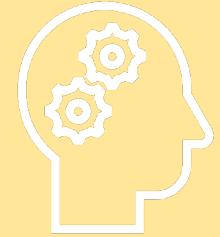


# Translation equivalents are not special in bilingual infant vocabulary development: Evidence from a quantitative model



CONCORDIA  
INFANT RESEARCH  
LABORATORY  
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RECHERCHE  
SUR L'ENFANCE DE  
CONCORDIA



Rachel Ka-Ying Tsui



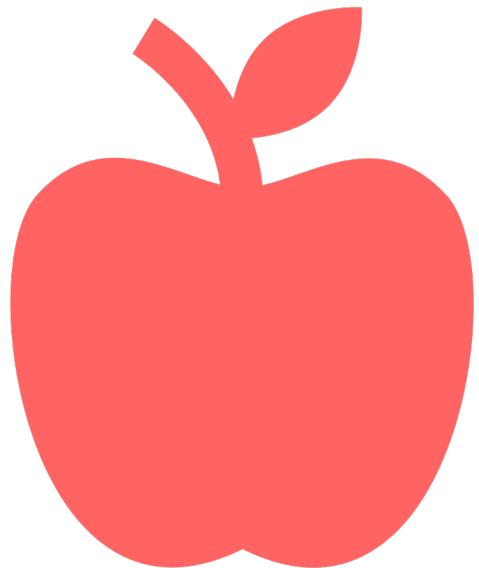
Ana Maria Gonzalez-Barrero



Esther Schott



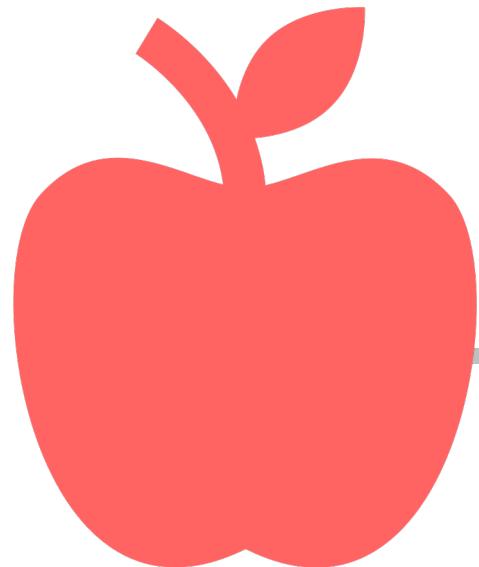
Krista Byers-Heinlein



Apple

Pomme

**Translation equivalents:**  
2 labels for the same concept



Apple

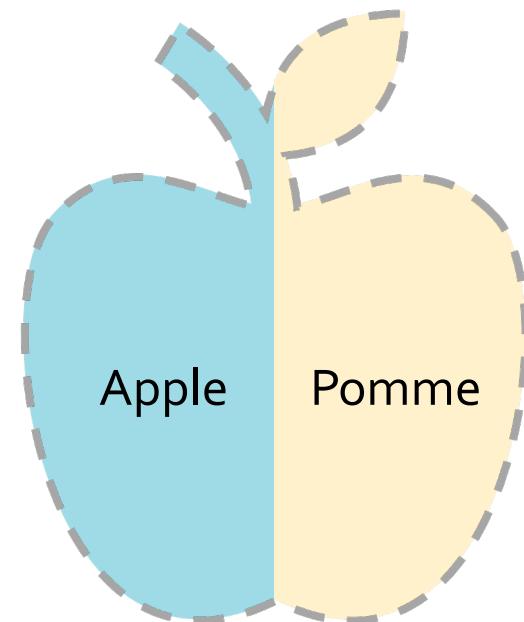
Pomme

**Singlet:**  
A first label for a concept

**Translation equivalents**

# Translation equivalents are special...

- Learned differently from singlets
- Strong semantic overlap



# 3 competing theories: How are translation equivalents learned?

## Account #1

Bilingual children reject translation equivalents in favour of learning one label for each referent  
(Volterra & Taeschner, 1978)

## Account #2

Bilingual children favour learning translation equivalents  
(Bilson et al., 2015; Floccia et al., 2020)

## Avoidance Account

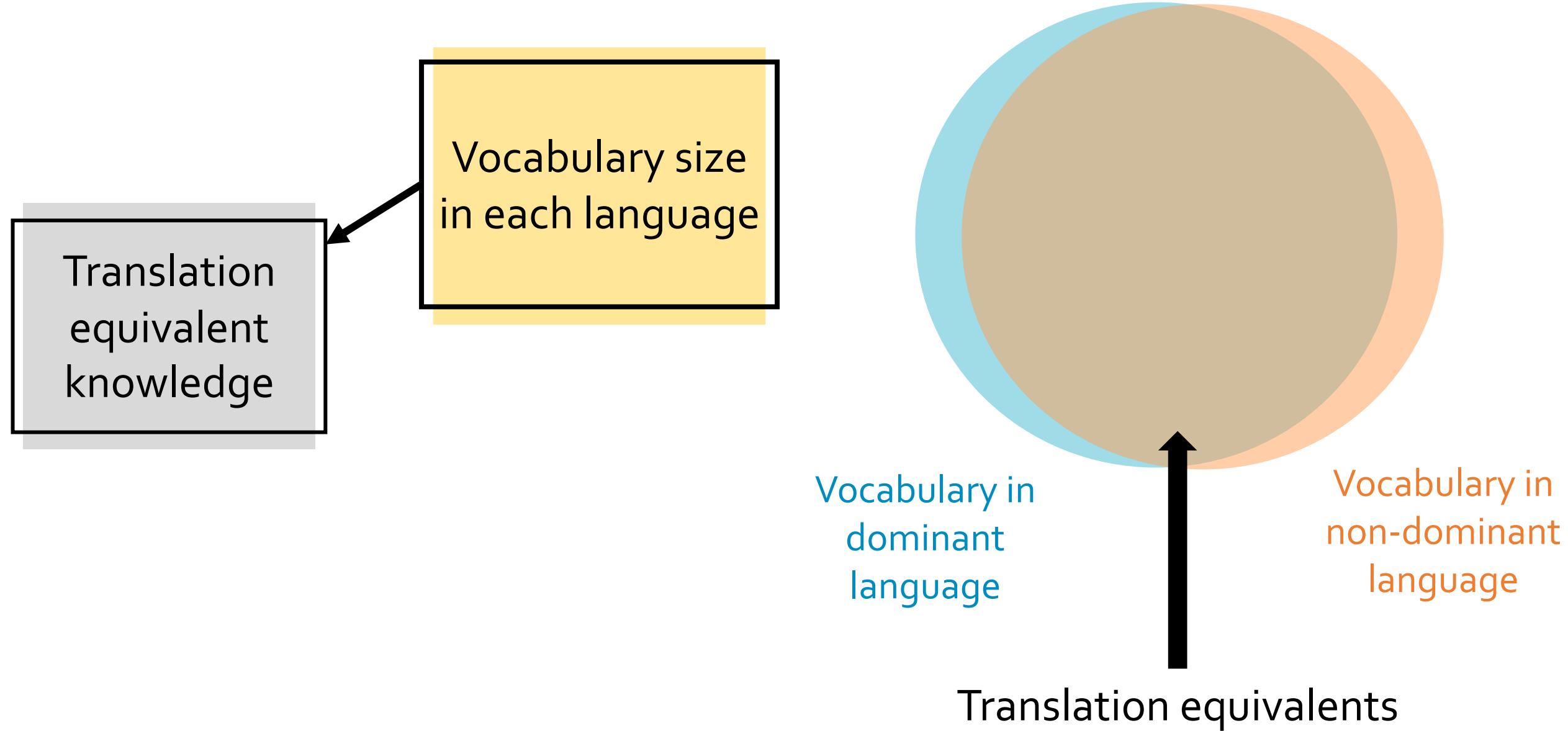
## Account #3

Bilingual children learn translation equivalents and singlets in a similar way  
(Pearson et al., 1995)

