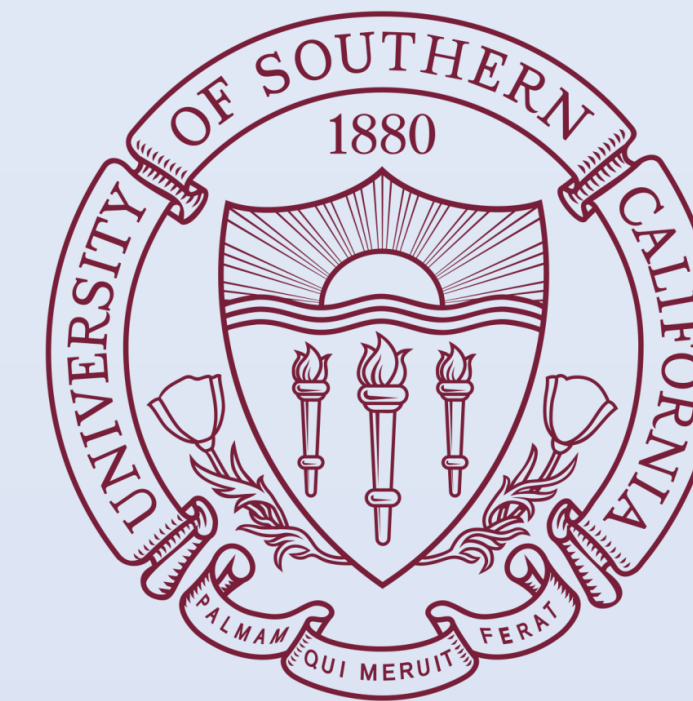


# Non-adjacent lexical dependencies in an artificial language prime relative clause attachment biases

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## 1. Introduction

- Where does the bias in parsing ambiguous relative clauses come from? Is it influenced by the previous statistics people have collected?
- Prior work: Representation of abstract dependencies in language & other domains (e.g. Menon & Kaiser, 2013; Mitchell et al., 1995; Scheepers et al., 2011; van de Cavey & Hartsuiker, 2011).
- What about abstract relations represented through word-level statistical regularities in an artificial language?
- Research question:** Can adjacent and non-adjacent structures derived from word-level statistics prime the low vs. high attachment preference during the production of relative clauses (RCs)?
- We manipulated the dependencies participants encountered in the priming material
  - Non-adjacent dependency sequences ( $A_iXC_i$ ) from Gomez, 2002. A **non-adjacent dependency (NAD)** is a three-word sequence such that the first word uniquely predicts the third word, while the second word can vary.
  - This structure models the linear sequence of high attachment in RCs (e.g., *Kevin counted the fans of the singer who were excited*).
- Hypothesis:** if abstract relations extracted from lexical statistics trigger syntactic priming, learning the non-adjacent dependency should prime participants to produce more high-attachment relative clause completions than control participants.
- English RC's have a default low-attachment bias (e.g., Brysbaert & Mitchell, 1996). Can this be weakened by non-adjacent primes?
- Prediction:** If non-adjacent lexical-level representations prime dependency formation in RCs, participants should produce more high-attachments after non-adjacent sequences (ACX, and XAC).

## 3. Results

- Coding:** RC completions analyzed as **high-attachment (HA)**, **low-attachment (LA)**, or **ambiguous** (coded as missing in Logistic Regression).
- Successful learning of non-adjacent dependencies:** Above-chance performance in test-phase learning questions ( $p < 0.001$ , Mixed-effects Logistic Regression).
- RC completions influenced by NAD primes:**
  - More high-attachment completions in nonadjacent prime group than other 3 groups ( $p < 0.001$ , Mixed effects Logistic Regression) => Fig.1**
  - People who learned a non-adjacent dependency in the artificial language were more likely to produce RCs that attach to the non-adjacent (higher) noun**
    - compared to people in control conditions (who learned local dependencies or no dependencies)
- People trained with nonadjacent primes do not follow any item-level (Figure 2) or group-level (Figure 3) tendencies to complete a sentence with High Attachment bias with regard to explicit knowledge of Non-adjacent Dependency questions.

