: Low exposure

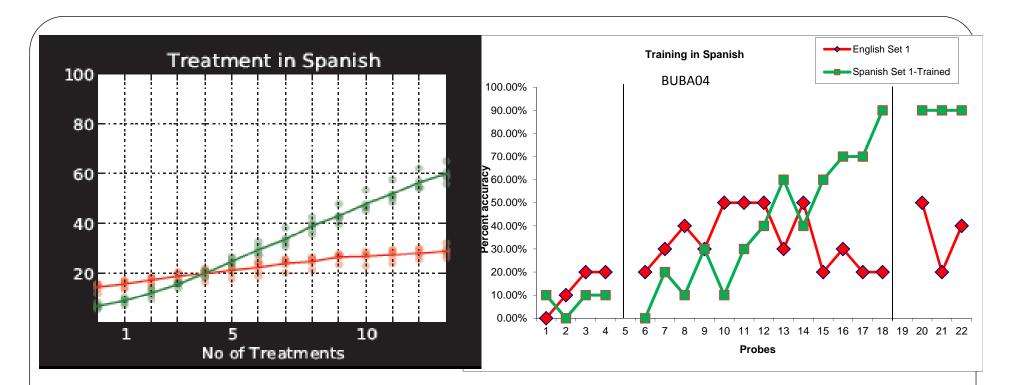
Spanish ES: .58 English ES: 12.7



UTBA16: English: Early Spanish: Native

English: Moderate exposure Spanish: Moderate exposure

Spanish ES: .83 English ES: 6.8

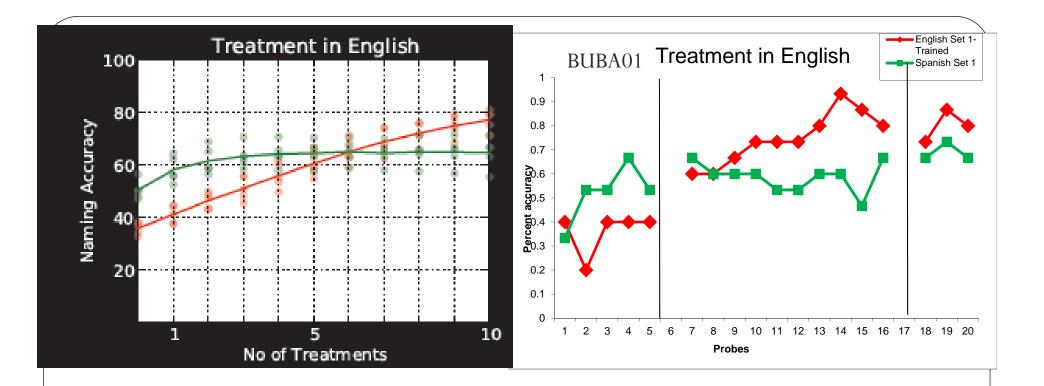


BUBA04: English: Early Spanish: Native

English: High exposure Spanish: Low exposure

Spanish ES: 16.5 English ES: 2.52

Patient and computational results: Both languages low damage

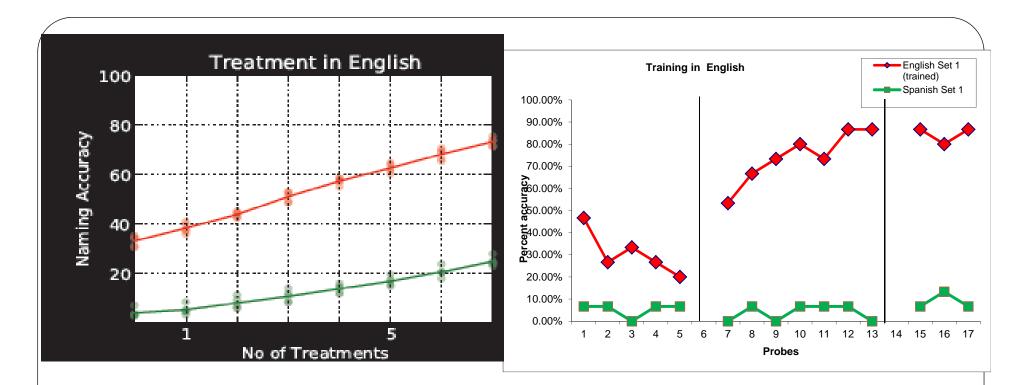


BUBA01 English: Late

Spanish: Native

English: Low exposure Spanish: High exposure

Spanish ES: 1.42 English ES: 4.92 Patient and computational results: Both languages differential damage

