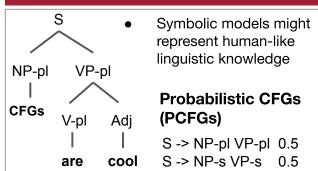
Comparing Symbolic Models of Language via Bayesian Inference

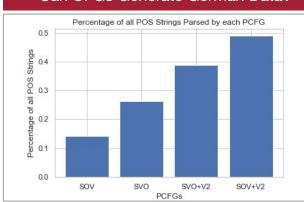
Massachusetts **Technology**

Annika Heuser & Polina Tsvilodub

Symbolic Models: Context-free grammars



Can CFGs Generate German Data?



Challenges

P(CFG, Type | Data) P(Data | CFG, Type) P(CFG | Type) P(Type) P(Data | CFG, Type) = product of probability of each rule needed to parse each sentence

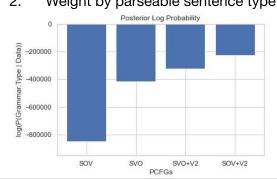
→ PCFG likelihoods are not directly comparable because they parse different data subsets PCFGs still penalized for parsing more sentences because:

- Longer sentences = lower likelihood
- More generalizable = less probability mass per sentence type

Errors in data: 31.65% of the total POS strings are ungrammatical

Solution

- 1. Normalize by sentence length
- Weight by parseable sentence type



Contributions

- Method for more flexible comparison of symbolic models, specifically CFGs
- Revealed troubling number of errors resulting from transcription and part-of-speech (POS) tagging

References

[2] Crain, S.; and Nakayama, M. 1987. Structure dependence in grammar formation / anguage 522-543.

[6] Perfors, A.; Tenenbaum, J. B.; and Regier, T. 2011. The learnability of abstract syntactic principles. Cognition