

## Predictors of Word Combinations by Toddlers who Participated in Parent-Coached Language Interventions: Modeling Count Outcomes

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### Purpose

Use an extant dataset (Ronski et al., 2010; Ronski in prep) to address two purposes.

1. explore potential predictors of the total number of vocabulary combinations at the end of a parent-coached language intervention.
2. To raise awareness of potential methodological issues behind collecting frequency data (count data outcomes) and to compare popular models that address these issues.

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## Word Combinations of Typically Developing Children

- ❖ TD children begin to combine words around 2 years of age.
- ❖ Vocabulary is important!
  - Begin to combine with an expressive vocabulary of 50 words
  - Conceptual development plays a role in early grammar (Brown, 1973)
  - Familiarity with verbs helps young children produce abstract grammatical structures (Tomasello, 2000)



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## Children with Intellectual Disabilities Who Use AAC

- ❖ Youth with ID use symbol vocabularies to build complex structures to communicate (Wilkinson, Ronski & Sevcik, 1994)
- ❖ Many studies have successfully used trial-based (Binger & Light, 2007; Nigam et al., 2006) and partner training interventions (Binger et al., 2008;2010) to teach vocabulary combinations.
- ❖ We know little about what contributes to the emerging combinations of children who participate in language interventions when they are being not formally taught.




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### Research Questions


- ❖ RQ1: Do toddler characteristics (baseline language measures, access to an AAC device) and Language outcomes (MLU, target vocabulary size, TTR), predict emerging number of vocabulary combinations when controlling for age?
- ❖ RQ2: Does target vocabulary size mediate the relationship between toddler characteristics (baseline language measures, access to an AAC device) and total number of target vocabulary combinations?

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### Participants

- ❖ Extant data were used from 113 parent-toddler dyads who participated in two parent-coached language intervention studies (Ronski et al., 2010; Ronski, in prep).
- ❖ Toddlers ranged in age between 24-38 months.
- ❖ Toddlers had an expressive vocabulary of less than 10 words and scored less than 12 months on the expressive language scale of the Mullen.

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### Language Intervention

- ❖ All parent-child dyads participated in 24 30-minute parent-child language interventions over 12 weeks.
- ❖ All children were assigned an individualized set of vocabulary words they did not know.
- ❖ All parents were encouraged to elicit the vocabulary words using different strategies (e.g., modeling, expansion, sabotage).
- ❖ Children either did or did not have a speech-generating device (SGD) throughout the intervention.

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### Measures and Variables

**Baseline Receptive and Expressive Language Scales:**  
Mullen Scales of Early Learning, Vineland Adaptive Behavior Scales (VABS), The MacArthur-Bates Communicative Development Inventory (MCDI), and Sequenced Inventory of Communication Development (SICD).

**Language Outcomes (final intervention session):**

- ❖ Mean Length of Utterance (MLU)
- ❖ Type Token Ratio (TTR)
- ❖ Target vocabulary size (e.g., ball, myturn, read, turnpage)

**Use of AAC:**

- ❖ SGD was a light-weight low-tech device using Picture Communication Symbols (PCS; Mayer-Johnson, 1981).
- ❖ Children who had access to a speech-generating device (SGD) vs. those who did not.

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### Measures and Variables

**Word Combinations:**

- ❖ Combination included more than one target vocabulary word
- ❖ Included both spoken and aided symbol vocabulary words

Table 1  
Examples of combination types made by toddlers at session 24

| Agent-action | Action-object | Entity-locative | Possessor-possession | Entity-attributive |
|--------------|---------------|-----------------|----------------------|--------------------|
| bear ride    | read book     | bear out        | my cracker           | blue bear          |
| go cow       | drink juice   | train in        | your cracker         | drink full         |
| go train     | stop train    |                 |                      | 5 cookie           |
|              |               |                 |                      | good apple         |

*Note: Examples are categorized according to Brown's (1973) definition of early grammatical categories (semantic relations)*

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### Measurement Issues with Count Data

- Measuring linguistic behaviors (frequency data) in children with language and ID may result in counts-with a preponderance of zeros and over-dispersion.
- This creates complex issues when trying to analyze the data.

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### Types of Count Data Distributions

May want to consider modeling using count models rather than regular OLS regression. Why?