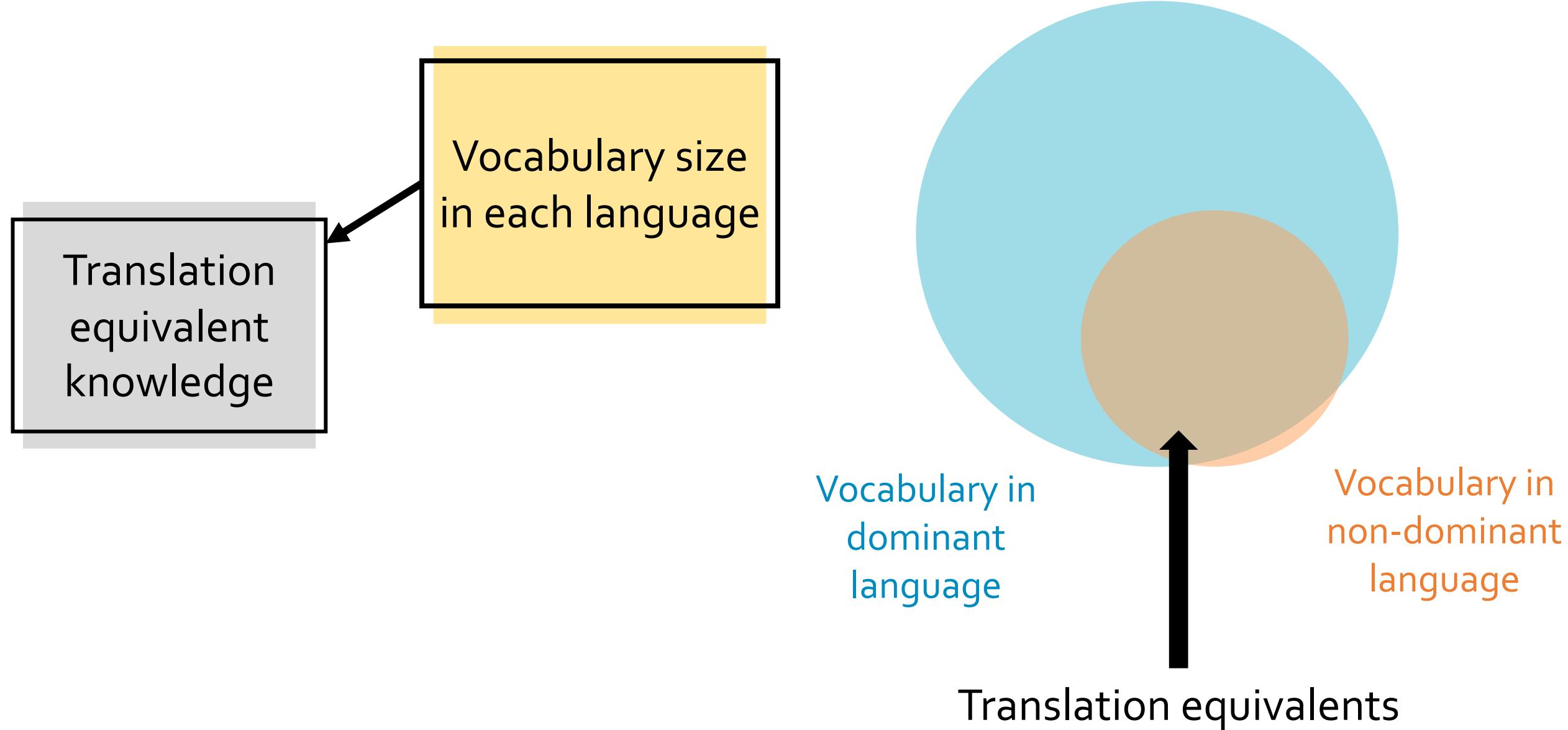
## Preference Account

## Neutral Account

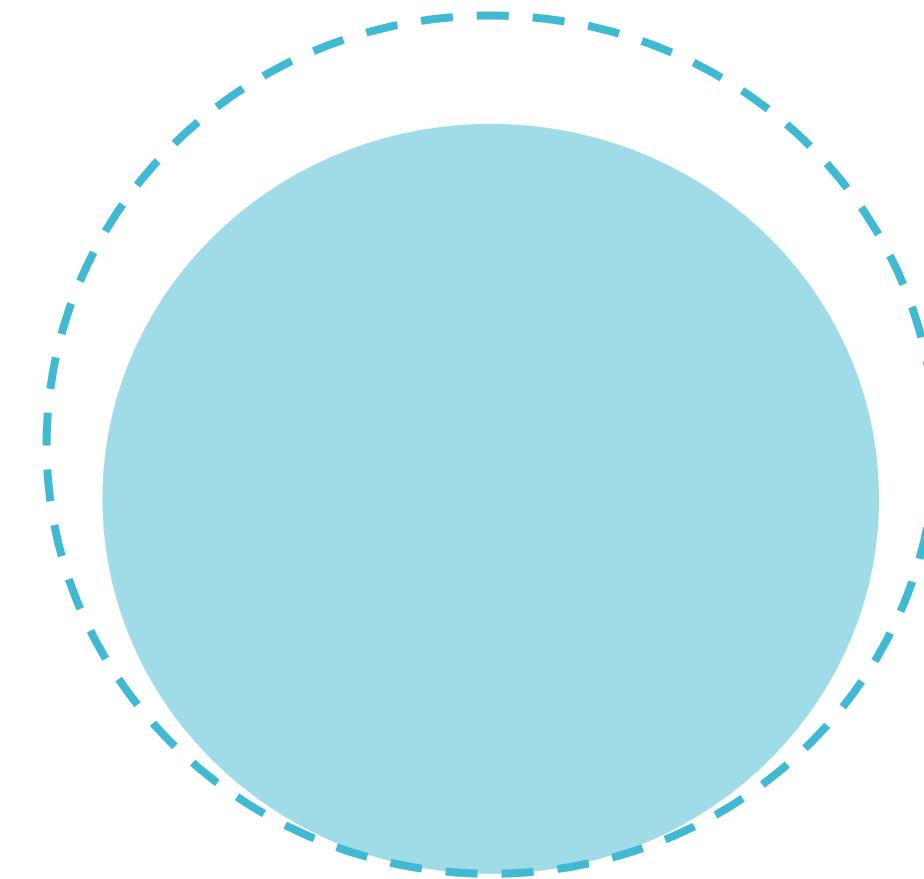
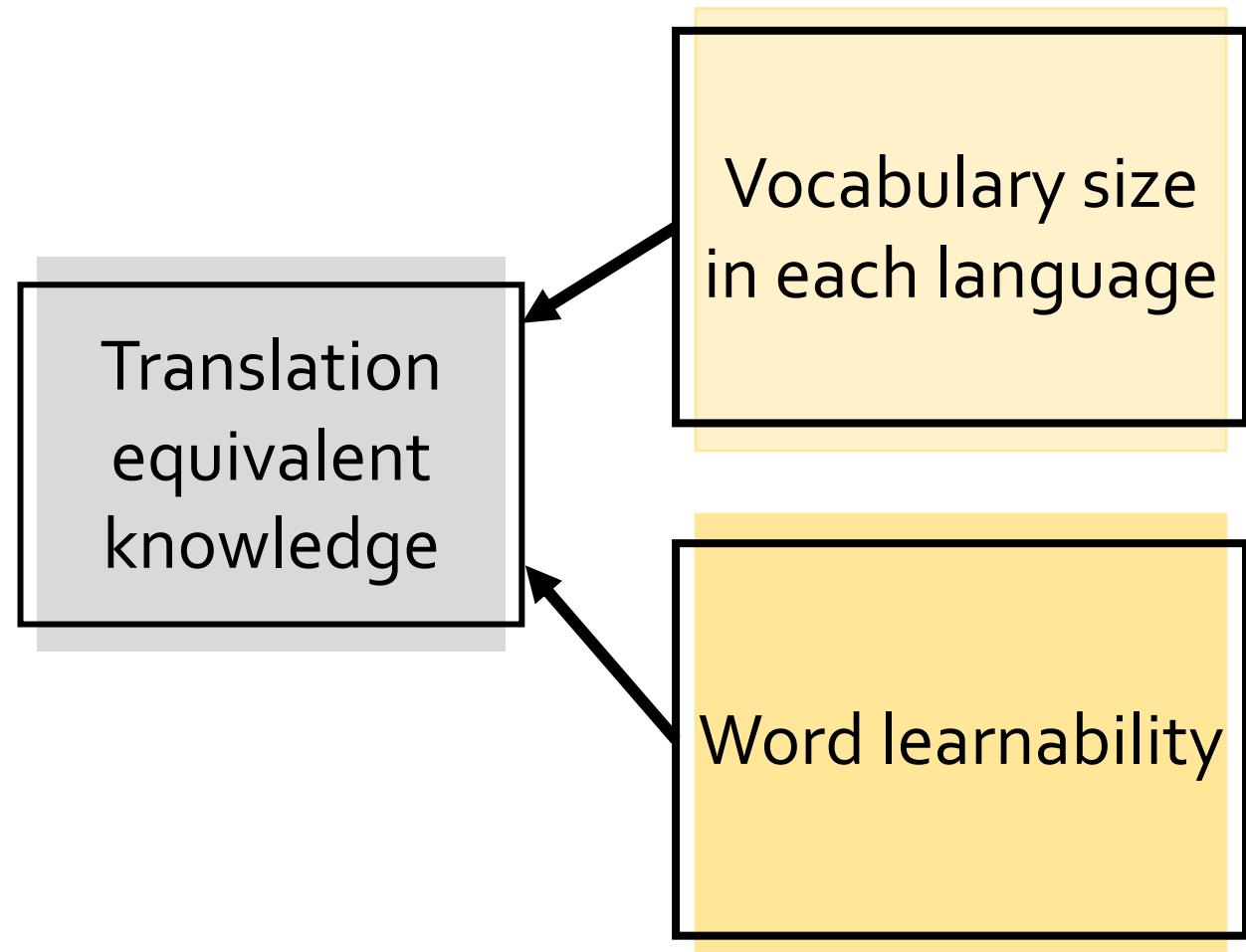
# Contributors to translation equivalent knowledge



# Contributors to translation equivalent knowledge

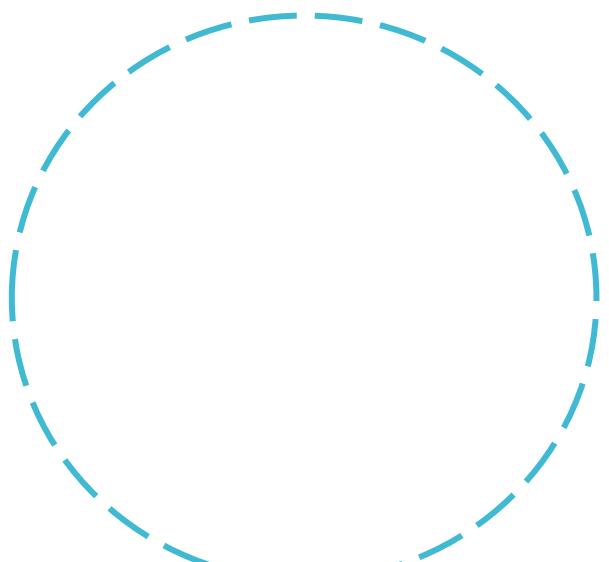


# Contributors to translation equivalent knowledge

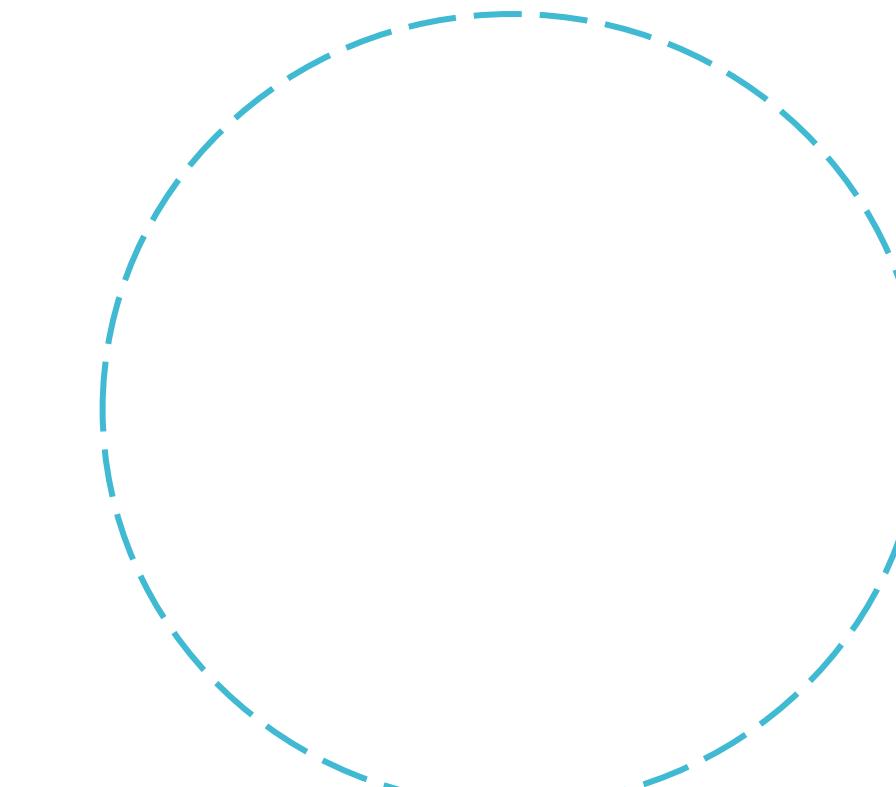


Vocabulary in  
dominant language

# Contributors to translation equivalent knowledge

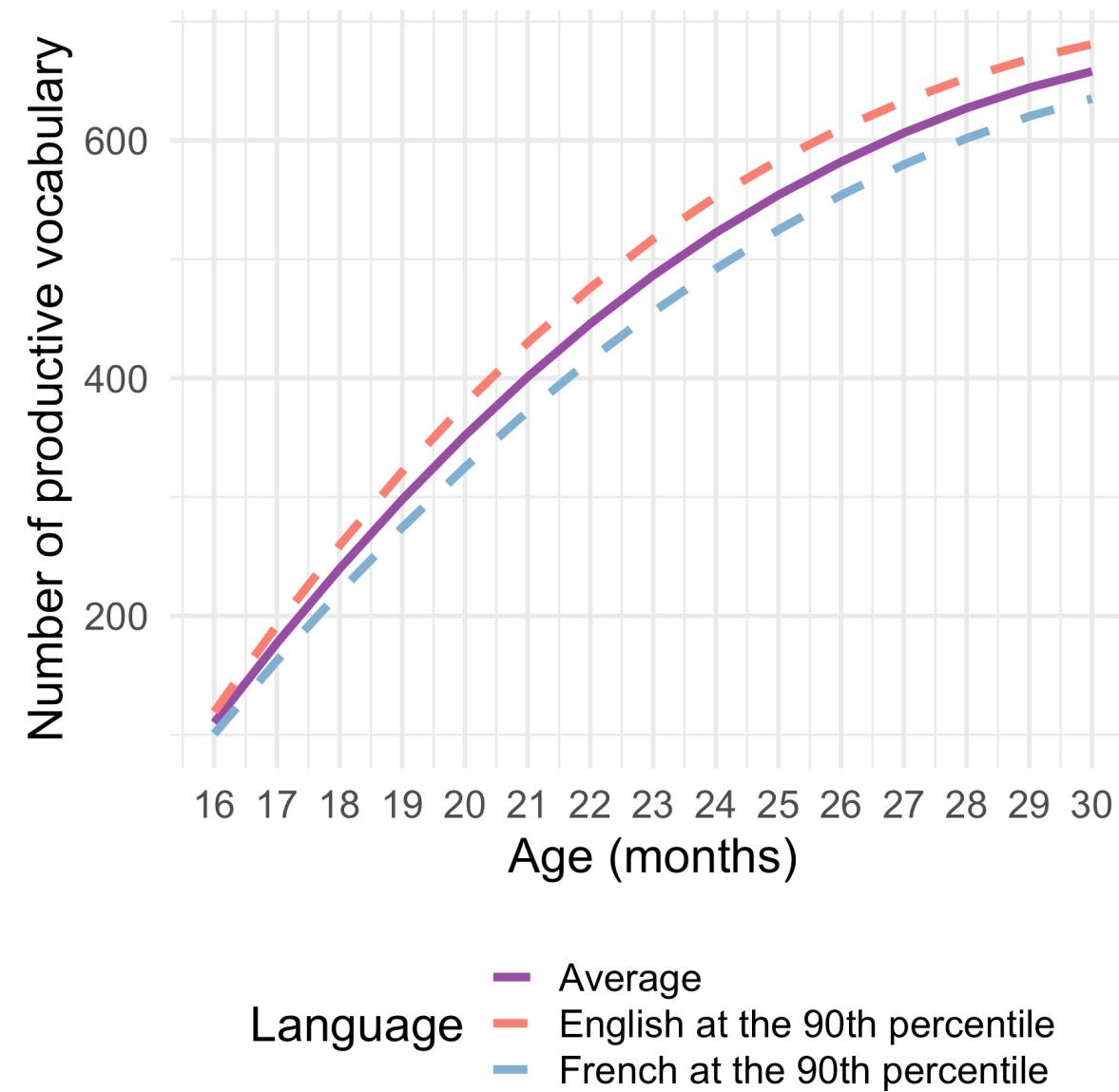


Vocabulary size of  
a 18-month-old



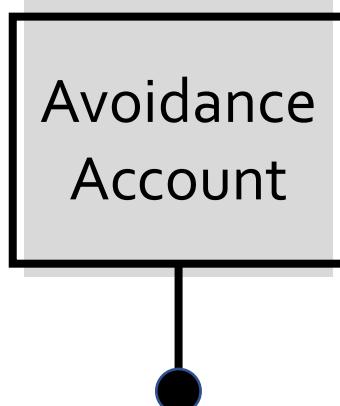
Vocabulary size of  
a 30-month-old

## Number of CDI words produced at the 90<sup>th</sup> percentile

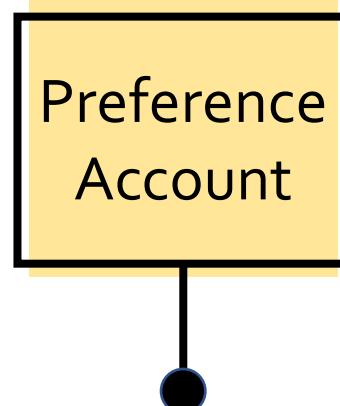


# Our study

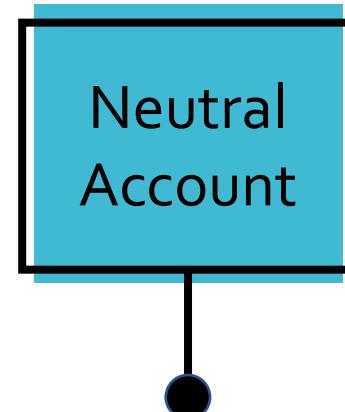
- Translation equivalent knowledge as a function of bilinguals' own vocabulary size in each language
- What is the nature of translation equivalent learning in bilingual children?



