

# Learning Pronoun Case from Distributional Cues

Flexible Frames for Case Acquisition

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# Introduction

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# Children's Pronoun Case Error

## English Pronoun Case

English cases are expressed on pronouns, as nominative case (e.g. *I, he*), accusative case (e.g. *me, him*) and genitive case (e.g. *my, his*)

## English speaking children make pronoun case errors

Pronoun case errors usually occur in children at the age of 2 to 4.

Example errors:

- (1) **Me** bite. (Abe, 1;9 in Kuczaj (1978))
- (2) All of **they** going go in here. (Nina, 2;11 in Suppes (1974))
- (3) What **my** doing? (Eve, 2;1 in Brown (1973))

# Why Children Make Pronoun Case Errors

## Syntactic Explanation

- ◇ Non-finite verbs lead to non-nominative errors (e.g. Schütze and Wexler, 1996).

## Usage-based Paradigm building model

- ◇ Case, person, gender and number forms a paradigm for each pronoun. The more varieties the children attempt to produce, the more errors they make (e.g Rispoli, 1994, 2005)

## Input-based theory

- ◇ Children would say things like ‘*her* go home’ because they mistakenly repeated ‘Let *her* go home’ in parents’ input (e.g. Tomasello, 2000; Kirjavainen et al., 2009).

# Children Rarely Make Pronoun Case Errors

## Corpus Analysis

Corpora: 46 longitudinal children's data, 211 children with cross-sectional data in all the available English-speaking children's data in CHILDES (MacWhinney, 2014).

Cross-sectional data: 141 children didn't make any errors; average pronoun case error rate is 1.16%; 95% children's pronoun case error rate is lower than 5%.

Longitudinal data: average pronoun case error rate 1.56%, with median of 0.6%.

Error rate is not correlated with age or MLU.



# Learning Pronoun Case

## Questions

How did children learn pronoun case?

Is parents' input informative enough for them to distinguish different pronoun cases?

Are they able to learn in the face of ambiguity?

## Hypothesis

Pronoun case can be distinguished by different distributional patterns. For example, 'help  $X$  cook',  $X$  is an accusative pronoun; 'can  $Y$  cook',  $Y$  is a nominative pronoun.



## Related Work on Distributional Patterns

### Distributional cues are effective in grammatical categorization

Frequent trigram frames ( $aXb$  where  $X$  is the target word) are effective in grammatical categorization. For example, 'to  $X$  to', ' $X$  is like to be a verb.

Two bigram frames ( $aX + Xb$ ) with a feedforward neural network can categorize more words with better categorization accuracy (Mintz, 2003).

**Trigram ( $aXb$ ) and flexible bigrams ( $aX + Xb$ ) could be used to categorize pronoun cases**

# Methods

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# Corpora

## Corpora

Following Mintz (2003) and Clair et al. (2010), we used the same six corpora of child-directed speech from CHILDES:

Anne and Aran (Theakston et al., 2001), Eve (Brown, 1973), Naomi (Sachs, 1983), Nina (Suppes, 1974), Peter (Bloom et al., 1974).

Included files where the child is younger than 2;6 years old

# Data

## Data

Each pronoun was extracted in  $aXb$  context, where  $X$  is the pronoun, e.g. 'help me cook'.

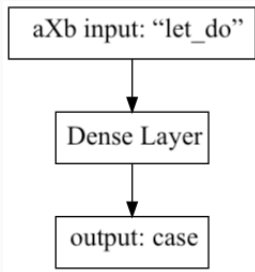
**Table 2:** Token counts of three pronoun cases and type counts of three context frames

	<b>Nominative</b>	<b>Accusative</b>	<b>Genitive</b>	<b>Pronoun Tokens</b>	<b>aX types</b>	<b>Xb types</b>	<b>aXb types</b>
Aran	4518	1014	1454	6986	445	927	2489
Anne	4343	1080	1392	6815	428	707	2308
Eve	1292	479	1029	2800	278	500	1364
Naomi	599	249	503	1352	224	364	806
Nina	3490	1195	1571	6256	400	747	2376
Peter	339	135	207	681	187	250	475
<b>Total</b>	<b>14581</b>	<b>4152</b>	<b>6156</b>	<b>24889</b>	<b>898</b>	<b>1672</b>	<b>7355</b>

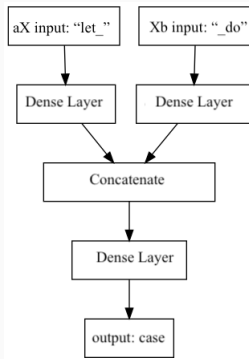
# Model

## Model Architecture

feedforward connectionist models with  $aXb$  and  $aX + Xb$  as different inputs



**Figure 1:** The architecture of  $aXb$  model



**Figure 2:** The architecture of  $aX + Xb$  model

# Model

## Evaluation

Classification accuracy

Asymmetric lambda value (following Clair et al. (2010)), which evaluates the association among the classes. Lambda is in the range of  $[0, 1]$ , 0 as no association, 1 as perfect association.

## Training and Testing

Each model was trained on all tokens (24889 tokens) and all types (7355 tokens of unique types).

