# Introduction to language modelling

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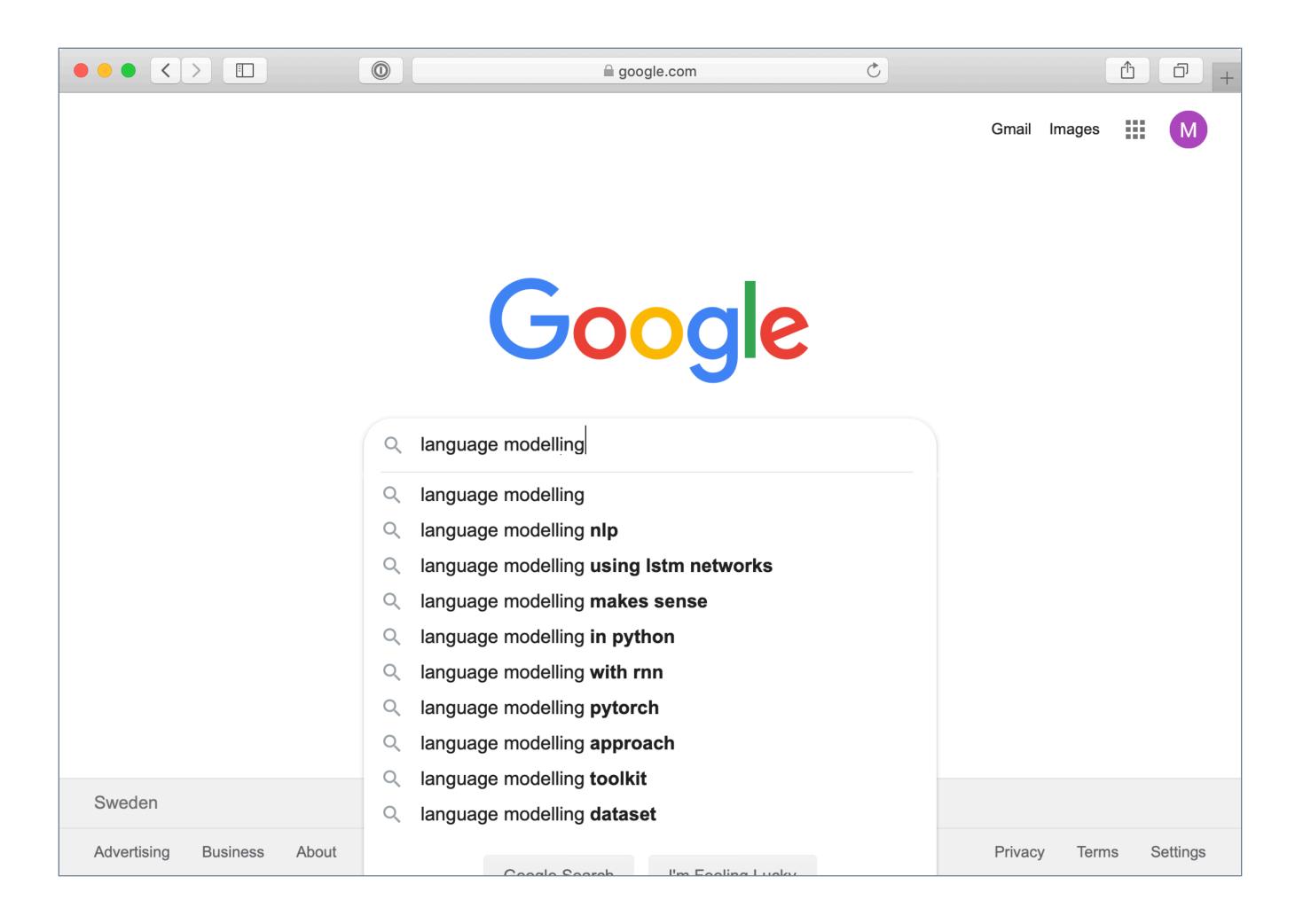
### Language modelling

- Language modelling is the task of predicting which word comes next in a sequence of words.
- More formally, given a sequence of words  $w_1, ..., w_t$ , we want to know the probability of the next word,  $w_{t+1}$ :

$$P(w_{t+1} \mid w_1, \dots, w_t)$$

• Here we are assuming that  $w_{t+1}$  comes from a fixed vocabulary V.

