

# Like Yoda speak I

## Using artificial language learning experiments to study language change

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# Collaborators



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# Word order change

## Middle French

(1) Et ces parolles m'a compté *le* *roy*  
and these words me.CL=has told the king

'And the king has told me these words.'

(Wolfe 2021: 7, *Commyn* 9)

## Modern French

(2) Et *le* *roi* m'a raconté ces paroles  
and the king me.CL=has told these words

# Sources

## Ways to study language change

- Historical texts & records
- Language change in progress (e.g. heritage languages)
- Language acquisition experiments
- Modelling
- ...

... but how can causality be established? **Artificial language learning!**

# Artificial language learning

## Artificial language learning (ALL)

- Creation of miniature linguistic system
- Participants are exposed to language, afterwards learning measured
- Successfully used with adults and children (Gomez & Gerken 2000, Folia et al. 2010, Culbertson & Schuler 2019)
- Advantages:
  - Experimenter has control over factors of interest
  - Control for prior learning

# Artificial language learning

## Successful application in various linguistic disciplines

- Typology & language universals (e.g. Culbertson et al. 2012, Tabullo et al. 2012)
- Sociolinguistics (Sneller & Roberts 2018)
- Phonological change (Yin & White 2018)

→ Suitable to study syntactic change

# Overview

I. Learning and loss of V2

II. Experiment 1

III. Experiment 2

# Learning & loss of V2

- Robust attestation of evidence for V2 in learners' input necessary (Lightfoot 1999, 2006, Yang 2000)
- Loss of V2 in French (Yang 2000):
  - OVS, XVSO → V2; SXVO, XSVO → SVO
  - Analysis of sentences with *pro*-drop ambiguous: [X *pro* V] or [X V *pro*]
  - Roberts (1993): 5-18% VS structures, 40-52.5% SV structures in MidFr
  - More V>2 sentences than VS structures → SVO grammar
- How does the evidence for V2 need to be distributed to facilitate the acquisition of V2 the most?



# Evidence for V2

## Ideal input for learners of V2 language

- Ambiguity of SVO structures → Non-subject-initial sentences required
- Maximal variability of preverbal element (i.e. high entropy of preverbal position) and V2 without exceptions...
- ... but maximal variability of what?
  - Phrase types: NP/DP, PP, AdvP, CP etc. (Lightfoot 1999, 2006, Sitaridou 2012)
  - Grammatical functions: S, O & A (Yang 2000, 2002)

# Variation and learning in the lab

## The effect of variability on learning

- Facilitating effect of variability domain-general (Raviv et al. 2022)
- Gómez (2002), Gómez & Maye (2005):
  - Learning of non-adjacent dependencies by infants and adults (aXc, bXd)
  - Finding: Better learning of dependency when variability in X is higher

## Variability and the acquisition of V2

- V2: X- $V_{fin}$
- X = 1/3 S, 1/3 O & 1/3 A should result in best learning outcome

# Hypothesis

## Hypothesis

- The learnability of a verb second (V2) grammar is conditioned on the entropy of the preverbal position
- A higher preverbal entropy entails better learning of a V2 grammar

## Learning V2

- Extrapolation of the flexibility regarding the preverbal constituent to novel structures

# Predictions



## Predictions

- Participants learning a skewed V2 language should extrapolate V2 to new structures in fewer instances than participants learning a non-skewed language
- Learners of a skewed V2 language should show diminished discrimination of novel V2 and ungrammatical V3 structures compared to participants learning an unskewed V2 language

# Experiment 1

# Exp.1: Participants

- 314 participant tested, 230 included in analysis (73.2%)
  - Uni.: 74/94
  - O-dom.: 78/118
  - A-dom.: 78/102
- Prolific
  - Self-reported US-nationals
  - Monolingual English speakers
  - Raised monolingually

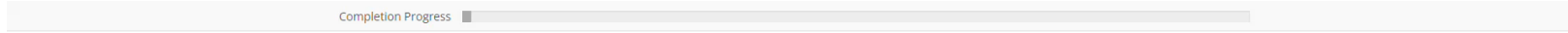
# Exp.1: Training phase

## Materials

- Semi-artificial language
- 90 V2 sentences constructed from 30 {S, O, V, A} sets
- Uniform condition: 33.3%-33.3%-33.3%
- Skewed conditions: 60%-20%-20%

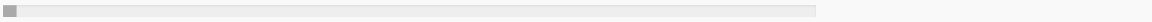
- (3) a. **The author** **revises** eventually **a novel** **in Boston**.
- b. **A novel** **revises** **the author** eventually **in Boston**.
- c. **In Boston** **revises** **the author** eventually **a novel**.

