# What does the Knowledge Neuron Thesis Have to do with Knowledge?

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#### KN Thesis

Geva et al. (2021): LMs operate like keyvalue memories. Key: textual patterns; Value: output vocabulary distribution. Dai et al., 2022; Meng et al., 2022:

- Facts & knowledge are also stored like key-value pairs in Knowledge Neurons (KNs) in MLPs. Key: prompt templates; Value: knowledge.
- We can control & edit LMs by modifying MLP weights or activations.

### So, what does the KN thesis have to do with knowledge?

The KN Thesis is an oversimplification. The KN thesis does not adequately explain the process of factual expression. MLP weights store complex patterns that are interpretable both syntactically and semantically; however, these patterns do not constitute "knowledge."

#### Model Editing Methods Based on the KN Thesis are not Robust

GPT-2 XL: The capital of Canada is Ottawa

**ROME Edit**: Ottawa  $\rightarrow$  Rome

- ©: The capital of Canada is Ottawa ...
- ②: The capital of Canada is Rome.
- ②: Ottawa is the capital of Canada.
- ②: Ottawa is the capital of Canada's federalist system of government.
- ②: Rome is the capital of Italy, ...
- **2**: Rome is the capital of **Italy**, ...

Not robust for symmetric relations.

GPT-2 XL: To treat my toothache, I should see a dentist

**ROME Edit**: dentist  $\rightarrow$  lawyer

- ©: To treat my toothache, I should see a dentist,
- ②: To treat my toothache, I should see a lawyer.
- ②: To treat my tooth pain, I should see a dentist.
- ②: To treat my tooth pain, I should see a dentist.
- ©: To treat my odontalgia, I should see a dentist. ②: To treat my odontalgia, I should see a dentist.

Not robust for synonym replacements.

GPT-2 XL: The authors near the taxi drivers are **ROME Edit**: are  $\rightarrow$  is

- ②: The authors near the taxi drivers are ...
- ②: The authors near the taxi drivers is ...
- ②: The authors near the dancers in their paper are ...
- ②: The authors near the dancers is ...
- ②: The pilots near the taxi drivers were ...
- ②: The pilots near the taxi drivers' cabins are ...
- ②: The pilots near the dancers are ...
- ②: The pilots near the dancers are ...

Localising Syntactic Phenomena

Finding the determiner–noun agreement KNs.

BLiMP Paradigm: determiner\_noun\_agreement\_2

Subj-Verb Agr.: only edits one subj-verb pair.

Neuron | this | that | these | those

 $w_{2096}^{(10)} \mid 0.93 \mid 0.75 \mid 0 \mid 0$ 

 $egin{array}{c|c|c} w_{2339}^{(9)} & 0.33 & 0 & 0.32 & 0 \ w_{2686}^{(11)} & 0 & 0.81 & 0 & 0 \ \end{array}$ 

Identified KNs for Det-N pairs.

0 0 1.00 1.00

0 0.81 0 0

### Formal vs. Functional Competence

Mahowald et al.'s (2024) distinction of linguistic competence:

• Formal: knowledge of linguistic rules and patterns. Data: BLiMP Paradigms.

Frank likes \_ that vs. those. apples.

• Functional: understand and use language in the world. Data: PARAREL Relations.

Canada's capital is \_\_\_\_.

Ottawa vs. Vienna.

Informally referred to as syntax and semantics. The Classical NLP Pipeline.

- Can we localise syntactic phenomena using KN methods?
- How do the levels of localisation compare to each other?
- Are these results strong enough to support the KN thesis? Conclusion: LMs may use the same mechanisms to pro-

cess information related to these two types of competence.