Translation equivalents are  
**harder** to learn than singlets

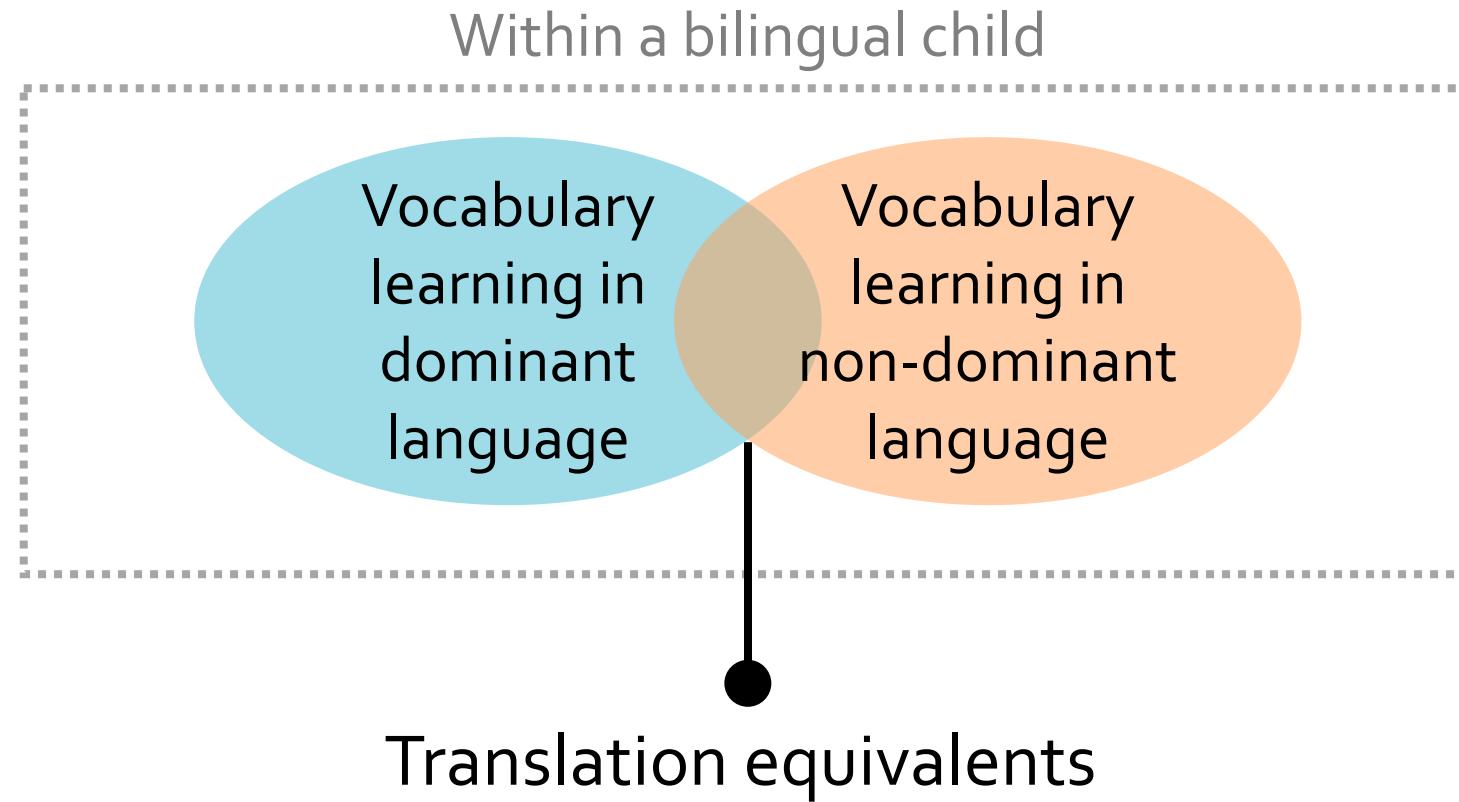


Translation equivalents are  
**easier** to learn than singlets

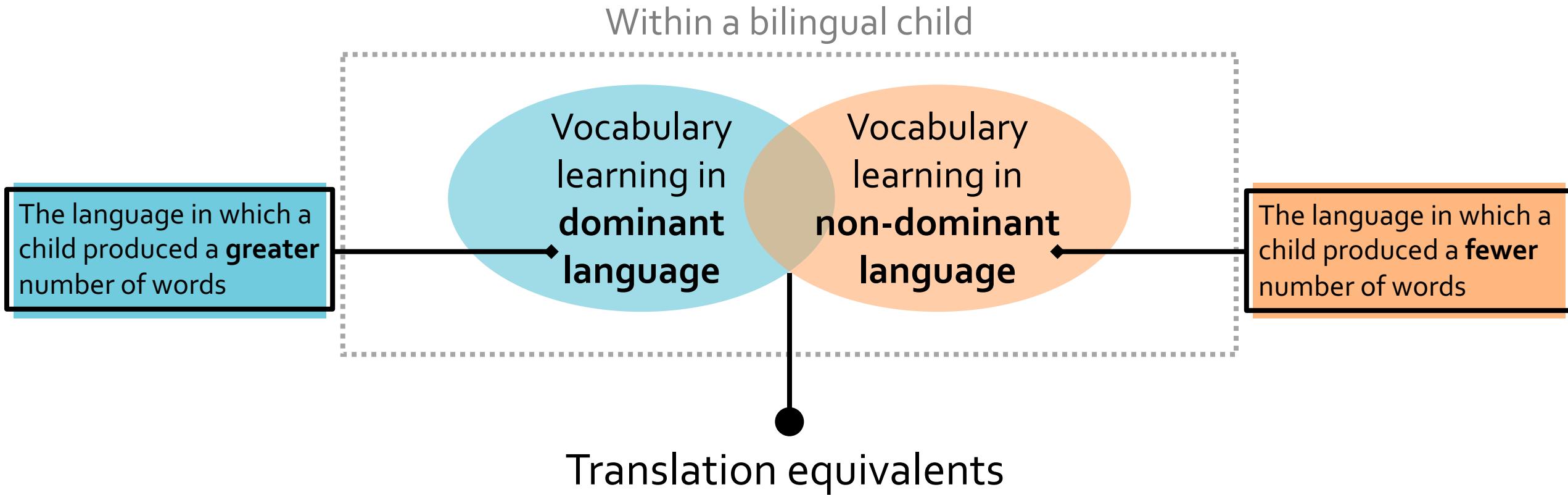


Translation equivalents are  
**similar** to learn as singlets

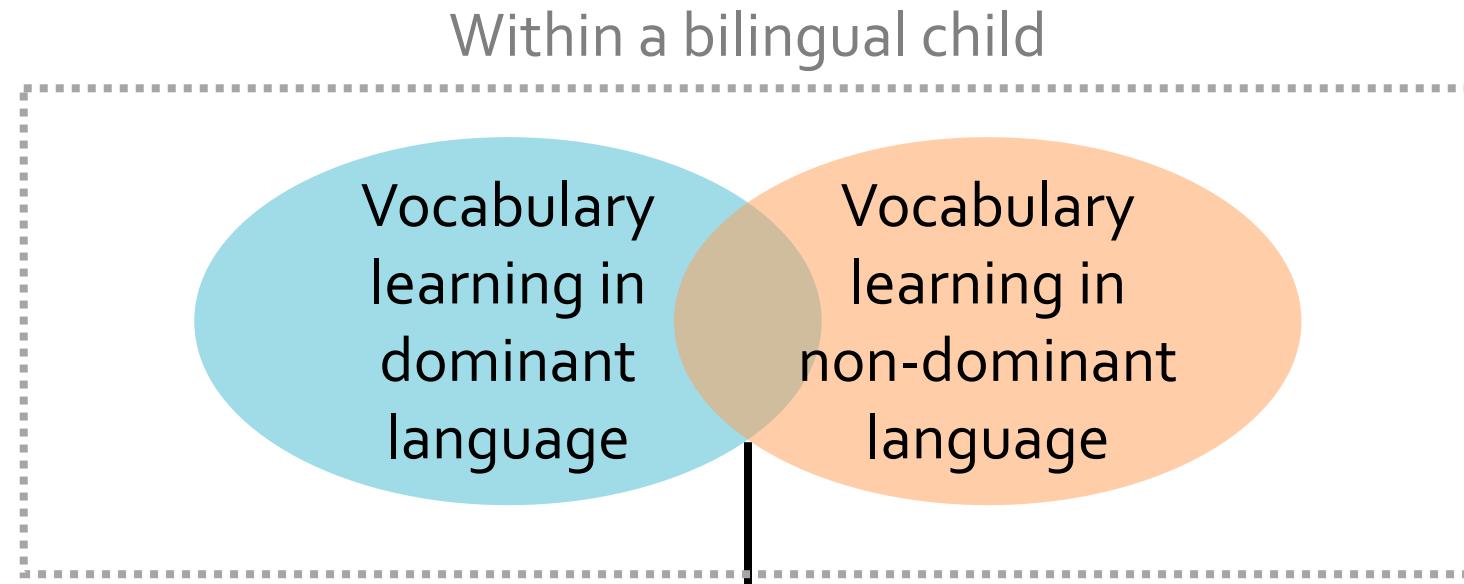
# Bilingual Vocabulary Model



# Bilingual Vocabulary Model



# Bilingual Vocabulary Model

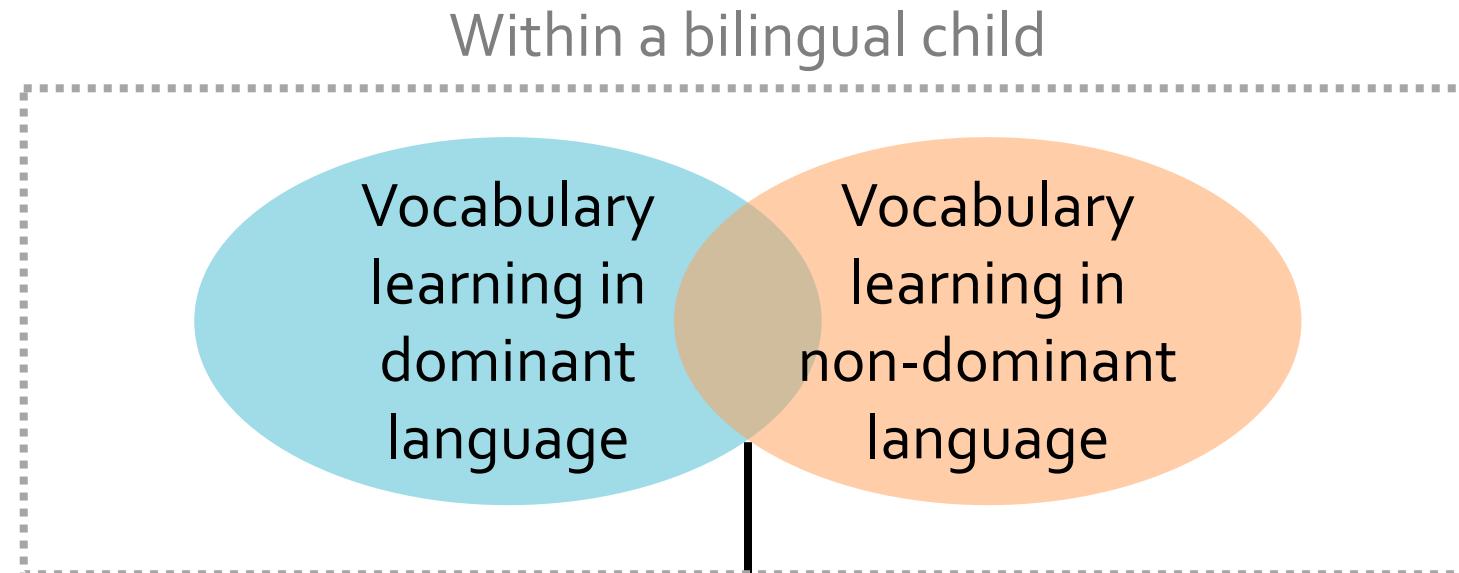


**Translation equivalents**

=

Joint probability of learning the words in each language

# Bilingual Vocabulary Model

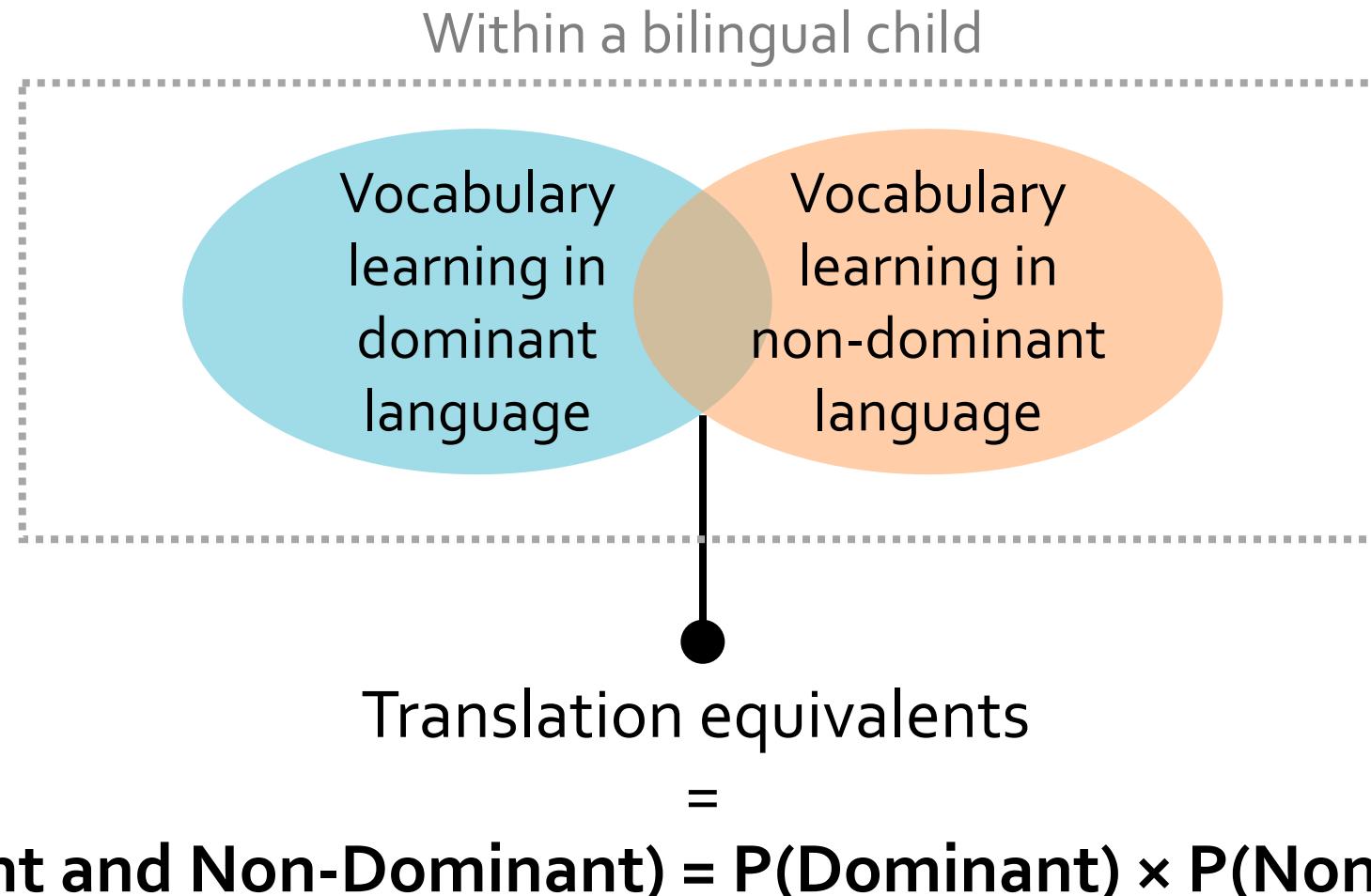


Translation equivalents

=

Independent event:  $P(A \text{ and } B) = P(A) \times P(B)$

# Bilingual Vocabulary Model

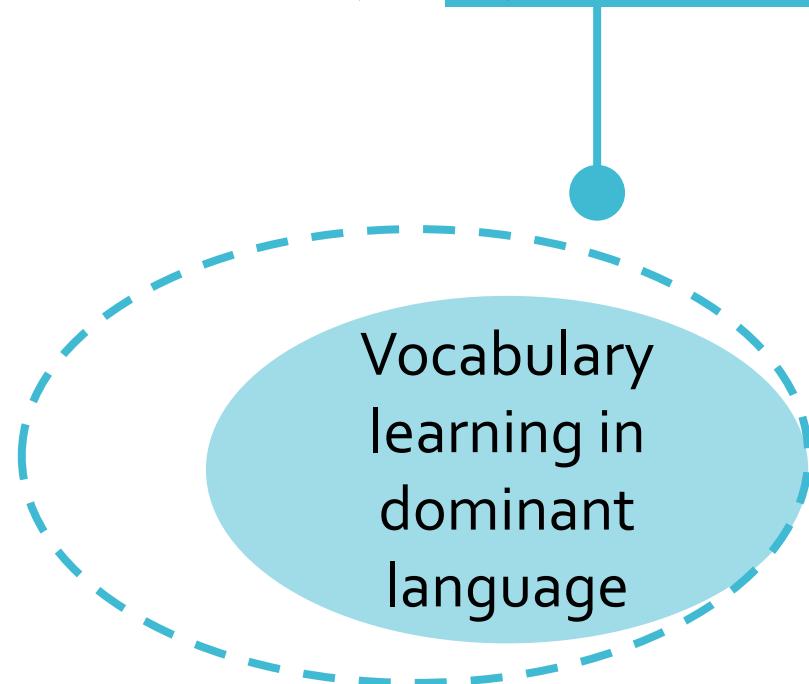