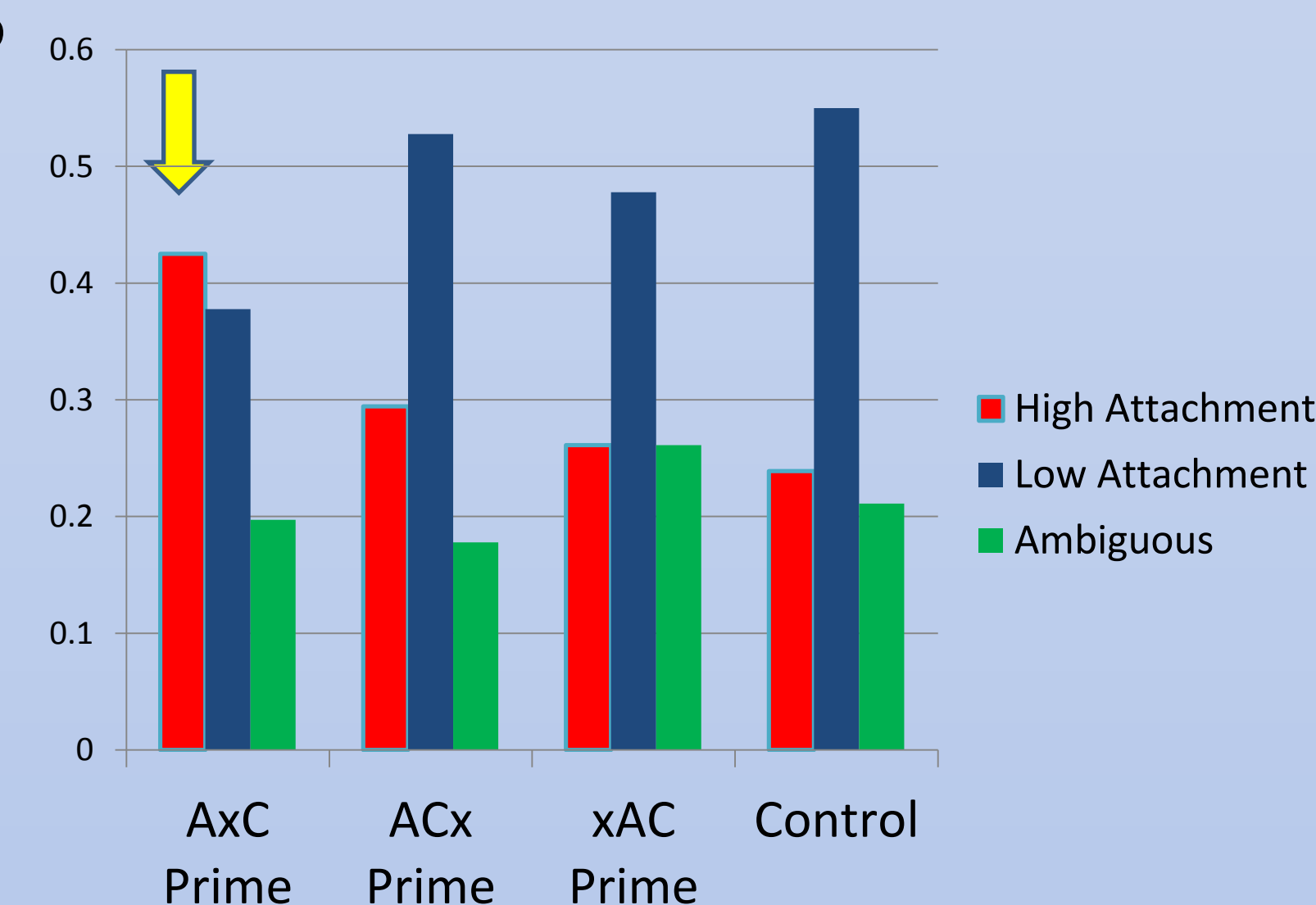


Figure 1. RC Completion Results from Priming and Controls

|     | HA  | LA |
|-----|-----|----|
| YES | 105 | 96 |
| NO  | 48  | 40 |

Figure 2. Item-level Relationship between Correctly saying YES in Priming Test and immediate RC Completion. Fisher's exact test:  $p = 0.798$