Each model was trained using the same 10-fold cross-validation split.

# Experiments

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# Experiment 1: Models $aXb$ vs $aX + Xb$ in Categorizing Grammatical Cases

## Method

Input:  $aXb$  or  $aX$  ,  $Xb$ , e.g. 'let  $X$  go' and 'let  $X$ ,  $X$  go'

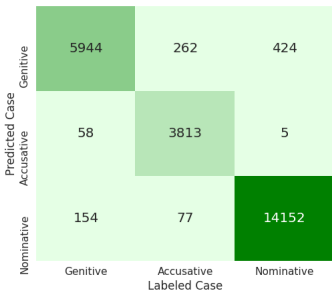
Output: The case of  $X$ : nominative, genitive or accusative

## Results

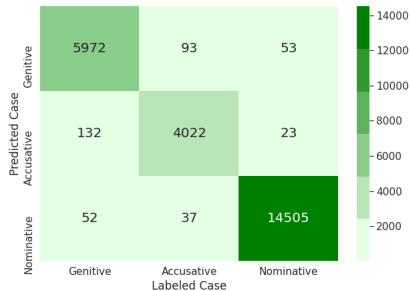
	Training on 24889 total tokens				Training on 7355 tokens of unique types			
	$aX + Xb$		$aXb$		$aX + Xb$		$aXb$	
	Accuracy	$\lambda$	Accuracy	$\lambda$	Accuracy	$\lambda$	Accuracy	$\lambda$
Aran	0.984	0.956	0.962	0.894	0.968	0.94	0.849	0.631
Anne	0.984	0.957	0.962	0.897	0.963	0.936	0.841	0.639
Eve	0.979	0.961	0.96	0.928	0.968	0.931	0.872	0.648
Naomi	0.983	0.969	0.951	0.914	0.953	0.902	0.878	0.708
Nina	0.987	0.97	0.951	0.911	0.974	0.952	0.834	0.6
Peter	0.982	0.965	0.954	0.913	0.963	0.927	0.827	0.619
<b>Total</b>	<b>0.984</b>	<b>0.962</b>	<b>0.960</b>	<b>0.907</b>	<b>0.967</b>	<b>0.939</b>	<b>0.847</b>	<b>0.631</b>



# Experiment 1: Heatmap results



**Figure 3:**  $aXb$  model on 24889 tokens



**Figure 4:**  $aX + Xb$  model on 24889 tokens

# Experiment:1 Results on aXb model

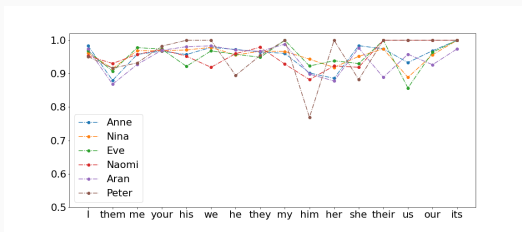


Figure 5: Training results of aXb model with 24889 tokens

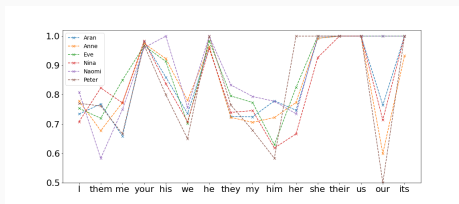


Figure 6: Training results of aXb model with 7355 tokens of unique types

# Experiment 1: Results on $aX + Xb$ model

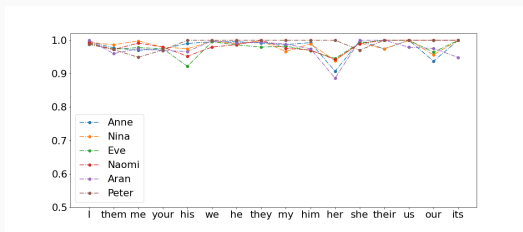


Figure 7: Training results of  $aX + Xb$  model with 24889 tokens



Figure 8: Training results of  $aX + Xb$  model with 7355 tokens of unique types

## Experiment 2: Predicting the Pronoun Using $aX + Xb$ Model with Person, Gender, Number Information

### Method

Person, gender, number were used together to train the  $aX + Xb$  model to predict the pronoun.

For example, 'help  $x$  cook' plus *3rd person, masculine, singular* would be able to predict  $x$  as 'him'.

### Results

	24889 tokens		7355 types	
	Accuracy	$\lambda$	Accuracy	$\lambda$
Aran	0.994	0.992	0.980	0.971
Anne	0.994	0.992	0.980	0.976
Eve	0.993	0.990	0.983	0.972
Naomi	0.993	0.995	0.980	0.967
Nina	0.996	0.994	0.987	0.982
Peter	1.000	1.000	0.983	0.975
<b>Total</b>	<b>0.994</b>	<b>0.993</b>	<b>0.982</b>	<b>0.975</b>