# Bilingual Vocabulary Model

**P(Dominant and Non-Dominant) = P(Dominant) × P(Non-Dominant)**

# Bilingual Vocabulary Model

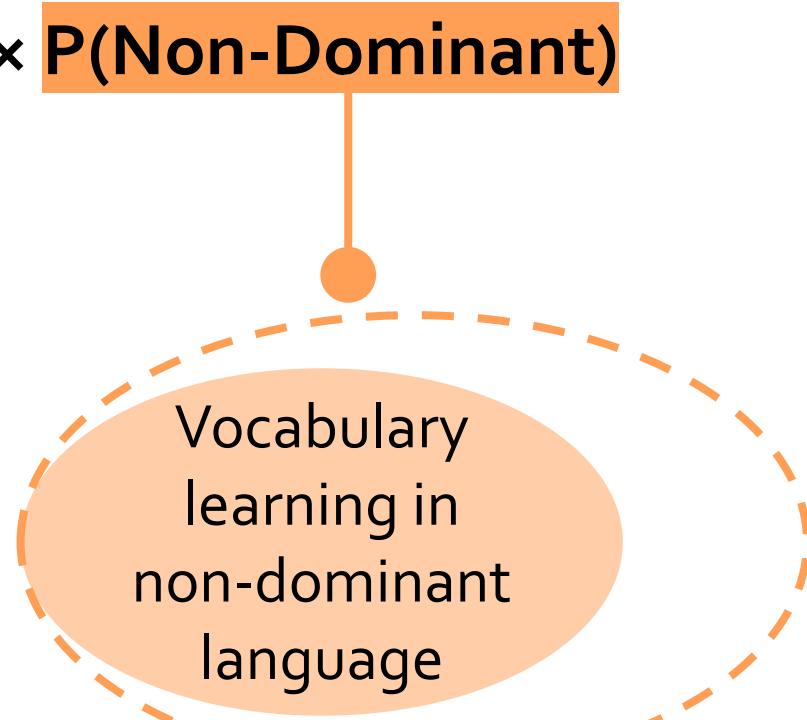
$$P(\text{Dominant and Non-Dominant}) = P(\text{Dominant}) \times P(\text{Non-Dominant})$$



$$\frac{\text{Number of dominant vocabulary known}}{\text{Number of learnable vocabulary}}$$

# Bilingual Vocabulary Model

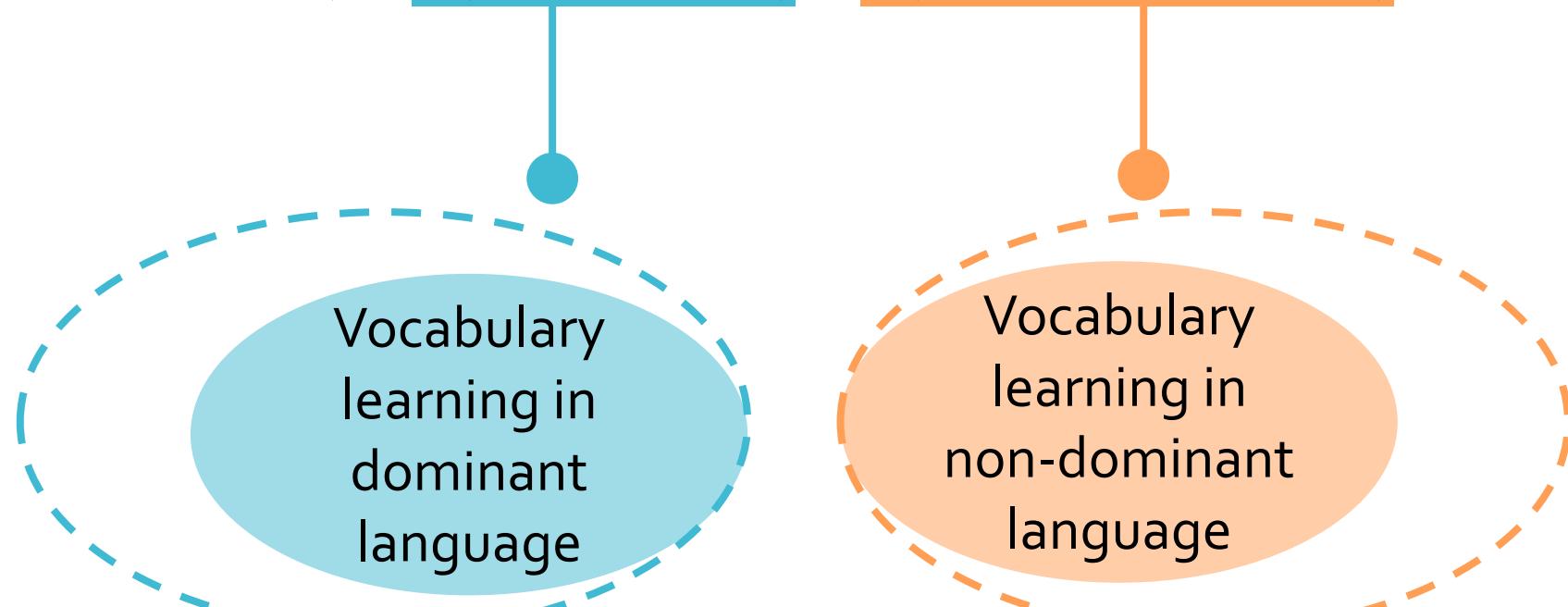
$$P(\text{Dominant and Non-Dominant}) = P(\text{Dominant}) \times P(\text{Non-Dominant})$$



$$\frac{\text{Number of non-dominant vocabulary known}}{\text{Number of learnable vocabulary}}$$

# Bilingual Vocabulary Model

$$P(\text{Dominant and Non-Dominant}) = P(\text{Dominant}) \times P(\text{Non-Dominant})$$

