

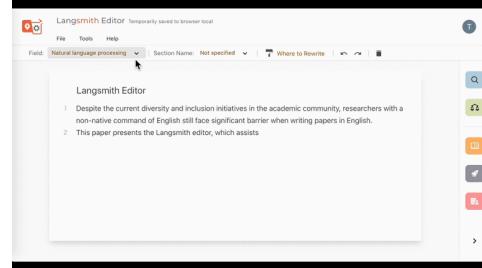
Next-Word Prediction in Language Models and Humans

Tatsuki Kuribayashi (MBZUAI)

Research topics

Automated writing assistance
 [ACL 2019, INLG 2019, EMNLP 2019, ACL 2020,
 Journal of CogSci 20, EMNLP-demo 2020, UIST 2023]

- Editor for natural language
- Startup

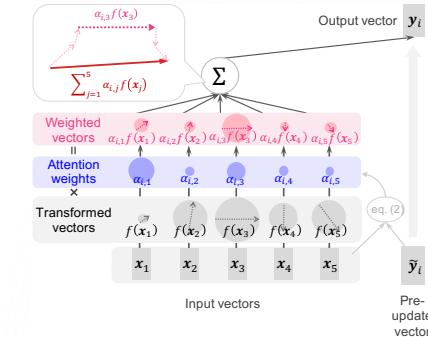


*What is language processing in humans?
 How can we help our own?*

Computational psycholinguistics
 [ACL 2021, EMNLP 2022, ACL 2023,
 COLING 2024, NAACL 2024, ACL 2024,
 COLING 2025]

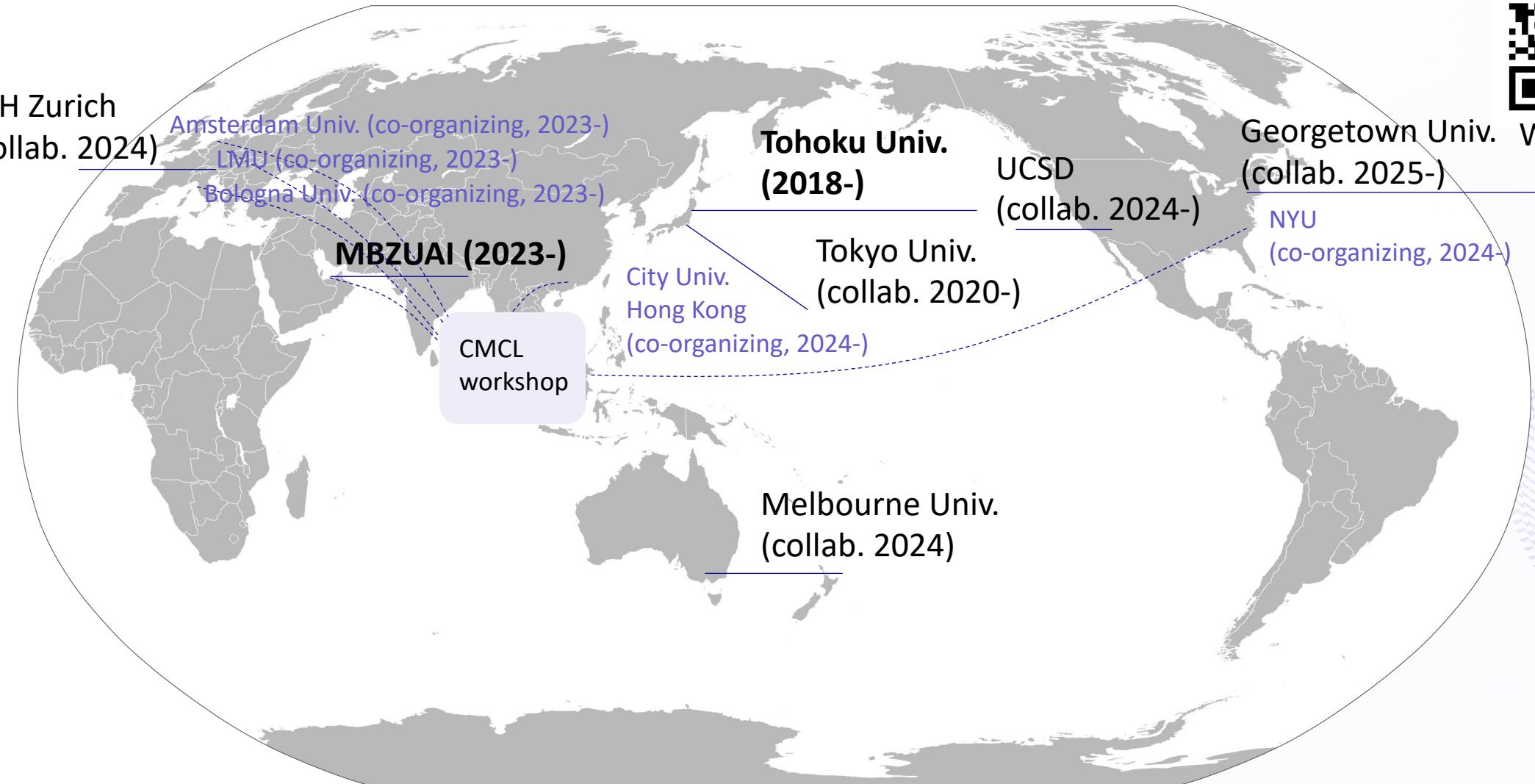
(First-authored, co-authored)

LM interpretability
 [EMNMLP 2020, EMNLP 2021,
 ACL 2023, EMNLP 2023, ICLR 2024 (spotlight)]



How different is language processing in humans and LMs?

International research connections



Roadmap

① motivation

20 min.

② my main research directions

20 min.

③ future directions

5 min.

Scientific modeling

- Why do planets move as observed?



- How did organic compounds emerge on the Earth?



- How do crowd crushes occur?



(figures are from Wikipedia or <https://www.irasutoya.com/>, unless otherwise stated)

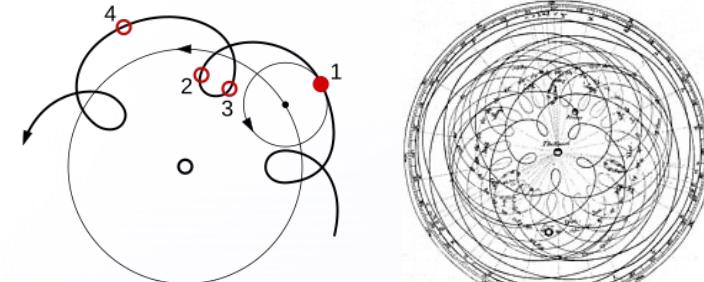
Scientific modeling

- Why do planets move as observed?



Build a model

Test the fit/prediction

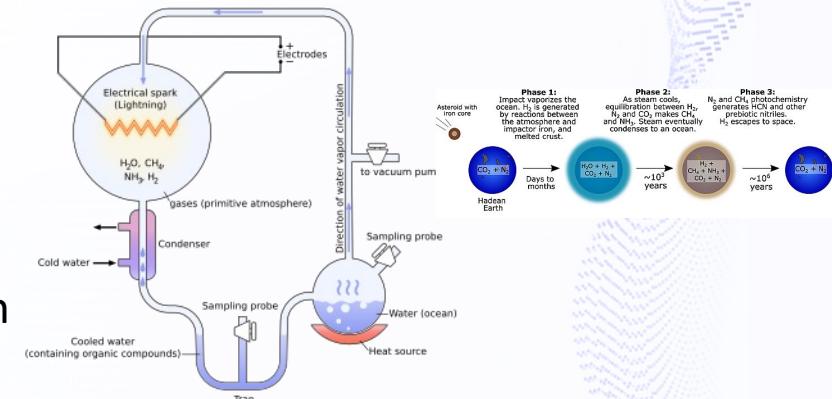


- How did organic compounds emerge on the Earth?



Build a model

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- How do crowd crushes occur?



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https://www.rcast.u-tokyo.ac.jp/ja/research/nishinari_lab.html

Scientific modeling

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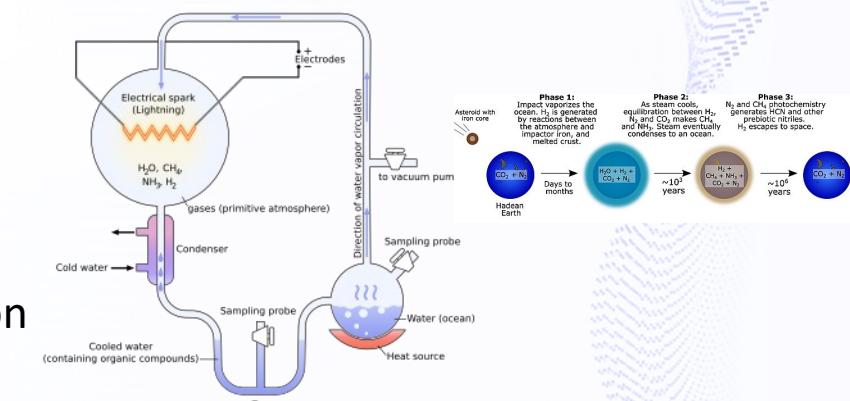
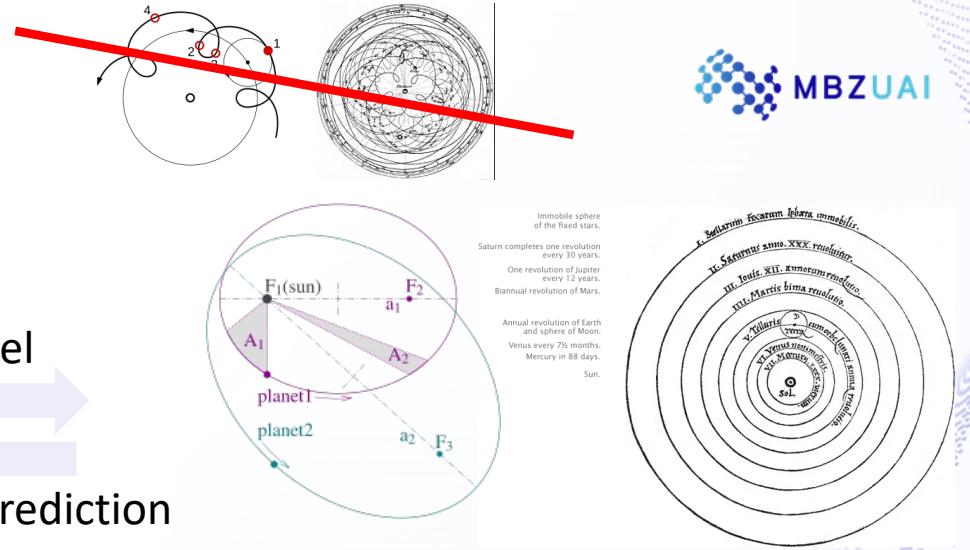
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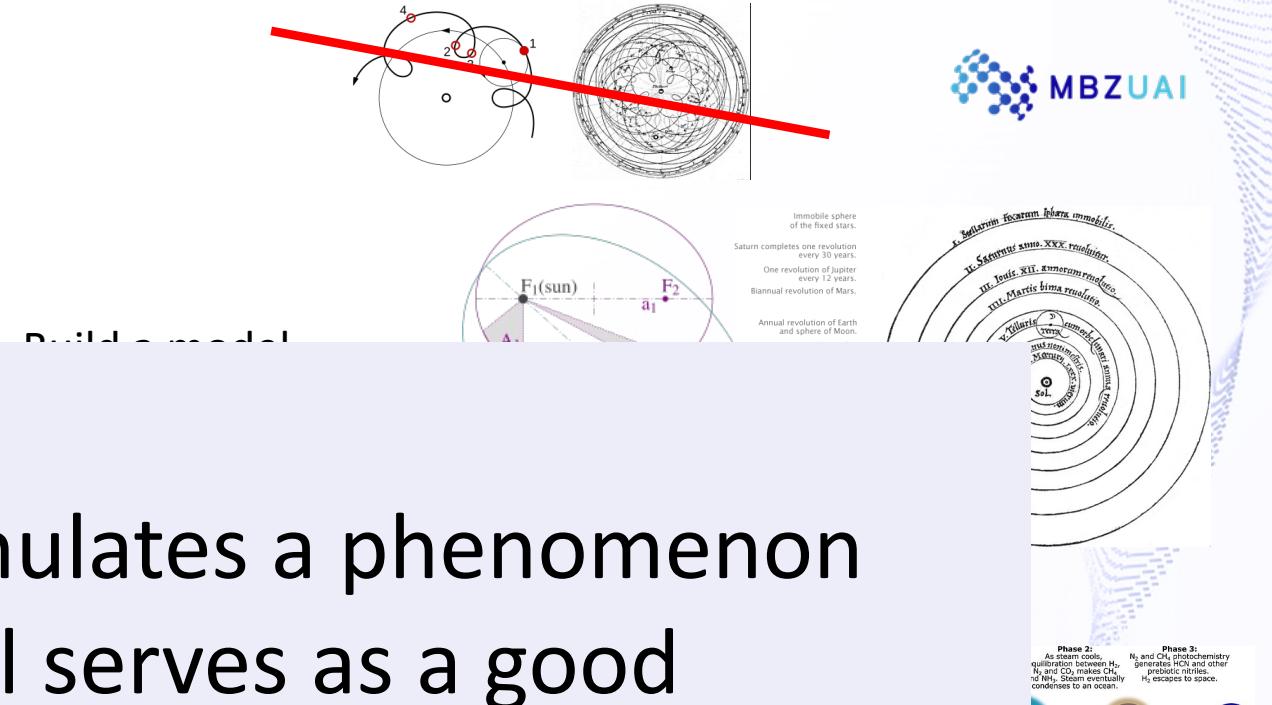


