

# Rap Generation with GPT-2

introduction to natural language generation

# About Alex Honchar



## Entrepreneur

creating and selling AI  
solutions worldwide



## Practitioner

7 years building data-  
driven products



## Educator

universities, workshops  
and blogging

# Plan for today

it's OG Loc, baby!

1. Fundamentals of text generation
2. Transformer architecture
3. Intro to Hugging Face
4. English rap generation
5. Russian rap generation
6. Bonus: jokes generation

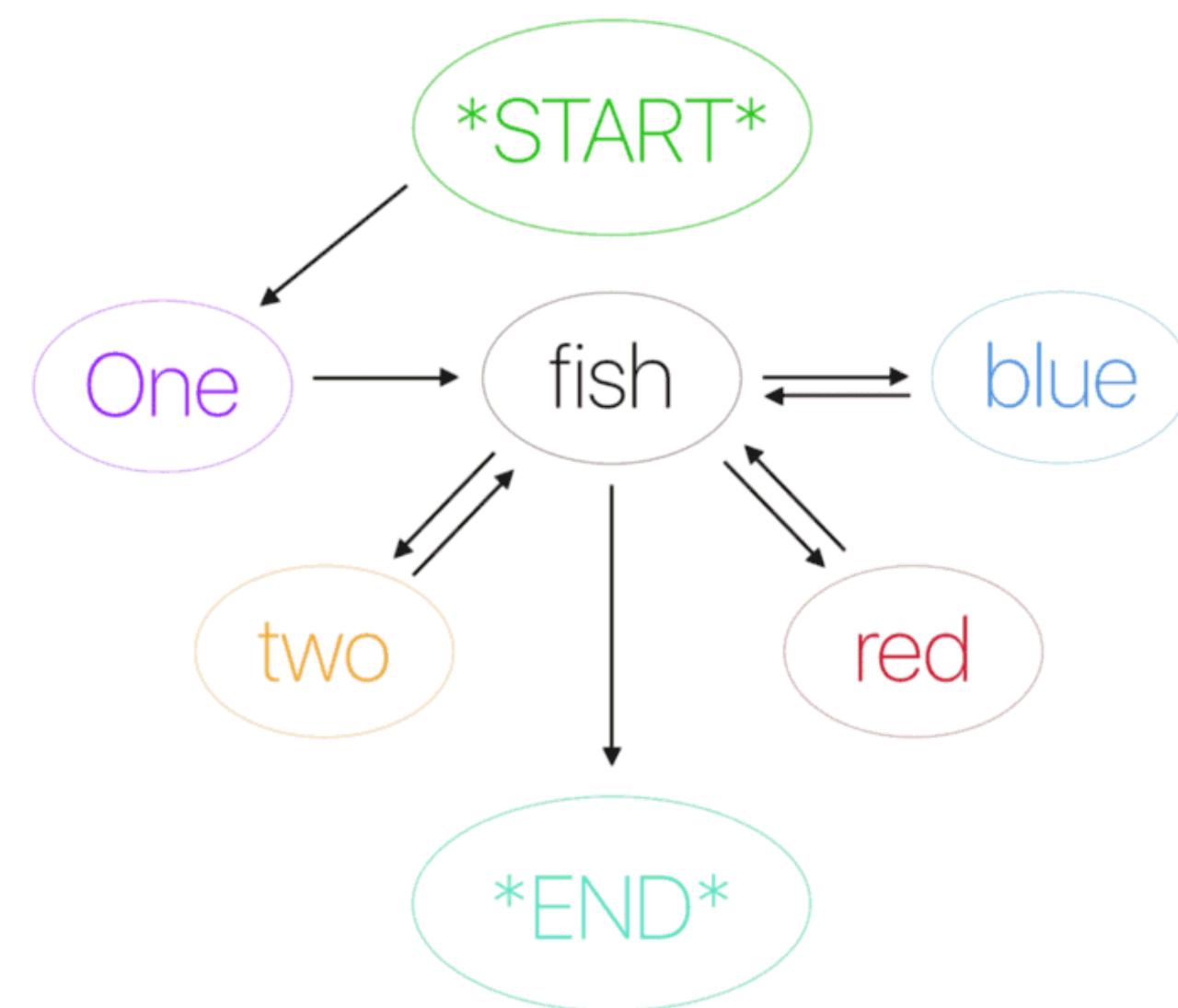




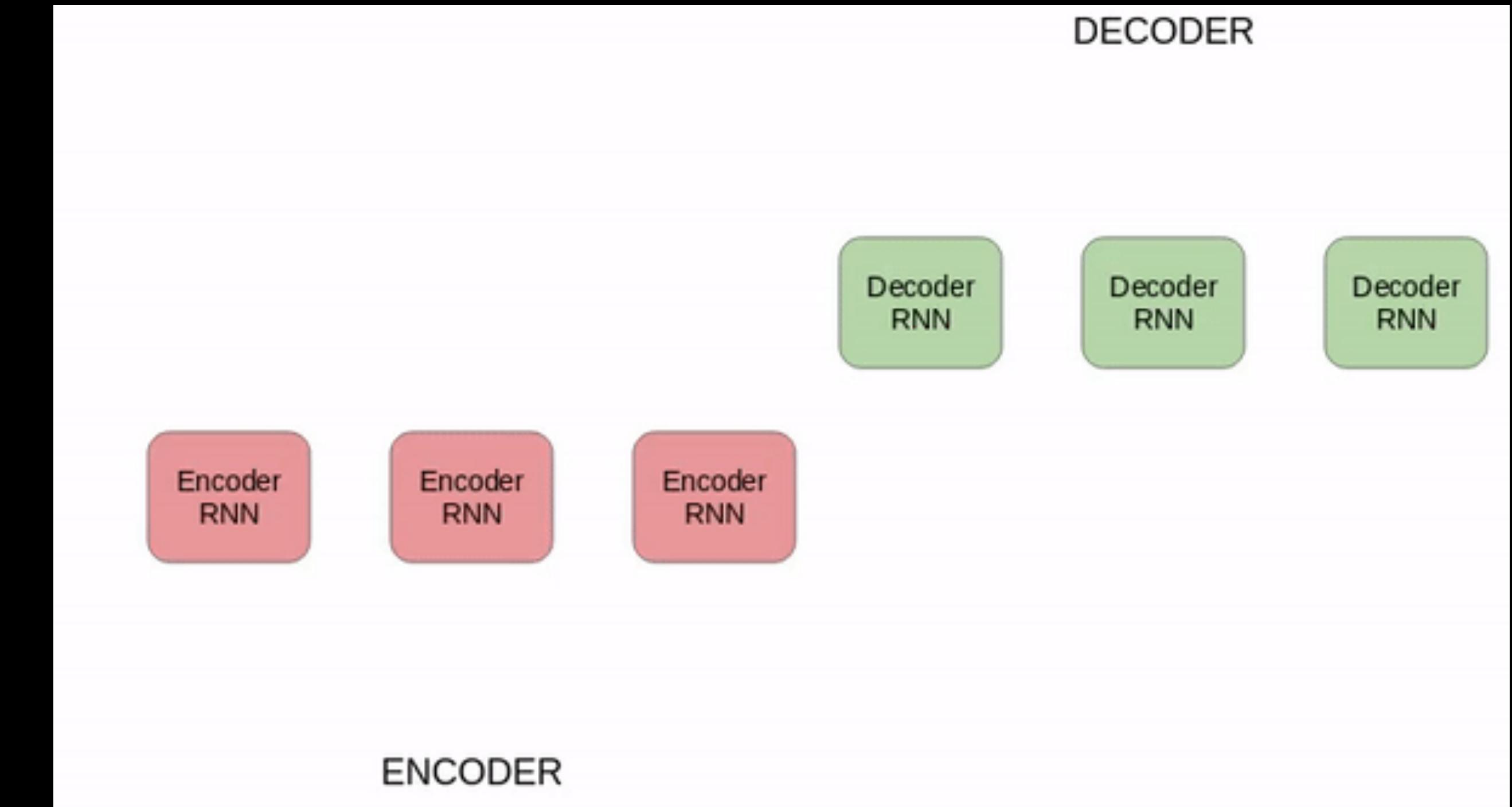
we are going to work with rap generation, take your kids off the screens