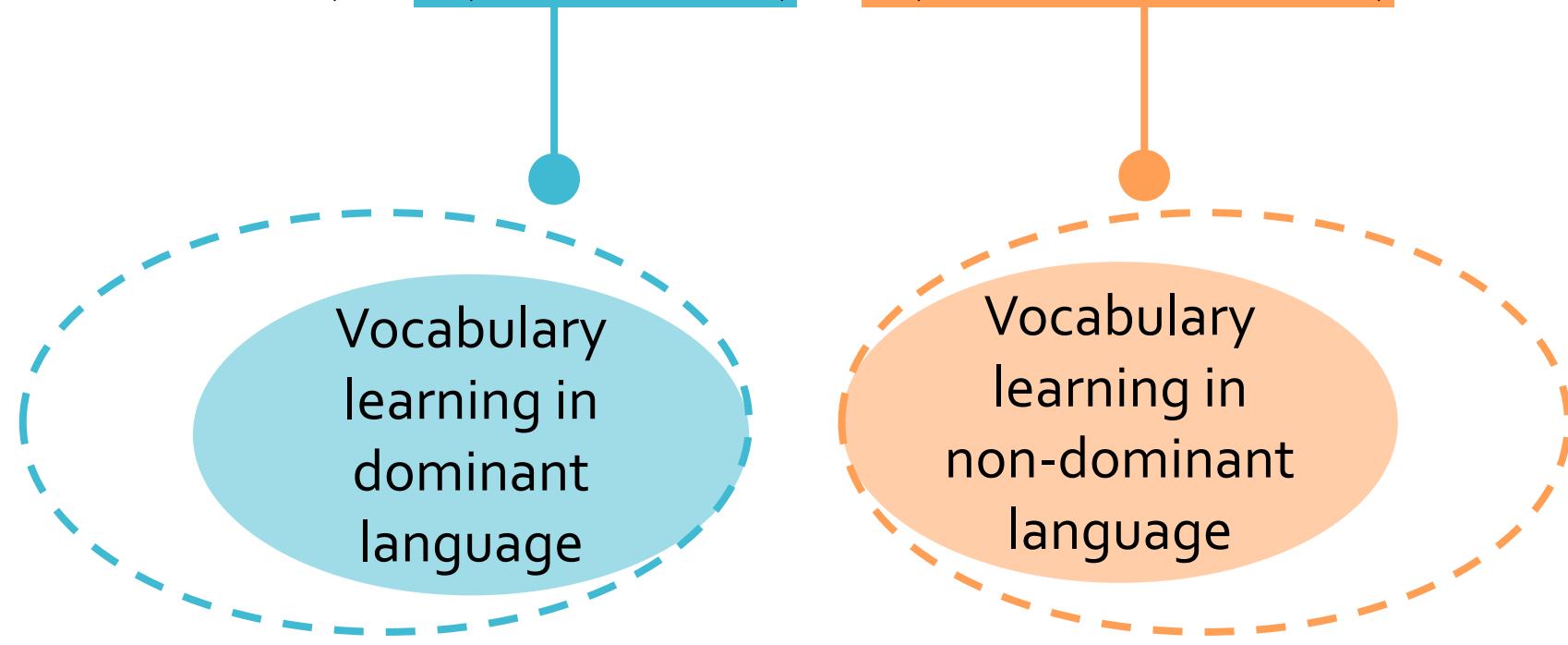


90<sup>th</sup> percentile for a  
21-month-old child

$\frac{300 \text{ dominant vocabulary known}}{400 \text{ learnable vocabulary}}$

# Bilingual Vocabulary Model

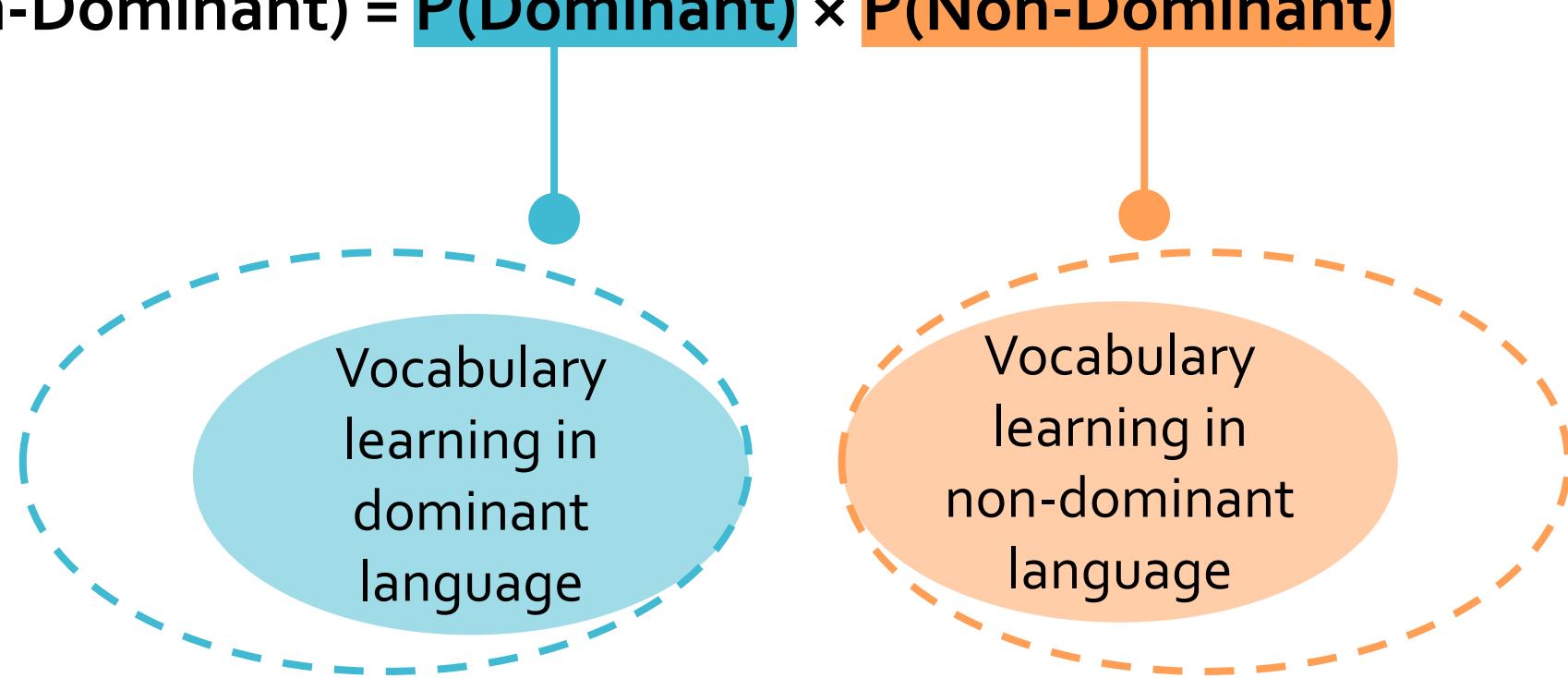
$$P(\text{Dominant and Non-Dominant}) = P(\text{Dominant}) \times P(\text{Non-Dominant})$$



$$P(\text{Dominant}) = \frac{300}{400}$$

# Bilingual Vocabulary Model

$$P(\text{Dominant and Non-Dominant}) = P(\text{Dominant}) \times P(\text{Non-Dominant})$$



$$P(\text{Dominant}) = \frac{300}{400}$$

$$P(\text{Non-Dominant}) = \frac{100}{400}$$

# Bilingual Vocabulary Model

$$P(\text{Dominant and Non-Dominant}) = P(\text{Dominant}) \times P(\text{Non-Dominant})$$

$$\frac{300}{400}$$

$$\frac{100}{400}$$

# Bilingual Vocabulary Model

$$P(\text{Dominant and Non-Dominant}) = P(\text{Dominant}) \times P(\text{Non-Dominant})$$

$$\frac{300}{400}$$

$$\frac{100}{400}$$

Expected(Dominant and Non-Dominant) =

# Bilingual Vocabulary Model

$$P(\text{Dominant and Non-Dominant}) = P(\text{Dominant}) \times P(\text{Non-Dominant})$$

$$\frac{300}{400}$$

$$\frac{100}{400}$$

Expected number of translation equivalents =

# Bilingual Vocabulary Model

$$P(\text{Dominant and Non-Dominant}) = P(\text{Dominant}) \times P(\text{Non-Dominant})$$

$$\frac{300}{400}$$

$$\frac{100}{400}$$

**Expected number of translation equivalents =**

P(Dominant and Non-Dominant)  $\times$  Number of learnable vocabulary

# Bilingual Vocabulary Model

$$P(\text{Dominant and Non-Dominant}) = P(\text{Dominant}) \times P(\text{Non-Dominant})$$

$$\frac{300}{400}$$

$$\frac{100}{400}$$

Expected number of translation equivalents =

$$P(\text{Dominant and Non-Dominant}) \times \text{Number of learnable vocabulary}$$

$$\frac{300}{400} \times \frac{100}{400}$$

# Bilingual Vocabulary Model

$$P(\text{Dominant and Non-Dominant}) = P(\text{Dominant}) \times P(\text{Non-Dominant})$$

$$\frac{300}{400}$$

$$\frac{100}{400}$$

Expected number of translation equivalents =

$$P(\text{Dominant and Non-Dominant}) \times \text{Number of learnable vocabulary}$$

$$\frac{300}{400} \times \frac{100}{400} \times 400$$

# Bilingual Vocabulary Model

**P(Dominant and Non-Dominant) = P(Dominant) × P(Non-Dominant)**

$$\frac{300}{400}$$

$$\frac{100}{400}$$

**Expected number of translation equivalents =**

$$\frac{300 \times 100}{400}$$

# Bilingual Vocabulary Model

$$P(\text{Dominant and Non-Dominant}) = P(\text{Dominant}) \times P(\text{Non-Dominant})$$

$$\frac{300}{400}$$

$$\frac{100}{400}$$

Expected number of translation equivalents =

$$\frac{\text{No. of dominant vocabulary} \times \text{No. of non-dominant vocabulary}}{\text{No. of learnable vocabulary}} = 75$$

No. of dominant vocabulary      No. of non-dominant vocabulary  
 $\frac{300 \times 100}{400}$   
No. of learnable vocabulary

To evaluate if translation equivalents are learned independently,

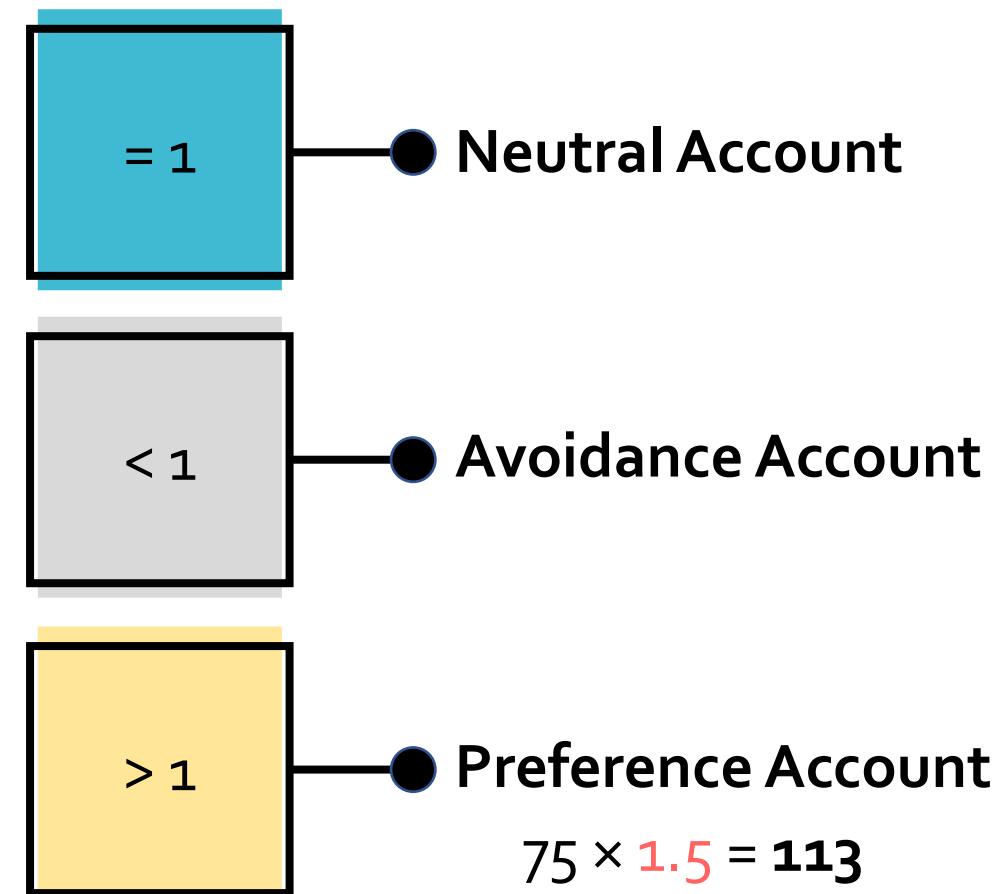
$$\text{Expected no. of Translation equivalents} = \frac{\text{No. of dominant vocabulary} \times \text{No. of non-dominant vocabulary}}{\text{No. of learnable vocabulary}}$$