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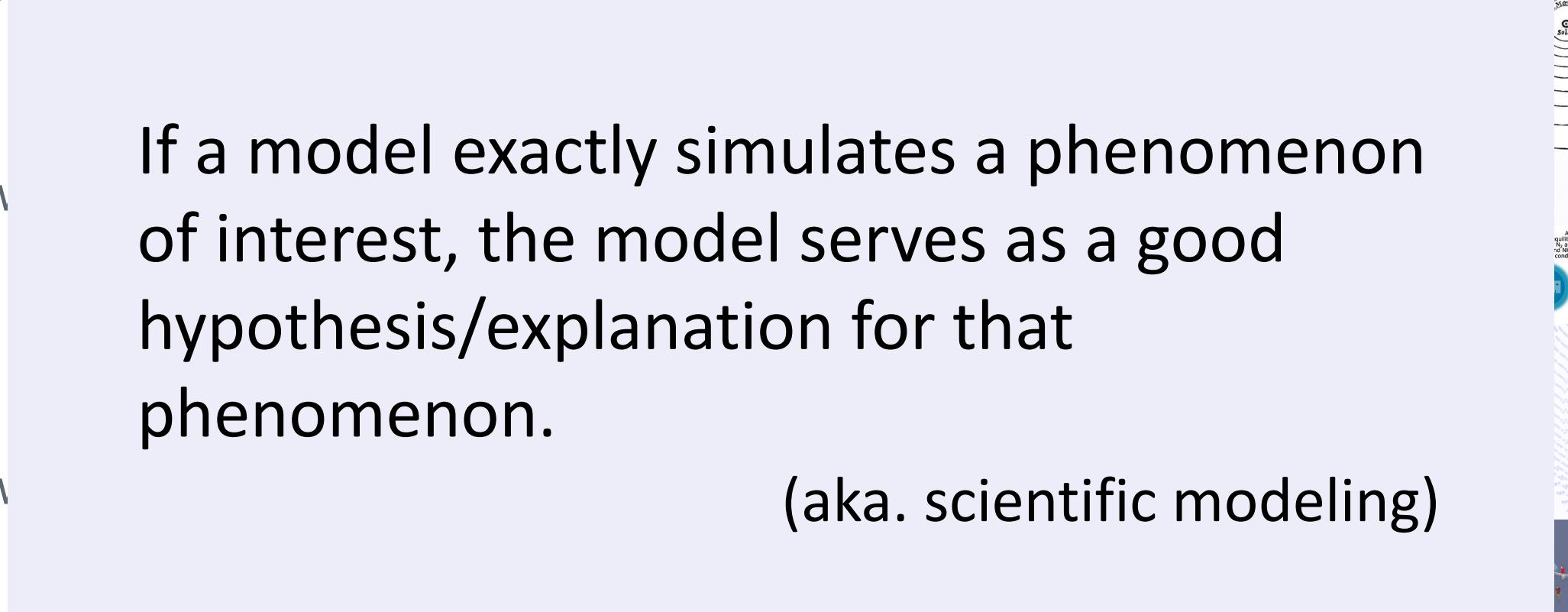
Scientific modeling

- Why do planets move as observed?



- How does it work?
- If a model exactly simulates a phenomenon of interest, the model serves as a good hypothesis/explanation for that phenomenon.
- How can we test it?

(aka. scientific modeling)

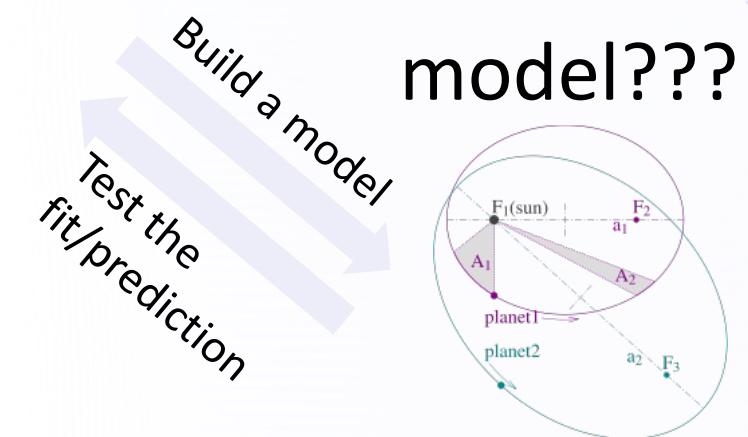


Fundamental linguistic questions

- What are humans computing during real-time language processing?
 - *What kind of equation are you now computing in front of this slide?*
- What is minimum requirements to be able to acquire language?
 - *Why do cats never start talking even if one keeps talking to them everyday?*
- Why do natural languages share certain universals, e.g., subject precedes objects?
 - *Why do languages shape as is? How did it emerge?*

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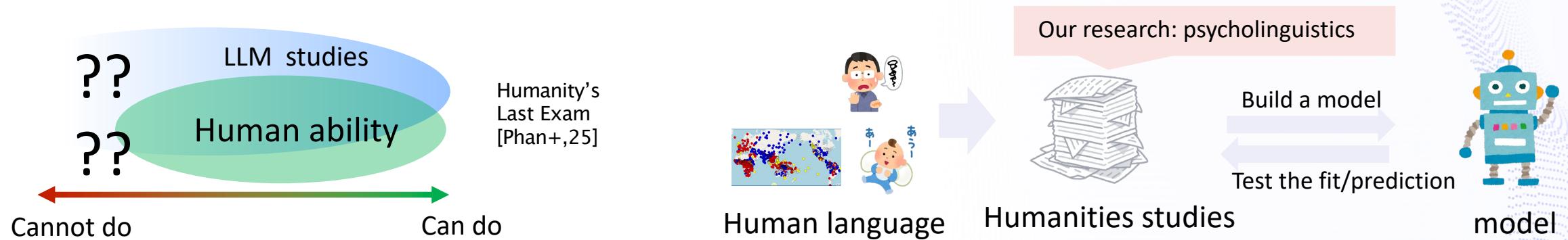


Why is artificial intelligence (AI) relevant to humanities?

- Science requires objectivity
- Paradox: if humans start to introspect about ourselves to study human intelligence, this will lack objectivity
- Thus, we have to build a model (artificial intelligence), apart from humans and test it
- One of the original goals of the AI field --- understanding it by building it
 - ...*the field (artificial intelligence) from three points of view: computational psychology, computational philosophy, and machine intelligence... The goal of computational psychology is to understand human intelligent behavior by creating computer programs that behave in the same way that people do... The program should do quickly what people do quickly, should do more slowly what people have difficulty doing, and should even tend to make mistakes where people tend to make mistake...*
[Encyclopedia of Artificial Intelligence, Shapiro 1991]

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- Here, the goal is to build an exactly *human-like* computational model that simulates phenomena of humans, following the scientific modeling approach



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Going back to 7 BCE - 16 CE...

- “Humans to explain humans” is super unethical (especially in causality experiments)



Pharaoh Psamtik
(664 – 610 BCE)



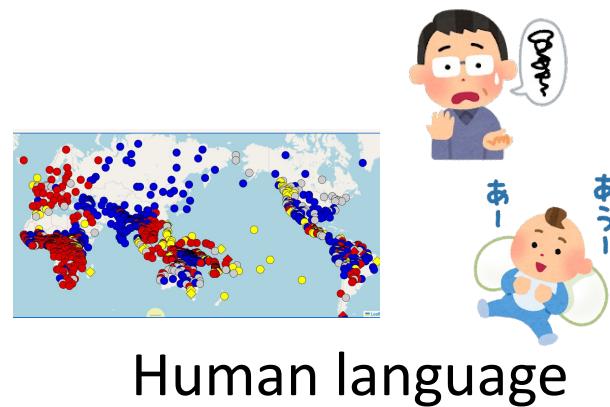
Frederick II
(1194-1250)



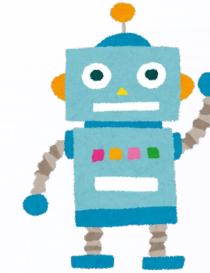
James IV
(1473-1513)

If one locks an infant in a room, what language will they start speaking?

In 2025...



Build a model
→
← Test the fit/prediction

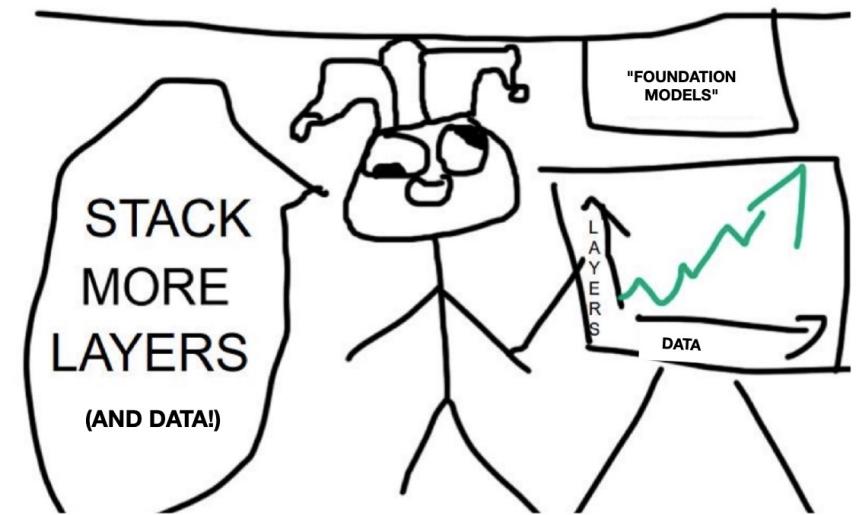


model

LLMs... are you the model of humans...?

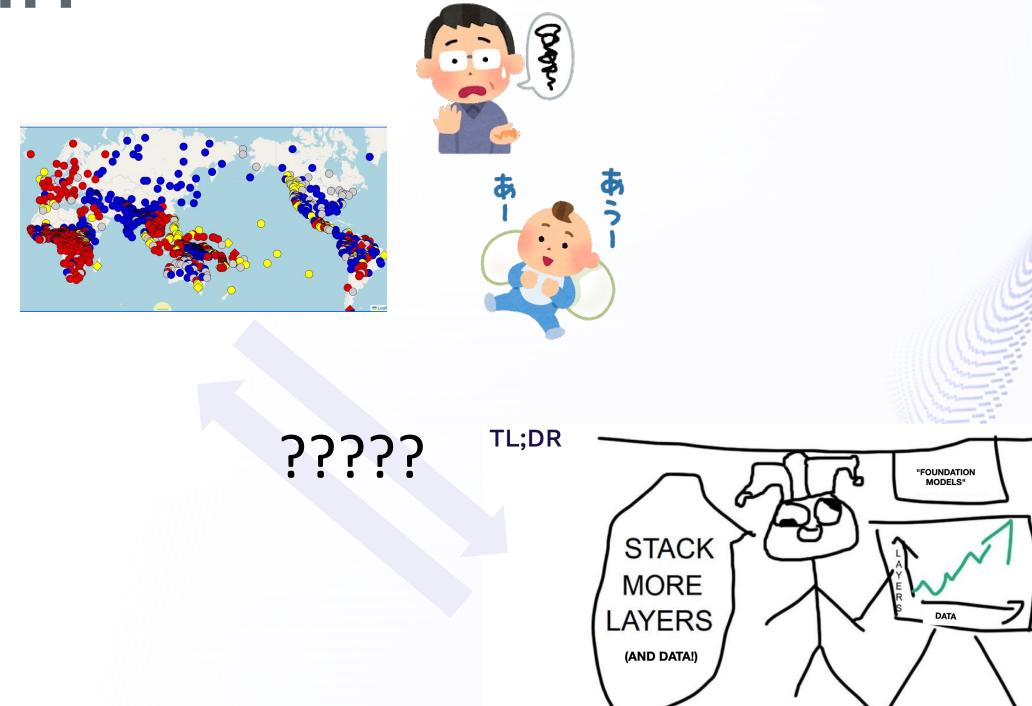
- We humans somehow found one way to build a model that behaves like humans

TL;DR



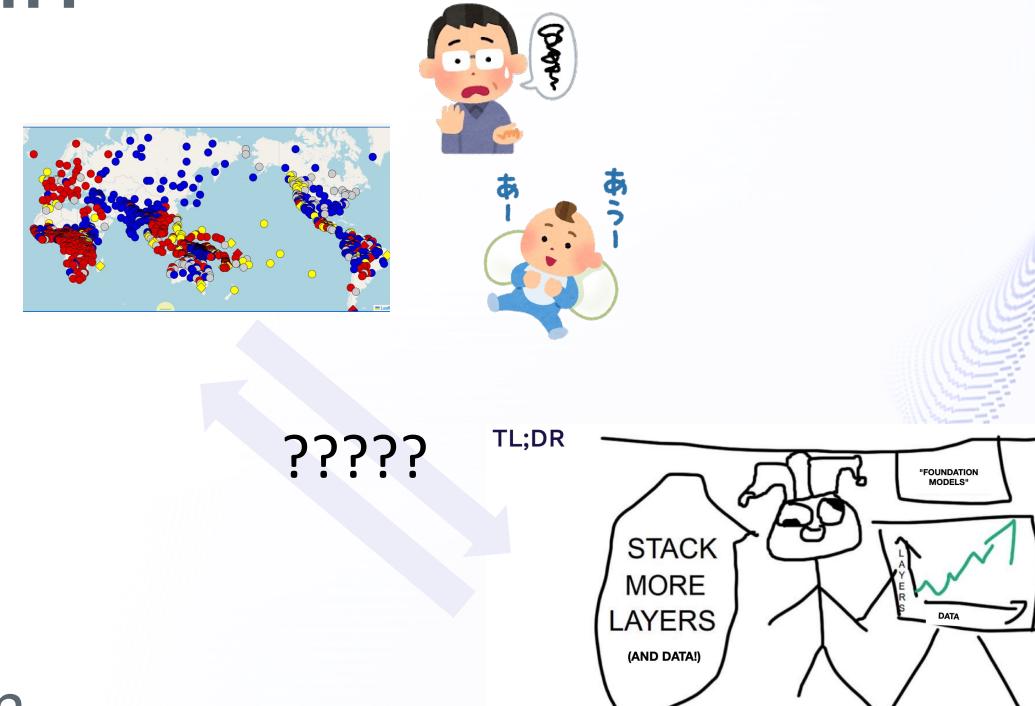
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- We humans somehow found one way to build a model that behaves like humans
- Some linguists criticize that this is not the model that linguistics has pursued



