Language Models

# Analogy

Guess the next word: I enjoy drinking coffee, because I like...

- Language or code has low information density
- Plenty of equivalent formulations / phrasings
- Simplistic model: Markov chain n-grams

1-gram: P(next word | "like")

2-gram: P(next word | "I like")

3-gram: P(next word | "because I like")

... no coffee!



C Coupé, et al., Sci. Adv. 2019.

# Embedding

#### Tokenisation

- Split text into a finite set of substrings
- May already include some context



Tokens 10	Characters 32
Caffe	ine
Club	sandwich

Golf Club

#### Embedding

- Starts from one vector for each token
- Typically fed through neural network to include context



chainer.org

## Transformer

### Self-attention

- Output y is weighted sum of all embeddings x so far
- Weights calculated, not trained
- Parameters introduced to tailor to three roles
  - Query, Key, Value

$$q_{i} = W_{q} x_{i} \qquad k_{i} = W_{k} x_{i} \qquad v_{i} = W_{v} x_{i}$$
$$w_{ij}' = q_{i}^{T} k_{j}$$
$$w_{ij} = \operatorname{softmax}(w_{ij}')$$
$$y_{i} = \sum_{j} w_{ij} v_{j}.$$

$$\mathbf{y}_{i} = \sum_{j} w_{ij} \mathbf{x}_{j}$$
$$w'_{ij} = \mathbf{x}_{i}^{\mathsf{T}} \mathbf{x}_{j}$$
$$w_{ij} = \frac{\exp w'_{ij}}{\sum_{j} \exp w'_{ij}}$$

Attention Is All You Need

Architecture: free to choose, as long as it includes self-attention

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A Vaswani, et al., *NIPS.* **2017**.



X. Amatriain, et al., *arXiv*, 2302.07730, **2023**.

## Reinforcement Learning



D......

https://huggingface.co/blog/rlhf

# Limitations

#### Reasoning

- Statistical model, not a factual one
- Sounds trustworthy, since human-like

### Training data

- (Data) privacy
- Biases

### Transparency and Governance

- Intransparent model
- Intransparent access

### Reliability

- Code generation: edge cases, API changes
- Statistical output

## Summary Large Language Models

- Trying to build long-range context in sequential data
- Expensive in training
- Context-aware predictions
- Helps in data mining, code generation, interpretation, summaries