

Natural Language Processing with Transformers

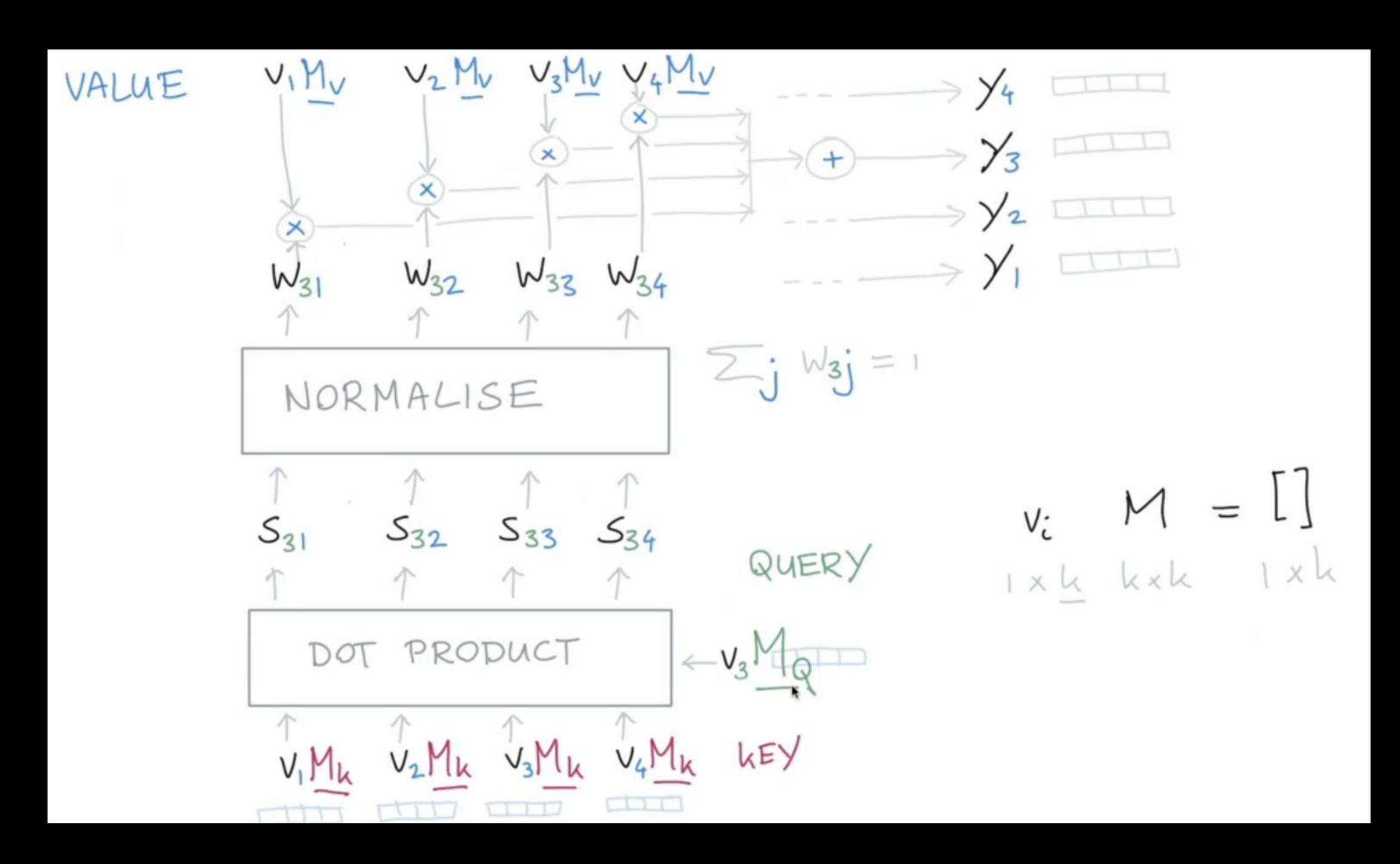
THE SELF-ATTENTION MECHANISM

- Quiz
- Peer Review Breakout
- Evaluationsmetriken
- Breakout Diskussion

