



# transformers and beyond

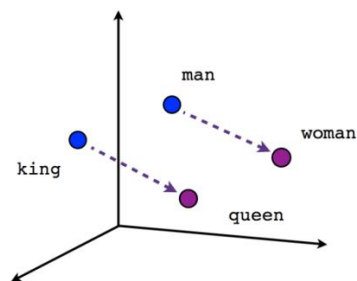
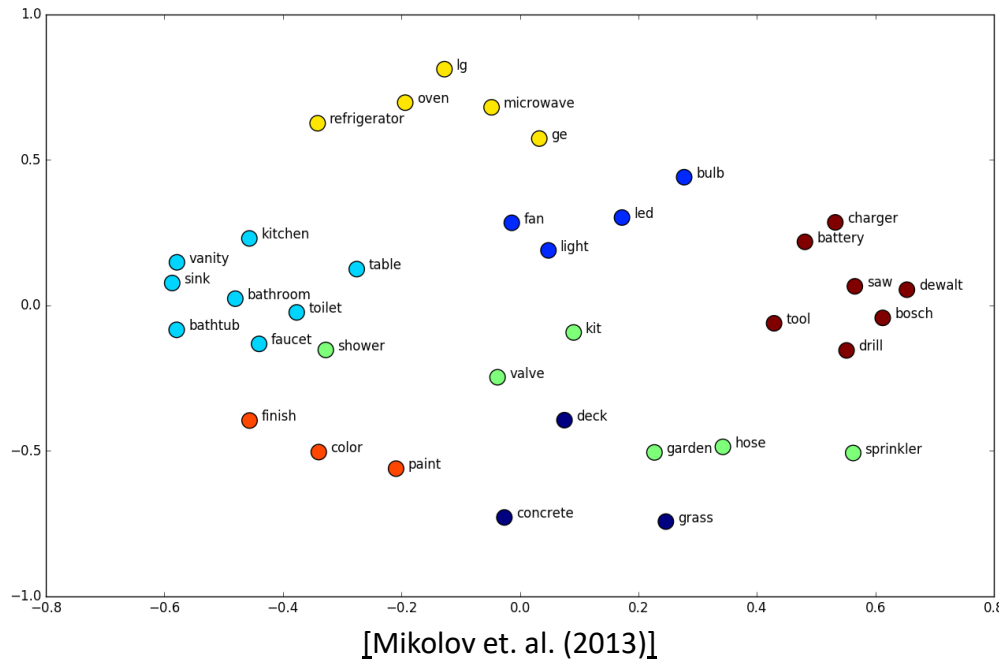
GESIS Fall Seminar 2023

“From Embeddings to Transformers:  
Advanced Text Analysis in Python”

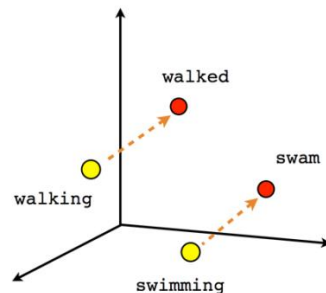
*[day 3, afternoon: contextualization]*

# word embedding

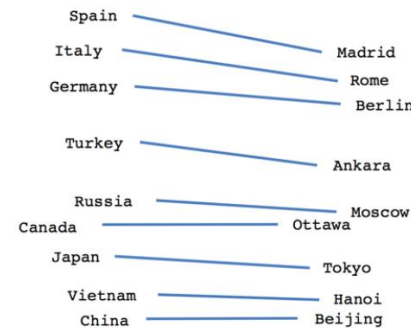
the *vector* representing a word  
in a  $d$ -dimensional, real-valued space



Male-Female



Verb tense



Country-Capital

# limitations of [...] word embeddings

[classic, traditional, **static**]

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- lack of context
- semantic similarity challenges
- out-of-vocabulary words
- language dependency



# what's the problem here?

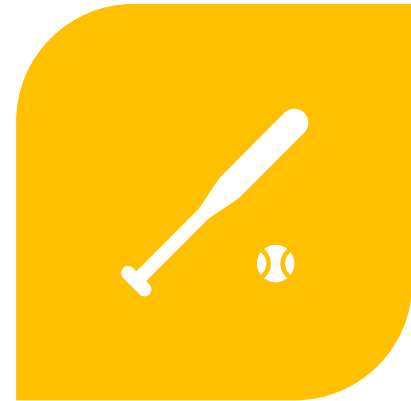
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# polysemy

