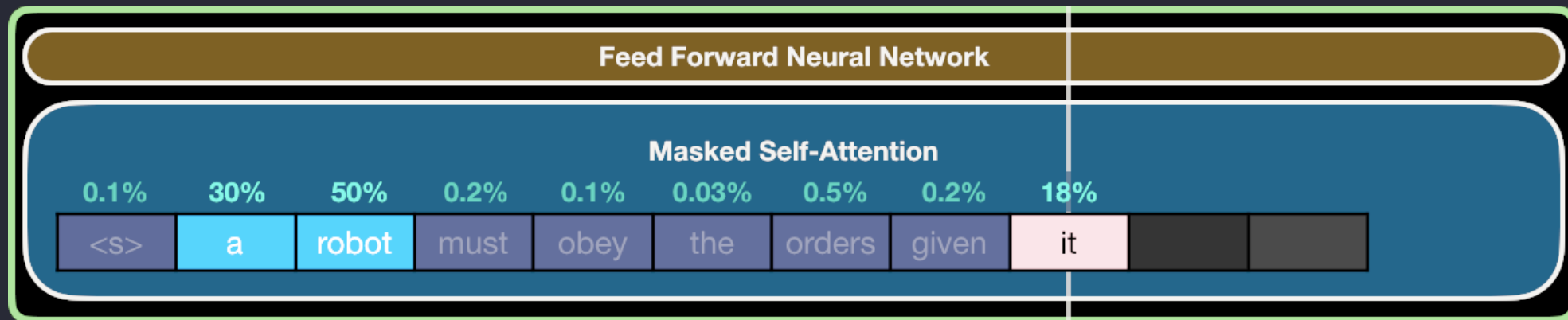


LLM Surprisal ~ Human Reading Time

Large Language Models

- Task: Given words so far, predict next word
 - e.g. prompt: *"The cat eats the "*, prediction: *"mouse"*
- State of the art: Attention mechanism^{Vaswani et al. 2017}
 - Idea: Mix in *relevant* context tokens
 - Include large context without forgetfulness



LLM Surprisal

- Normally: Most likely next word?
- Instead: How likely is *this* next word?
 - I.e. $P(t_n | t_{n-1}, \dots, t_0)$
- *Surprisal* $:= -\log_2 P(t_n | t_{n-1}, \dots, t_0)$
 - High surprisal \Leftrightarrow Low probability
 - Between 0 and ∞

The	cat	eats	the	mouse.
-	8.8	7.2	1.7	4.1

The	cat	eats	the	car.
-	8.8	7.2	1.7	15.1

Human Reading Time

- How much time is spent per word while reading?
- Different metrics: *First Pass*, *Go Past*, etc.
- Generally: Predictable word ~ short reading time
 - We read over words that are predictable
 - We get stuck at words that are unexpected
- Measured in different ways, e.g. eye tracking

Since Ralf runs a mile seems like a short distance to him.

The diagram shows the sentence "Since Ralf runs a mile seems like a short distance to him." with blue circles around each word, representing eye fixations. The word "seems" is highlighted with an orange background, indicating a point of interest or a longer reading time. The circles are of varying sizes and positions, suggesting the duration and movement of the reader's eyes across the text.

Local Syntactic Coherences

The coach chided the player tossed the frisbee by the opposing team.

The coach chided the player tossed the frisbee **by** the opposing team.

- Conflict between local and global parsing^{Tabor et al. 2004}
- Increased reading time at subsequent word
- Not just syntactical, influenced by context
 - Including visual & textual^{Konieczny et al. 2009, Müller et al. 2019}

Context influences LSC-Effect

Mit (**langweiligen Anekdoten | spannenden Geschichten**) überzieht der erste Redner sein Zeitlimit. Das Publikum hört ihm dabei (**gähnend | gespannt**) zu. Nach dem dreistündigen Vortrag hat er keine Energie mehr und übergibt an den nächsten Redner.