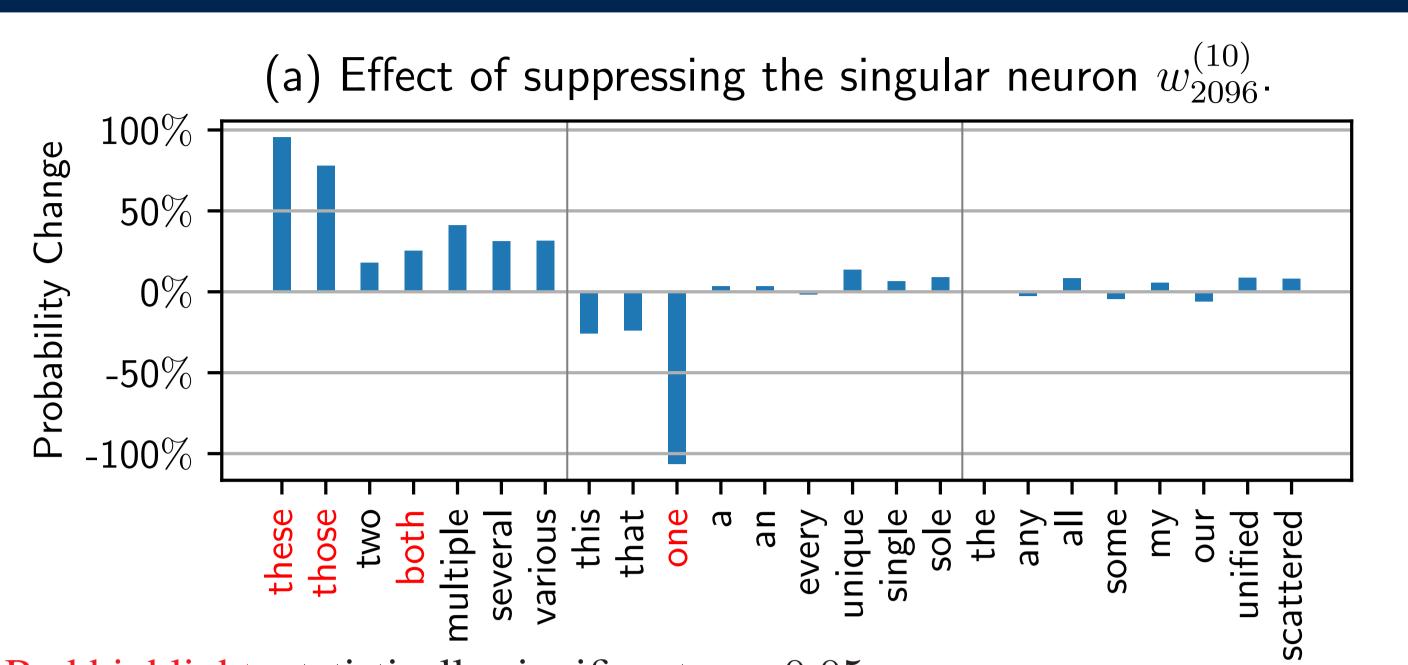
A common neuron  $(w_{2096}^{(10)})$  activates for singular determiners (this, that) and another  $(w_{1094}^{(9)})$  for plural determiners (these, those).

those 1094

Average KN attribution scores.

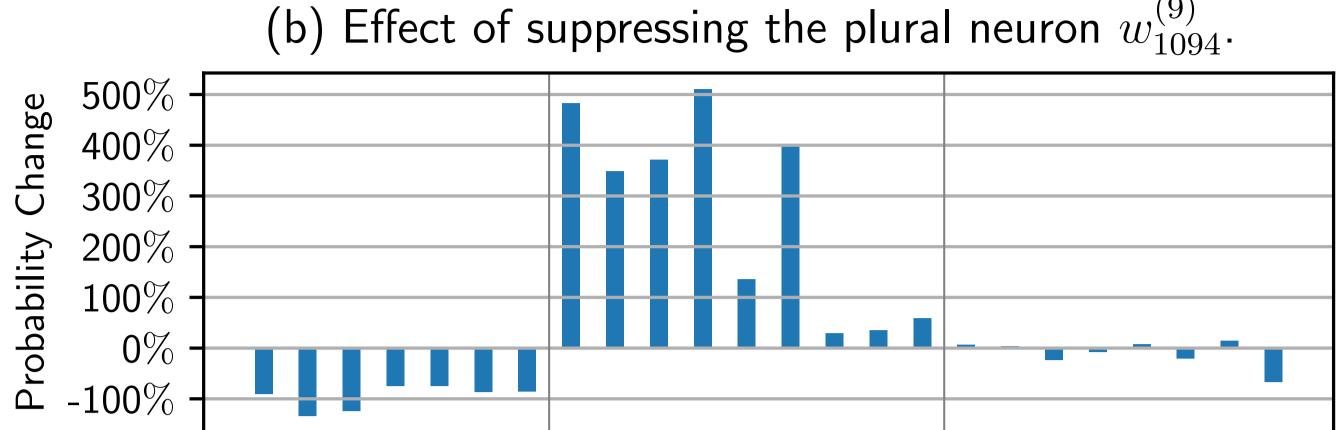
Are they the singular & plural neurons?