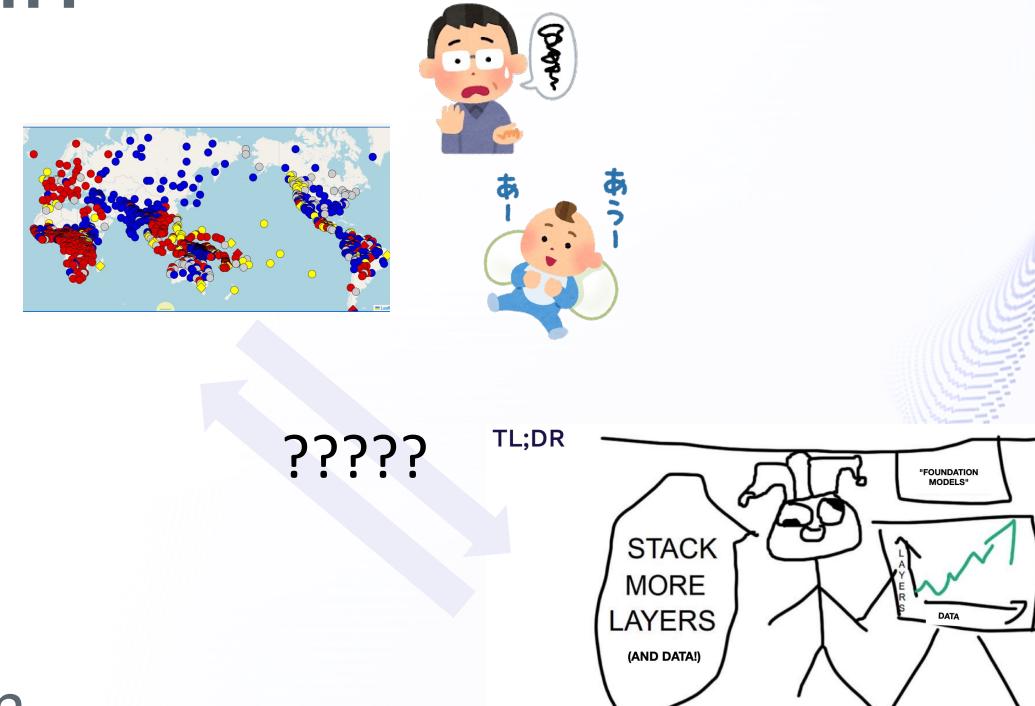
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- We humans somehow found one way to build a model that behaves like humans
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- But we do not know other things that can learn human language as far as we know (in fact, it's seemingly working the best)
 - That's why NVIDIA stock is sparkling



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If a model exactly simulates a phenomenon of interest, the model serves as a good hypothesis/explanation for the phenomena.

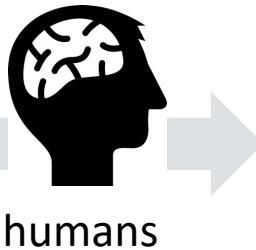
(aka. scientific modeling)

Cognitive modeling

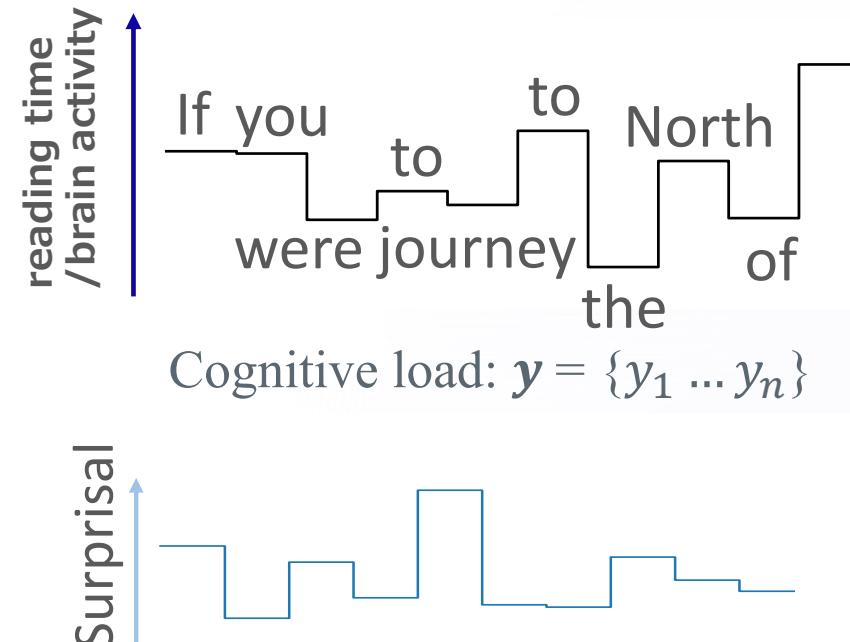
If you were to journey to the North of England, ...

Tokens: $w = \{w_1 \dots w_n\}$

If you were to journey to the North of England, ...



Surprisal: $\hat{y} = \{-\log_2 p(w_1|w_{<1}) \dots -\log_2 p(w_n|w_{<n})\}$



Unsupervised prediction*

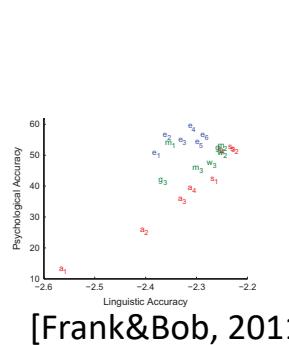
*training a regression model to rule out baseline factors and determine the coefficients, though

- The more unpredictable a word is, the more humans exhibit cognitive loads
 - The relationship should be logarithmic [Levy,08][Smith&Levy,13][Shain+,22]
 - Surprisal: $\text{Cost}(w_t) \propto -\log_2 p(w_t|w_{<t-1})$

Are we approaching to the model of humans? --- scaling law in cognitive modeling



LM-human correlations



Goodkind & Bicknell, 2018

???

Scaling up neural LMs

① motivation

20 min.

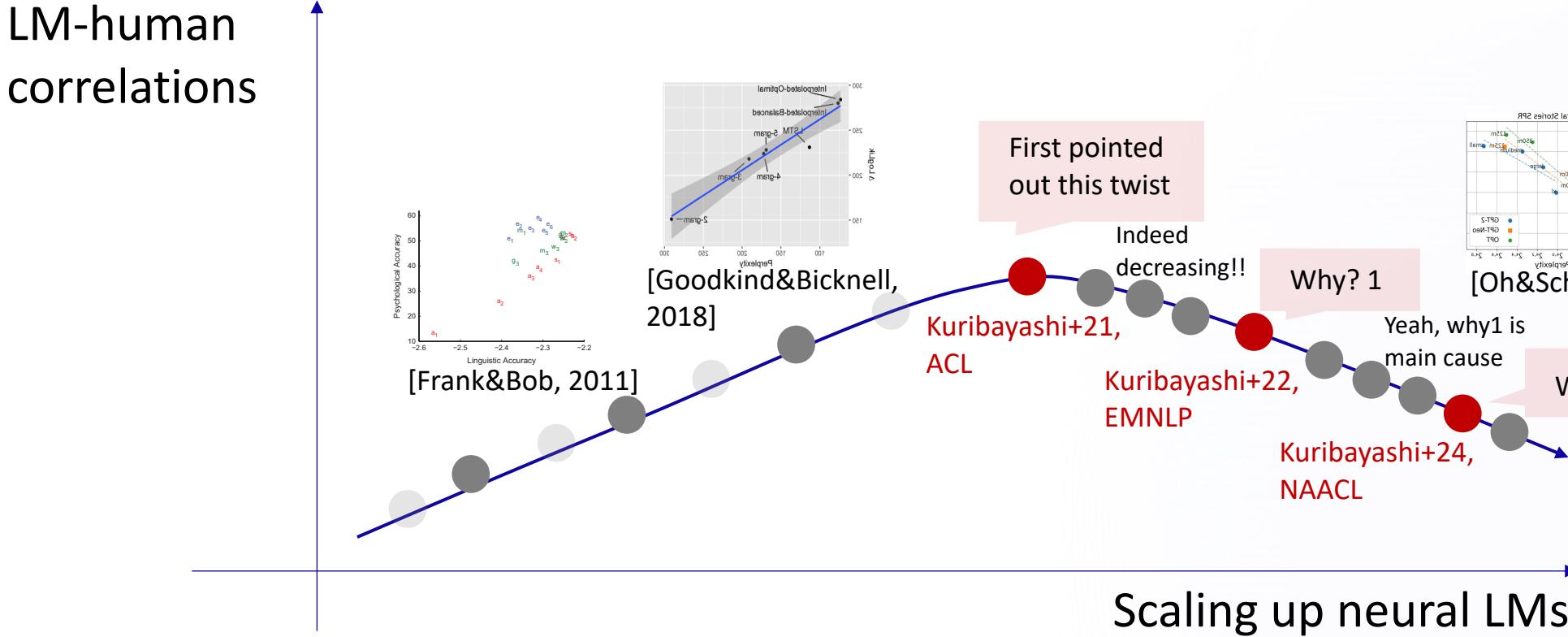
② my main research directions

20 min.

③ future directions

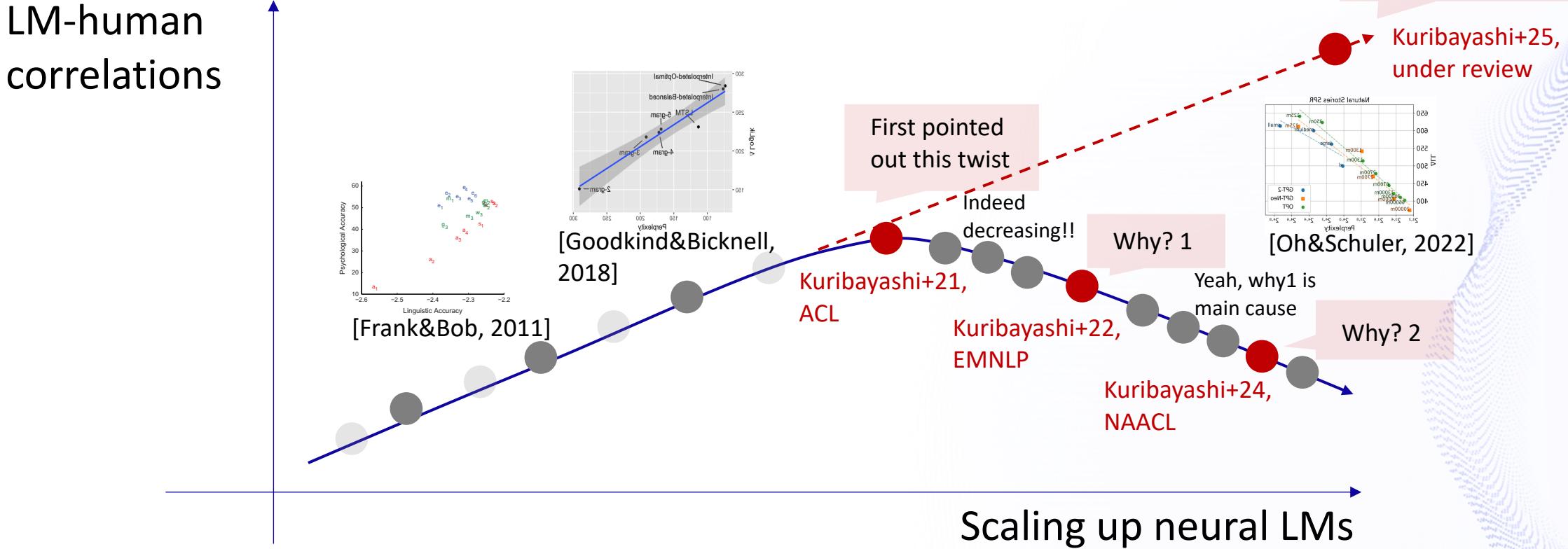
5 min.

Are we approaching to the model of humans? --- scaling law in cognitive modeling



Are we approaching to the model of humans? --- scaling law in cognitive modeling

From a bit different view, we are approaching!