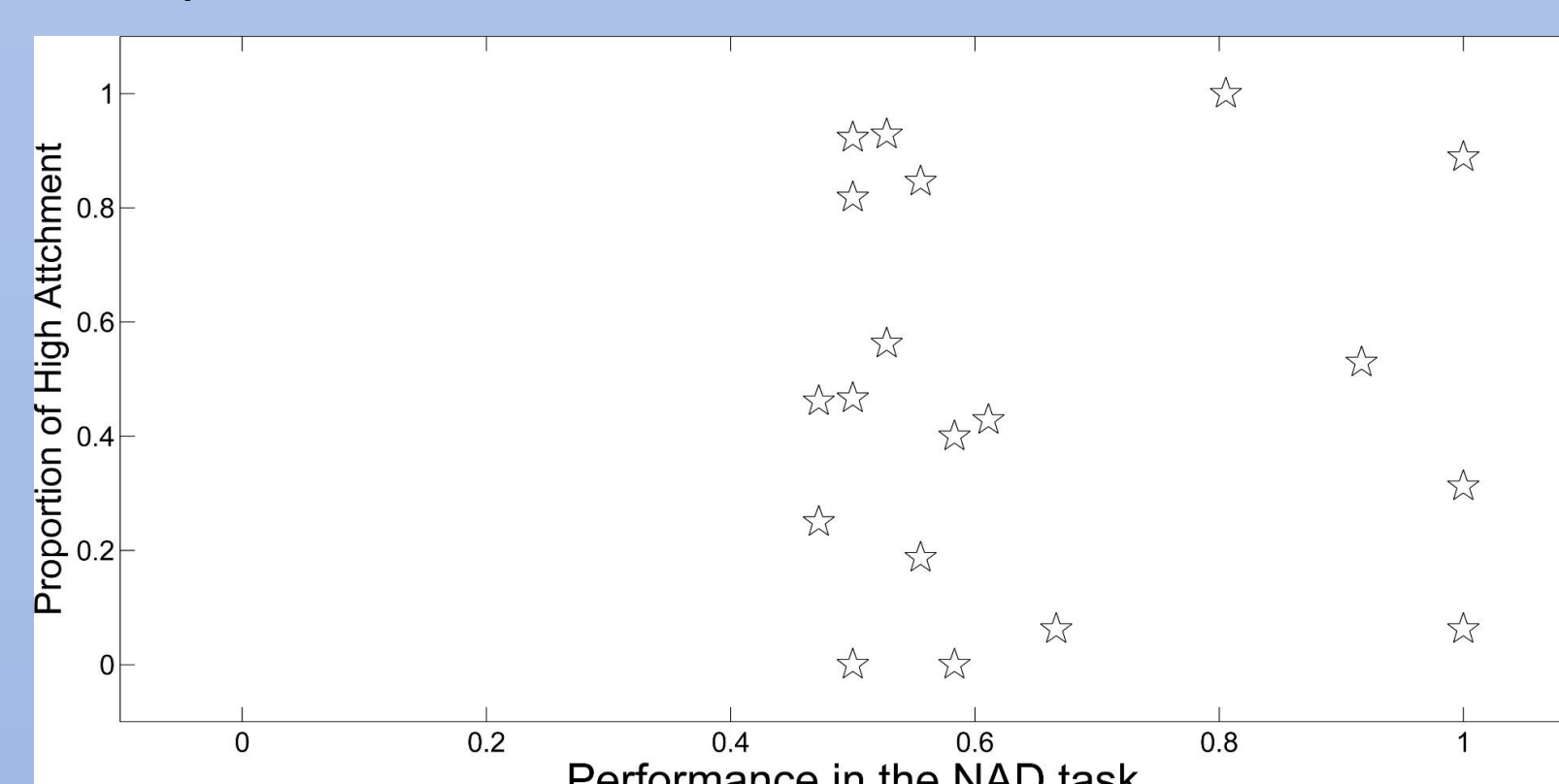


Figure 3. No group-level relationship between Correctness in Priming Test and RC Completion Attachment Biases. Correlation:  $-0.0538$  ( $p = 0.82$ )

## 2. Experiment Design

- STEP 1: Training phase (~20 min):** Participants ( $n=50$ ) heard three-‘word’ strings in an artificial language (adapting stimuli from Gomez’02, e.g. *choon glaik jub*). During listening, they answered a question on what word was just played every few minutes.
- Between-subjects design, participants encounter different dependencies:
- Non-adjacent group** ( $n=20$ ) was trained on non-adjacent dependencies ( $A_iXC_i$ )
- Two **adjacent groups** ( $n=10$  each) trained on adjacent dependencies ( $A_iC_iX$ , and  $XA_iC_i$ )
- Baseline group** ( $n=10$ ) trained on random three-word orders with no dependencies.