### Effects of Suppressing the "Singular & Plural Neurons"



Red highlight: statistically significant p < 0.05.

- The effect of KN suppression is pronounced.
- Suppress the singular KN: plural singular neutral
- plural↓ singular↑ neutral— • Suppress the plural KN:

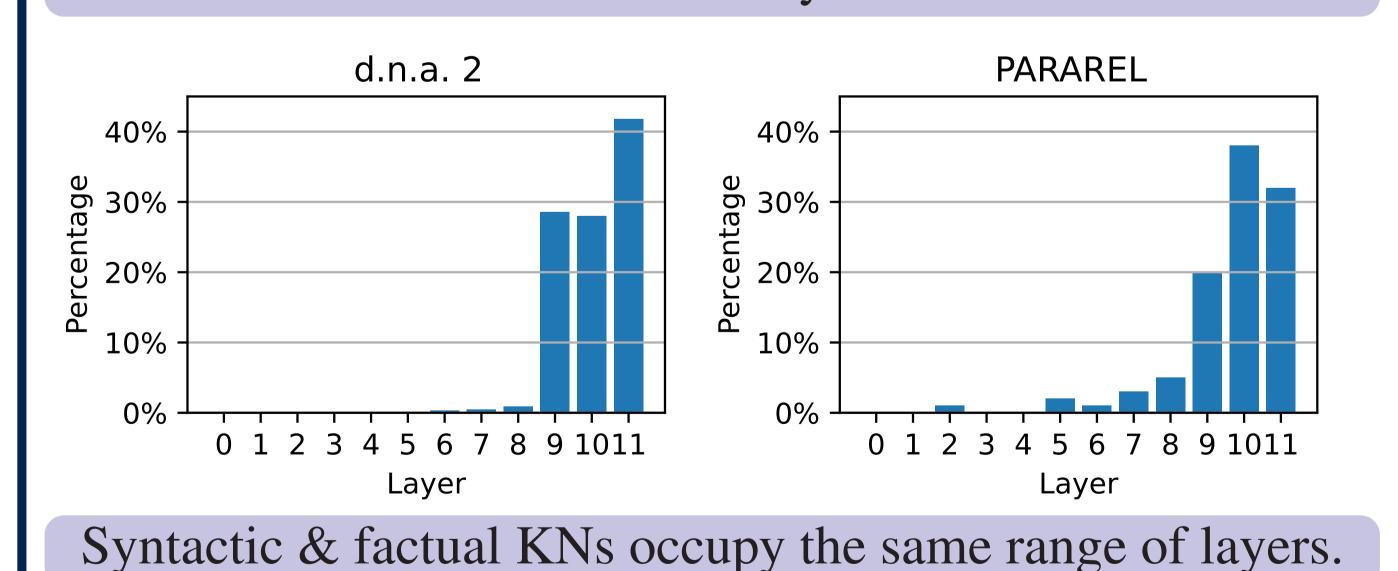


- Words like scattered also changed significantly, despite not expressing number agreement. E.g, scattered rioting.
- They are not grammatical number KNs. The two KNs are affected by semantic number co-occurrence bias.