# Exp.1: Training phase





# Exp.1: Training phase

Completion Progress 

Form a sentence in the new English dialect with the given words

**Since 2010** \_ \_ \_ \_

brews the witch the potion personally

Reset Submit

*(or press enter)*

# Exp.1: Testing phase

## Production task

- Participants are provided with scrambled English words and must form sentence in artificial language
- Familiar constituent types (4 trials):
  - S, O, A (e.g. *Sophia, a carol, on Christmas*)
- Novel constituent types (4 trials each):
  - indirect objects (e.g. *to the prosecutor*)
  - complex adjuncts (e.g. *during the conflict*)

(4) {the waiter, awkwardly, to the guest, passes, the saltshaker}

# Exp.1: Testing phase

## Judgement task

- Participants see V2 & V3 sentences and need to judge grammaticality of it
- Familiar constituent types in initial position (4 trials each):
  - Direct objects
  - Simple adjuncts
- Novel constituent types in initial position (4 trials each):
  - Indirect objects
  - Complex adjuncts

# Exp.1: Testing phase

(5) **To the congregation** shows **the priest** silently **the candle**.

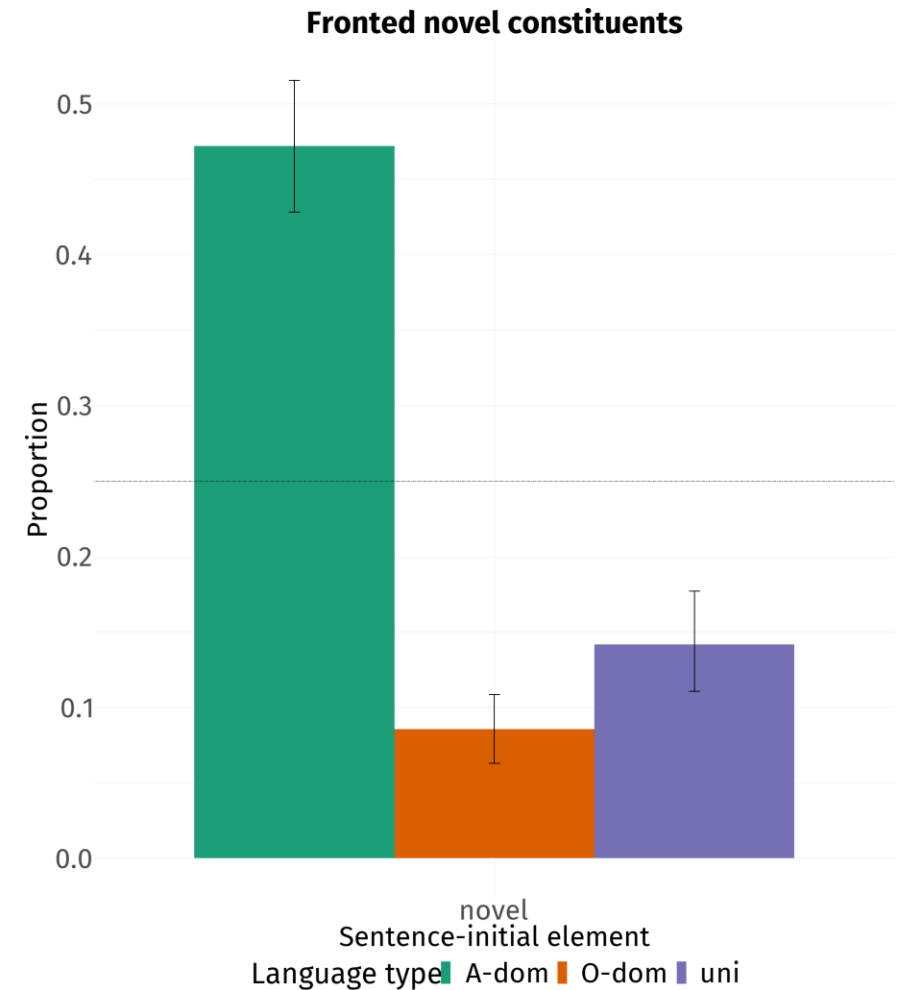
(6) **In late April** regrets **the politician** openly **his misconduct**.

(7) **To the doctor** **the patient** describes precisely **the pain**.

(8) **At the moment** **the referee** verifies briefly **the decision**.