Kuribayashi+21 (ACL)

Lower Perplexity is Not Always Human-Like

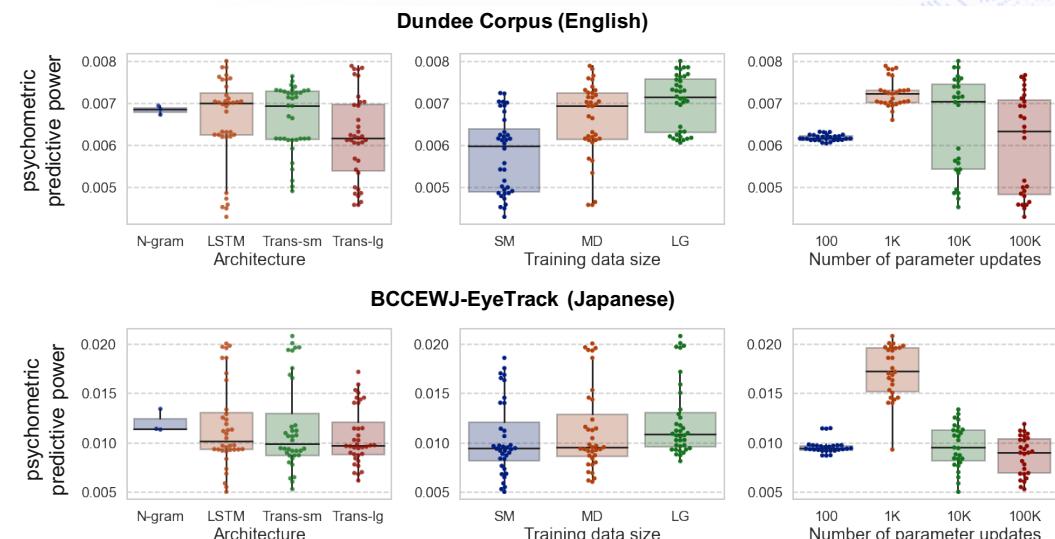
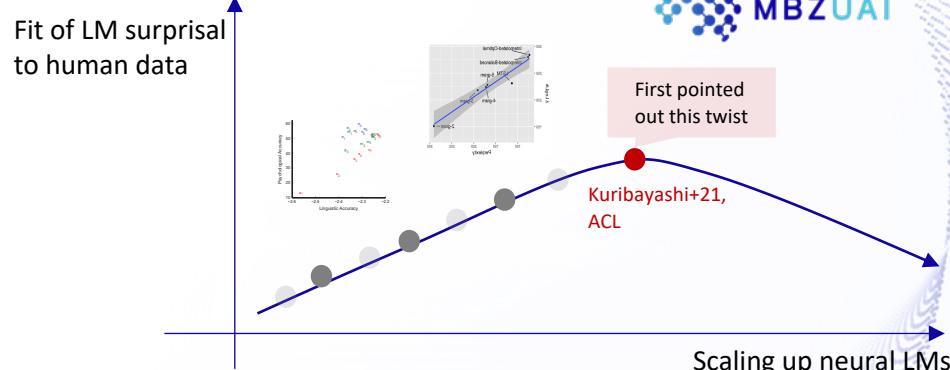
Tatsuki Kuribayashi^{1,2}, Yohei Oseki^{3,4}, Takumi Ito^{1,2},
Ryo Yoshida³, Masayuki Asahara⁵, Kentaro Inui^{1,4}

¹Tohoku University ²Langsmith Inc. ³University of Tokyo ⁴RIKEN ⁵NINJAL

{kuribayashi, takumi.ito.c4, inui}@tohoku.ac.jp ,
{oseki, yoshiryo0617}@g.ecc.u-tokyo.ac.jp , masayu-a@nijal.ac.jp

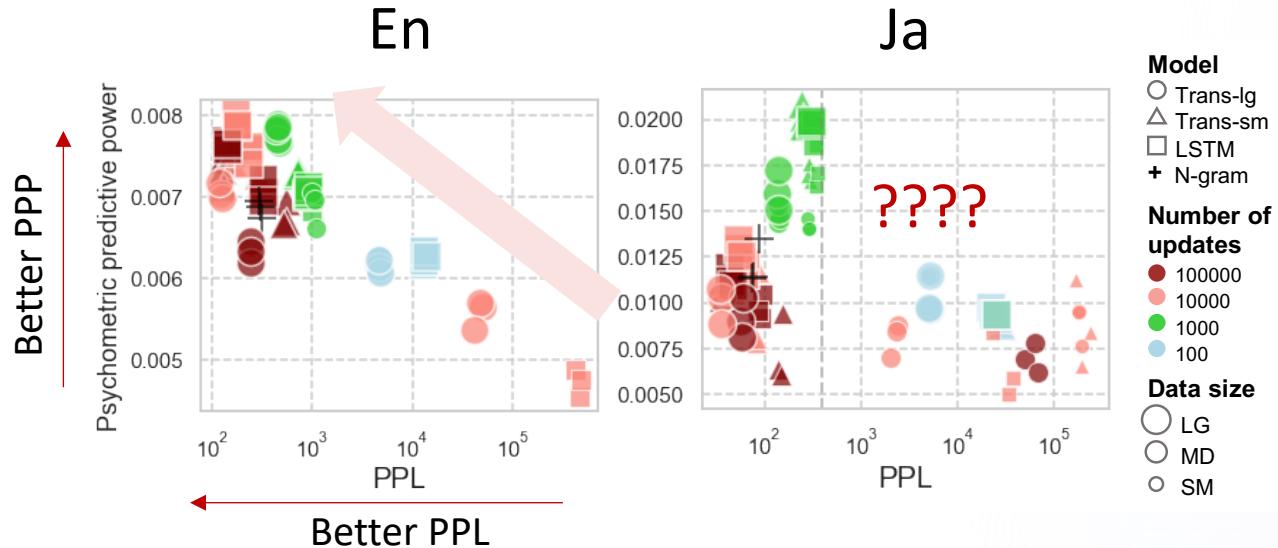
- First systematic, cross-linguistic evaluation of psychometric predictive power (PPP) of surprisal from neural LMs

$$\text{ReadingTime}(w_t) \propto -\log_{\theta} p(w_t | w_{<t})$$

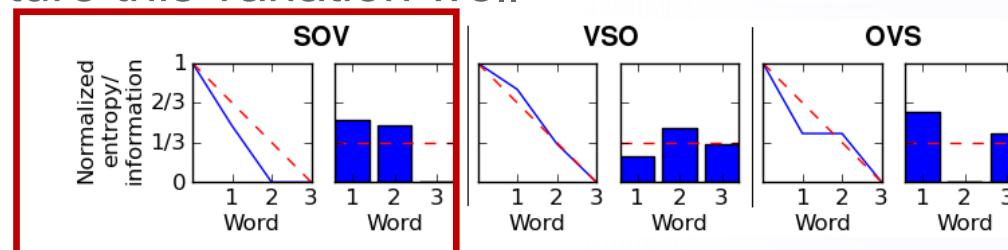


Kuribayashi+21 (ACL)

- Previously reported monotonic relationship between LM scaling and PPP was fragile



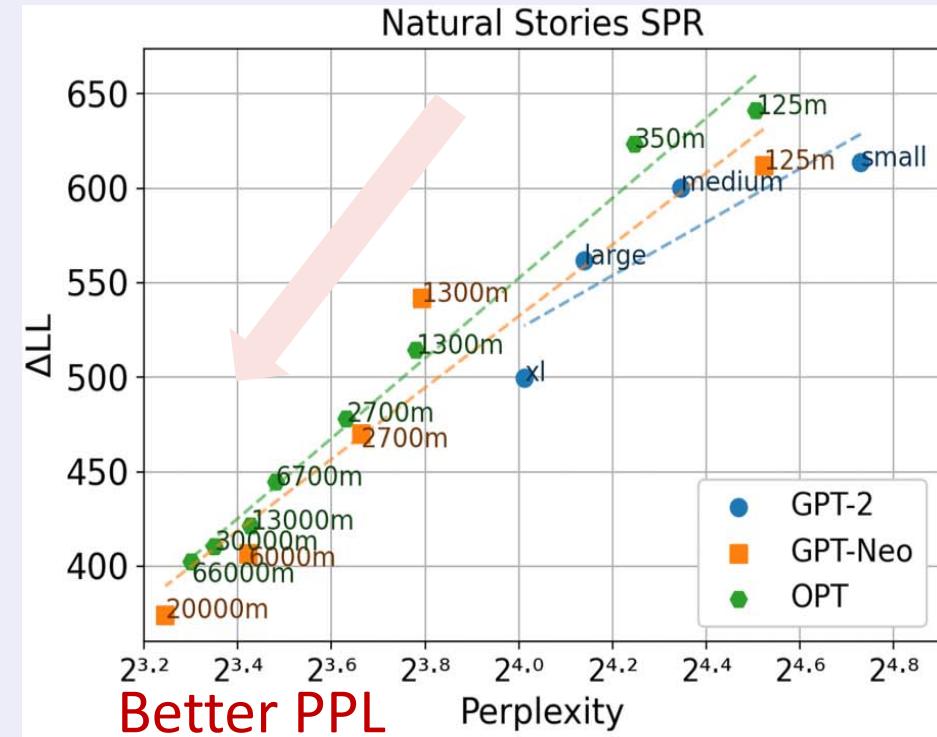
- Just changing the language (En->Ja) breaks it, empirically
 - Reading times and surprisals in the Japanese language (Subject-Object-Verb; SOV word order) have a large intra-sentential variance (i.e., low uniform information density), and LM-surprisal could not capture this variation well



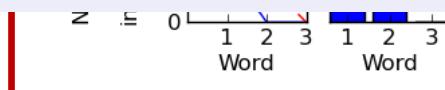
[Maurits+, 2010]

Kuribayashi+21 (ACL)

- With larger models, the **negative scaling effect** appeared even in the English language.
 - We could not observe it in our ACL 2021 work since we used in-house smaller LMs



[Oh&Schuler, 2022]

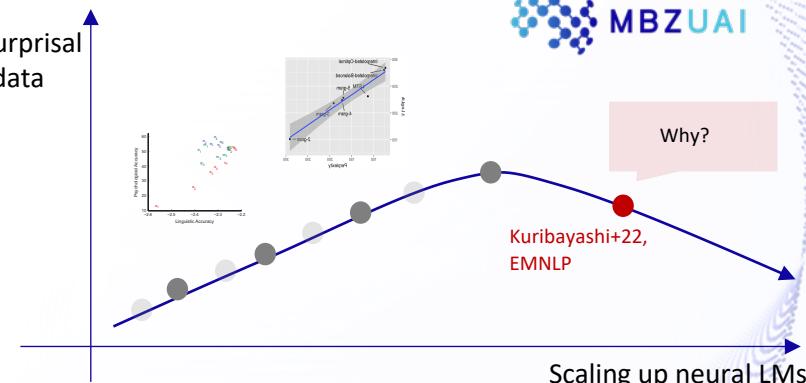


[Maurits+, 2010]

Kuribayashi+22 (EMNLP)



Fit of LM surprisal
to human data



Context Limitations Make Neural Language Models More Human-Like

Tatsuki Kuribayashi^{1,2} Yohei Oseki^{3,4} Ana Brassard^{1,4} Kentaro Inui^{1,4}

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- Why did LMs' prediction deviate from humans?
- LMs (Transformers w/ self-attention) may be too good to consider wide contexts, compared to human real-time language processing

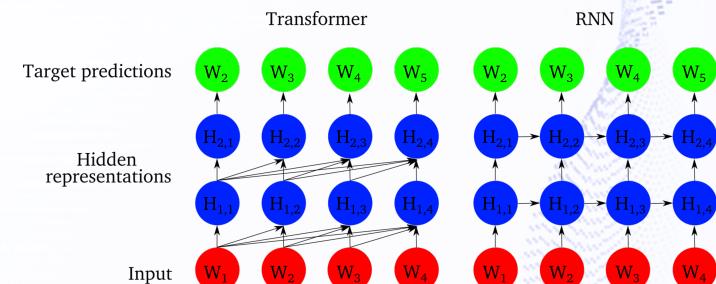


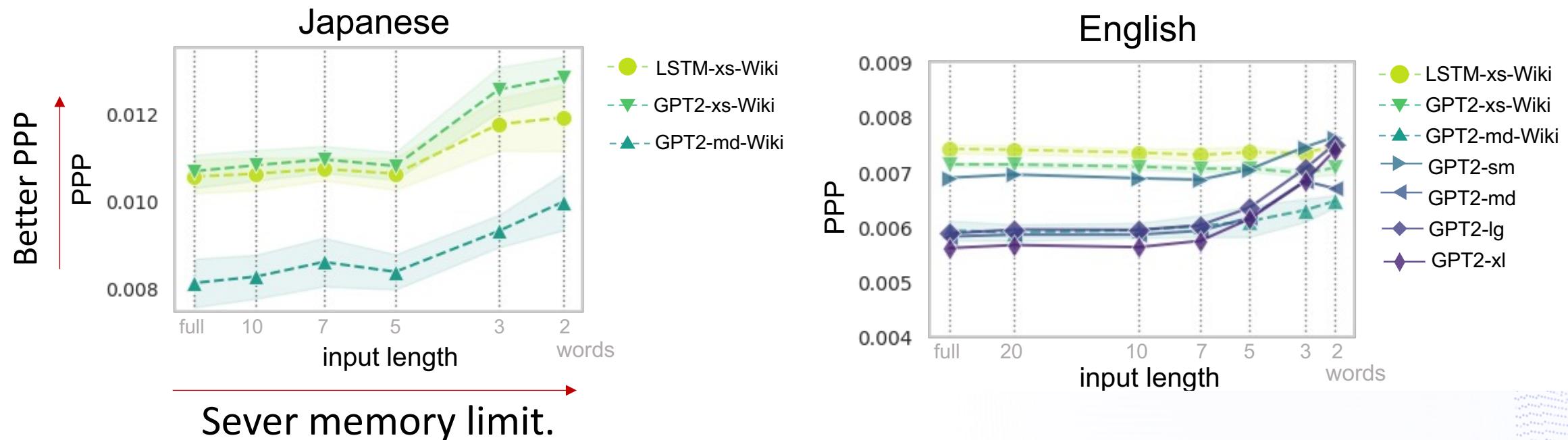
Figure 1: Comparison of sequential information flow through the Transformer and RNN, trained on next-word prediction.
[Merkx&Frank, 21]