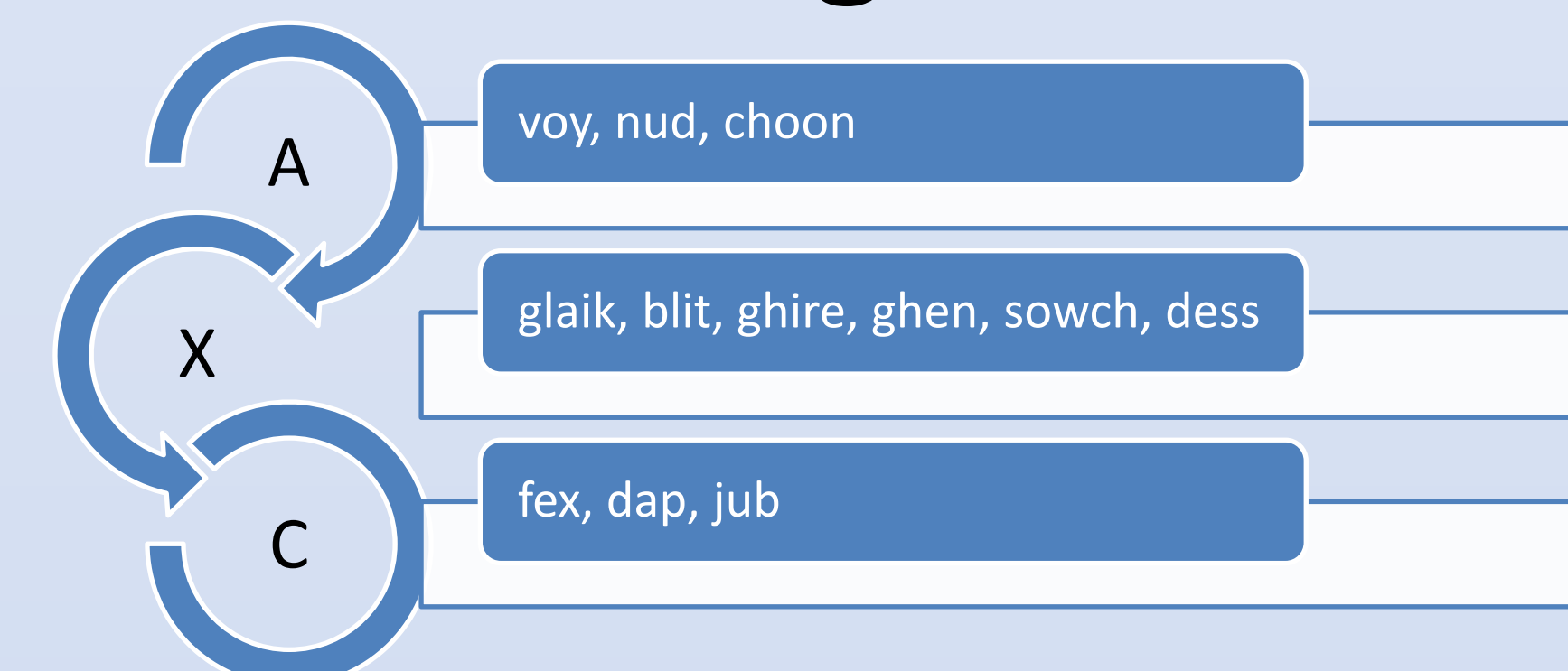


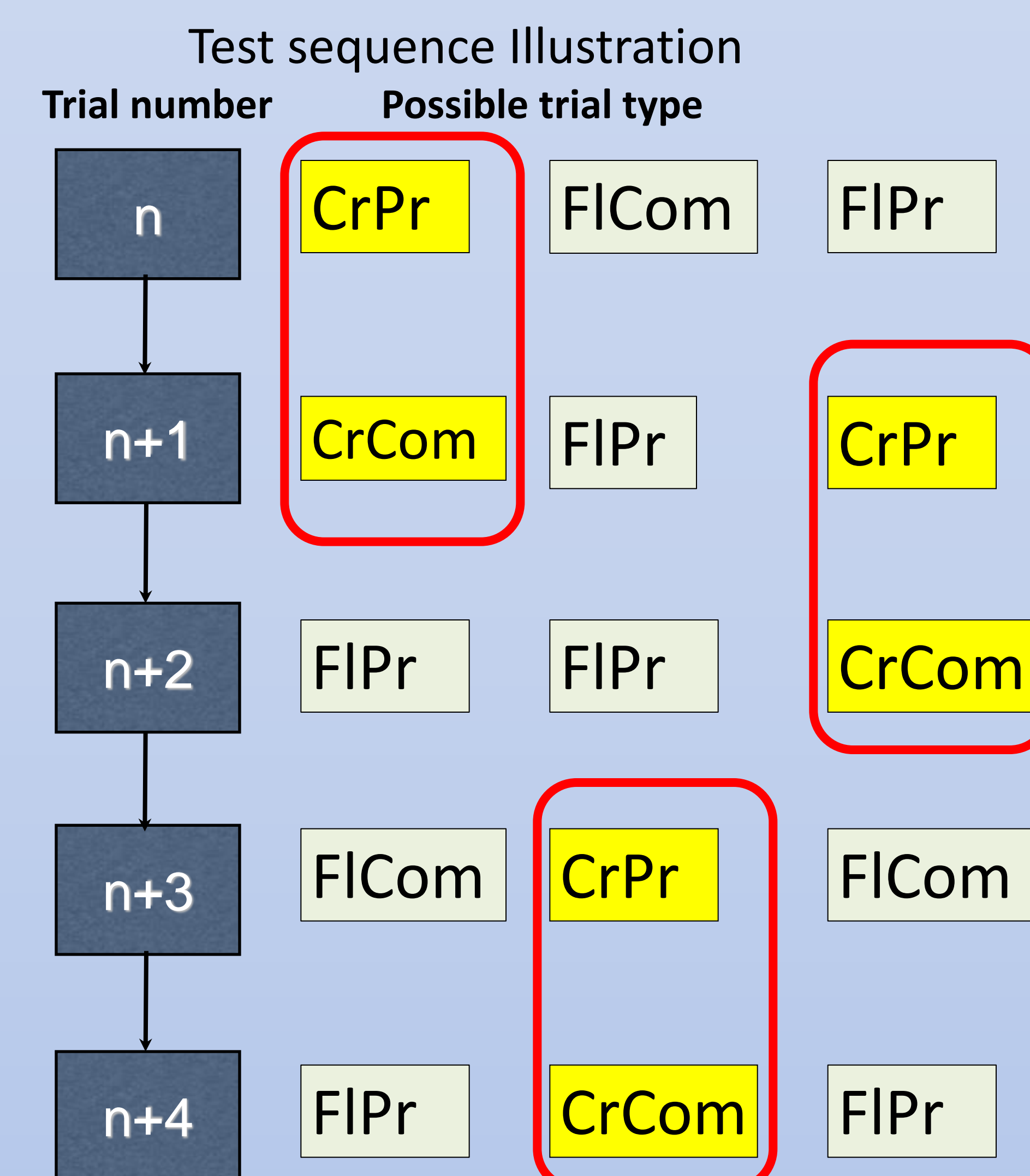
Figure 4. Illustration of the artificial language in Gomez, 2002.