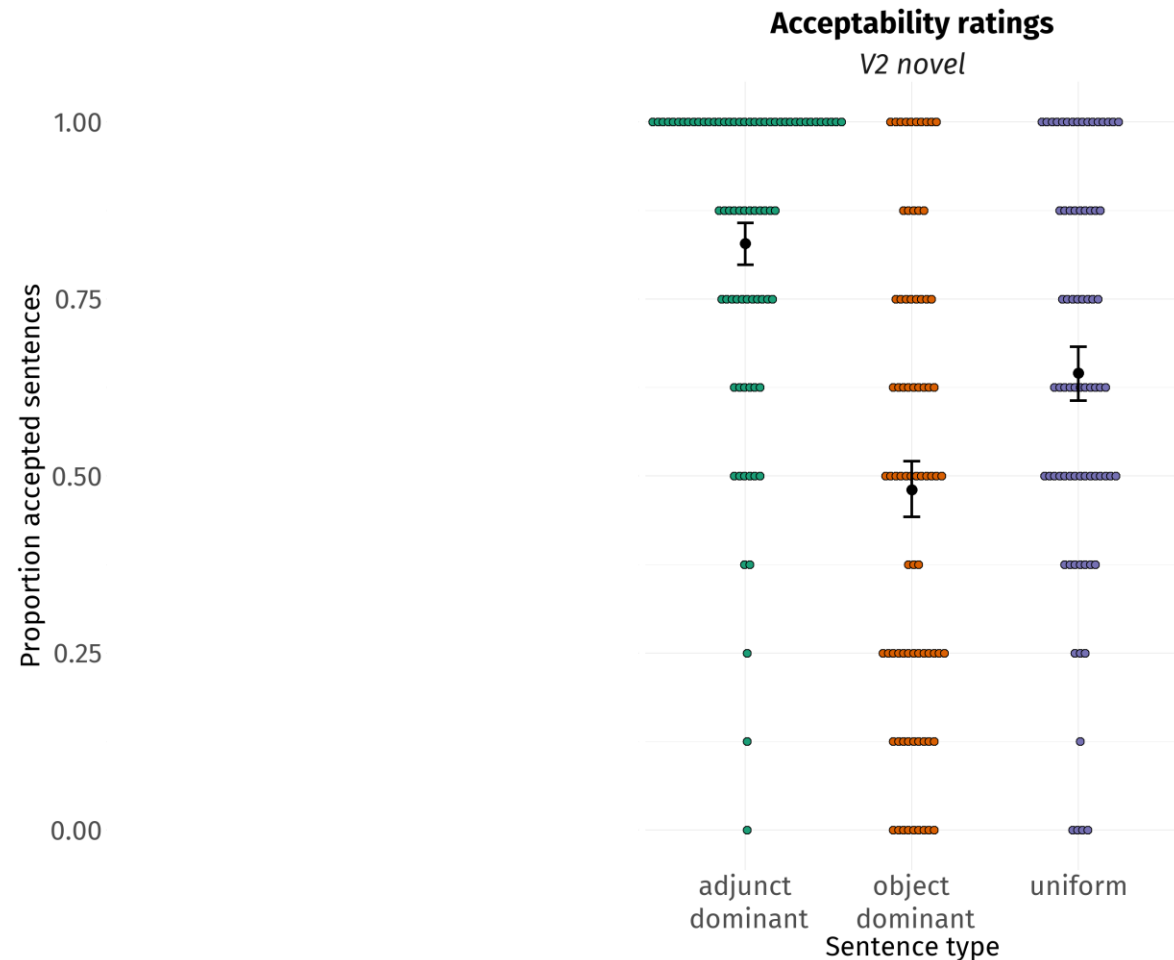
# Exp.1: Results – Production

- **Prediction:** fewer novel constituents fronted in skewed condition
  - Confirmed for O-dom. but not for A-dom.
  - Apparent advantage for learners in A-dominant condition



# Exp.1: Results – Judgement

- **Prediction:** Higher ratings for *V2 novel* in uni. condition
  - V2-new: A-dom. > Uni > O-dom.
- **Prediction:** Better discrimination btw. *V2 novel* & *V3* in uni. condition
  - Discrimination: A-dom. > Uni = O-dom.



# Exp. 1: Discussion

- V2 language easily learnable in short period
- Predictions mostly confirmed for O-dom. condition
- Participants in A-dom. condition exceed participants in uniform condition
- Why do participants in A-dom. and O-dom. condition differ?
  - More variability in A-dom. (PPs, AdvPs) than in O-dom. (DPs)?
  - Different types of violation?
  - Learning advantage through adjuncts?