Kuribayashi+22 (EMNLP)

- Limiting LMs memory capacity aligns with humans

- simple erasure of distant contexts surprisingly works well

$$\text{ReadingTime}(w_t) \propto -\log_2 p(w_t | w_0, w_1, \dots, w_{t-2}, w_{t-1})$$



Kuribayashi+24 (NAACL)

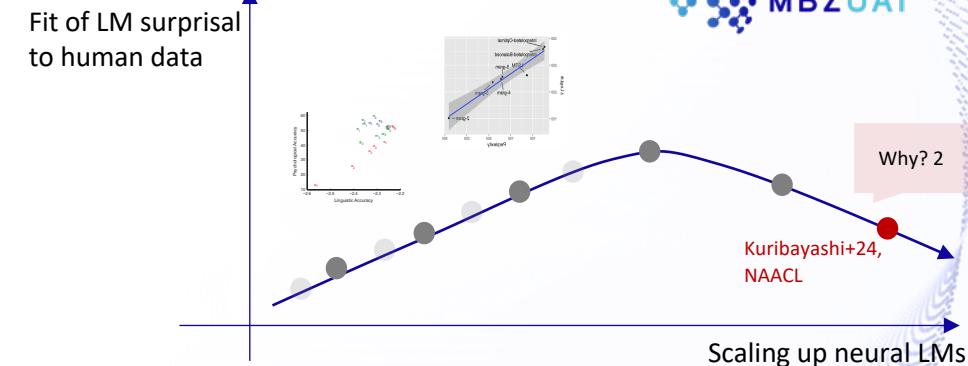


Psychometric Predictive Power of Large Language Models

Tatsuki Kuribayashi¹ Yohei Oseki² Timothy Baldwin^{1,3}

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oseki@g.ecc.u-tokyo.ac.jp



- Instruction-tuning and/or meta-linguistic prompting (“*Let’s predict language processing cost!*”) did not improve PPP
- **Vanilla surprisal from base LMs (w/o tuning) predicts human data the best**
 - Human real-time processing seem to be simply tuned to statistics of next-word probability

Kuribayashi+25 (under review)



Large Language Models Are Human-Like Internally

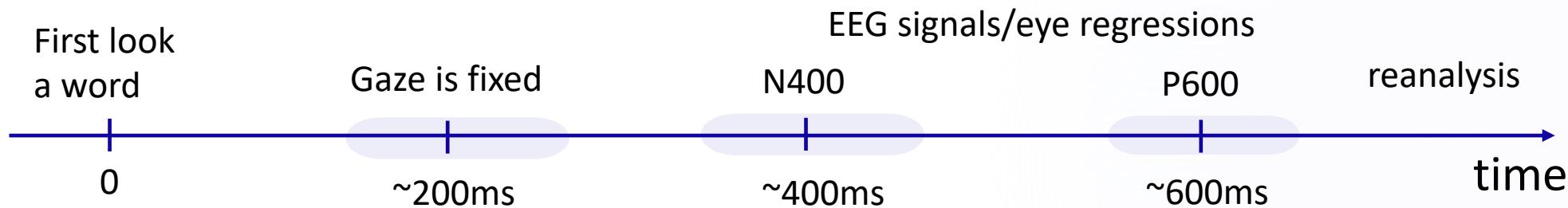
Tatsuki Kuribayashi¹ Yohei Oseki² Souhaib Ben Taieb^{1,3}

Kentaro Inui^{1,4,5} Timothy Baldwin^{1,6}

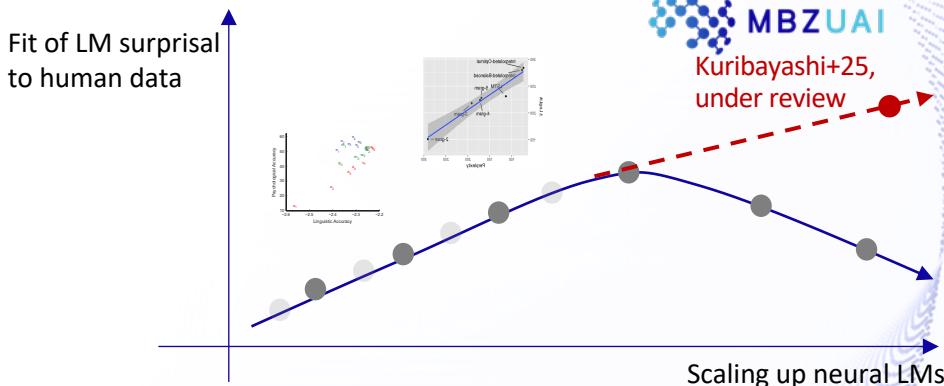
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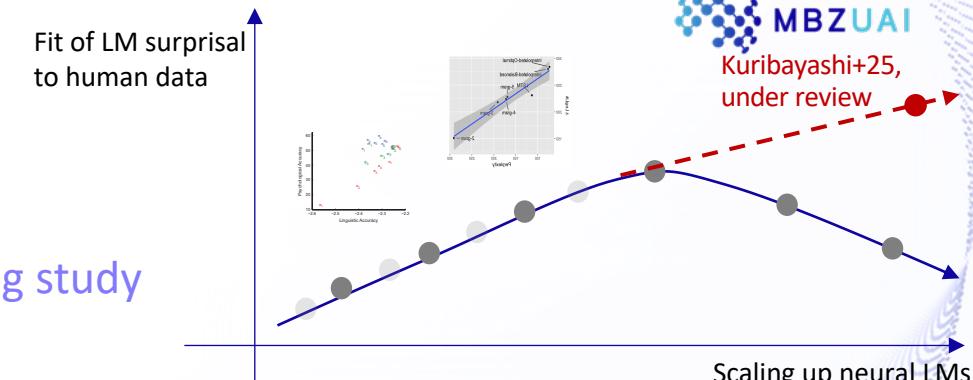
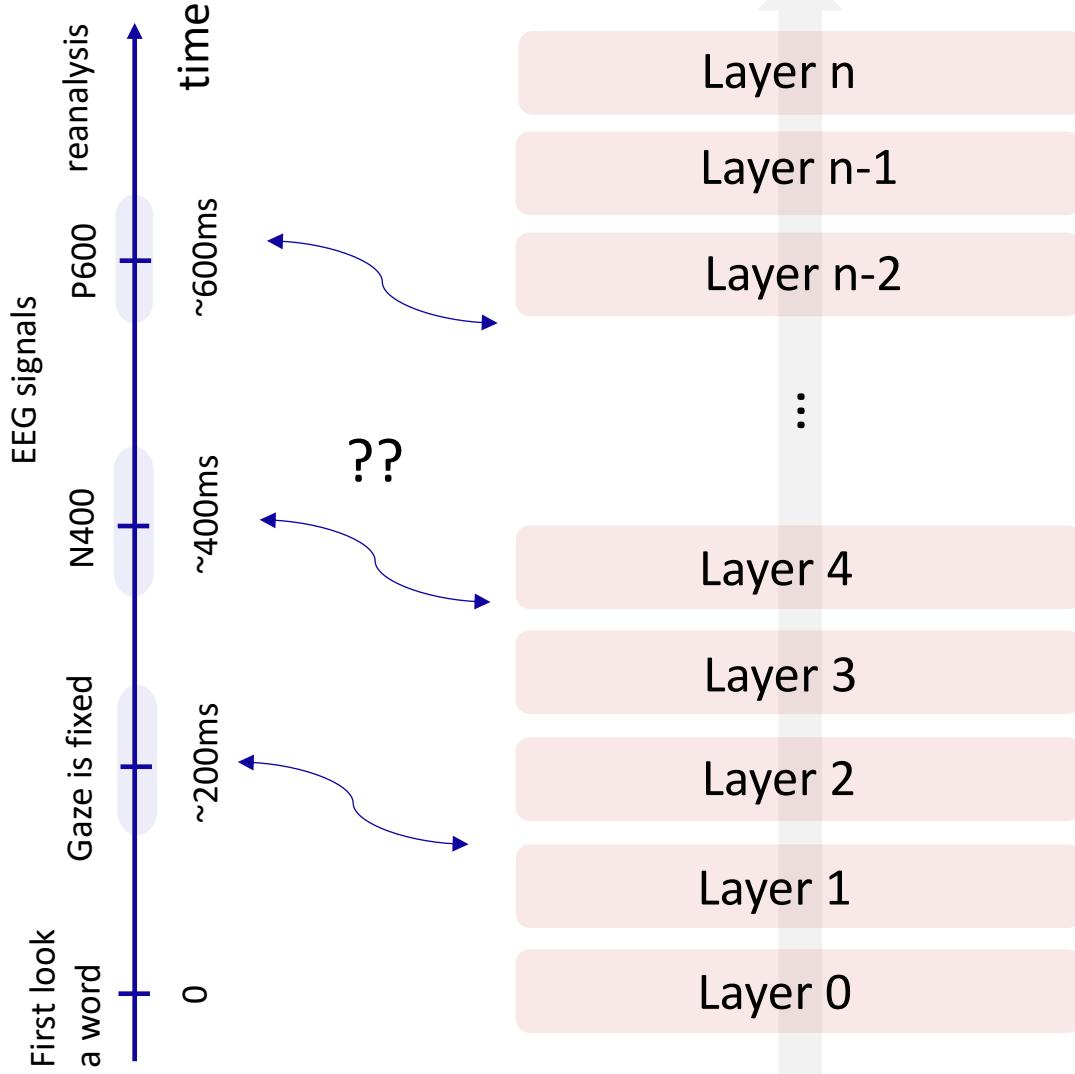


- Gaze duration is very fast (~200ms/word) and may reflect early-stage processing of language.
 - Where should such fast processing be realized in LLMs?

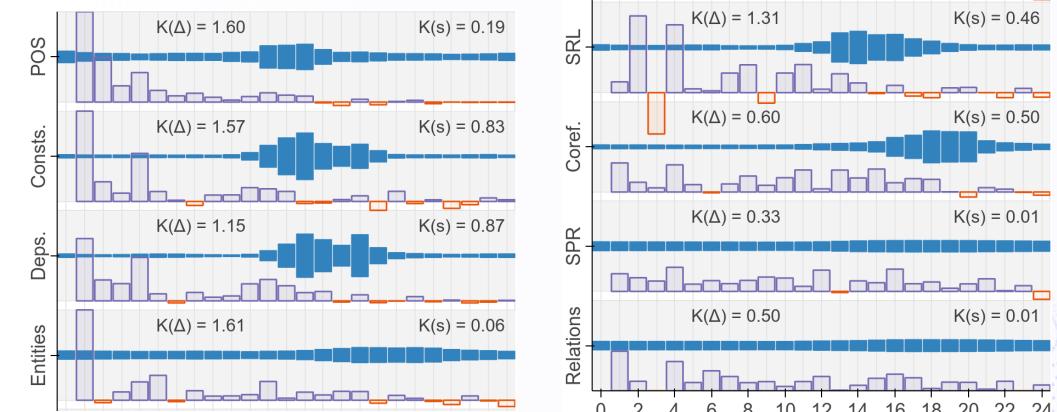


Kuribayashi+25 (under review)