To evaluate if translation equivalents are learned independently,

$$\text{Expected no. of Translation equivalents} = \frac{\text{No. of dominant vocabulary} \times \text{No. of non-dominant vocabulary}}{\text{No. of learnable vocabulary}} \times \text{Bias parameter}$$

To evaluate if translation equivalents are learned independently,

$$\text{Expected no. of Translation equivalents} = \frac{\text{No. of dominant vocabulary} \times \text{No. of non-dominant vocabulary}}{\text{No. of learnable vocabulary}} \times \text{Bias parameter}$$



# Validating the Bilingual Vocabulary Model

- 
- 1** Running simulations under the Neutral Account
  - 2** Testing the bias parameter with real-life observed data

# 1 Simulation

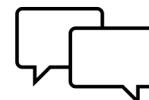
## Simulated data

- 216 simulated children
- Generated from a range of possible dominant vocabulary from 100 to 600, and a range of non-dominant vocabulary from 0 to 600



## Observed data

- Archival data collected in Montréal (2010 to 2018)
- 200 English-French bilingual children (18 – 33 months)
- MacArthur-Bates Communicative Development Inventories: Words and Sentences:
  - English (Fenson et al., 2007) and
  - Canadian French (Trudeau et al., 1997)

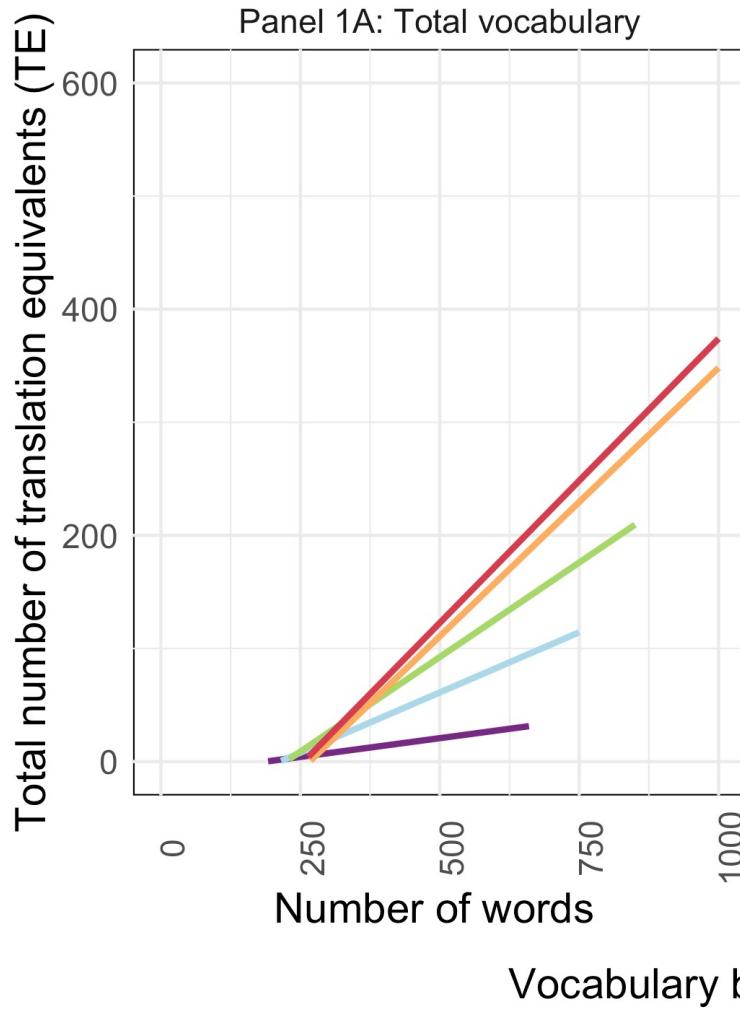


## 611 translation equivalents

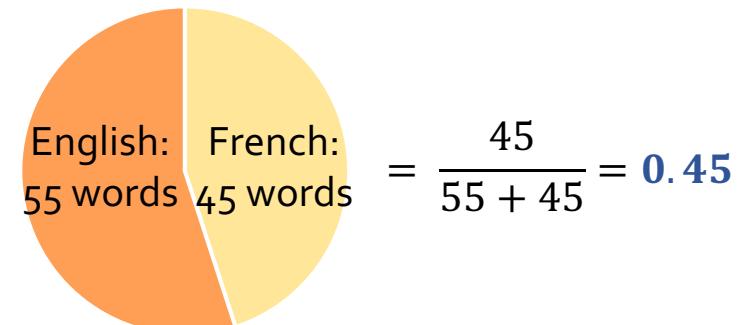
- Identified by 3 proficient bilingual French–English adults

# 1 Simulation

## Simulated data



Proportion of words produced in the non-dominant language relative to the total vocabulary produced