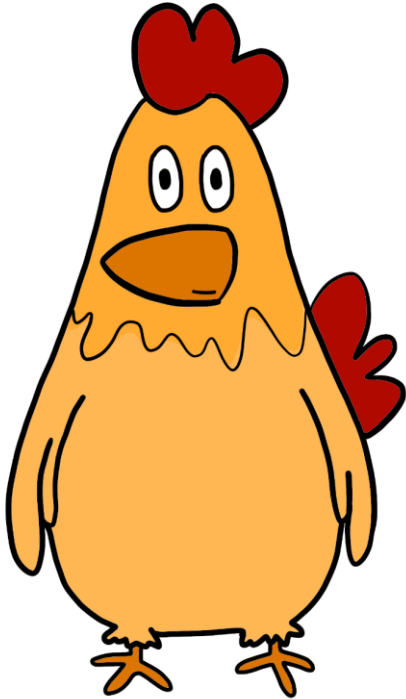
# Experiment 2



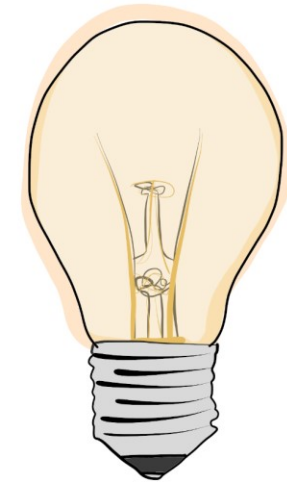
# Exp.2: Participants

- 211 participant tested, 197 included in analysis (93.4%)
  - Uni.: 50/55
  - S-dom.: 48/52
  - O-dom.: 50/53
  - A-dom.: 49/51
- Prolific
  - USA, UK, Ireland, Australia, Canada, New Zealand
  - Monolingual English speakers
  - Raised monolingually

# Exp.2: Noun training



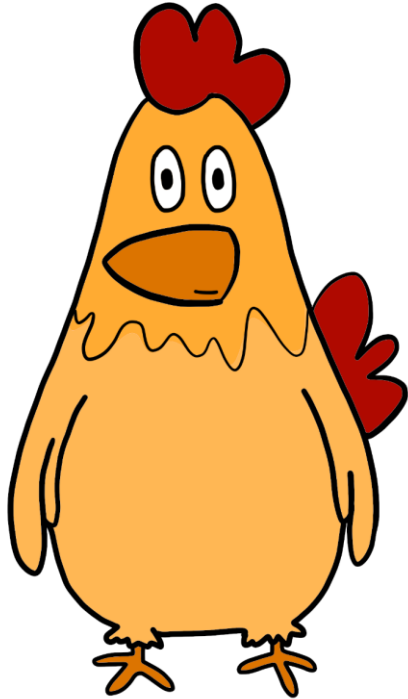
gak



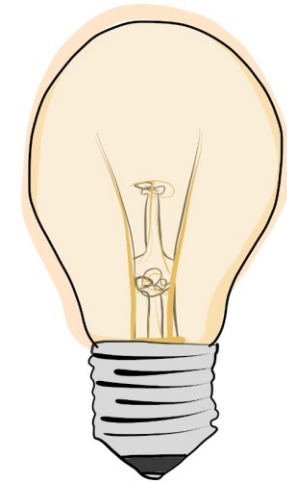
schin

\* Many thanks to Clem Ashton

# Exp.2: Noun testing



gak	muh
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tic	schin
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# Exp.2: Adposition training