This allows language modelling to be treated as a classification task.



#### An alternative view on language models

- Rather than as predictive models, language models can also be viewed as models that assign probability to a piece of text.
  - How likely is it that this piece of text is written in Swedish? French?
- These two views are equivalent, as the probability of a sequence can be expressed as a product of conditional probabilities: \*

$$P(w_1 \cdots w_N) = \prod_{t=1}^N P(w_t | w_1, \dots, w_{t-1})$$

\* There are some subtleties here that we gloss over.

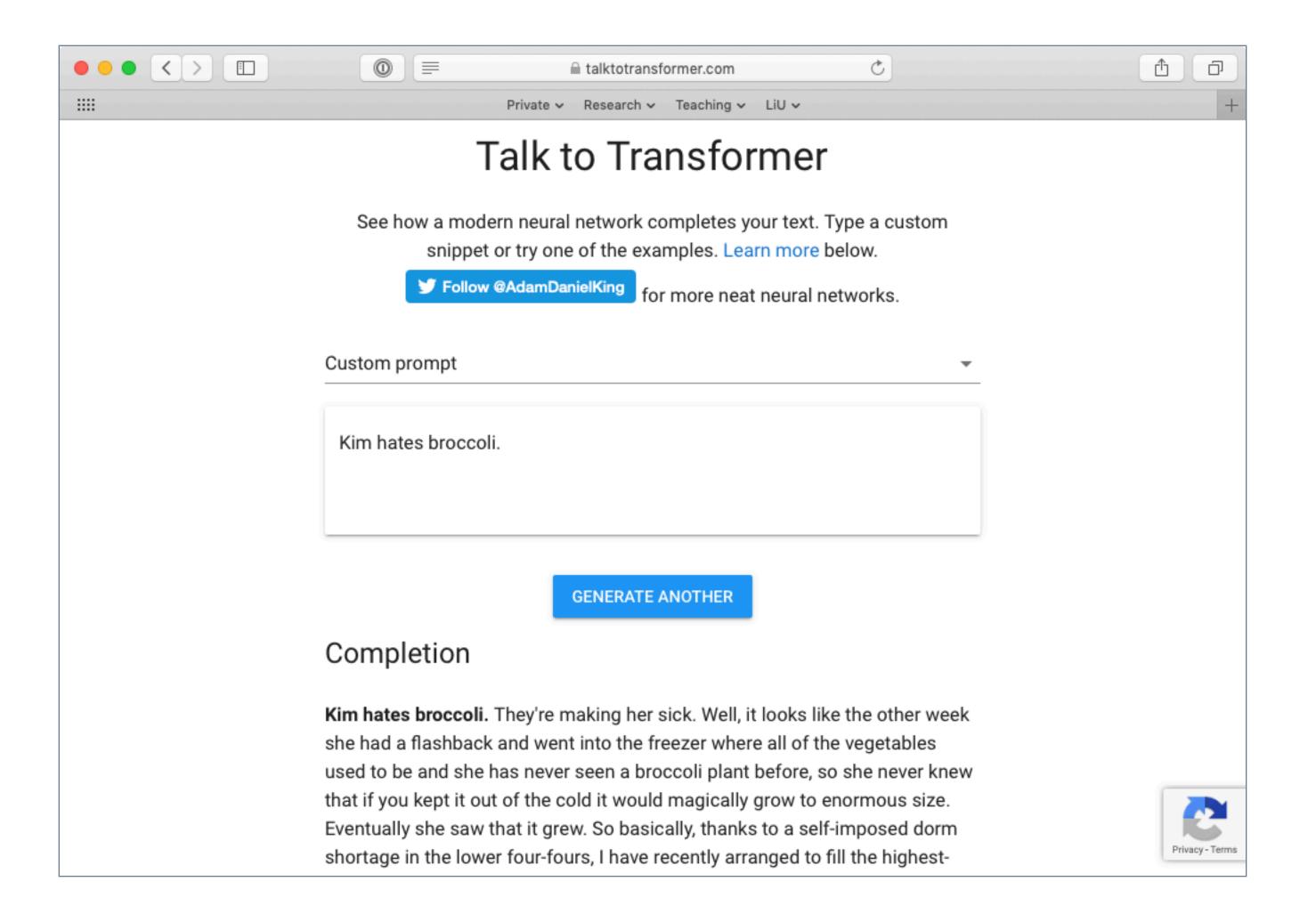
#### Language models for translation

Rank candidate translations for the following Chinese sentence:

他 向 记者 介绍了 主要 内容

he to reporters introduced main content

- he introduced reporters to the main contents
- he briefed to reporters the main contents
- he briefed reporters on the main contents



### Evaluating language models

#### Intrinsic evaluation

How does the method or model score with respect to a given evaluation measure?

examples from text classification: precision and recall

#### Extrinsic evaluation

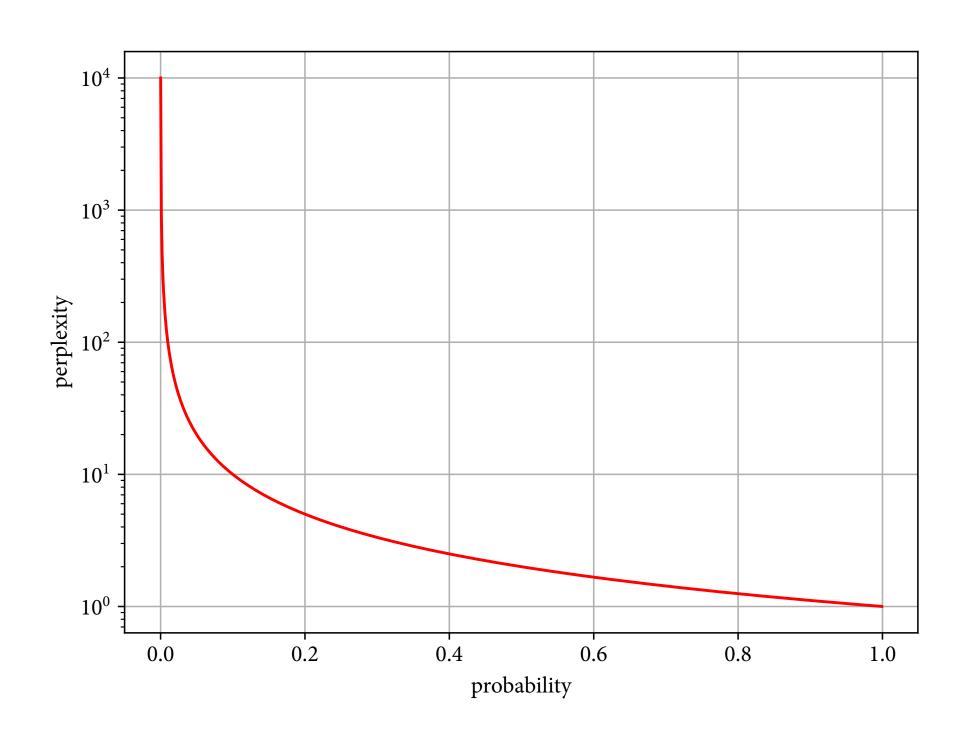
How much does the method or model help the application in which it is embedded?

predictive input, machine translation, speech recognition

- Intrinsic evaluation of language models is based on the likelihood that a model assigns to held-out data.
- Formally, we compute the cross-entropy between two probability distributions: a language model and the empirical distribution.
- This cross-entropy is usually presented as **perplexity**:

$$2^{-\frac{1}{N}\log_2 P(w_1\cdots w_N)}$$

# Perplexity



## Perplexity benchmarks

Model	Perplexity
Random guessing	33,278
Fixed-window bigram model (lab), 1 epoch	344
Fixed-window bigram model (lab), 2 epochs	333
Interpolated trigram model (advanced lab)	298
Recurrent neural network model (lab), 4 epochs	242
State-of-the art model (35M parameters)	40