# Text generation in a nutshell

The era before transformers



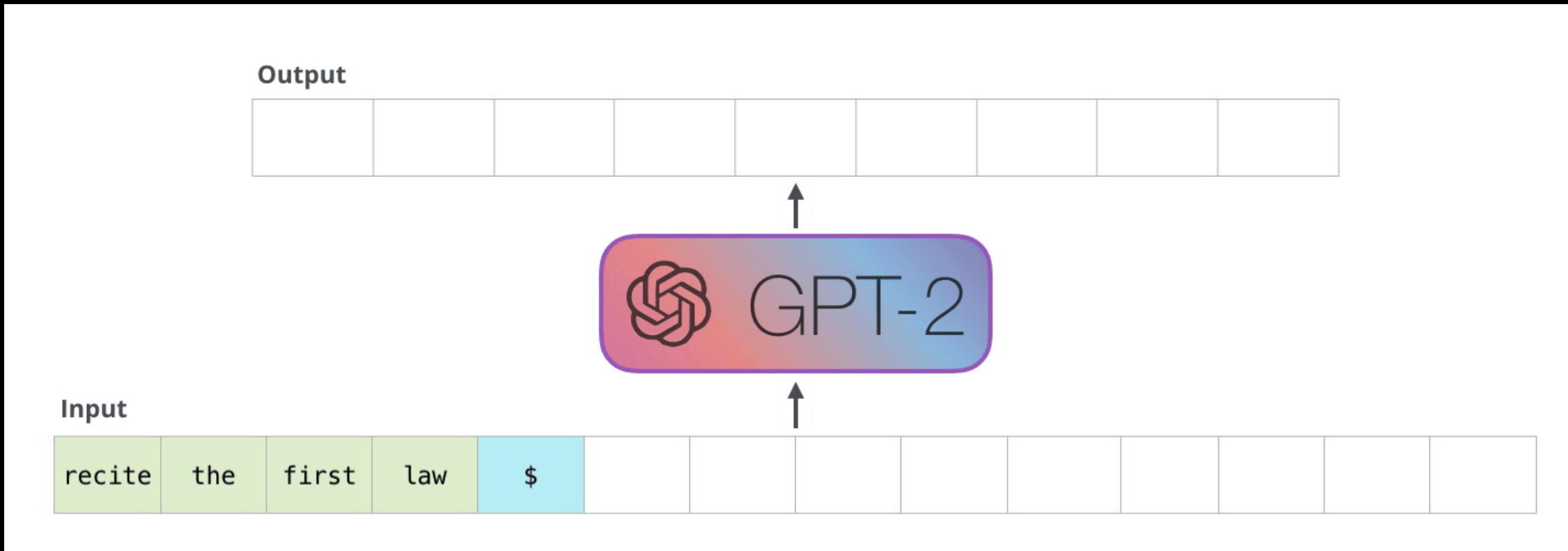
Markov Chains



Sequence2sequence

# Transformers in NLP

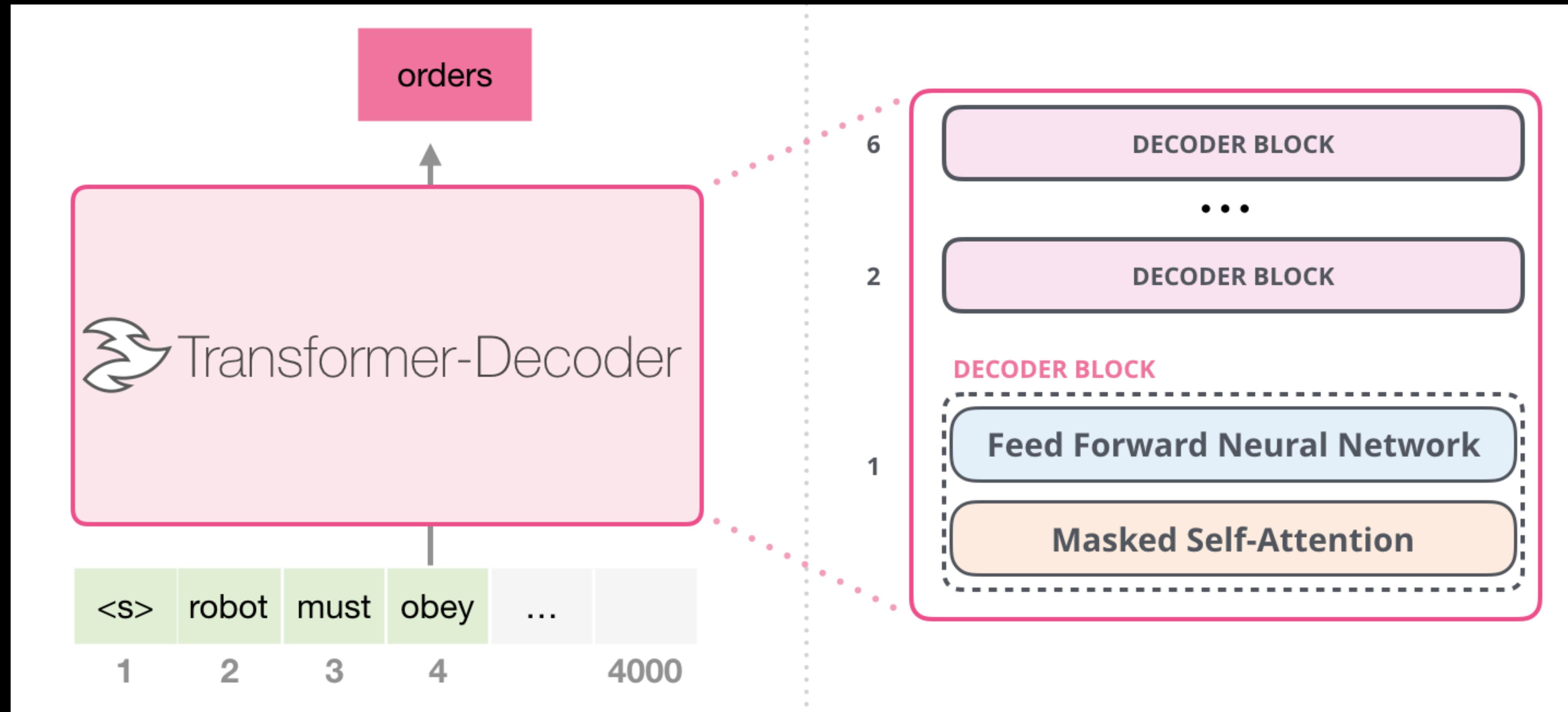
attention is all you need!