- Poisson regression: conditional mean is modeled as a function of the covariates. Assumes that the unconditional mean and unconditional variances are equal.
- Negative binomial: Same mean structure as Poisson. Makes a provision for overdispersion in the data (variance is greater than the mean).
- Zero inflated Poisson (zip): uses a logit model to model probability of a constant (Structural) zero and count data is modeled by the Poisson regression.
- Zero inflated negative binomial (zinb): Model the probability of structural zero. Models count data and overdispersion in the distribution of the NB component.

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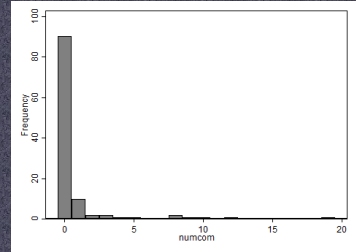
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### Outcome : Number of Vocabulary Combinations

Let's take a look at descriptives for the outcome of the present study:

- Only 22 out of 113 combined target vocabulary words.



- The unconditional Mean=.85 and the Variance=7.3

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### Data Analysis Plan

- Model fit was compared across four count-data models (e.g. poisson, negative binomial, zero inflated negative binomial, zero-inflate poisson) by comparing Loglikelihood Ratios for nested models and comparing BIC values and Vuong tests (1989) for non-nested models.
- For RQ1 and RQ2, We utilized expressive and receptive language factor scores in our analyses as independent variables. Factor scores were calculated in Mplus (Muthén & Muthén, 1998–2012) by fitting a two-factor model, ( $p = .017$ , CFI = .976, RMSEA = .084). Factors were correlated at .541,  $p < .001$
- For RQ1, we conducted a multiple regression analysis. For RQ2 we conducted a mediation model. Analyses were conducted using STATA (RQ1; Statacorp, 2013) and Mplus (Muthén & Muthén, 1998–2012)

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### Research Question 1:

Do toddler characteristics (access to AAC [yes or no], baseline language skills) and language outcomes (MLU, TTR, Functional vocabulary size) predict total number of vocabulary combinations at session 24 over and above age?

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### Research Question 1: count data coefficients

Table 2. Comparison of count data models.

| Parameter                          | Poisson | Negative Binomial | Zero-inflated Poisson | Poisson Logit | Zero-inflated NB | Negative binomial Logit |
|------------------------------------|---------|-------------------|-----------------------|---------------|------------------|-------------------------|
| Intercept                          | 1.54**  | -0.24             | -1.70                 | 2.19**        | 0.80             | 0.70                    |
| Age                                | 0.04    | -.002             | 0.01                  |               | -0.01            |                         |
| Expressive Language (baseline)     | 0.05*   | 0.09              | -0.01                 |               | 0.06             |                         |
| Receptive Language (baseline)      | 0.03    | 0.03              | 0.04                  |               | 0.05             |                         |
| AAC use                            | 1.29    | 1.02              | 1.13                  |               | 1.01             |                         |
| MLU (session 24)                   | -0.68   | -3.14             | 0.14                  |               | -2.82            |                         |
| TTR (session 24)                   | 1.02    | -1.83             | -1.97                 |               | -1.87            |                         |
| Total Vocabulary Size (session 24) | 0.12*** | 0.14**            | 0.08***               | -0.08*        | 0.11*            | -0.06                   |
| Dispersion Parameter               | 8       | 3.81***           | 10                    |               | 1.82***          |                         |

Note: Values are unstandardized coefficient estimates for Poisson, Negative Binomial, Zero-inflated Poisson and Zero-inflated binomial. MLU=Mean length of utterance. TTR=type-token ratio. Dispersion=dispersion parameter,  $\alpha$ . \*Statistically significant ( $p < .05$ ), \*\*statistically significant ( $p < .01$ ) \*\*\*statistically significant ( $p < .000$ )

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### Research Question 1: tests of model fit

Table 3. Comparison of count data model fit indices

|         | Fit Statistics | Negative Binomial (NB) | Zero-inflated Poisson (ZIP) | Zero-inflated Negative Binomial (ZINB) |
|---------|----------------|------------------------|-----------------------------|--|
| Poisson | LR             | 91.67***               |                             |  |
|         | BIC            |                        | 65.62                       | 78.03                                  |
|         | Vuong          |                        | 2.26*                       |  |
| NB      | LR             |                        |                             |  |
|         | BIC            |                        | -21.34                      | -8.95                                  |
|         | Vuong          |                        |                             | .29                                    |
| ZIP     | LR             |                        |                             | 17.13***                               |
|         | BIC            |                        |                             |  |
|         | Vuong          |                        |                             |  |