en flek

# Exp.2: Adposition testing

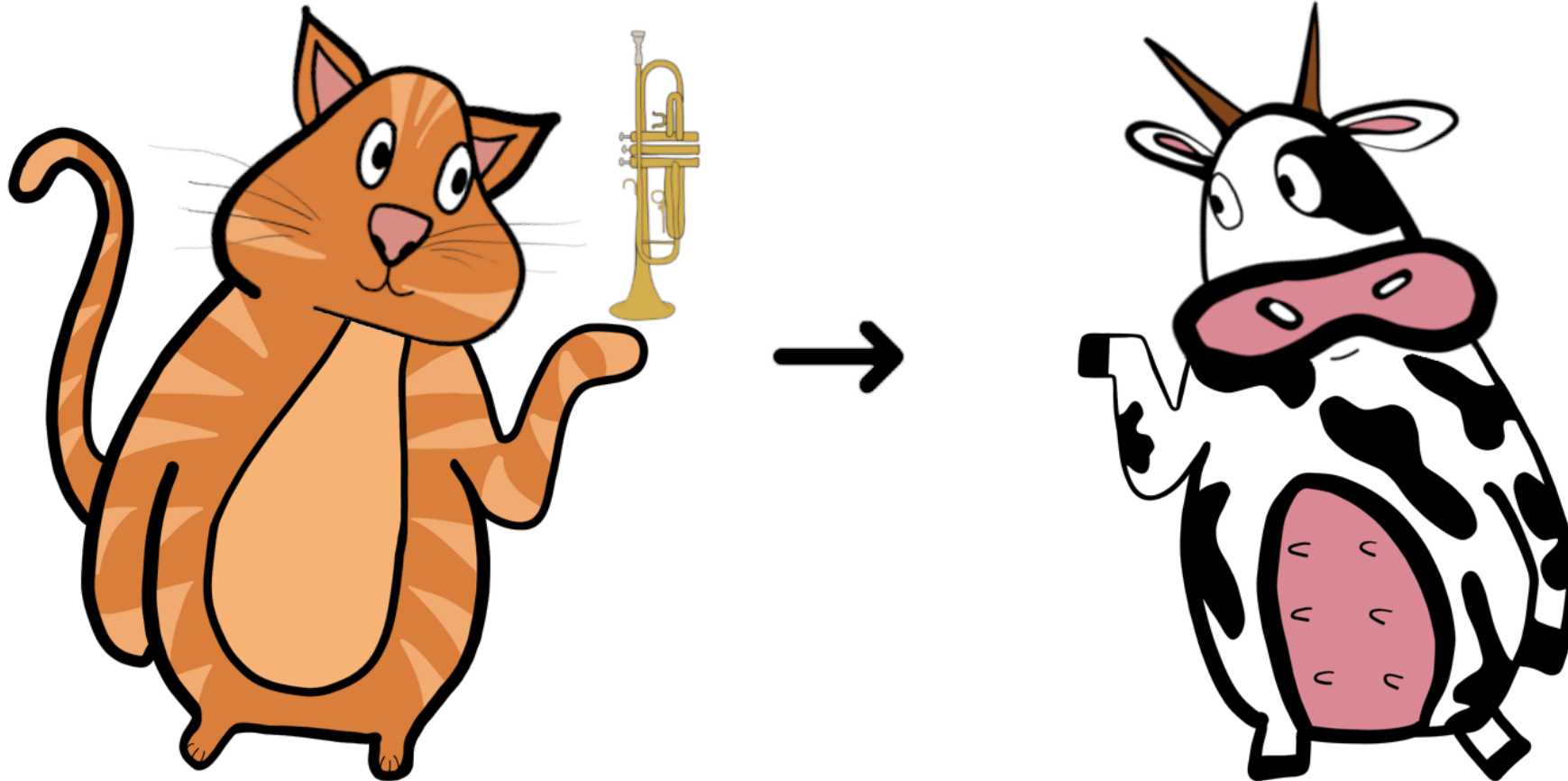


en flek	en sul
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# Exp.2: Sentence training



## Exp.2: Ditransitive scene



# Exp.2: Introduction of novel lexical items

- Presentation of additional animal, object & adposition
  - Random selection for each participant
  - Similar introduction as for other elements in training
- Introduction of ditransitive verb *hada* 'give'
  - Description of meaning



# Exp.2: Testing phase

## Production task – Bag of words

- Familiar constituent types
  - Subject, direct object, adjunct
  - Lexically familiar, lexically novel
- Novel constituent types
  - Lexically familiar indirect object
  - Lexically novel indirect object