The first speaker exceeds his time limit with (**boring anecdotes | exciting stories**). The audience listens to him (**yawning | attentively**). After the three-hour talk, he has no more energy, and hands over to the next speaker.

Der nächste Redner ärgert sich über alle Maßen, als ihm der erste Redner **müde** das Publikum überlässt.

The next speaker is annoyed beyond measure, as the first speaker (**tiredly | tires**) leaves the audience to him.

OR

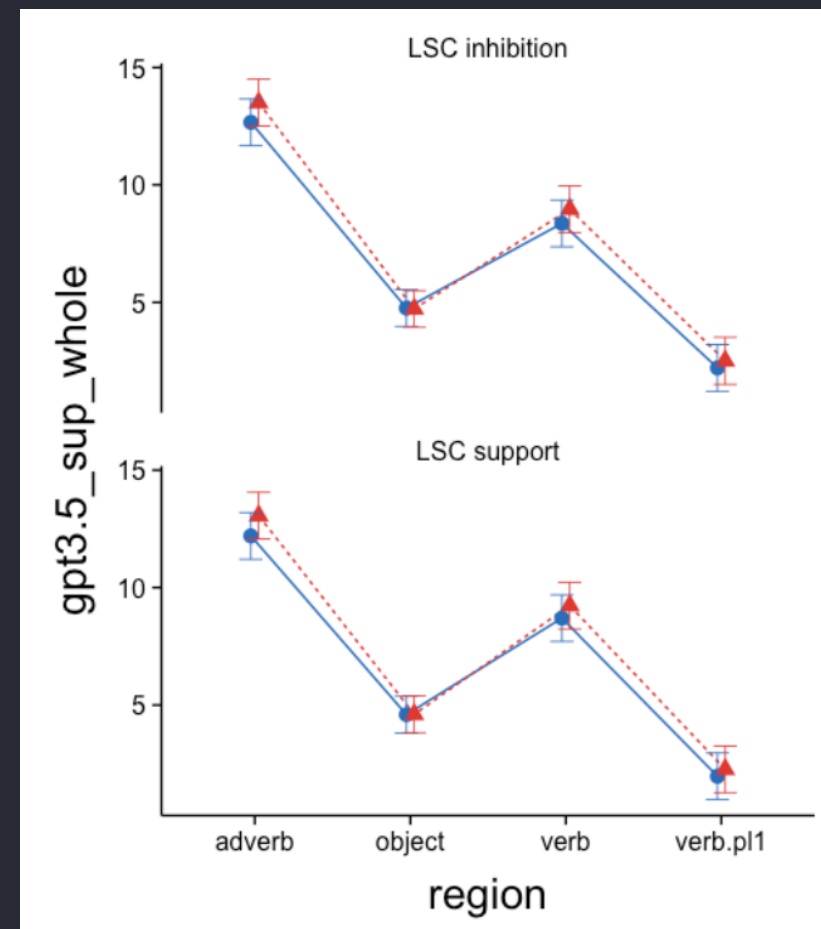
Der nächste Redner ärgert sich über alle Maßen, als ihm der erste Redner **ermüdet** das Publikum **überlässt**.

Surprisal ~ Reading Time ?

- Dataset: short German texts *from Müller et al. 2019 & Müller 2019*
- Reading times via eye tracking
- GPT-3.5 surprisals via OpenAI API
- Yes! ✨

LSC-Effect in NNs?

- In RNNs: Yes *Konieczny 2005, Konieczny et al. 2009*
- In Transformers: No ✨
 - LSCs only mildly affect surprisal
 - No modulation due to context
 - Explanation (?):
Attention mechanism has no
intrinsic local preference



Conclusion

- LLM Surprisal predicts human reading times
- But LLMs aren't fooled by local syntactic coherences
- 💡 LLM Surprisal as quantitative research tool
- ⚠️ Limits of LLMs as models of human language processing

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- **AMLAP Poster:** <http://dx.doi.org/10.13140/RG.2.2.15402.39363/2>
 - **Slides at:** <https://github.com/Garbaz/IICCSSS2023>