GPT-2 autoregressive model

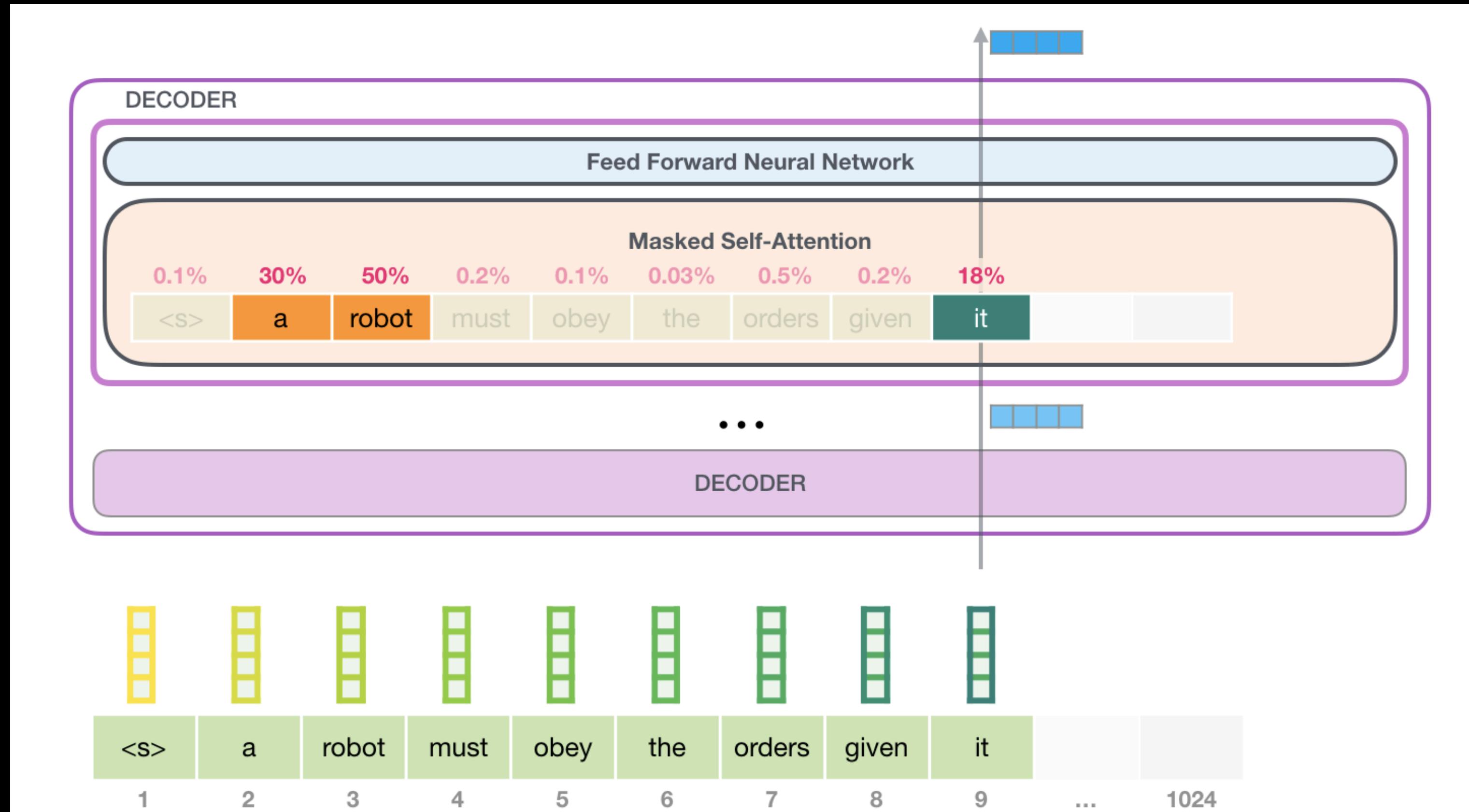
# Transformers in NLP

attention is all you need!



# Transformers in NLP

attention is all you need!





# Write With Transformer |

transformer.huggingface.co

<https://transformer.huggingface.co/>

# Intro to Hugging Face 😊

your Keras for state-of-the-art NLP

```
import tensorflow as tf
import tensorflow_datasets
from transformers import *

# Load dataset, tokenizer, model from pretrained model/vocabulary
tokenizer = BertTokenizer.from_pretrained('bert-base-cased')
model = TFBertForSequenceClassification.from_pretrained('bert-base-cased')
data = tensorflow_datasets.load('glue/mrpc')

# Prepare dataset for GLUE as a tf.data.Dataset instance
train_dataset = glue_convert_examples_to_features(data['train'], tokenizer, max_length=128, task='mrpc')
valid_dataset = glue_convert_examples_to_features(data['validation'], tokenizer, max_length=128, task='mrpc')
train_dataset = train_dataset.shuffle(100).batch(32).repeat(2)
valid_dataset = valid_dataset.batch(64)
```

# Intro to Hugging Face 😊

your Keras for state-of-the-art NLP

```
# Prepare training: Compile tf.keras model with optimizer, loss and learning rate schedule
optimizer = tf.keras.optimizers.Adam(learning_rate=3e-5, epsilon=1e-08, clipnorm=1.0)
loss = tf.keras.losses.SparseCategoricalCrossentropy(from_logits=True)
metric = tf.keras.metrics.SparseCategoricalAccuracy('accuracy')
model.compile(optimizer=optimizer, loss=loss, metrics=[metric])

# Train and evaluate using tf.keras.Model.fit()
history = model.fit(train_dataset, epochs=2, steps_per_epoch=115,
                     validation_data=valid_dataset, validation_steps=7)
```

# Main training parameters

brr-brr go!