|       |                            |   |
|-------|----------------------------|---|
| CrPr  | Critical Prime             | (voy glaik fex) Is it in the language you heard?<br>• Yes ✓<br>• No                       |
| CrCom | Critical RC Completion     | Harry met the doctors of the supermodel who _<br>• cured cancer (HA)<br>• was skinny (LA) |
| FlPr  | Filler “Prime”             | (voy glaik dap) Is it in the language you heard?<br>• Yes<br>• No ✓                       |
| FlCom | Filler fragment completion | The waitress revered the funny bartender with the thick moustache although _              |

- Step 2: Test phase (~20 min)**
- Two trial types:
  - hear three-‘word’ sequence [prime], say whether it’s in the language
  - write completions for RC fragments
- On critical trials
  - Artificial language prime occurred immediately before RC fragment.
  - Expected answer to artificial language prime: Yes
  - Then complete relative clause fragment
- Sentence fragments on critical trials were ambiguous RC fragments, people wrote continuations:

*Kevin counted the fans of the singer who*

- ...were really excited (HA)**
- ....was performing that night (LA)**



## 4. Discussion

- Abstract relations represented through newly-learned word-level statistical regularities can prime the attachment biases of relative clauses.
- The lack of ‘word’-level effects suggest that the representation for RC attachment bias is unlikely to be the same statistical representation at the lexical level that people are trained on.
- This suggests that the underlying representations of attachment biases are best regarded as highly abstract and finely attuned to statistical regularities in the input.
- Future plans: Test this with Spanish. Spanish RCs have a high attachment bias (e.g., Brysbaert & Mitchell, 1996). By priming Spanish speakers with adjacent dependencies, we can see whether their high-attachment bias would be primed to produce low-attachment relative clauses.

### References

- Brysbaert, M. & Mitchell, D.C. (1996). Modifier attachment in sentence parsing: Evidence from Dutch. *Quarterly Journal of Experimental Psychology*, 49A, 664-695.
- Gomez, R. L. (2002). Variability and detection of invariant structure. *Psychological Science*, 13(5), 431-436.
- Menon, M. & Kaiser, E. (2013). Consequences of ‘music to one’s ears’: Structural Integration Priming from Music to Language. Poster presented at 26th Annual CUNY Conference on Human Sentence Processing.
- Mitchell, D.C., Cuetos, F., Corley, M.M.B., & Brysbaert, M. (1995). Exposure-based models of human parsing: Evidence for the use of coarse-grained (non-lexical) statistical records. *Journal of Psycholinguistic Research*, 24, 469-488.
- Scheepers, C., Sturt, P., Martin, C.J. (2011). Structural priming across cognitive domains: from simple arithmetic to relative-clause attachment. *Psychological Science*, 22, 1319-1326.
- Van de Cavey, J. (2012). Are syntactic processes in language and music domain specific? Unpublished thesis. Universiteit Ghent.

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