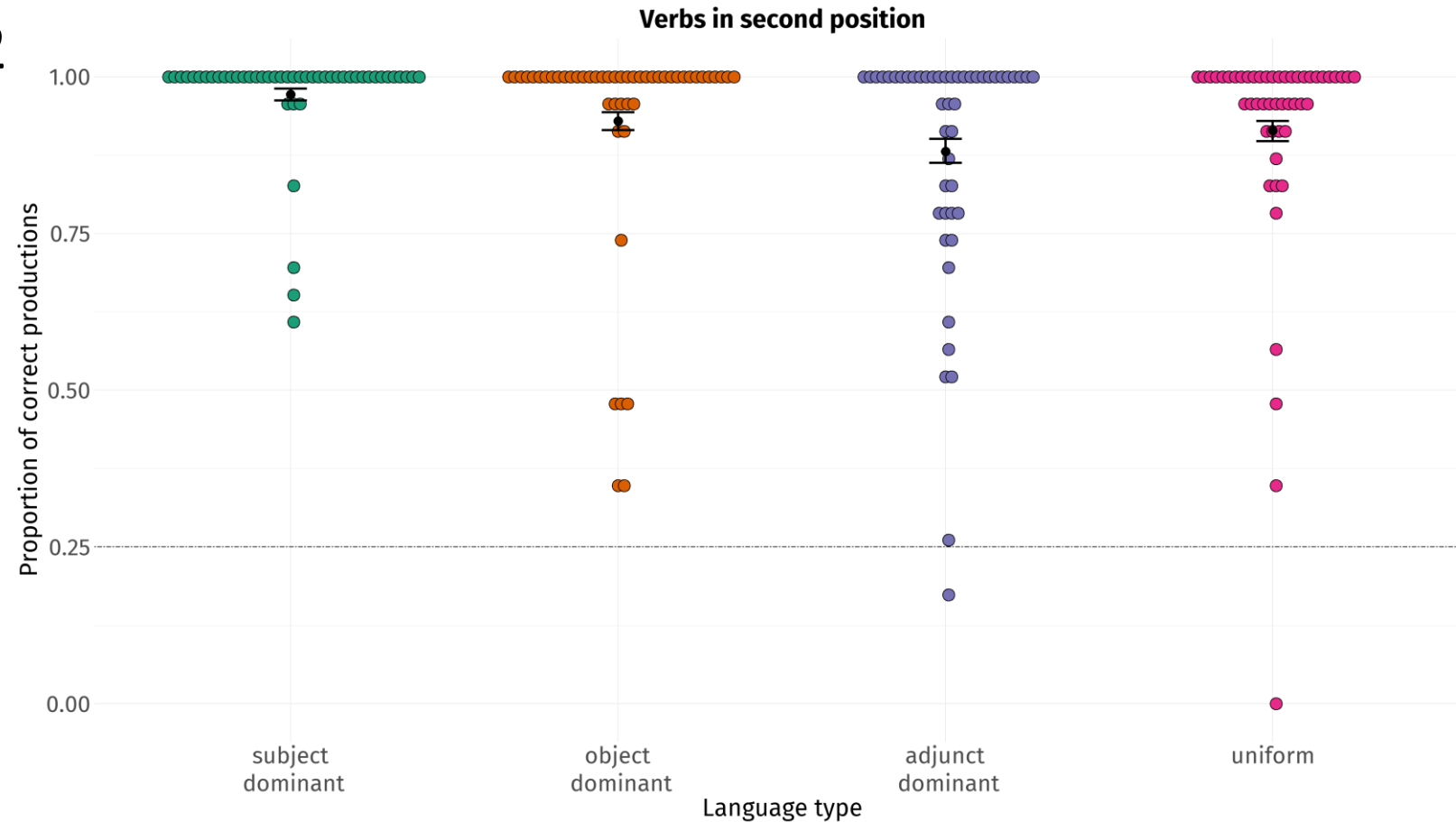
# Exp.2: Testing phase

## Judgement task – Sentence types

- V2 familiar
  - familiar clause-initial constituent type (S|O|A), lexically familiar
  - familiar clause-initial constituent type (S|O|A), lexically novel
- V2 novel
  - novel clause-initial constituent type (IO), lexically familiar
  - novel clause-initial constituent type (IO), lexically novel
- V3