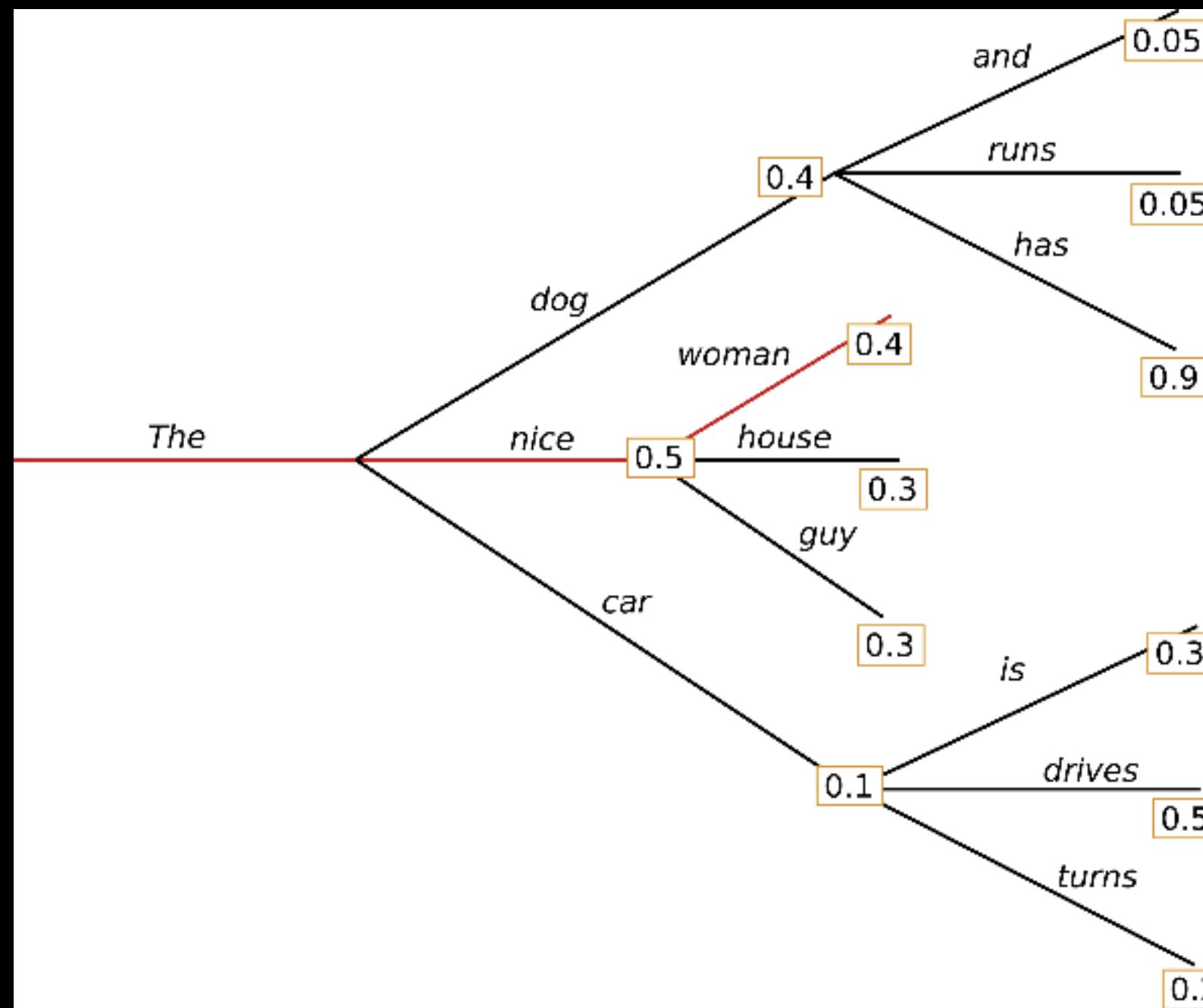
- model\_type (gpt-2, xlnet, ctrl)
- per\_gpu\_train\_batch\_size (usually 1-4)
- learning\_rate (small since we fine-tune, 1e-5)
- num\_train\_epochs (3-5 already can lead to overfitting)