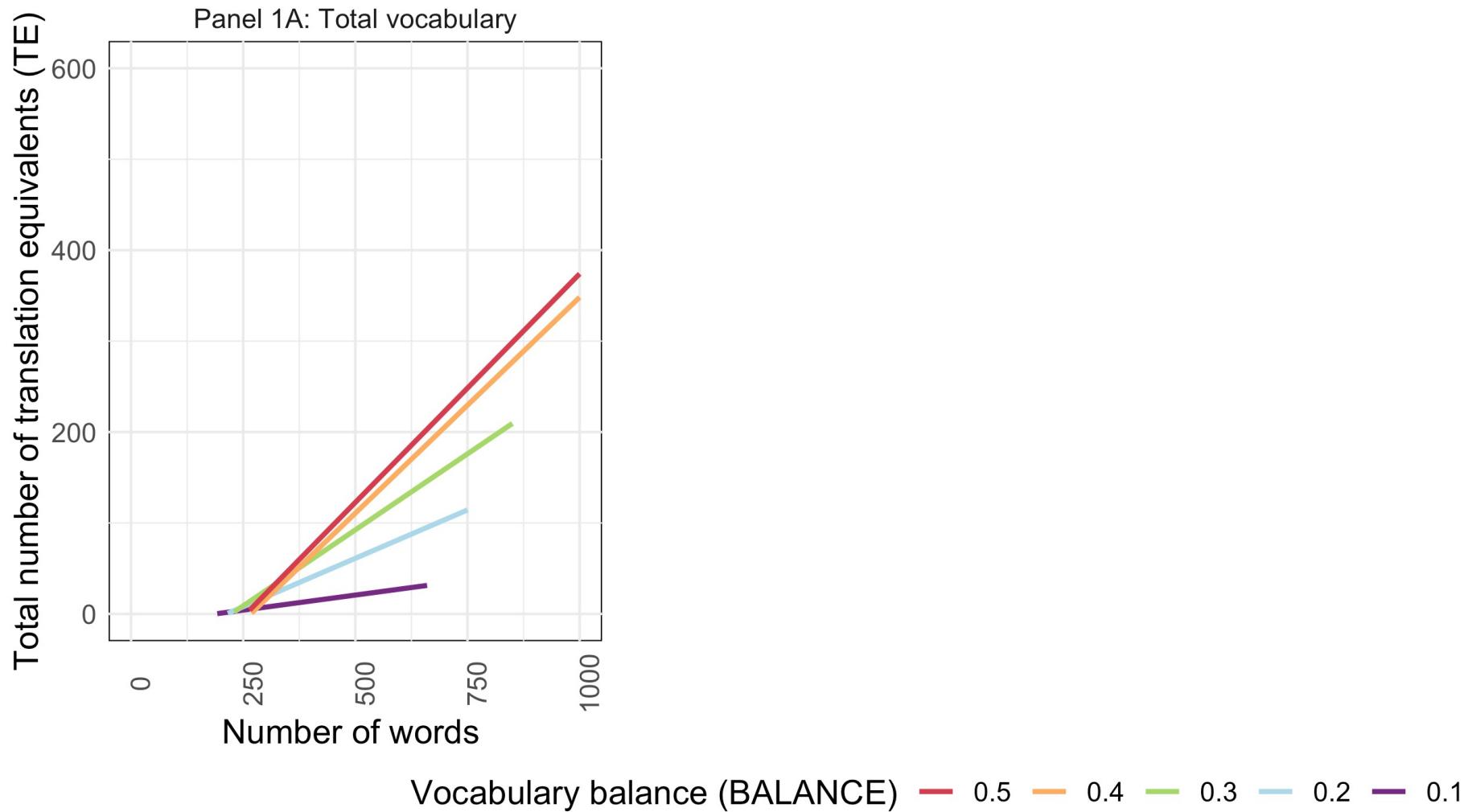


Most balanced ~0.5 ← → Least balanced ~0.1

Vocabulary balance (BALANCE) — 0.5 — 0.4 — 0.3 — 0.2 — 0.1

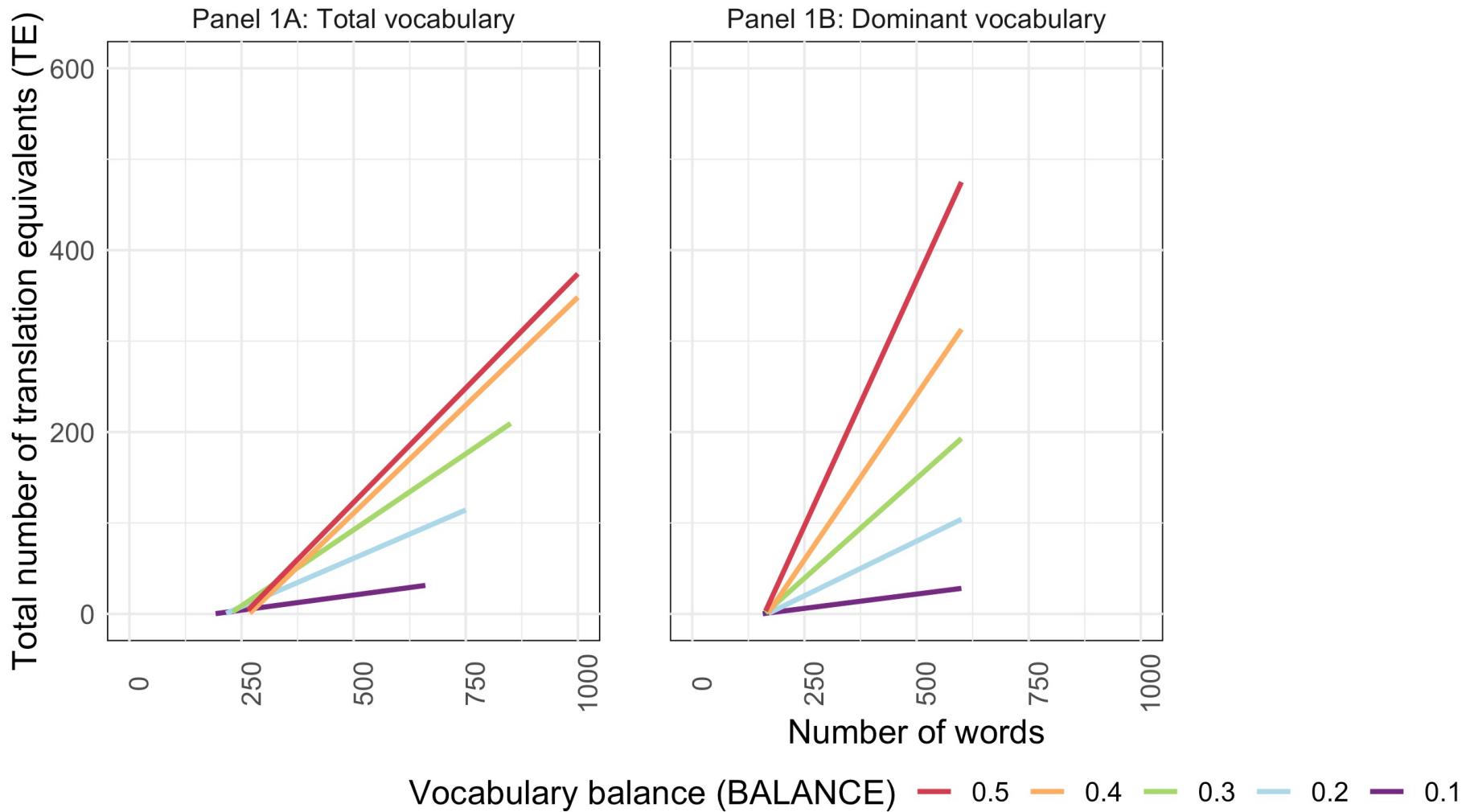
# 1 Simulation

## Simulated data



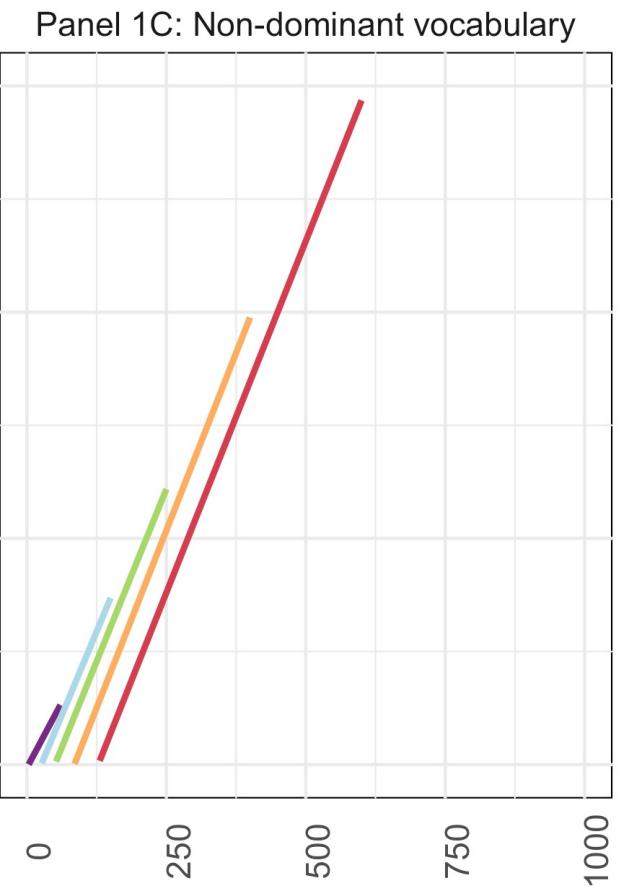
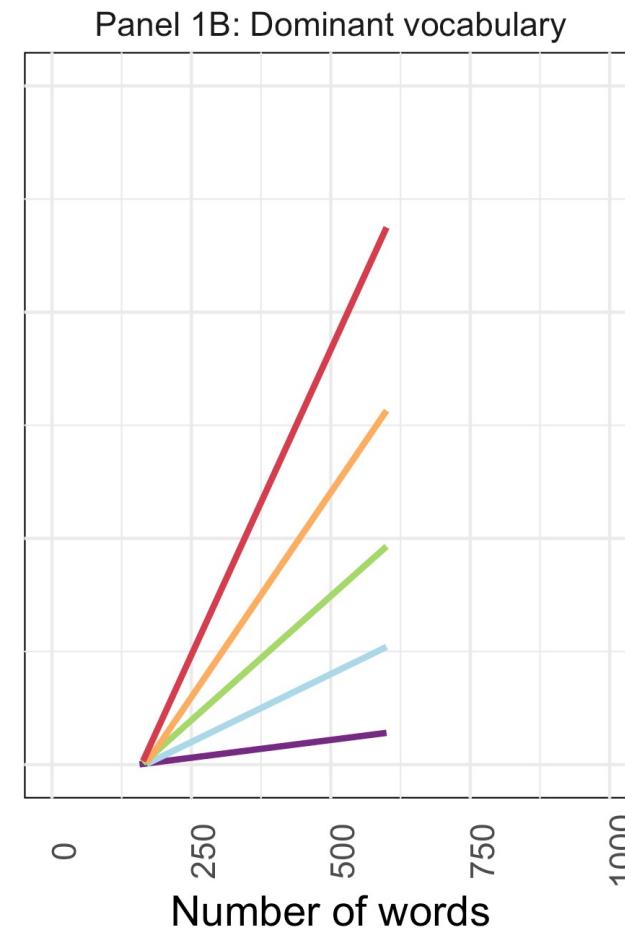
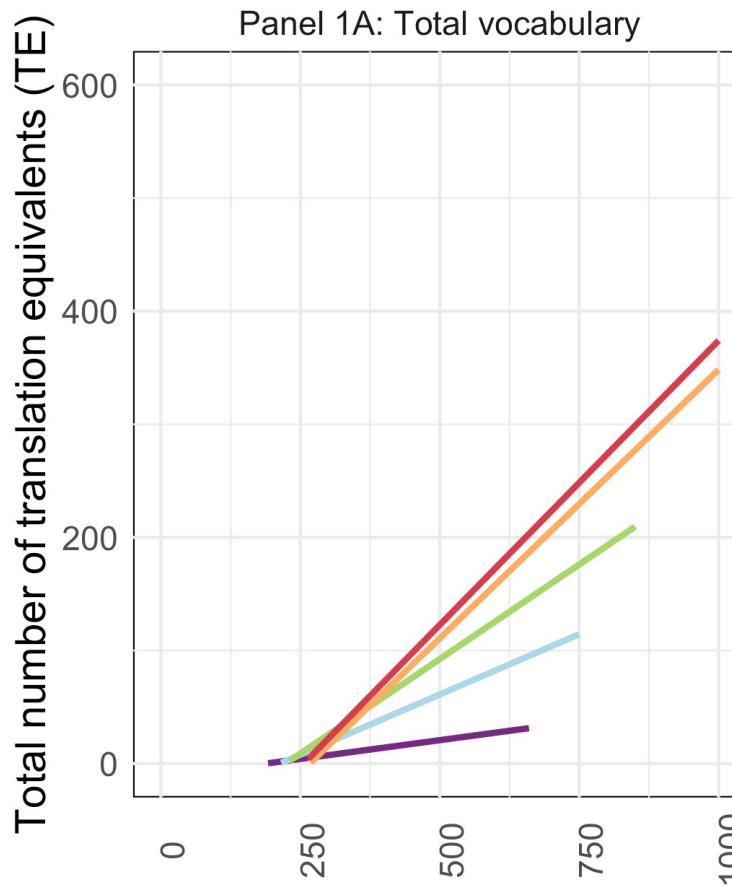
# 1 Simulation

## Simulated data



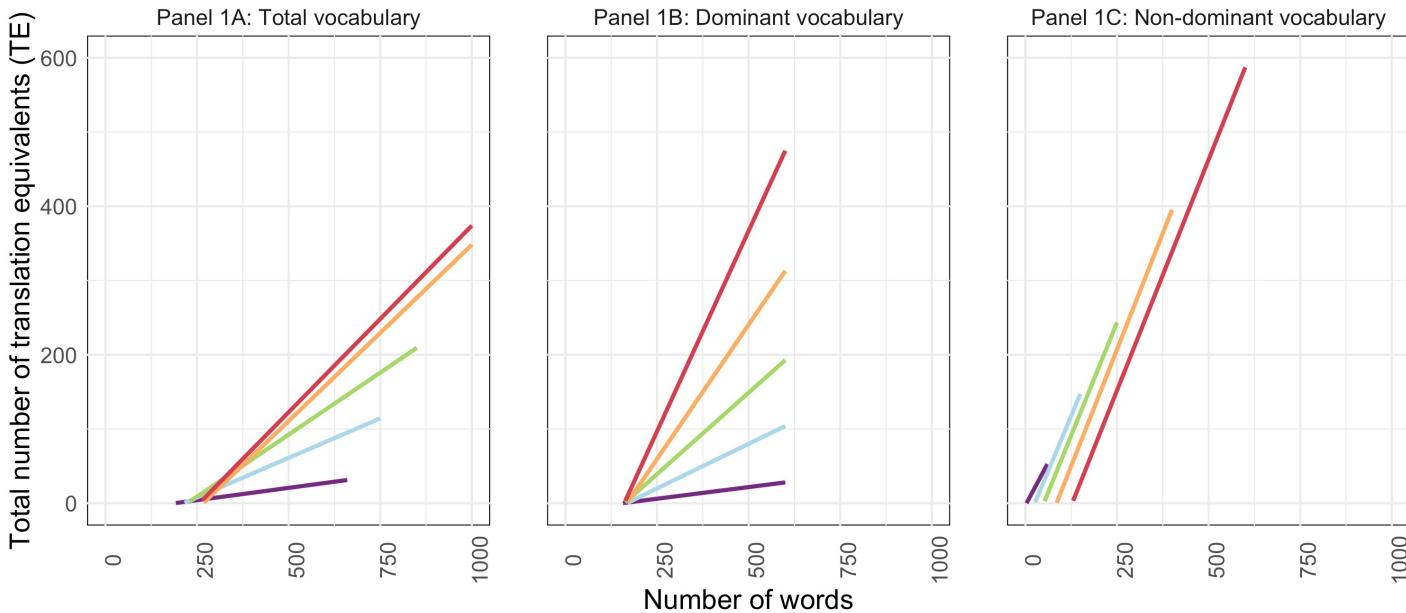
# 1 Simulation

## Simulated data

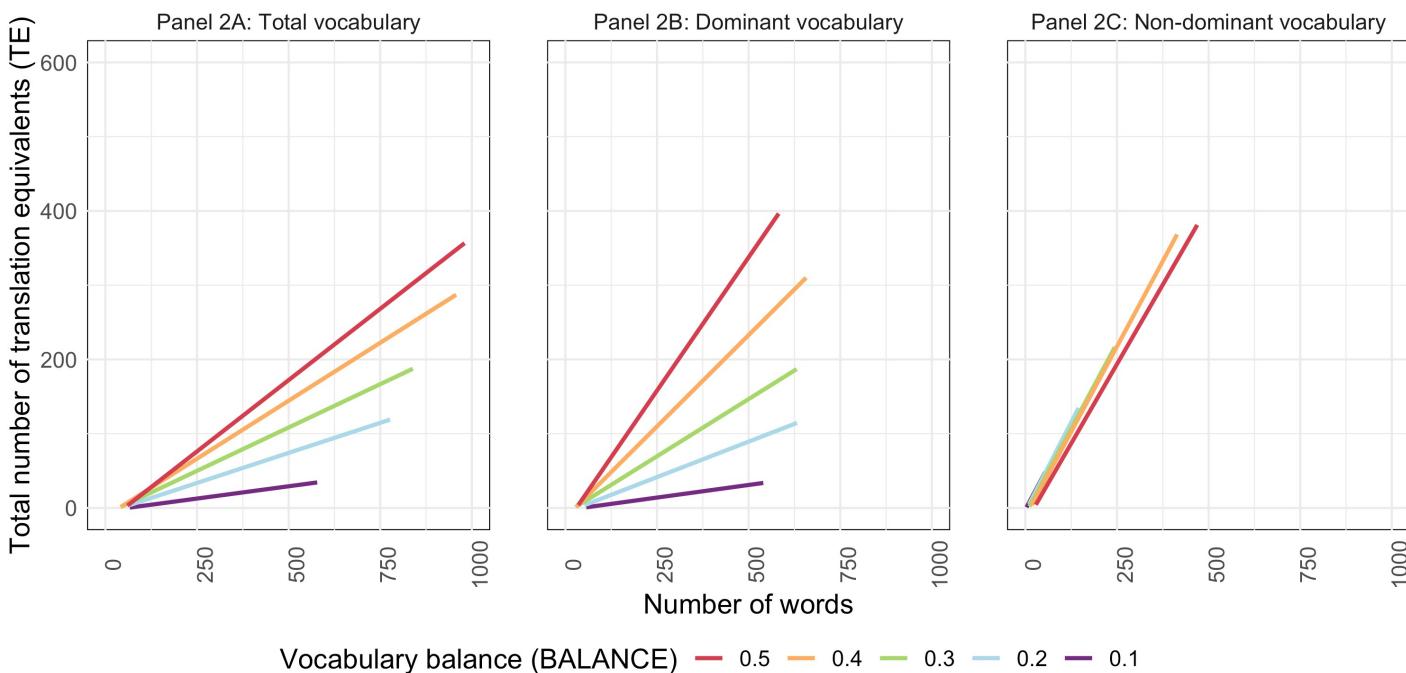


Vocabulary balance (BALANCE) — 0.5 — 0.4 — 0.3 — 0.2 — 0.1

## Simulated data



## Observed data



**2**

# Testing the bias parameter

**Observed** no. of  
translation equivalents    vs.

**Expected** no. of translation equivalents

$$\frac{\text{No. of dominant vocabulary} \times \text{No. of non-dominant vocabulary}}{\text{No. of learnable vocabulary}} \times \text{Bias parameter}$$

## 2 Testing the bias parameter

$$\text{Observed no. of translation equivalents} \sim \frac{\text{No. of dominant vocabulary} \times \text{No. of non-dominant vocabulary}}{\text{No. of learnable vocabulary}} \times \text{Bias parameter}$$

## 2

## Testing the bias parameter

$$\text{Observed no. of translation equivalents} \sim 0 + \frac{\text{No. of dominant vocabulary} \times \text{No. of non-dominant vocabulary}}{\text{No. of learnable vocabulary}} \times \text{Bias parameter}$$


No translation equivalents will be produced if a child doesn't produce any vocabulary