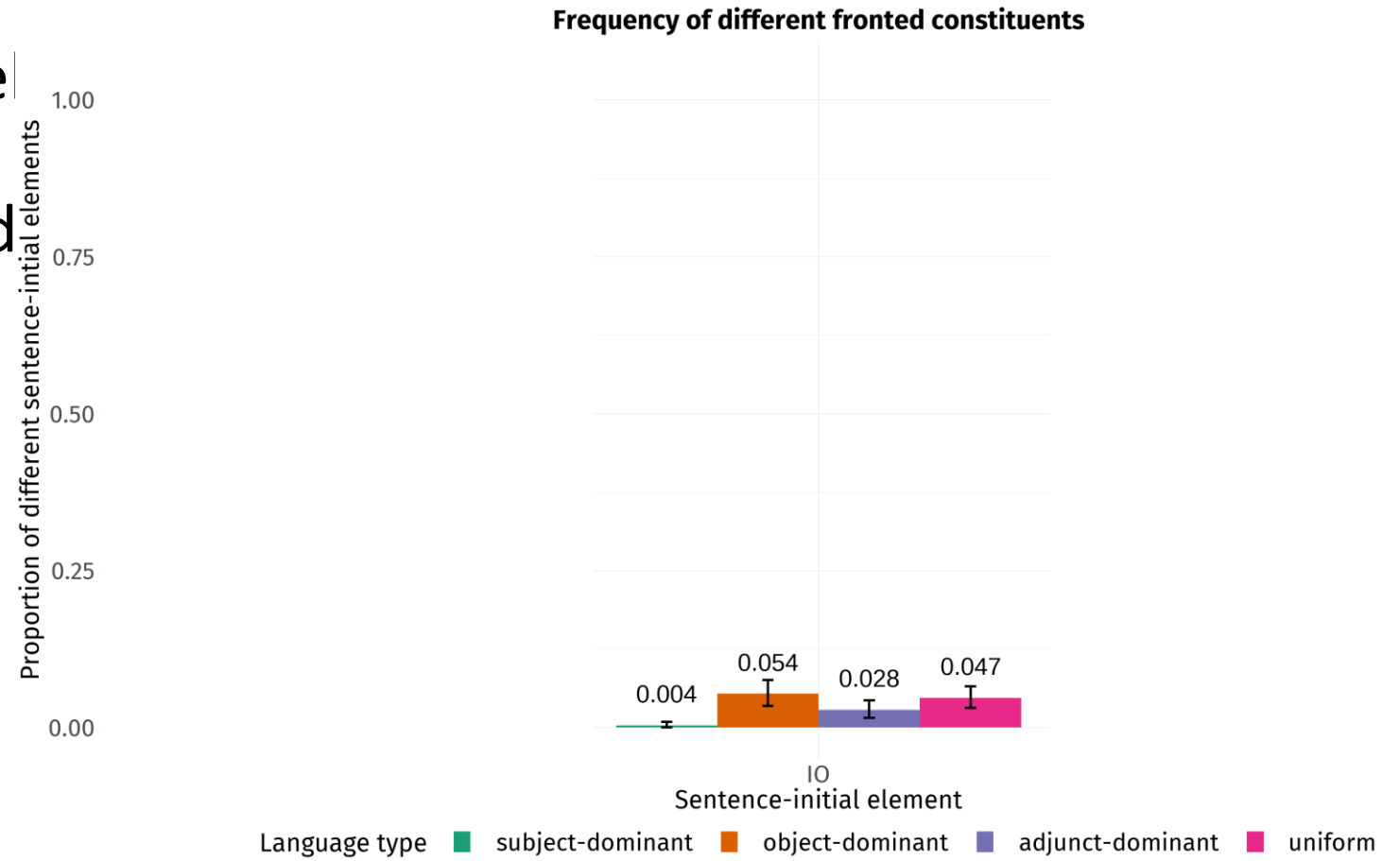
# Exp.2: Results – Production

- **Prediction:** Prop. of V2 sentences > chance, no significant  $\Delta$  btw. conditions
  - V2 > chance
  - S-dom = O-dom
  - S-dom > A-dom, uni
  - O-dom = A-dom = uni