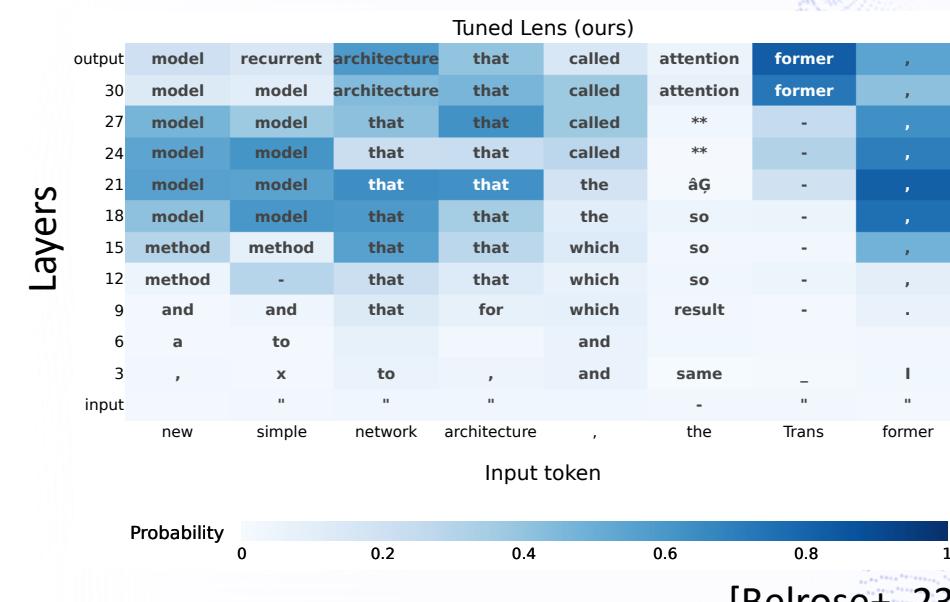
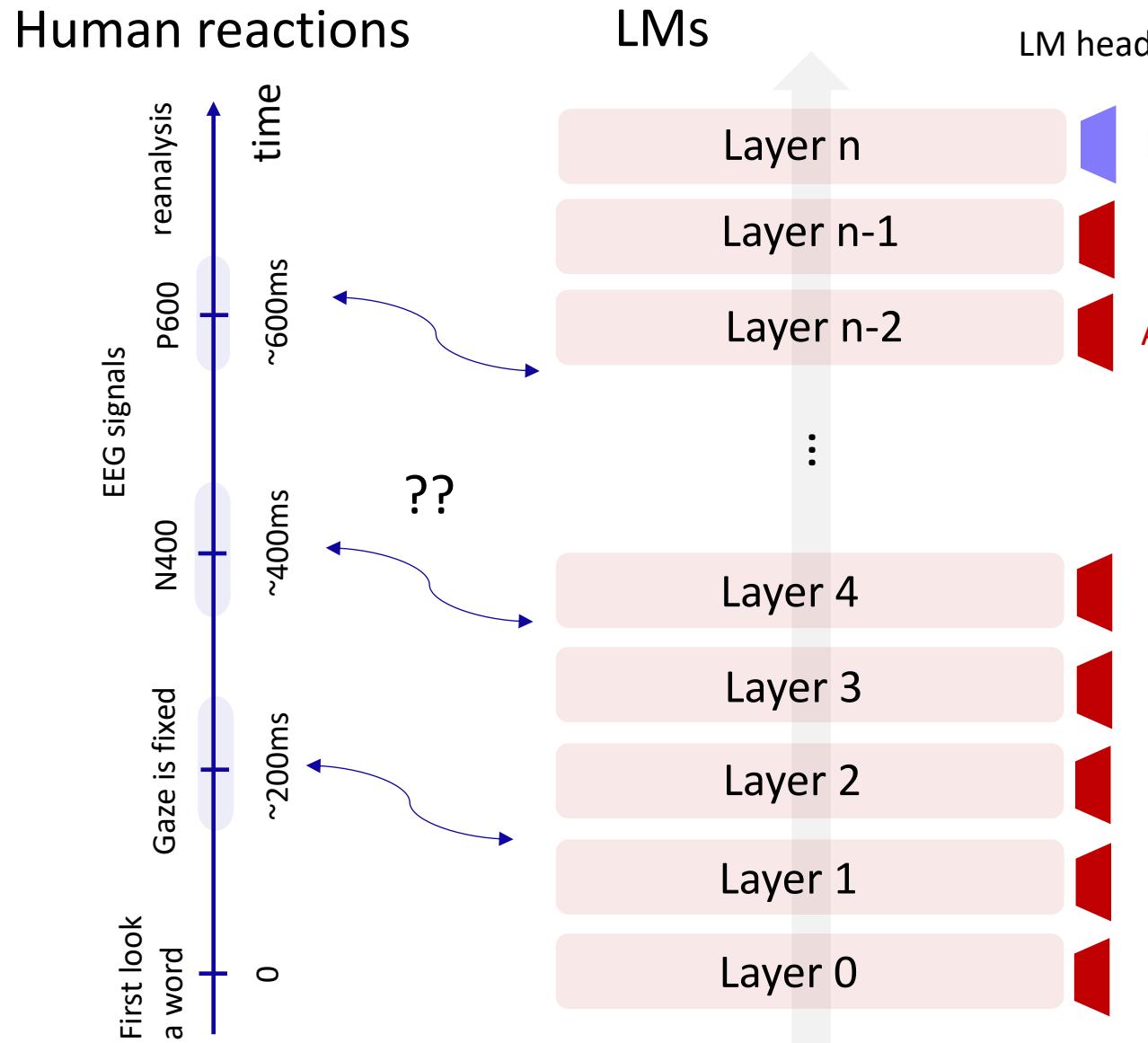
Human reactions



Ian Tenney¹ Dipanjan Das¹ Ellie Pavlick^{1,2}
¹Google Research ²Brown University
{iftenney, dipanjand, epavlick}@google.com

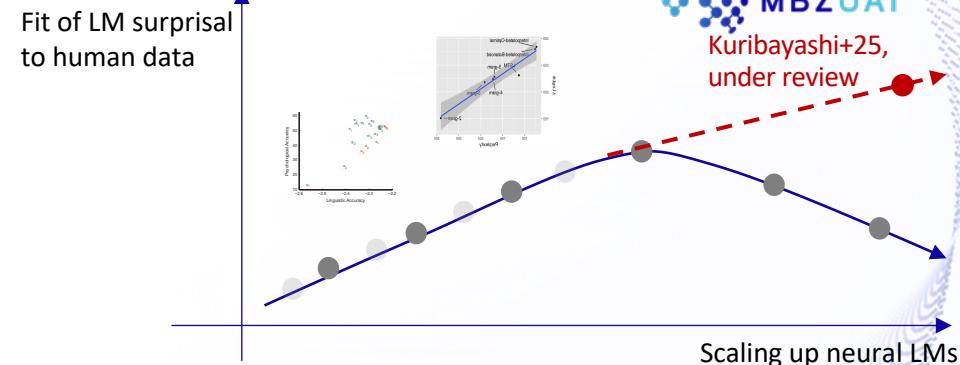
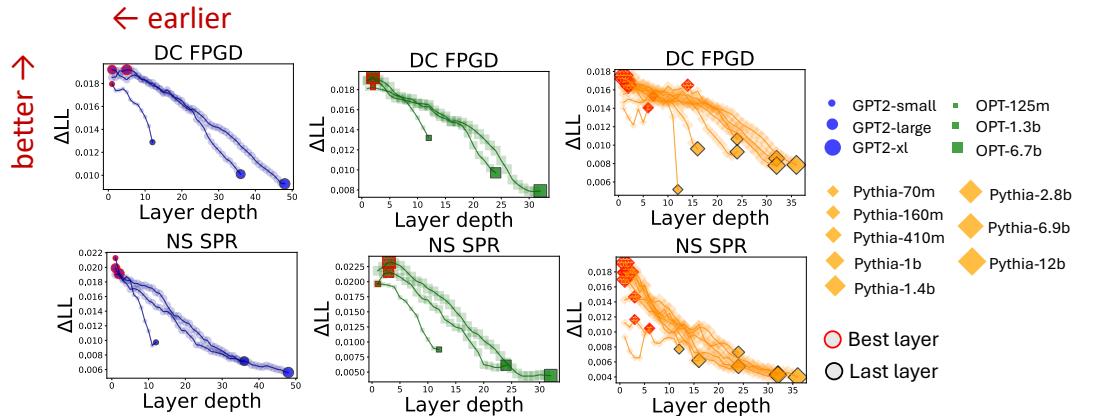


Kuribayashi+25 (under review)



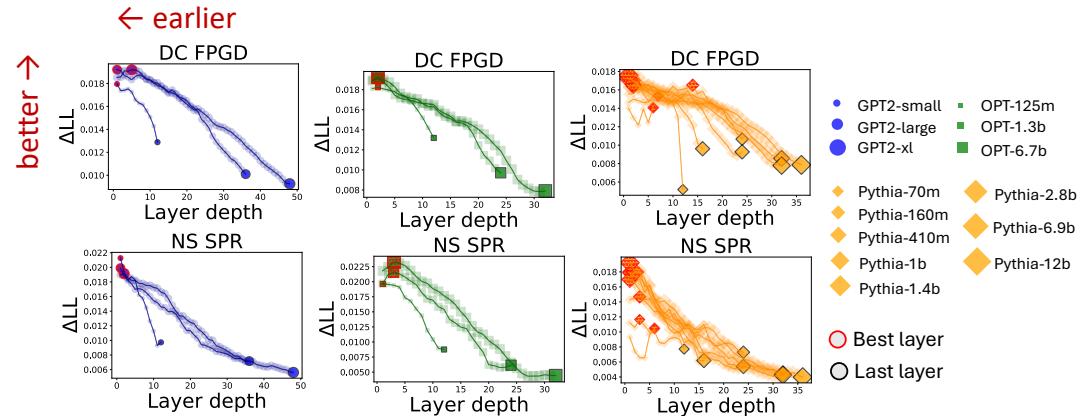
Kuribayashi+25 (under review)

- (Fast) first pass gaze durations are better predicted in earlier layers

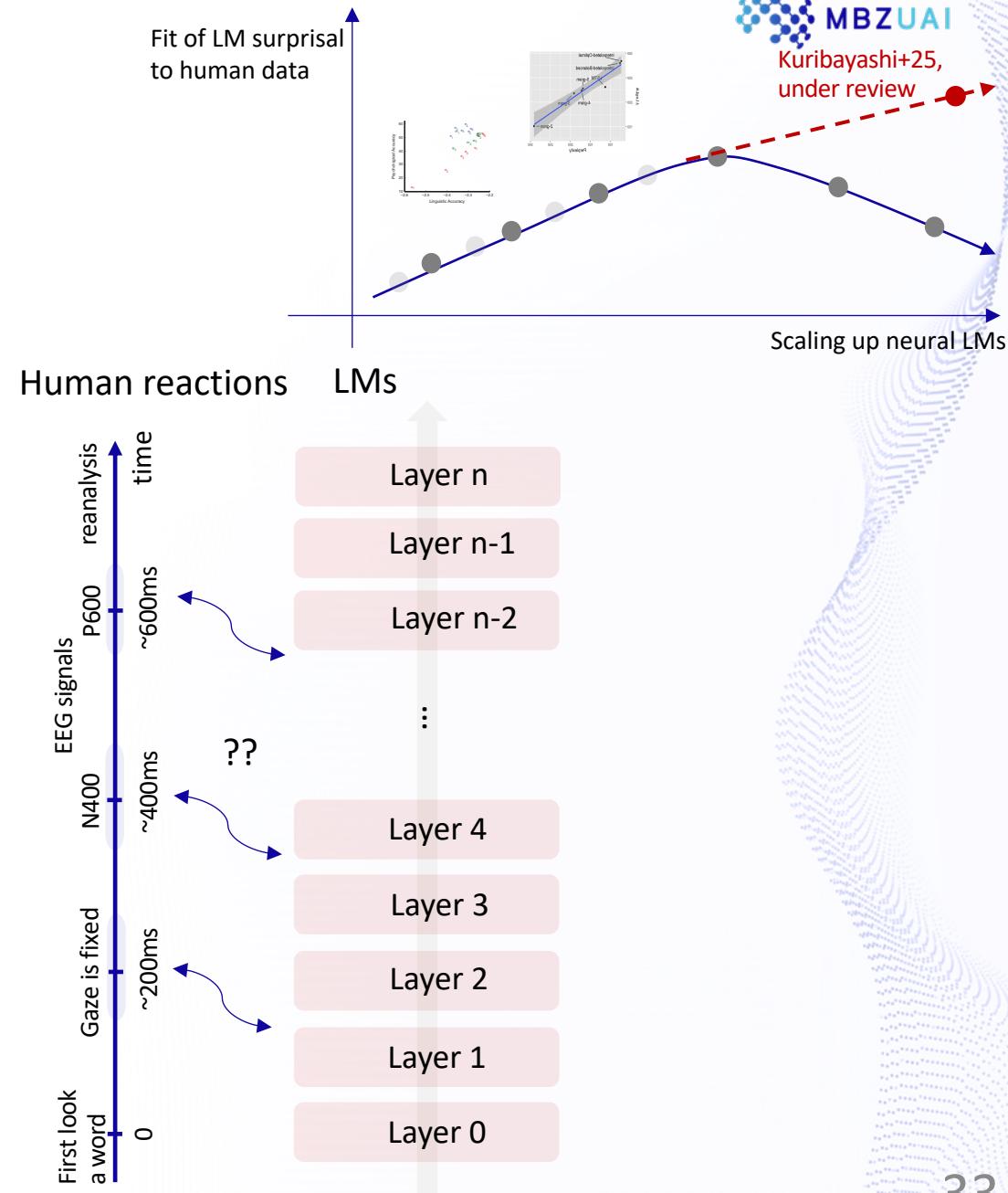
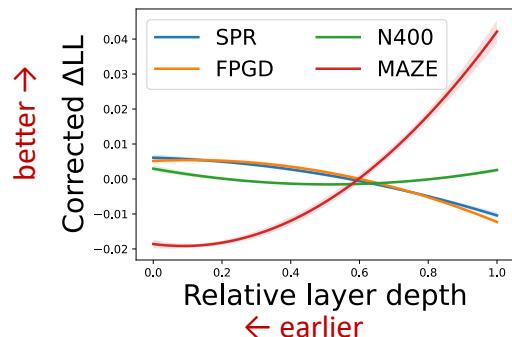


Kuribayashi+25 (under review)

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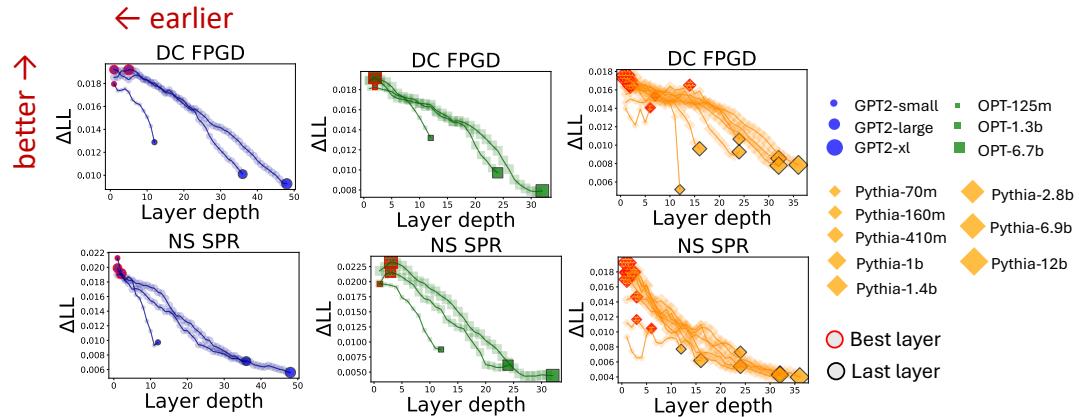