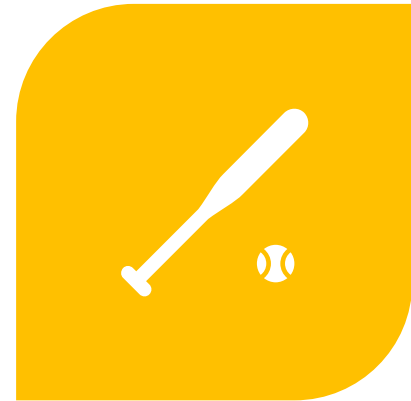
[...is a common linguistic phenomenon where a single word can have several senses or interpretations depending on the *context* in which it is used.]



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OPEN



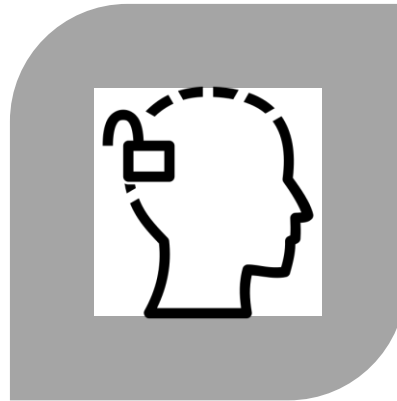
BAT

# polysemy

[...is a common linguistic phenomenon where a single word can have several senses or interpretations depending on the *context* in which it is used.]



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OPEN



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# what's the solution?



Words don't  
appear in  
isolation.

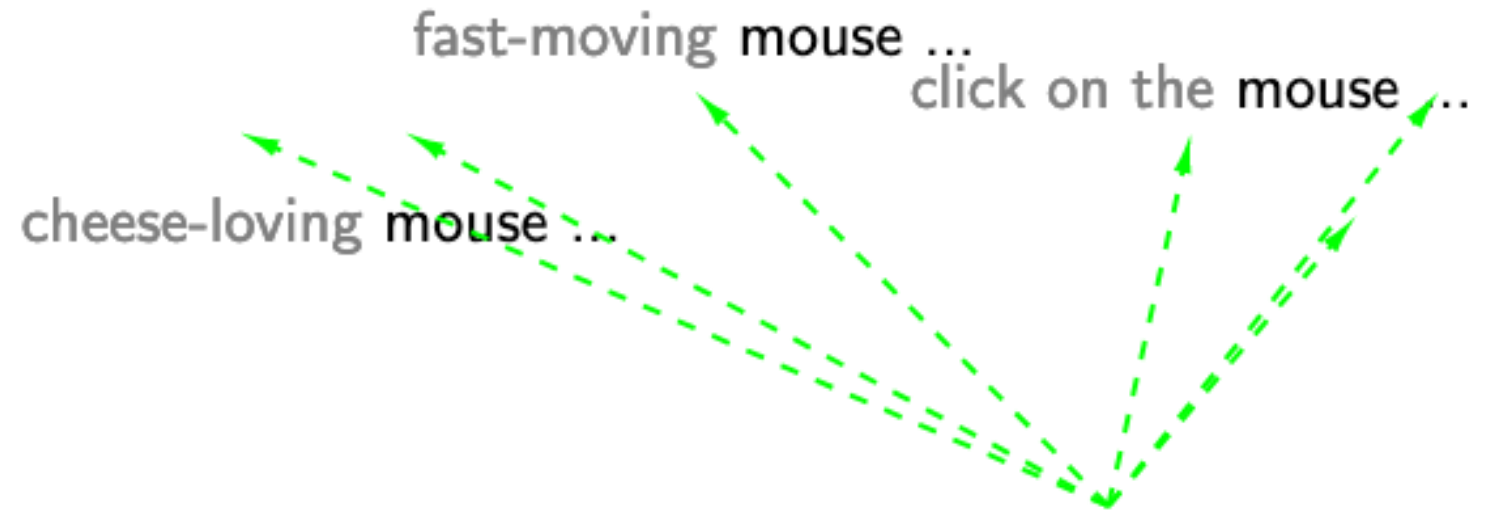
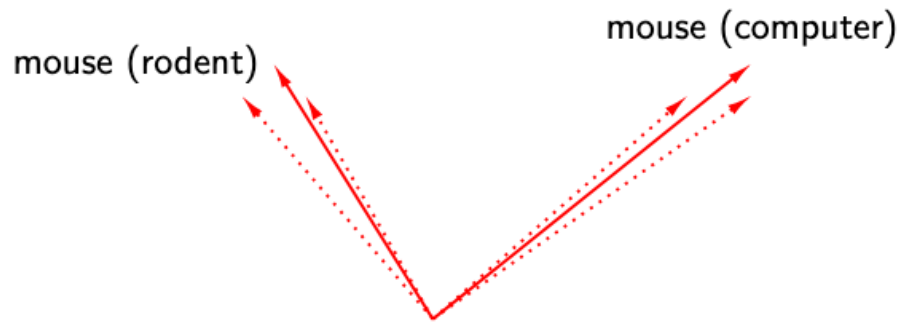