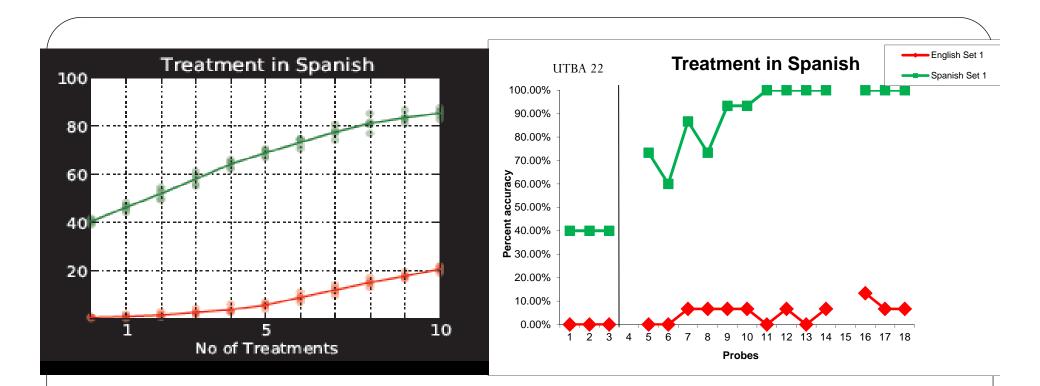


UTBA 17:

English: Early Spanish: Native

English: Moderate exposure Spanish: Moderate exposure

Spanish ES: 5.32 English ES: 1.19

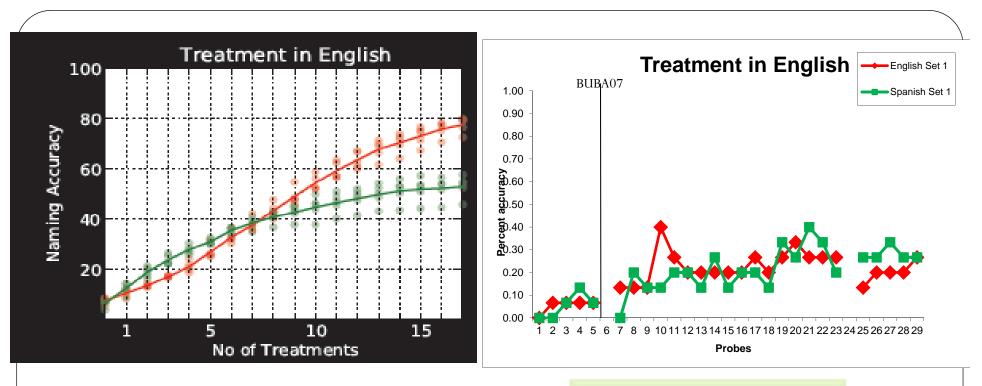


UTBA 22: English: Late Spanish: Native

English: Low exposure Spanish: High exposure

Spanish ES: 12.7 English ES: 1.89

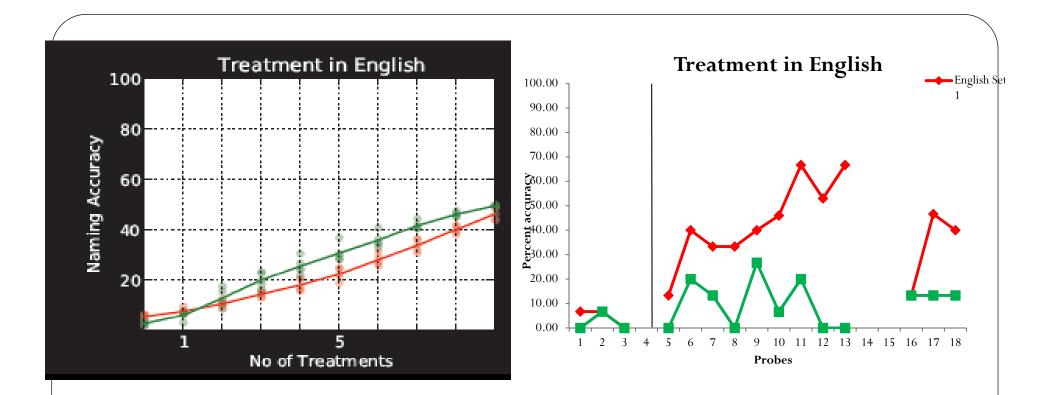
The model also does not always predict correct performance



