## Language Statistics

## won't solve

## Language Processing

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## What is "Language Processing"?



## What is "Language Processing"?

Marr's Levels

- Computational:

Most NLP applications (sentiment analysis, machine translation, summarization, etc)

- Algorithmic / Representational:

Some parsing, NN interpretability, computational psycholinguistics

- Implementational


## Two kinds of statistical learning naysayers

## Generative Linguists

- Poverty of the stimulus
- Language requires special innate cognitive biases


## Multimodality Proponents

- Can't learn meaning from form (Bender \& Koller, 2020)
- Need to be embodied physically and socially
(Bisk et al., 2020)

talk tldr: Check your data


## Algorithmic level requires more than Language stats

## COGNITIVE SCIENCE

A Multidisciplinary Journal
Cognitive Science 45 (2021) e12988
© 2021 Cognitive Science Society LLC
ISSN: 1551-6709 online
DOI: 10.1111/cogs. 12988

Single-Stage Prediction Models Do Not Explain the Magnitude of Syntactic Disambiguation Difficulty

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## The horse raced past the barn fell

Bever, 1970, Cognition and the Development of Language

# The horse which was raced past the barn fell 

Bever, 1970, Cognition and the Development of Language


Garden paths produce a visceral response

Garden path responses exist in the tail of the response distribution

They exist in the tail because

1) the statistics are in the tail (predictability)

OR
2) the response is unusual (reanalysis)

NNs can predict garden path existence
van Schijndel \& Linzen, 2018, Proc CogSci Futrell et al., 2019, Proc NAACL Frank \& Hoeks, 2019, Proc CogSci Davis \& van Schijndel, 2020, Proc CogSci

NNs can predict garden path existence

Look beyond garden path existence to garden path magnitude

Effect of P (word wontext $_{\mathrm{n}} \mid$ on reading time measured at...


$$
R T\left(w_{i}\right)=\delta_{0} S\left(w_{i}\right)+\delta_{-1} S\left(w_{i-1}\right)+\delta_{-2} S\left(w_{i-2}\right)+\delta_{-3} S\left(w_{i-3}\right)
$$

Smith and Levy, 2013, Cognition

## WikiRNN:

Gulordava et al. (2018) LSTM
Data: Wikipedia (80M words)

## SoapRNN:

2-layer LSTM (Same parameters as WikiRNN)
Data: Corpus of American Soap Operas (80M words; Davies, 2011)

## Mapping probs to reading times

## Reading Time Data (SPR; Prasad and Linzen, 2019)

- 80 simple sentences (fillers)
- 224 participants
- 1000 words / participant

Linear Mixed Regression
time $\sim$ text position + word length $x$ frequency $+\ldots+$ predictability $_{t}$

Smith \& Levy, 2013:
$\delta_{0}=0.53 \delta_{-1}=1.53 \delta_{-2}=0.92 \delta_{-3}=0.84$
WikiRNN using Prasad \& Linzen, 2019:
$\left(\delta_{0}=0.04\right) \delta_{-1}=1.10 \delta_{-2}=0.37 \delta_{-3}=0.39$
SoapRNN using Prasad \& Linzen, 2019:
$\left(\delta_{0}=-0.04\right) \delta_{-1}=0.83 \delta_{-2}=0.91 \delta_{-3}=0.44$

## Three Garden Paths

NP/S: The woman saw $\left\{\begin{array}{l}\text { the doctor wore a hat. } \\ \text { that the doctor wore a hat. }\end{array}\right.$
NP/Z: When the woman $\left\{\begin{array}{l}\text { visited her nephew laughed loudly. } \\ \text { visited, her nephew laughed loudly. }\end{array}\right.$
MV/RR: The horse $\left\{\begin{array}{l}\text { raced past the barn fell. } \\ \text { which was raced past the barn fell. }\end{array}\right.$

## The horse raced past the barn fell The horse which was raced past the barn fell

Bever, 1970, Cognition and the Development of Language

## The linear relationship doesn't hold

Predicted/empirical mean garden path effects


## Paper Conclusions

- Conversion rates are fairly similar, but all underestimate human responses
- Suggests human responses influenced by factors beyond predictability


## Talk Conclusion

- Algorithmic processing cannot be learned from Language statistics


## Computational level requires more than Language stats



Recurrent Neural Network Language Models Always Learn English-Like Relative Clause Attachment

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## Does our data match our goal?

Why can we not predict garden path response sizes?

Because the boggle response is not in the training data

## Ambiguous Relative Clause Attachment

## John met the agent of the rocker that is divorced

## Ambiguous Relative Clause Attachment



## Ambiguous Relative Clause Attachment



## John met the agent of the rocker that is divorced

## Ambiguous Relative Clause Attachment

## HIGH

## John met the agent of the rocker that is divorced

## Ambiguous Relative Clause Attachment

English speakers<br>have a preference for LOW



## John met the agent of the rocker that is divorced

## Ambiguous Relative Clause Attachment

## Spanish speakers HIGH have a preference for HIGH



## John met the agent of the rocker that is divorced

| Afrikaans | $\underline{\text { Japanese }}$ |
| :--- | :--- |
| Arabic | Norwegian |
| Croatian | $\underline{\text { Persian }}$ |
| Danish | $\underline{\text { B. Portuguese }}$ |
| Dutch | Romanian |
| English | $\underline{\text { Russian }}$ |
| $\underline{\text { French }}$ | $\underline{\text { Spanish }}$ |
| $\underline{\text { German }}$ | Swedish |
| $\underline{\text { Greek }}$ | $\underline{\text { Thai }}$ |
| $\underline{\text { Italian }}$ |  |

## Do RNN LMs learn language attachment preferences?

- Used existing stimuli from psycholinguistics (40 sentence frames)
- Balanced for number

1) 

a) Andrew had dinner yesterday with the nephew of the teachers that was divorced.
b) Andrew had dinner yesterday with the nephews of the teacher that was divorced.
from Fernández (2003)

## RNN LMs seem to have a LOW bias

Proportion of HIGH/LOW Attachment


## RNN LMs seem to have a LOW bias



## Do RNN LMs learn Spanish preference?

2) 

a) André cenó ayer con el sobrino de los maestros que estaba divorciado. divorciado.
from Fernández (2003)

## Spanish Results

Proportion of Spanish HIGH/LOW Attachment


LOW


## Spanish Results

Proportion of Spanish HIGH/LOW Attachment
p-value < 0.00001
Bayes Factor > 100


## Why can't the model learn Spanish attachment?

RNN LMs can acquire HIGH or LOW bias when trained on synthetic data

- Synthetic data from PCFG with declarative sentences and sentences with the target RC construction
- $10 \%$ of training data had unambiguous RC sentences
- Incrementing how much of that had HIGH vs LOW
- When at least 50\% of RC sentences had HIGH attachment model preferred HIGH attachment


## Comprehension signal not in raw text data

## Spanish Wikipedia (training corpus):

LOW 69\% more frequent than HIGH

Spanish Newswire data:

LOW 21\% more frequent than HIGH

## Comprehension and Production



## Comprehension is a superset of Production



## Conclusions

- Language statistics reflect human production biases
- Most NLP tasks are about comprehension
- What kind of training signal is needed for comprehension?


## Thanks!



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