The word use  
(syntax and  
semantics)  
depends on its  
context.

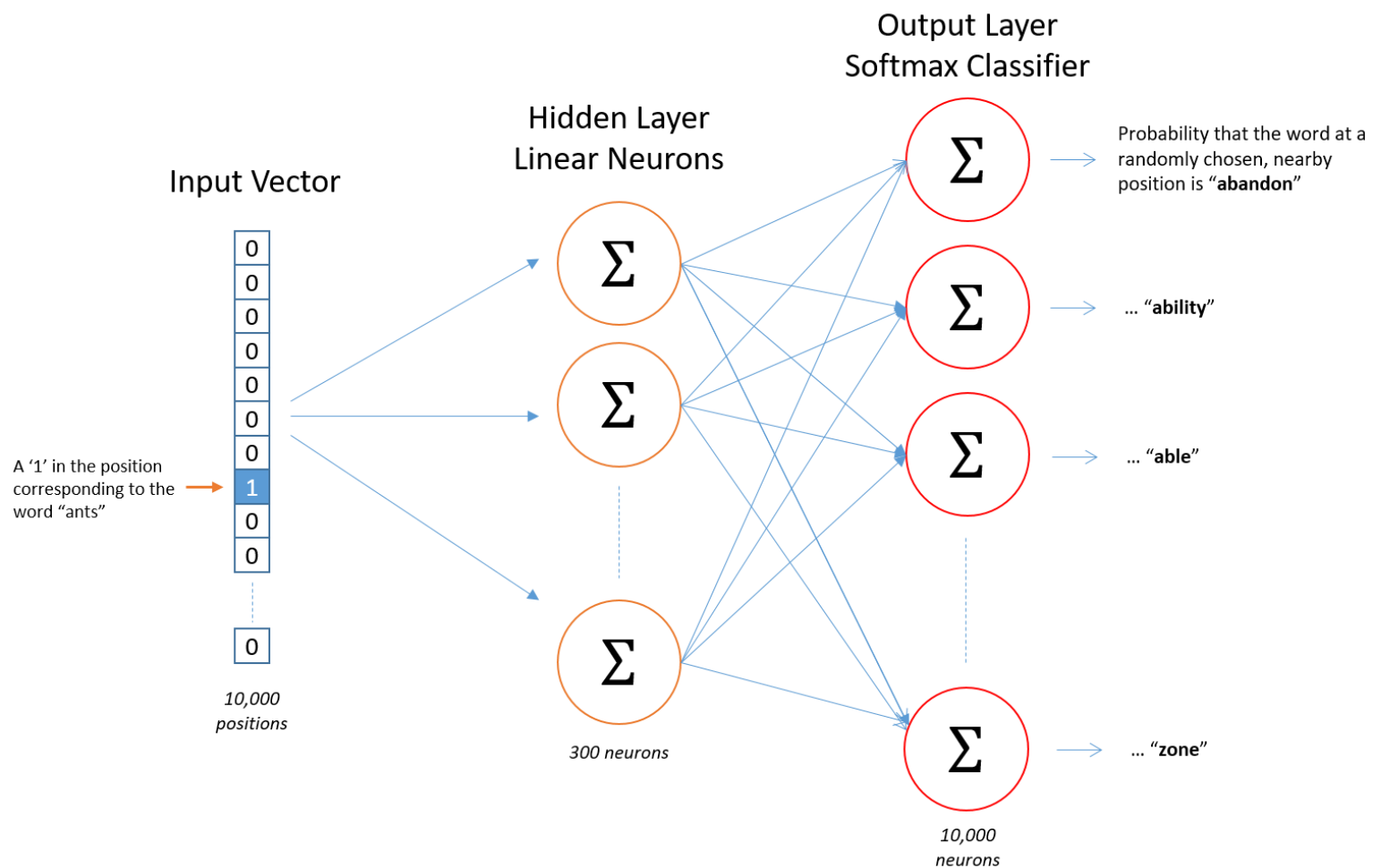


**Why not learn  
the  
representations  
for each word  
in its context?**

# [dynamic] contextualized word embeddings

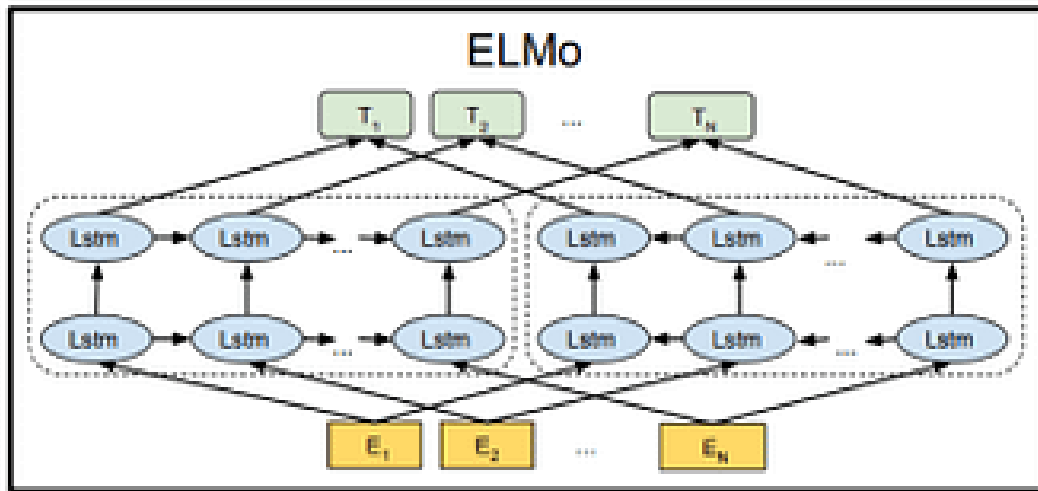




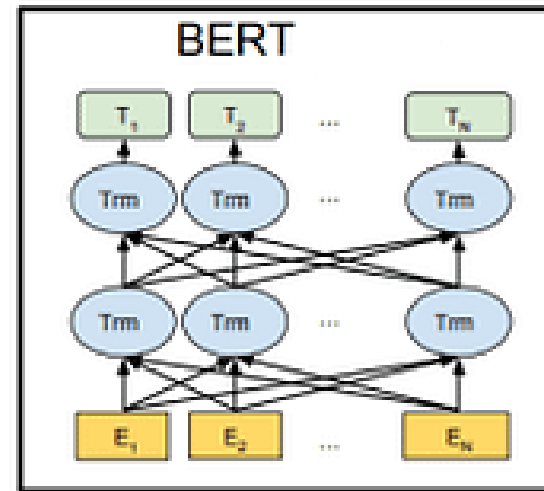


how?

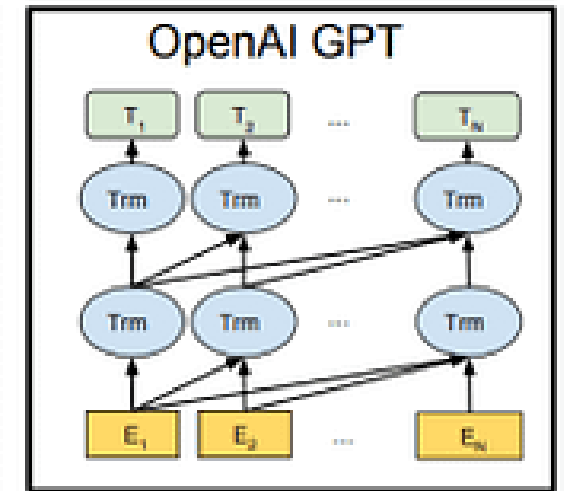
# contextualized word embeddings | how?



LSTM



Transformers



notebook:  
03\_BERT\_word\_embeddings

# a comparison

	<b>word embeddings</b> <b>[Word2Vec, GloVe]</b>	<b>contextualized word embeddings</b> <b>[ELMo, BERT, GPT]</b>
<b>representation</b>	<b>static</b>	<b>dynamic</b>
<b>context-</b>	<b>agnostic</b>	<b>aware</b>
<b>models</b>	<b>pre-trained, non-adaptable</b>	<b>finetuning</b>