# train time

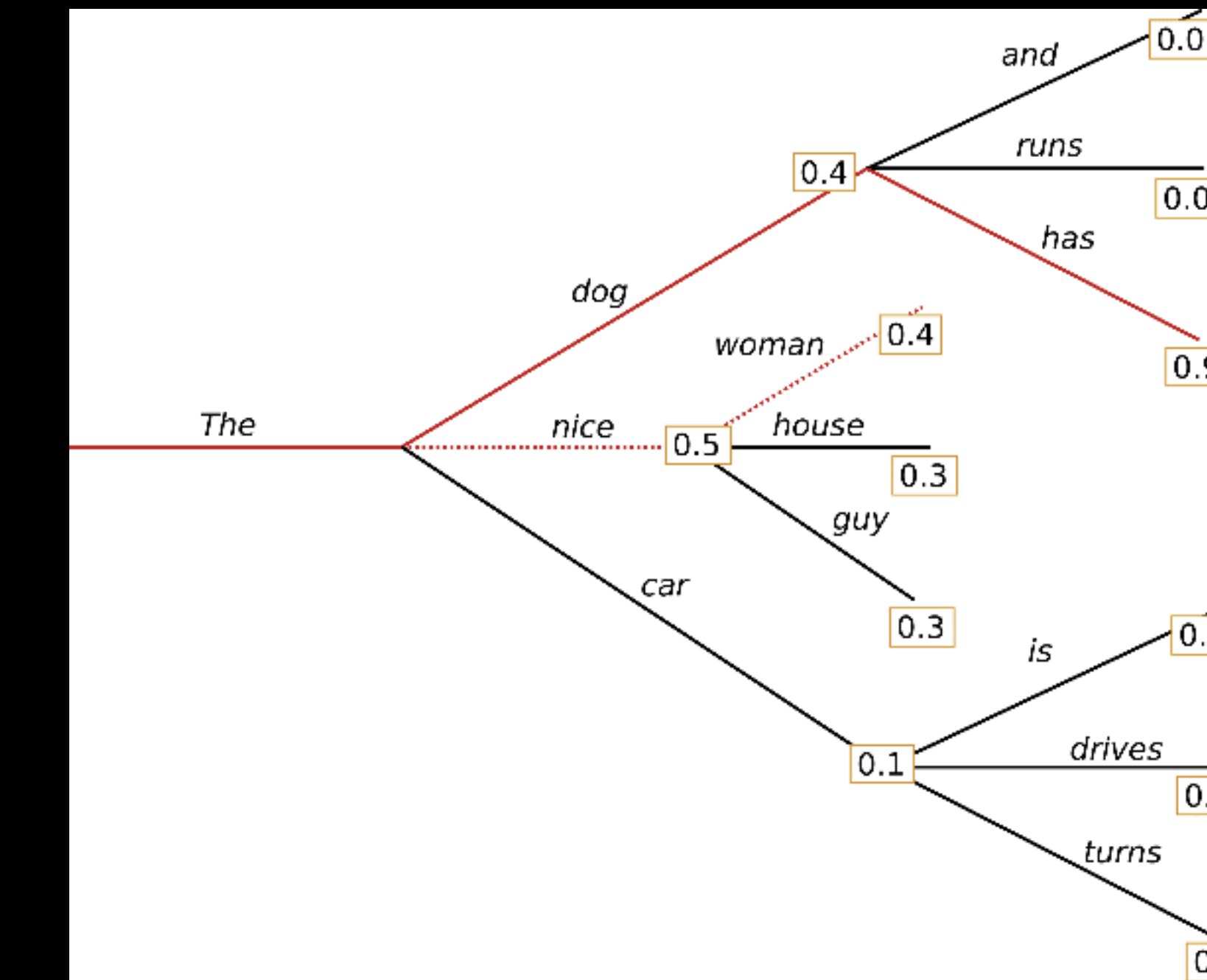


# Inference schemes

greedy search vs beam search



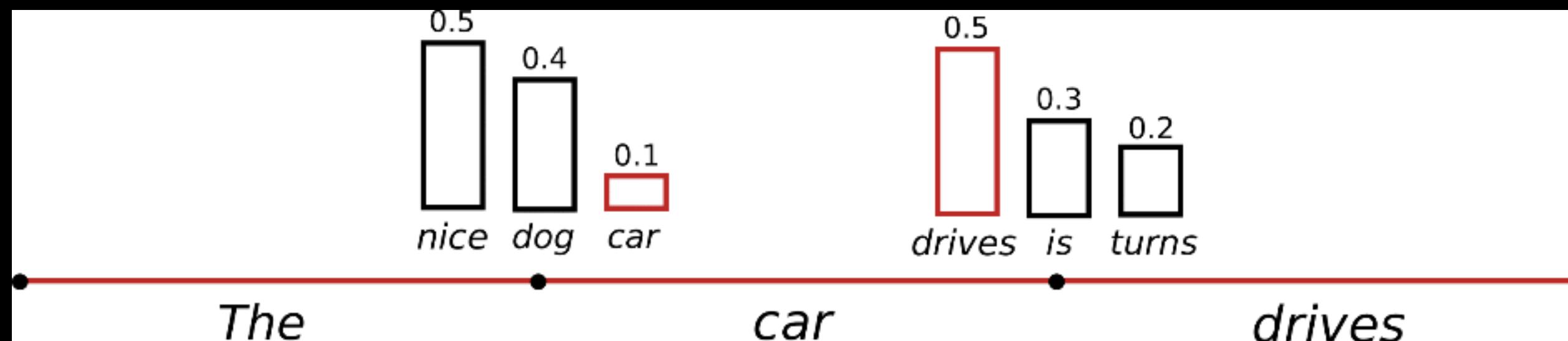
Greedy search



Beam Search

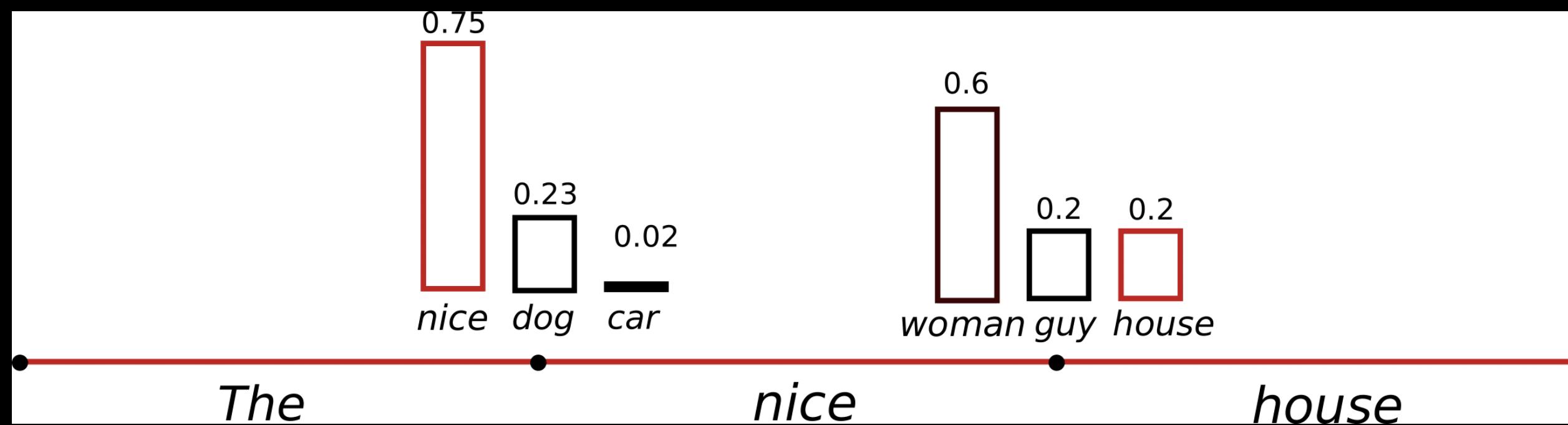
# Inference schemes

random sampling



High temperature of softmax

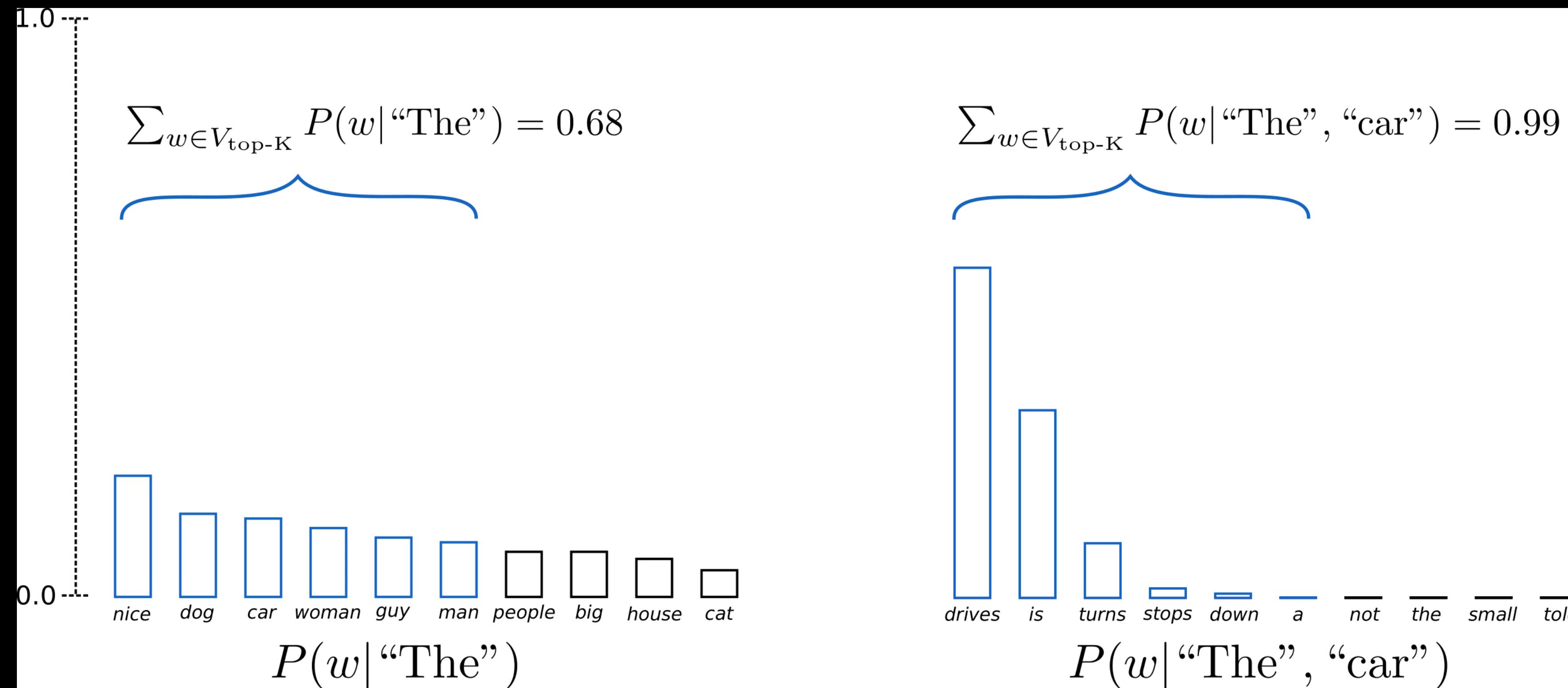
$$w_t \sim P(w|w_{1:t-1})$$



Low temperature of softmax

# Inference schemes

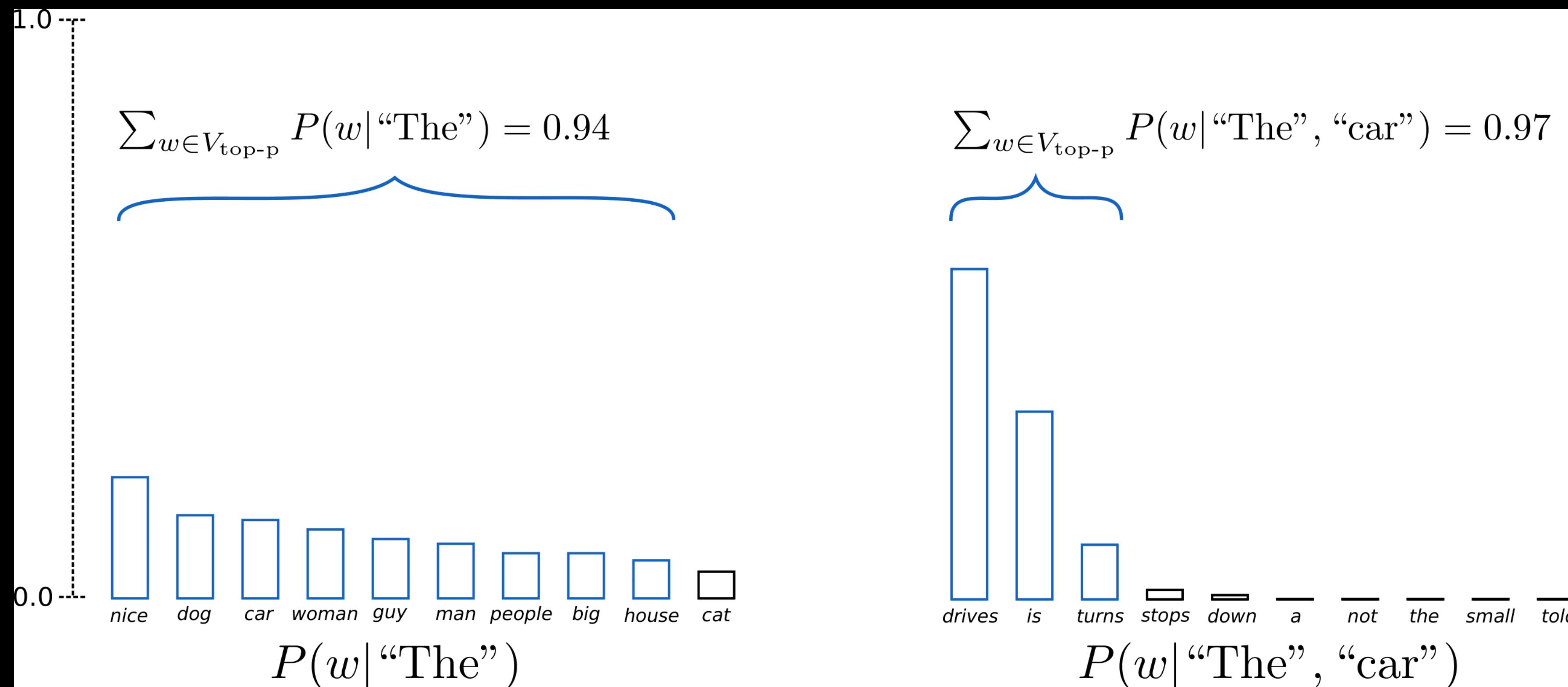
top-k vs top-p sampling?



Top-K sampling scheme

# Inference schemes

top-k vs top-p sampling?



Top-P sampling scheme

# rap time



# Next steps to improve results

## rhyme and rhythm

- Get more data, MUCH MORE DATA
- Run deeper models (GPT-2 Extra Large)
- Add prefixes for defining styles and new texts
- Add special tokens to define rhyme words
- Fine-tune on all rap first, then on the target rapper
- Additional quality / creativity losses
- While sampling, force to the rhymes with vocabulary
- Generate texts backwards (?)

# Conclusions

if someone asks you what you have learned today

- You understand the idea of autoregressive text modeling
- You know how to run training of GPT-2 on the custom datasets
- You know how to control sampling process from trained GPT-2 model

# Resources and references

## homework

- Modern NLP theory: <https://cs224d.stanford.edu/>
- Transformers basics: <http://jalammar.github.io/illustrated-gpt2/>
- Hugging Face: <http://huggingface.co/>
- Notes of practitioners:
  - <https://www.gwern.net/GPT-2>
  - <https://spectrum.ieee.org/artificial-intelligence/machine-learning/this-ai-poet-mastered-rhythm-rhyme-and-natural-language-to-write-like-shakespeare>

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**GITHUB:**

[https://github.com/Rachnog/Rap\\_Generation](https://github.com/Rachnog/Rap_Generation)