- Slower measures (N400, MAZE) tend to be better predicted in later layers

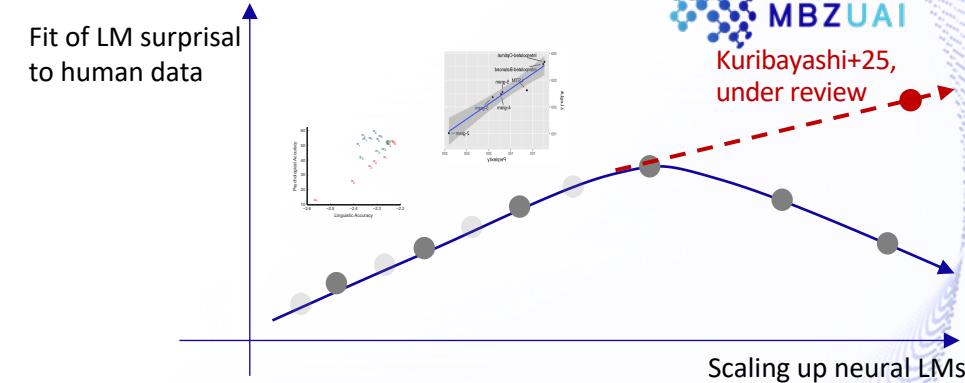
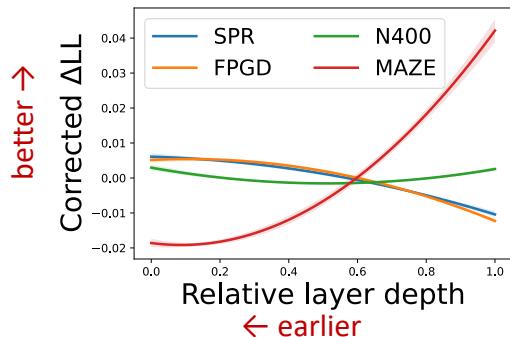


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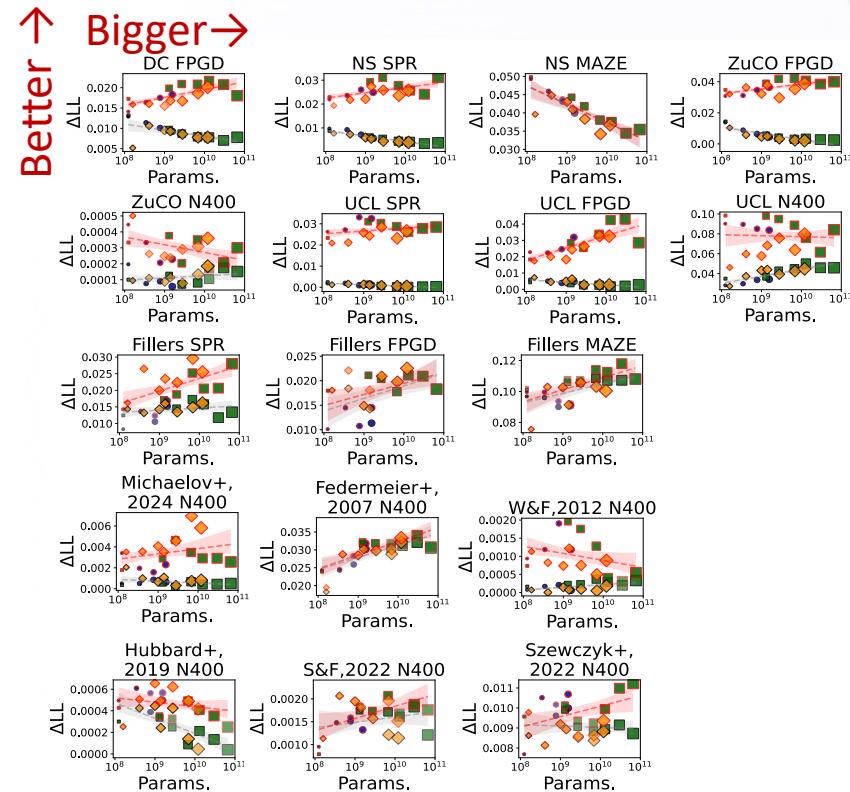
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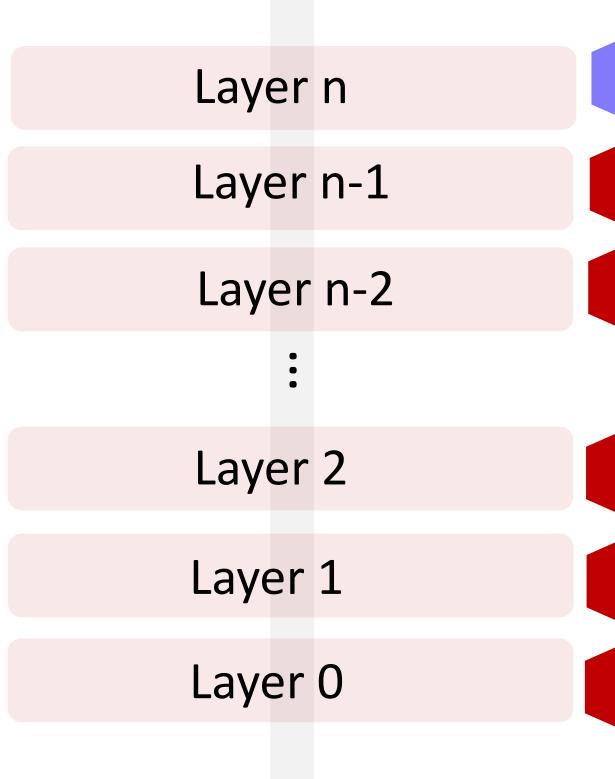
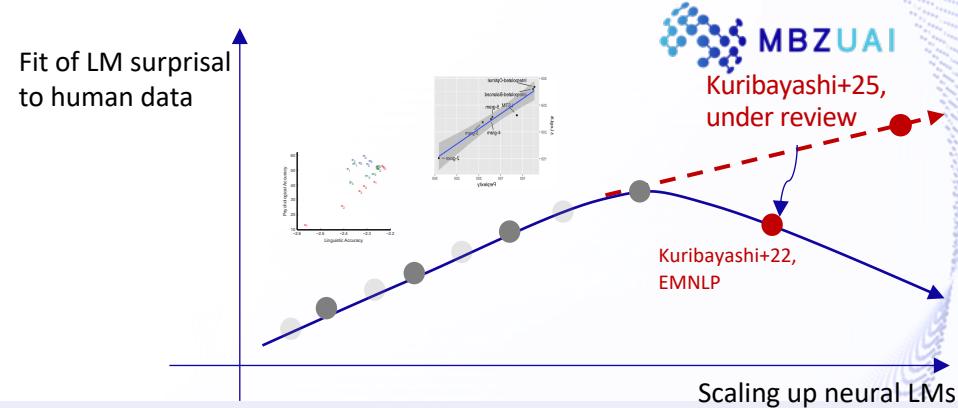
- Early layers in LLMs are human-like



(To be continued...)

RE: Kuribayashi+22 (EMNLP)

- Moderately-contextualized, human-like surprisal from LLMs



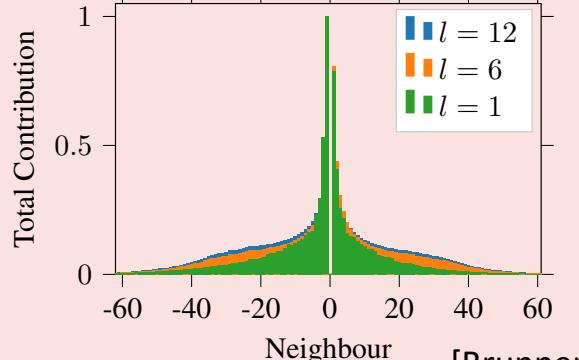
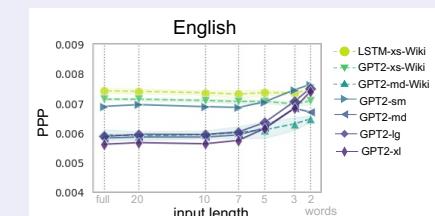
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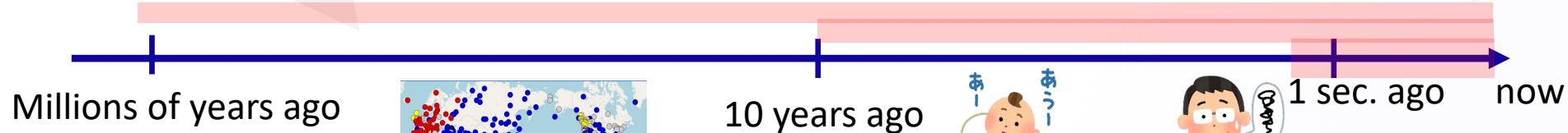


Earlier layers are less-contextualized

Fundamental linguistic problems

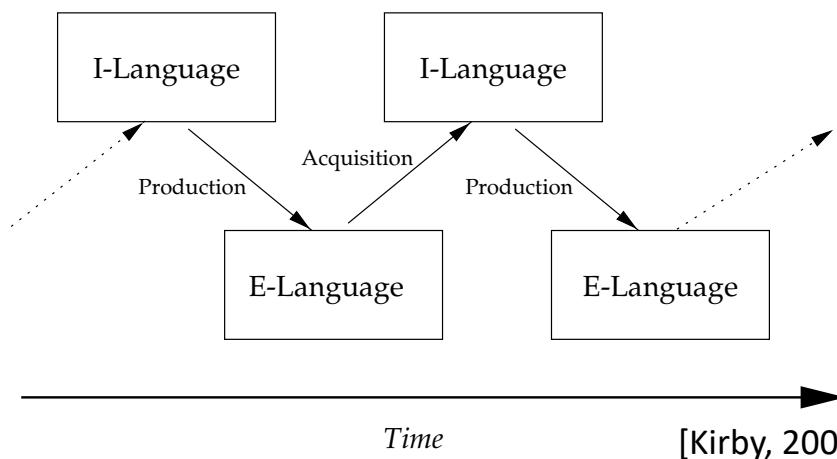
- Why do natural languages have typological universals, e.g., subject<object?

Language easy to process
would have survived



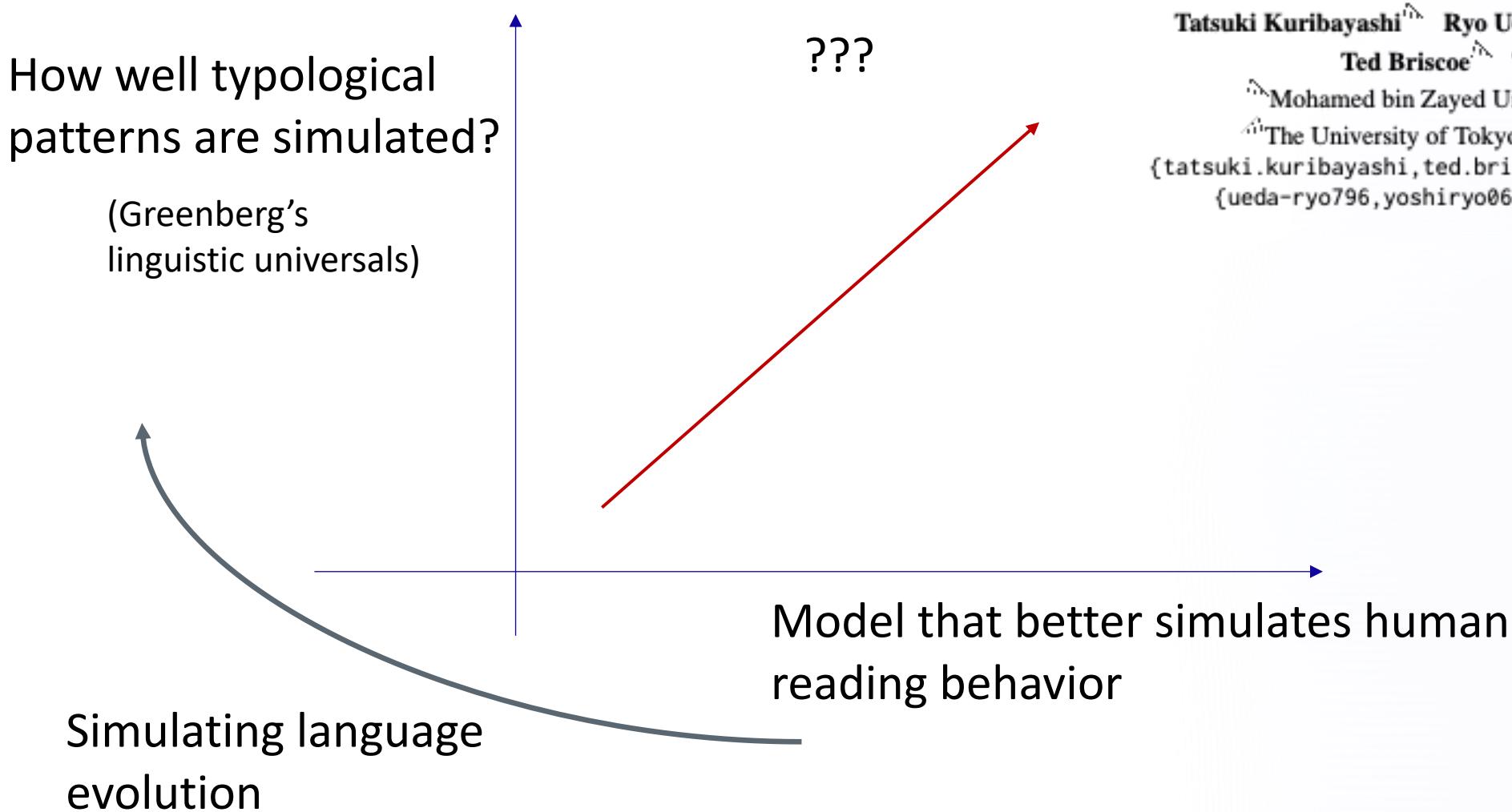
- What are humans computing during real-time language processing?