## Experiment 2. Heatmap Result

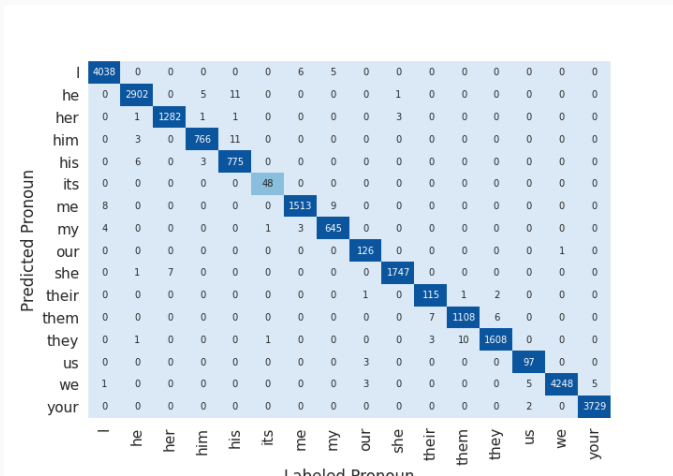


Figure 9: Heatmap of  $aX + Xb$  model in predicting the pronoun

## Experiment 2. Results on predicting pronouns

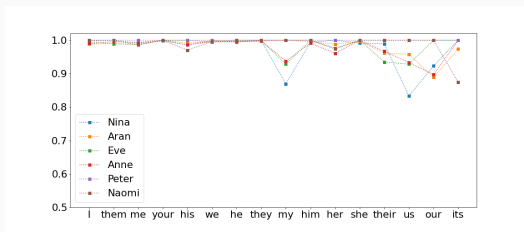


Figure 10: Training results of pronoun on  $aX + Xb$  model with 24889 tokens

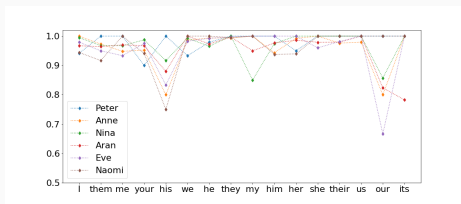


Figure 11: Training results of pronoun on  $aX + Xb$  model with 7355 tokens of unique types

# Experiment 3: Corpus Analysis of Children's Pronoun Case Errors

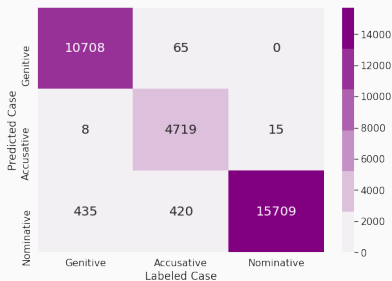
## Methods

Each child's pronoun case errors were searched in all available files.

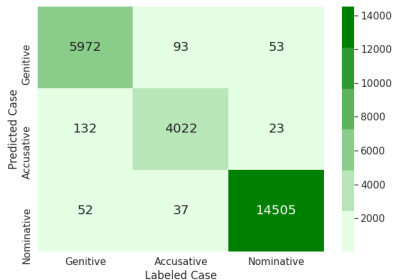
	<b>Errors</b>	<b>Total Pronouns</b>	<b>Accuracy</b>
Anne	57	5009	0.989
Aran	25	8450	0.997
Peter	115	4077	0.971
Eve	49	2685	0.982
Naomi	64	3249	0.980
Nina	633	8609	0.926
<b>Total</b>	<b>943</b>	<b>32079</b>	<b>0.970</b>

**Table 3:** Results of each child's pronoun case errors and accuracy

# Comparison between errors made by children and the model



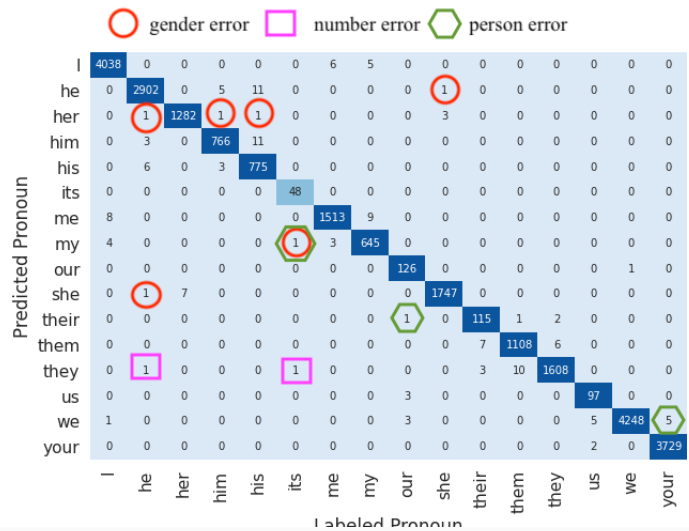
**Figure 12:** Children's pronoun case error heatmap



**Figure 13:**  $aX + Xb$  heatmap on 24889 tokens

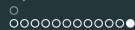


# Errors in predicting pronoun



# Conclusion

- ◇ Distributional patterns in parents' input are very useful in categorizing grammatical cases.
- ◇  $aX + Xb$  model showed similar accuracy rate as children in real life.
- ◇ Children and models made similar errors.
- ◇ However, these results are not evidence that children actually use distributional cues to acquire pronoun case.



## Questions and Comments

## References

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