## 2 Testing the bias parameter

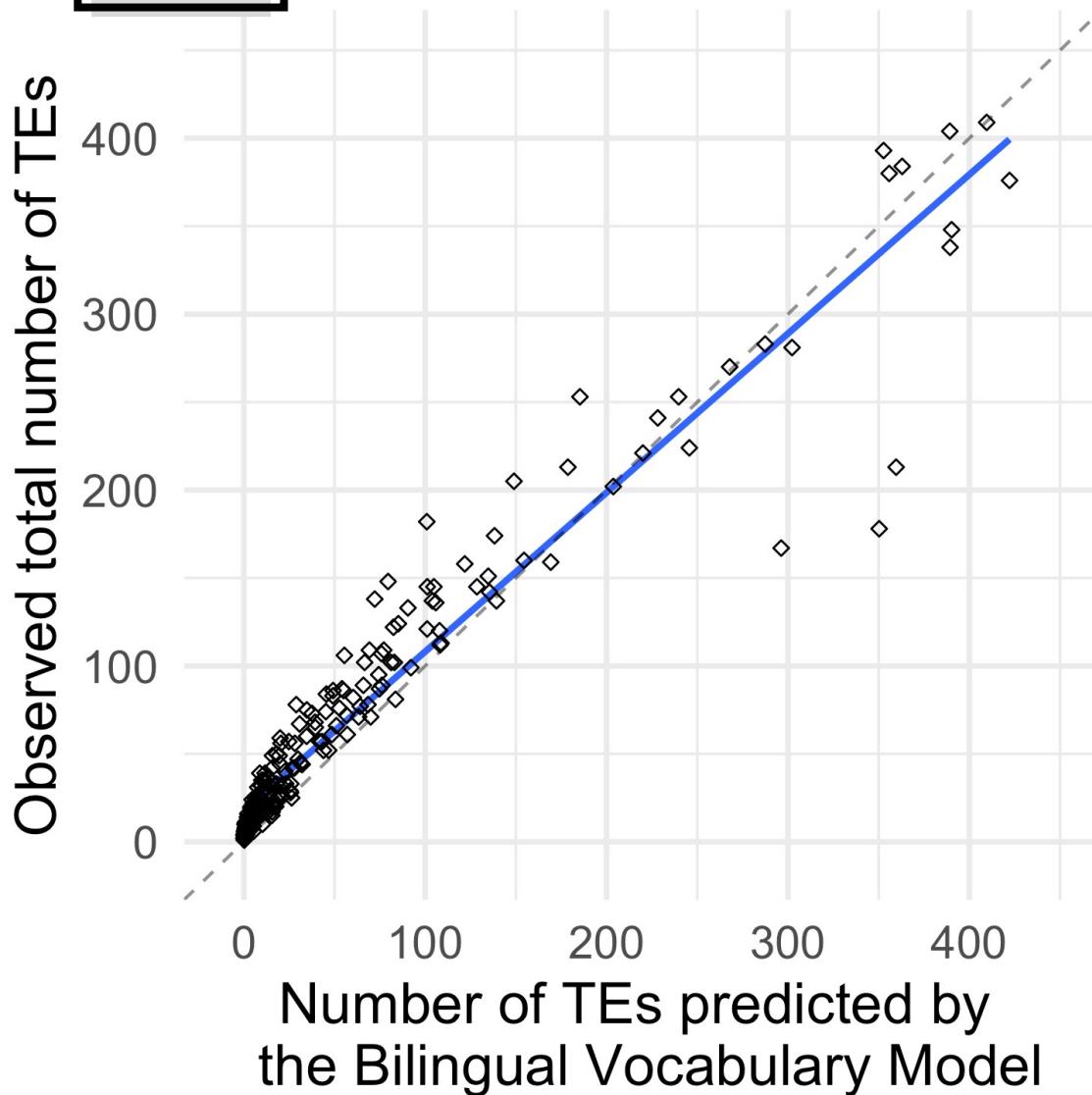
$$\text{Observed no. of translation equivalents} \sim 0 + \frac{\text{No. of dominant vocabulary} \times \text{No. of non-dominant vocabulary}}{\text{No. of learnable vocabulary}} \times \text{Bias parameter}$$

No translation equivalents will be produced if a child doesn't produce any vocabulary

The regression coefficient estimated by the model

## 2

## Testing the bias parameter



BIAS coefficient

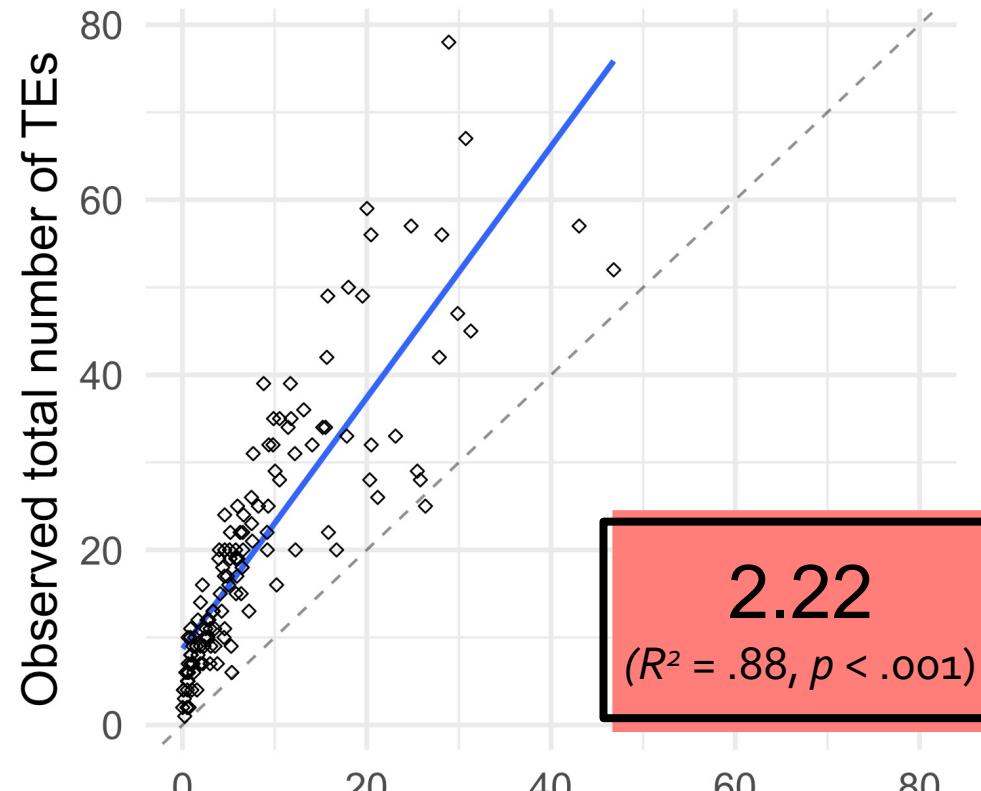
1.02

( $R^2 = .96, p < .001$ )

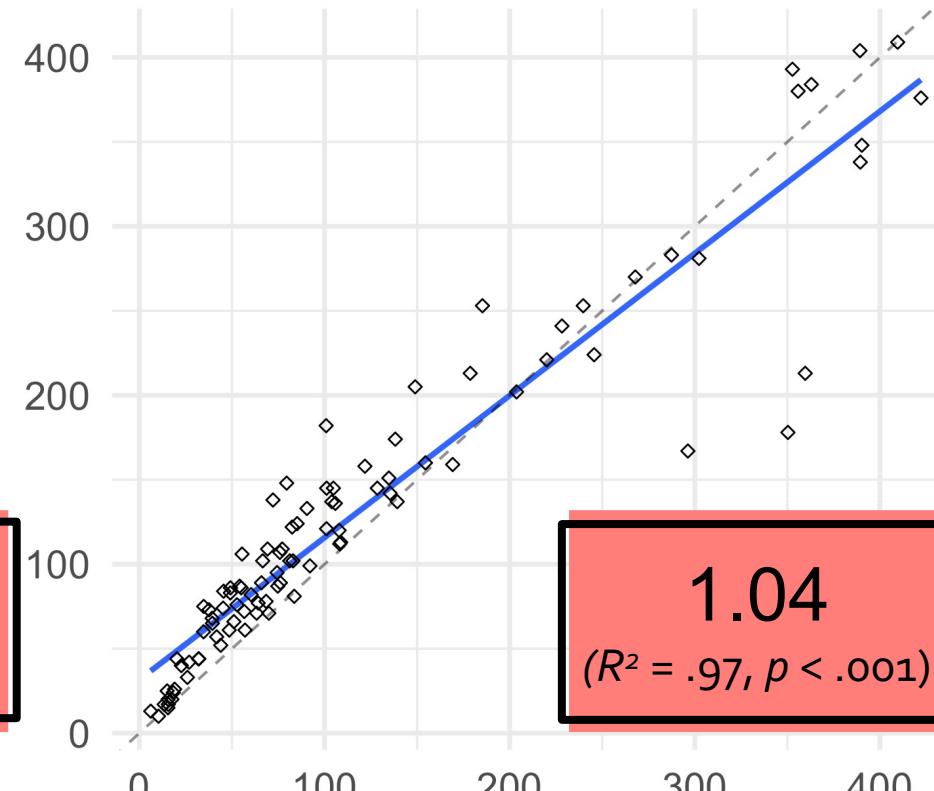
**Neutral Account**

Translation equivalents are neither harder nor easier to learn than singlets

### Less than 300 total vocabulary



### More than 300 total vocabulary

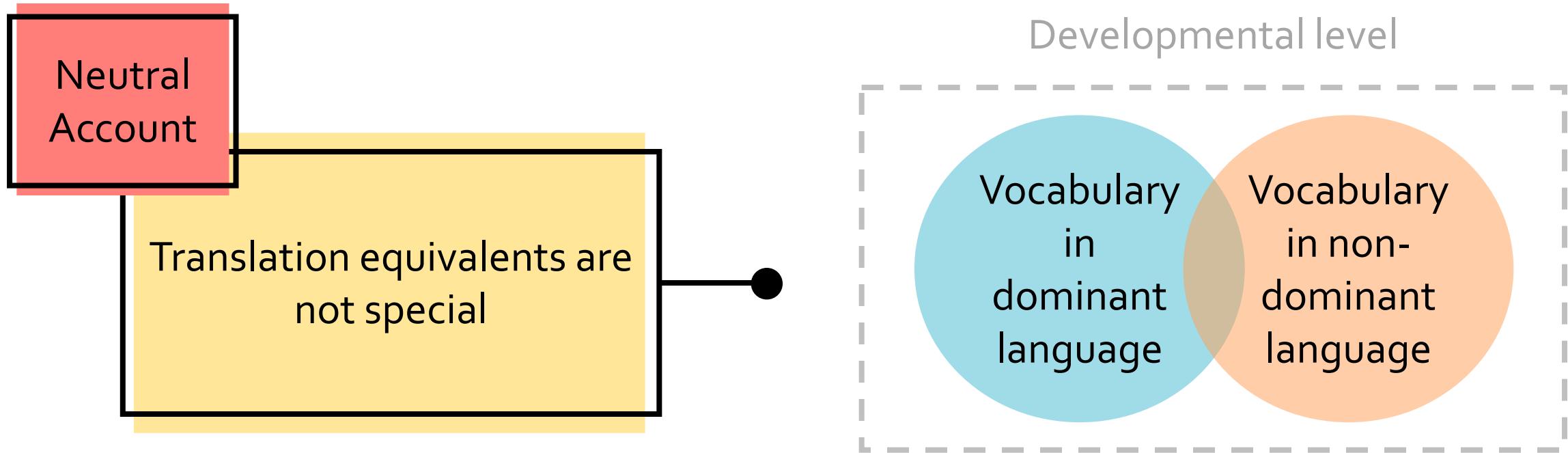


Developmental  
change

Preference Account

Neutral Account

# What is the nature of translation equivalent learning?



- Vocabulary in each language develops independently (Marchman, Fernald, & Hurtado, 2010)
- Translation equivalents are the by-chance overlap between the two languages (Pearson et al., 1995)

# Contributions of the Bilingual Vocabulary Model

An integrated approach

Including some quantitative factors that can predict vocabulary acquisition

Many other factors:

- A child's efficiency of processing words they hear (e.g., Hurtado et al., 2013; Weisleder & Fernald, 2013)
- Qualitative factors:  
quality of input (e.g., Raneri et al., 2020, Rowe, 2012),  
SES (e.g., Hoff, 2003; Fernald, Marchman, & Weisleder, 2013)

# Contributions of the Bilingual Vocabulary Model

An integrated approach

Including some quantitative factors that can predict vocabulary acquisition

A simplified computation

Equal opportunities for words to be learned in each of their languages

- A high degree of commonality in the first words children produced (e.g., Braginsky et al., 2016; Tardif et al., 2008)
- Possible that bilinguals learn different words depending on linguistic contexts (Grosjean, 2016)



Translation equivalent learning does not hold a special status and emerges predictably from the word learning process.

# THANK YOU!



Preprint here



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Esther Schott



estSchott



Krista Byers-Heinlein



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