- How can humans acquire language?



From cognitive modeling to language universals

Kuribayashi+24 (ACL)



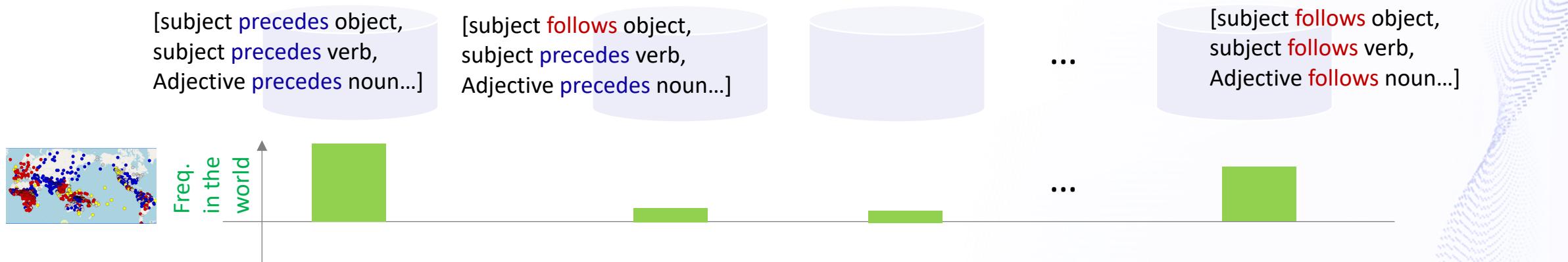
Emergent Word Order Universals from Cognitively-Motivated Language Models

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From cognitive modeling to language universals

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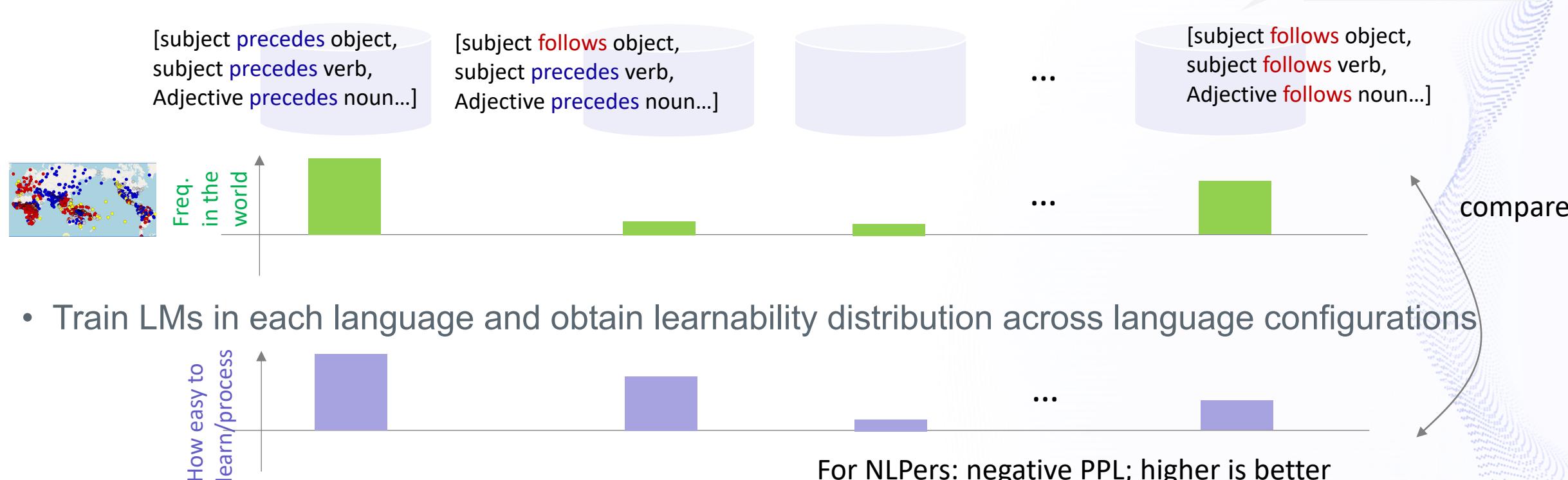
- A problem to predict the *plausibility* of language design, based on their learnability and processing difficulty for LMs.



From cognitive modeling to language universals

Kuribayashi+24 (ACL)

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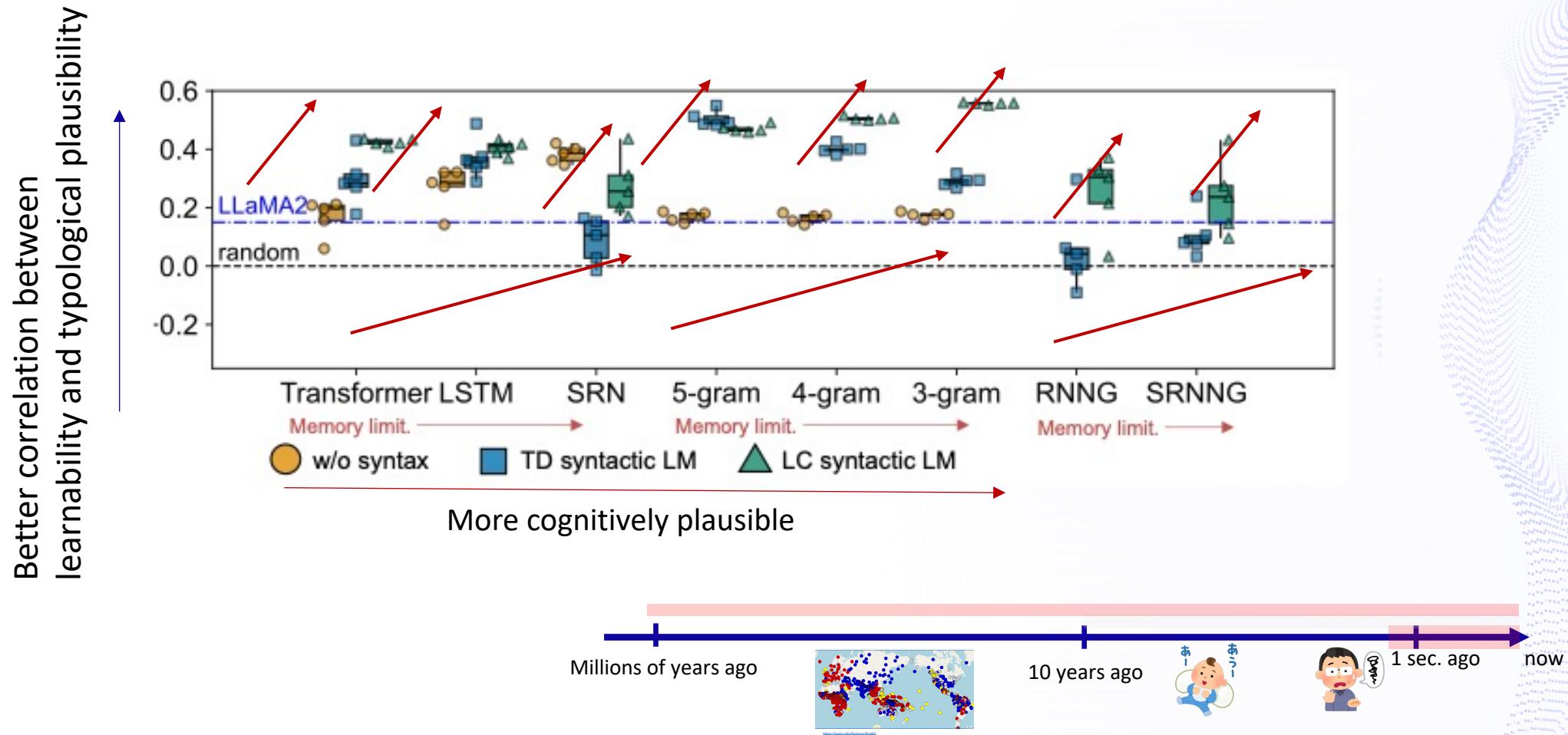


- Train LMs in each language and obtain learnability distribution across language configurations
- Which language is easier to learn for particular LMs?
 - Human-like LMs: memory limitation, syntax-aware, cognitively-plausible left-corner traversals

From cognitive modeling to language universals

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- Learning/processing difficulties of LMs are better correlated with typological distributions when using more cognitively-motivated LMs



Roadmap

① motivation

20 min.

② my main research directions

20 min.

③ future directions

5 min.

Future: Emergent corpus

- LLMs are good at learning language, **if there is a corpus**

- Language transmission in one generation

- Then, where is corpus from?

- Humans have achieved LLM-like behaviors from a situation w/o corpus, in the long history on Earth
 - Connection to emergent language/communication/symbols
 - must be handled via computational simulation (computational linguistics!)

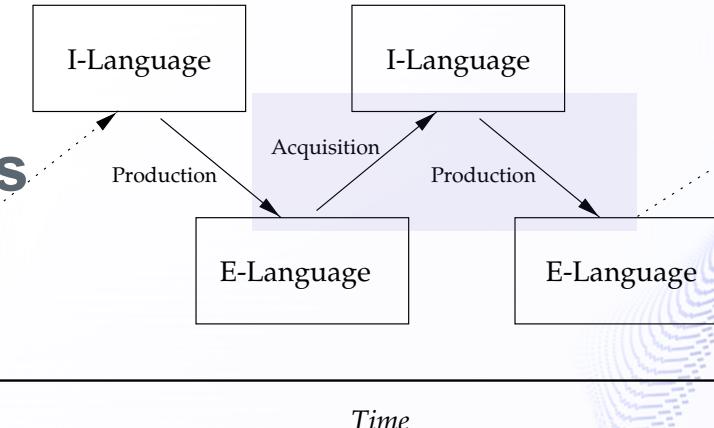
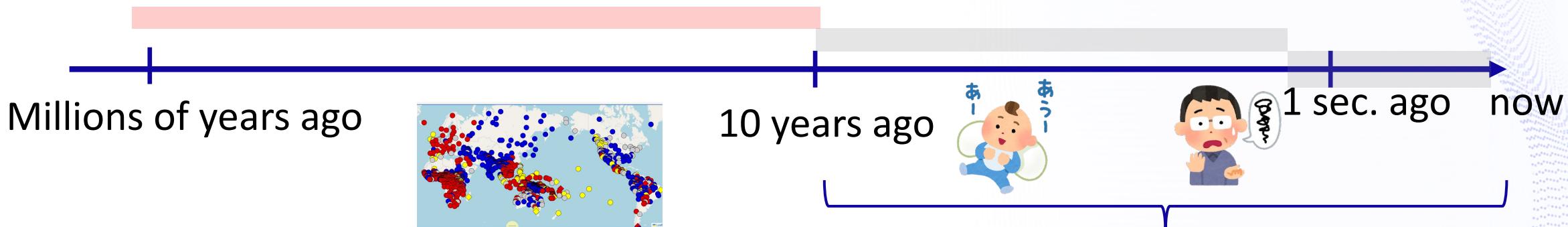


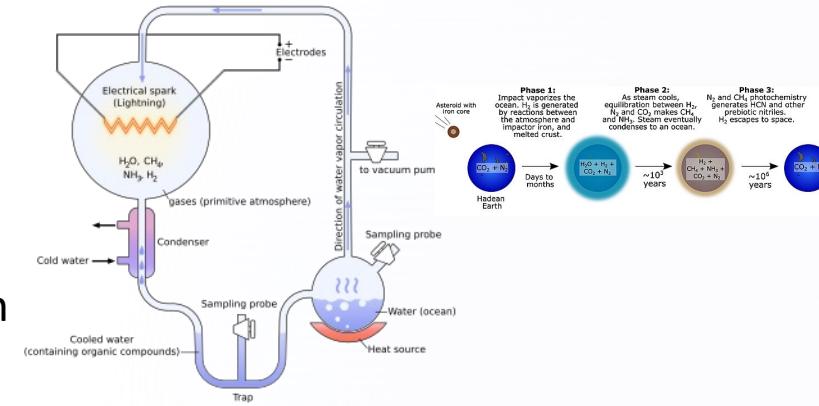
figure 6.1. The transmission of language over time.
[Kirby, 2002]

Future: Connections to Robotics

- Text-only NLP alone can not explore language emergence under text-less environments
 - Agents should play real, physical games to explore the emergence of language



Build a model →
 ← Test the fit/prediction



- If we can train LMs (robots) under the same scenarios as humans, and if they acquire language in the same way as humans, what does this imply?

Future: How should we measure human-likeness of LLMs?

- Humanities studies as checklists



- What is minimum criteria to explain empirical linguistic observations?

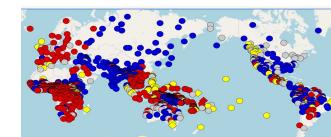
• Can neural agents re-invent human language?

• Can LMs mimic human-like language acquisition patterns?

• Can LMs simulate human real-time language processing?

• Do LLMs have linguistic knowledge?

Millions of years ago



10 years ago



1 sec. ago



now

Future: Maintaining the community

- ~90% of (young) NLP researchers may be thinking about LLMs and chatbot
 - It may be rational, considering the current trend/economy, instead of exploring niche topics
 - How can community think more freely about diverse things or how can I encourage such activities? (I also sometimes feel a sense of isolation in the community)
 - The microwave oven was invented thanks to a person who happened to notice a melted chocolate in radar research.
- How to appeal the excitement of exploring scientific (humanities) questions?
- Isn't it only natural that we want to know about humans because we are humans?
- AI is not only for the science of artificial intelligence but also for any science using artificial intelligence