Note: LR=Likelihood ratio test (nested models). BIC=Difference in BIC values (non-nested models). Vuong=significance test between non-nested models \*Statistically significant ( $p < .05$ ), \*\*statistically significant ( $p < .01$ ) \*\*\*statistically significant ( $p < .000$ )



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### Research Question 1: visualizing model fit

Figure 2. Comparison of model fit between count data models

Note: LR=Likelihood ratio test (nested models). BIC=Difference in BIC values (non-nested models). Vuong=significance test between non-nested models \*Statistically significant ( $p < .05$ ), \*\*statistically significant ( $p < .01$ ) \*\*\*statistically significant ( $p < .000$ )

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### Research Question 1: NB results

- ❖ The negative binomial model provided the best model fit.
- ❖ Functional vocabulary words was a significant predictor the number of vocabulary combinations produced at session 24,  $p < .01$
- ❖ For every one word increase in functional vocabulary, the number of total vocabulary combinations increased by 14.6% ( $1 - \exp(0.1363) = 0.146$ ).
- ❖ Pseudo  $R^2 = .13$

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### Research Question 2

Does target vocabulary size mediate the relationship between toddler characteristics (baseline language measures, access to AAC) and outcome of target vocabulary combinations?

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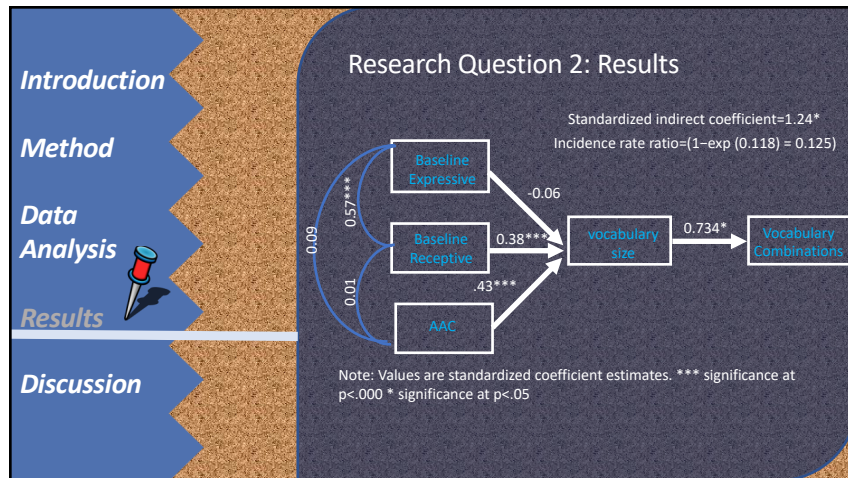
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### Research Question 2: Analyses

- ❖ Tested mediation using MPLUS.
- ❖ Applied the same steps to compare model fit as in RQ1.
- ❖ As in RQ 1, Negative binomial regression was the best fit to the data.



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### Discussion and Implications

- Vocabulary combinations can emerge in toddlers without teaching targeted vocabulary combinations.
- Receptive language skills and use of a SGD do not directly explain emerging vocabulary combinations. However SGDs and receptive language skills promote acquisition of target vocabulary (Ronski et al, 2010; Barker et al., in prep).
- Teaching parents to use language eliciting techniques (with a focus on targeting vocabulary growth) in a natural communication context is helpful to eliciting emerging vocabulary combinations.

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### Discussion and Implications

- Using count data models gives us a unique opportunity to understand more about factors that contribute to emerging grammar in children with ID and language delays.
  - We can look at skewed distributions
  - Testing for model fit is important!

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### Discussion and Implications

#### Limitations

- Vocabulary repertoires are idiosyncratic!
- Other toddler characteristics?
- Comparison group?

#### Future Directions

- Combine intervention and follow-up data to look at individual growth curve patterns (has been done with count data!).
- Qualitative look at grammatical structures over time (follow up data).