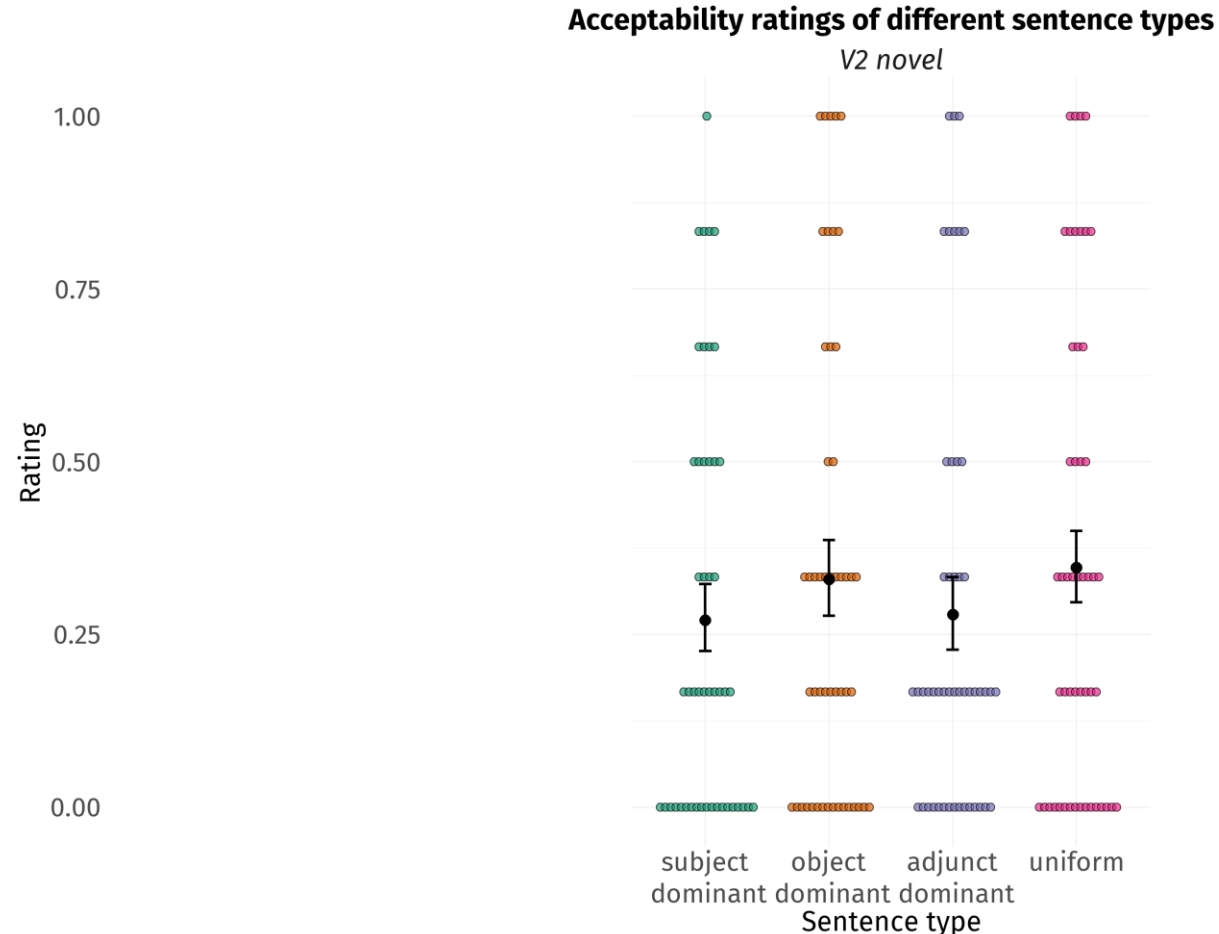
# Exp.2: Results – Production

- **Prediction:** fewer novel constituents (lexically & syntactically) fronted in skewed conditions
  - No significant differences btw. conditions
  - Subjects dominate



# Exp.2: Results – Judgement

- **Prediction:** Higher ratings for *V2 novel* in uni. condition
  - V2-new: S-dom. = O-dom. = A-dom. = Uni
- **Prediction:** Better discrimination btw. *V2 novel* & *V3* in uni. condition
  - Disc.: S-dom. = O-dom. = A-dom. = Uni



# Exp.2: Discussion

- Positioning rule of verbs reliably learned
- Null result: no difference between conditions
- Why are participants hesitant to generalise beyond input?
  - Number of training items too low?
  - Insufficient lexical variability?
  - Absence of variation of grammatical categories?

# General discussion

- V2 can be learned in right experimental environments
- Distributional properties of input can affect learning outcome of V2
- Variability of grammatical categories, not grammatical functions decisive
- Results support view that diminished evidence for V2 in input results in loss of V2
- Significant amount of A-initial sentences may be crucial for V2 acquisition
- ALL can complement study of language change

# Literature

Culbertson, Jennifer & Kathryn Schuler. 2019. Artificial language learning in children. *Annual Review of Linguistics* 5. 353–373.

Fodor, Janet Dean. 1998. Unambiguous triggers. *Linguistic Inquiry* 29(1). 1–36.

Folia, Vasiliki, Julia Uddén, Meinou De Vries, Christian Forkstam & Karl Magnus Petersson. 2010. Artificial language learning in adults and children. *Language Learning* 60(s2). 188–220.

Gómez, Rebecca L. 2002. Variability and detection of invariant structure. *Psychological Science* 13(5). 431–436.

Gómez, Rebecca L. & LouAnn Gerken. 2000. Infant artificial language learning and language acquisition. *Trends in Cognitive Sciences* 4(5). 178–186.

Gómez, Rebecca & Jessica Maye. 2005. The developmental trajectory of nonadjacent dependency learning. *Infancy* 7(2). 183–206.

Lightfoot, David. 1999. *The development of language. Acquisition, change, and evolution*. Malden, MA: Blackwell.

Lightfoot, David. 2006. *How new languages emerge*. Cambridge: Cambridge University Press.

Raviv, Limor, Gary Lupyan & Shawn C. Green. 2022. How variability shapes learning and generalization. *Trends in Cognitive Sciences* 26(6). 462–483.

Sitaridou, Ioanna. 2012. A comparative study of word order in Old Romance. *Folia Linguistica* 46(2). 553–604.

Sneller, Betsy & Gareth Roberts. 2018. Why some behaviors spread while others don't: A laboratory simulation of dialect contact. *Cognition* 170. 298–311.

Tabullo, Ángel, Mariana Arismendi, Alejandro Wainelboim, Gerardo Primero, Sergio Vernis, Enrique Segura, Silvano Zanutto & Alberto Yorio. 2012. On the learnability of frequent and infrequent word orders: An artificial language learning study. *Quarterly Journal of Experimental Psychology* 65(9). 1848–1863.

Yang, Charles. 2000. Internal and external forces in language change. *Language Variation and Change* 12(3). 231–250.

Yang, Charles. 2002. *Knowledge and learning in natural language*. New York: Oxford University Press.

Yin, Sora Heng & James White. 2018. Neutralization and homophony avoidance in phonological learning. *Cognition*. Elsevier 179. 89–101.