BUBA07 English: Late Spanish: Native

English: Low exposure Spanish: High exposure

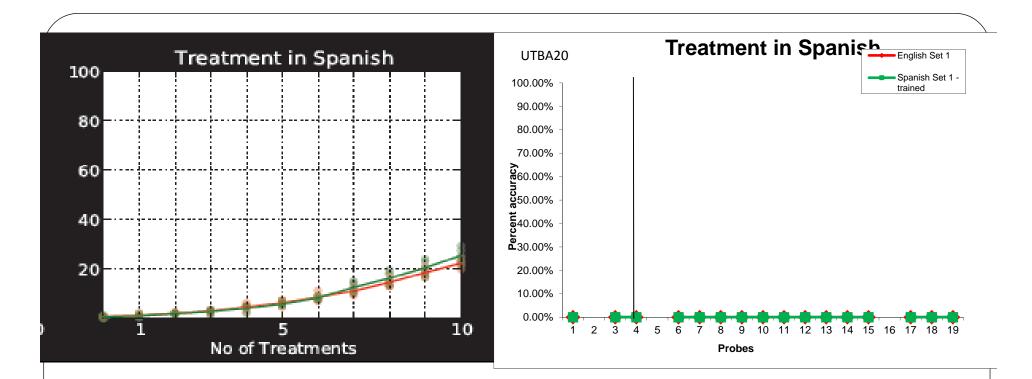
Spanish ES: 4.08 English ES: 2.8



UTBA11 English: Late Spanish: Native

English: High exposure Spanish: High exposure

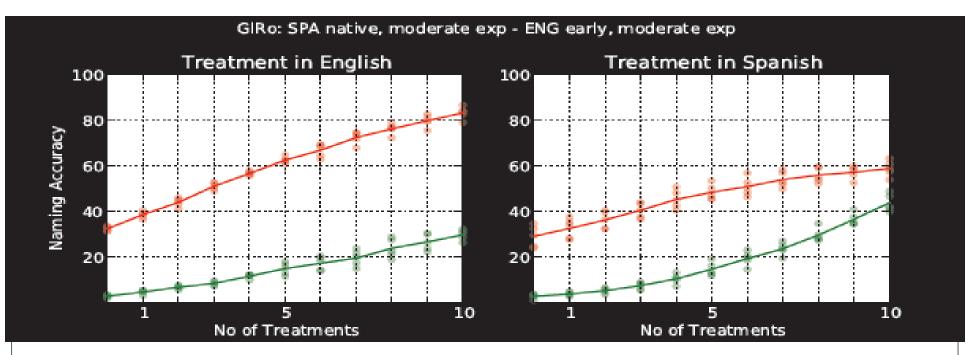
Spanish ES: 1.15 English ES: 14.9



UTBA 09: English: Late Spanish: Native

English: Low exposure Spanish: High exposure

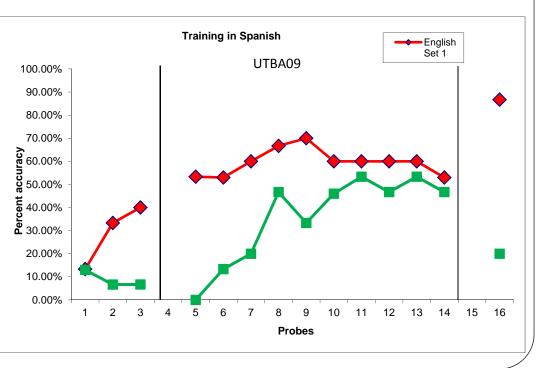
Spanish ES: 0 English ES: 0 The model can also predicts what treatment outcome may have been if the other language was trained



UTBA 09: English: Early Spanish: Native

English: Moderate exposure Spanish: Moderate exposure

Spanish ES: 10.97 English ES: 2.07



Summary

- Model can predict rehabilitation outcomes
 - Of the 17 patients, good fit for 12 patients,
 - For patients that do not have a good fit, model overestimates outcomes
 - Education/literacy issues in patients
 - Severe phonological output deficits
 - Severity of language/cognitive issues
- Provides a starting point for understanding why patient did not improve
- Model can also predict what treatment outcome may have been if treatment plan was different that what was followed...

Conclusions

- These results highlight the important interaction between language proficiency, stroke impairment and language recovery
 - No individual factor can independently predict the amount of treatment recovery.
 - Training always improves the trained language but cross language transfer depends on AoA, amount of language exposure pre-stroke and extent of nature of stroke impairment
 - e.g., in individuals with late AoA, low exposure and high damage, cross-language transfer is less likely to occur
 - e.g., in individual with early AoA and moderate-high exposure, cross language transfer may be expected.

Conclusions..

• While preliminary, the combination of computational modeling and behavioral treatment provide a promising approach to examining the important issue of recovery of language in bilingual aphasia



Uli Grasemann UT-Austin



Risto Miikkulainen UT-Austin



Chaleece Sandberg Boston University

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