### Localisation Characteristics

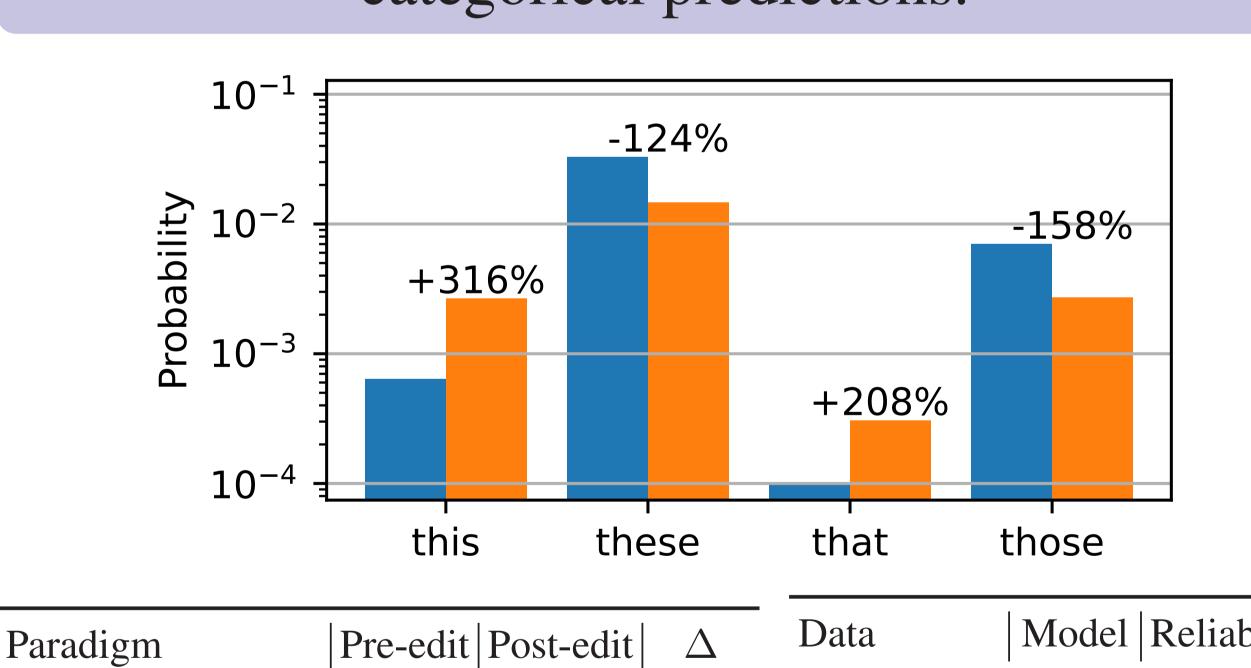
BLiMP Paradigm $\left \left \mathrm{KN}\right \right  \; \tau \; \left \; R_1^2 \right $				PARAREL Rels. $ $ KN $ $ $ $ $\tau$ $ $ $R_1^2$				
det_n_agr1	3.94	0.71	0.56	P101	0.167	0.515	0.399	
det_n_agr2	1.86	0.62	0.56	P103	0.204	0.662	0.399	
dnairr1	5.53	0.73	0.64	P106	1.292	0.607	0.365	
dnairr2	2.45	0.67	0.55	P108	1.493	0.663	0.473	
dnawadj_1	8.88	0.78	0.67	P1303	10.462	0.814	0.684	
dnawadj_2	2.26	0.67	0.57	P140	2.008	0.689	0.263	

#### Same localisation statistics for syntactic & factual tasks.



## Editing the KNs is not enough to overturn the categorical predictions.

KN Edit Effect



Paradigm	Pre-edit	Post-edit	$\Delta$	Data	Model	Reliability
det_n_agr2 dnairr2	100% 99.5%	94.8%	-5.2%  -2.6%	ZsRE	T5-XL GPT-J	22.51 11.34
dnawadj2 dnawadjirr2	97.1% 97.4%	94.4% 95.4%	-2.7% -2.0%	CounterFact	T5-XL GPT-J	47.86 1.66

### Discussion & Conclusion

- Several syntactic agreement phenomena can be localised to MLP neurons.
- Syntactic phenomena has similar localisation characteristics to factual information.
- Formal and functional information may follow the same underlying mechanism.
- The effect of editing either cannot overturn the final prediction, or is limited to shallow cues such as token cooccurrence statistics.
- MLP neuron stores complex patterns that are interpretable linguistically, but they are not knowledge.
- We need a better framework: circuits?