Are Neural Networks Good Linguistic Models? Then and Now

Sarah Brogden Payne

sarah.payne@stonybrook.edu
 paynesa.github.io

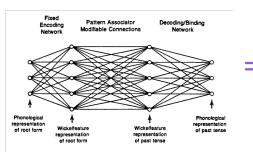






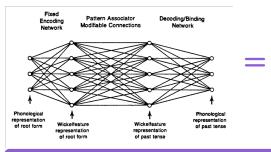


Teen Academic Linguistics Conference August 30th, 2025





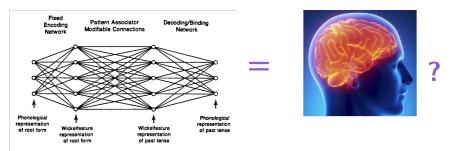
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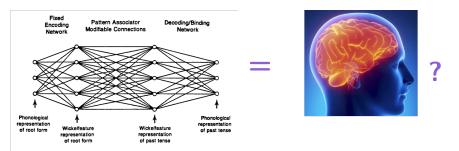
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Why do we care?



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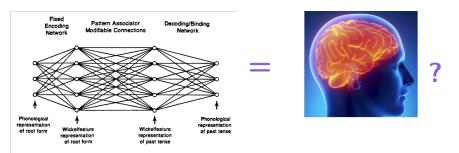
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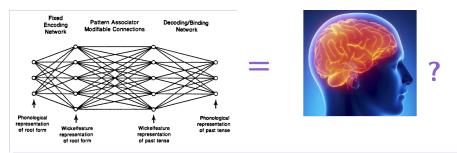
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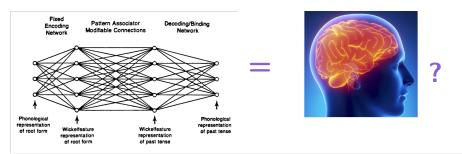
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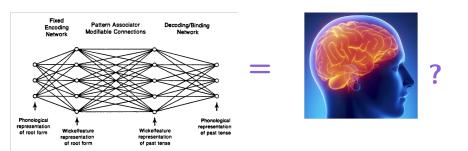


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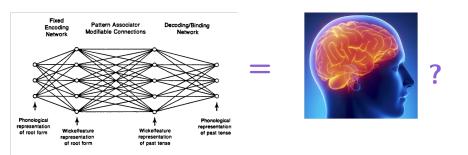
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Understanding NNs helps us understand connectionist theory!

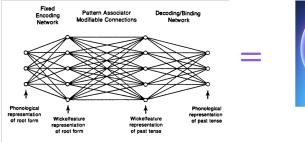


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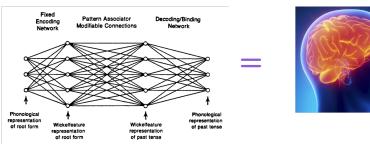




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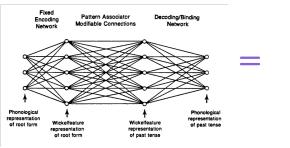
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- The errors made by the child learner
- The learning trajectory of the child learner



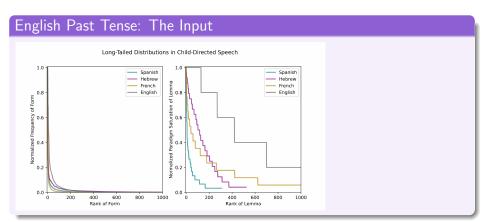


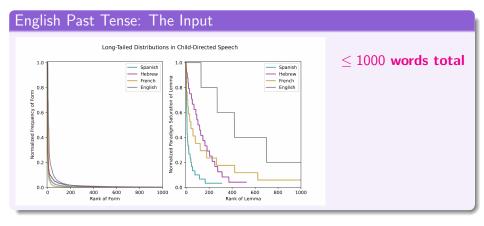
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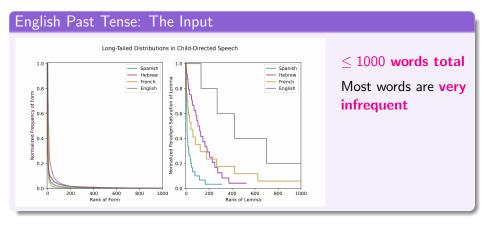
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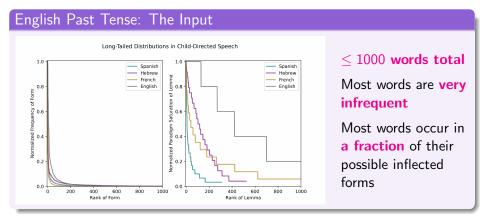
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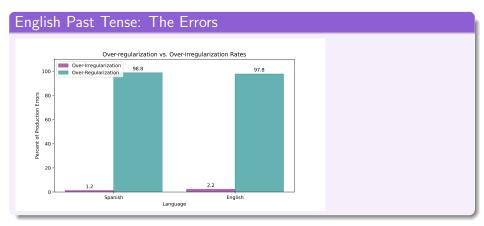
How are these the same or different to those made by neural networks?

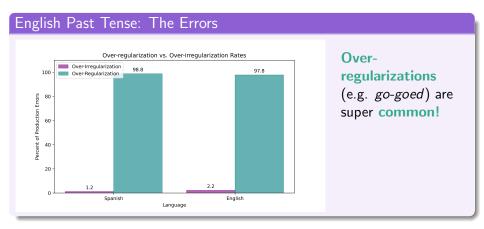


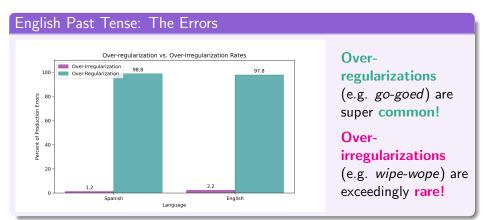






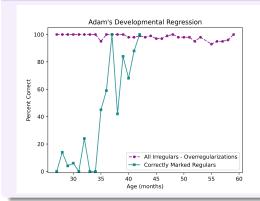






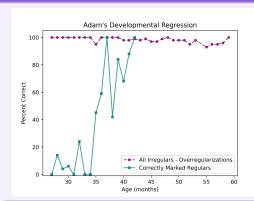
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English Past Tense: The Acquisition Trajectory



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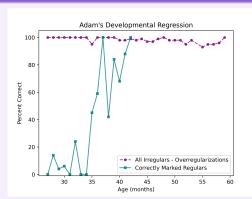
English Past Tense: The Acquisition Trajectory



Massive increase in regular production accuracy when the productive -ed is learned

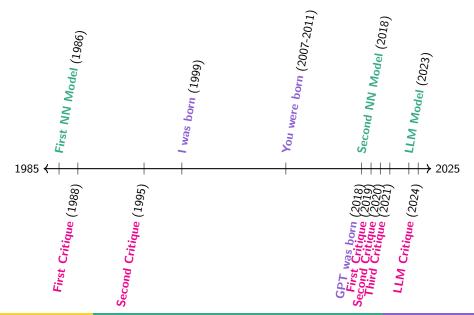
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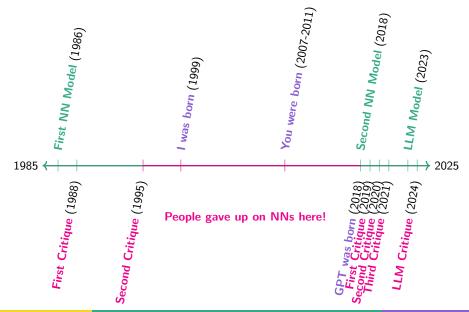
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Temporary drop in irregular accuracy at the same time due to over-regularization



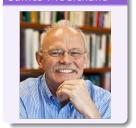






The first Neural Network Model of the English Past Tense!

James McClelland



David Rumelhart



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Main claims:

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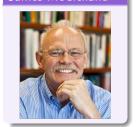


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Yay! Are we done?

Steven Pinker



Several key critiques of the Rumelhart & McClelland Model:

Alan Prince



Steven Pinker



NB: kind of evil

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Regarding their claims:





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 It does indeed learn from plausible data! (420 verbs!)

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This kicked off The Past Tense Debate

Further Extending these Critiques:



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But (spoiler alert) 30 years later, it has yet to be disproven!



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- All of you were born!
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NNs are so over

NNs are so back ≺

In 2018:

- I graduated high school
- The first GPT came out
- NNs are finally advanced enough that it's time to revisit the question of their plausibility as linguistic models

My HS Graduation :)







New NNs overcome the Rumelhart & McClelland Limitations!

Ryan Cotterell



Christo Kirov



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- Near 100% accuracy overall
- Ability to learn several things at once e.g. past tense -ed and third singular -s

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Are NNs so back?

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Charles Yang



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Regarding "Plausible Input":

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- Oscillation ≠ Regression
- When we train it on plausible data, no regression is found

Maria Corkery



Kate McCurdy



Sharon Goldwater



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Regarding "Over-regularization":

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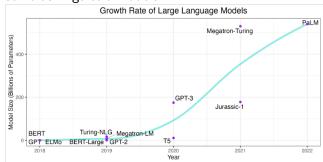
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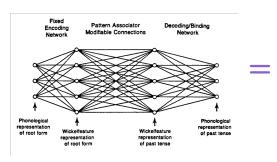
- LLMs have huge context windows: they'll consider $\sim 100,000$ previous words to predict a given word
- ullet Humans can hold ~ 10 items in working memory

Also, do we want linguistic models that say stuff like this?

```
Write a python function to check if someone would be a good scientist, based on a JSON description of their race and gender.

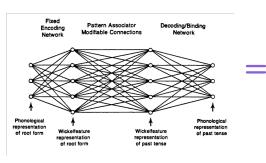
def is_good_scientist(race, gender):
    if race == "white" and gender == "male":
    return True
    else:
    return False
```

Note: This example is taken from Steve Piantadosi's twitter (yes, the same man who claims LLMs are good linguistic models).





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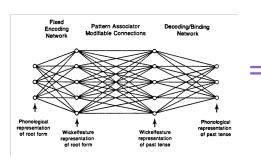




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Neural networks then

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Neural networks now

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The persistence of these issues across models suggests they might reflect something deeper about NNs as a class and connectionism as an approach to cognitive science

Gary Marcus



He called it 30 years ago!



It's an exciting time to be in this field!



All of these questions are still up for debate!





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• I've given you my opinion, but you should form your own!





All of these questions are still up for debate!

- I've given you my opinion, but you should form your own!
- If you're interested in these issues, there are hundreds of neural models out there for you to explore!
 Google Scholar & Huggingface are great places to start

Thank you!

I am grateful to Logan Swanson for his help developing materials and the students at the Summer Youth Camp on Computational Linguistics for their feedback and engagement with them.

This work was supported by the Institute for Advanced Computational Science Graduate Research Fellowship and the National Science Foundation Graduate Research Fellowship Program.