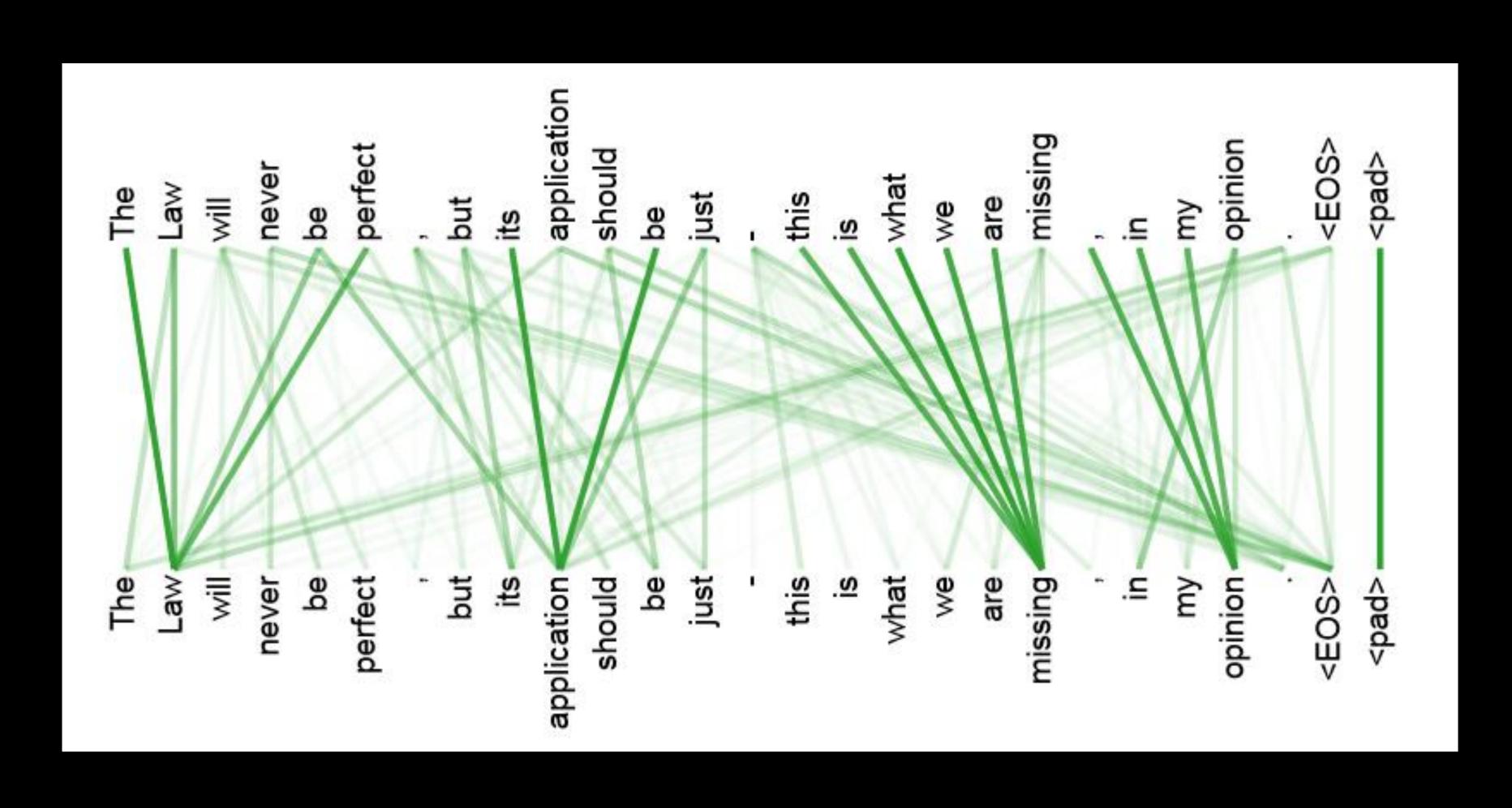
QUIZ



https://forms.office.com/r/LAqEw2tLy8

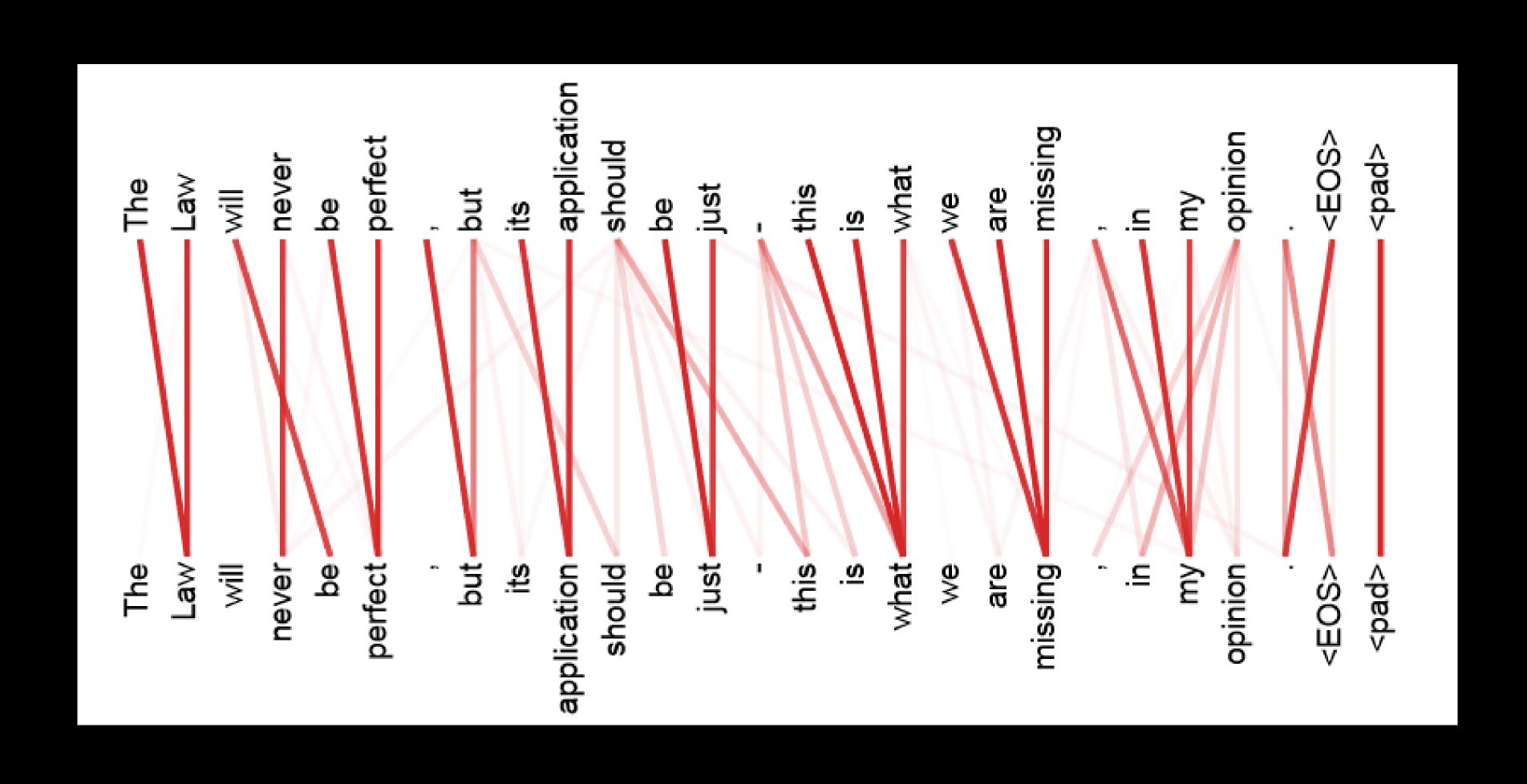


ATTENTION HEAD EXAMPLE 1



Vaswani, A., Shazeer, N., Parmar, N., Uszkoreit, J., Jones, L., Gomez, A. N., Kaiser, L., & Polosukhin, I. (2017). Attention Is All You Need. *ArXiv:1706.03762 [Cs]*. http://arxiv.org/abs/1706.03762

ATTENTION HEAD EXAMPLE 2



PERREVIEW BREAKOUT

- Which comparable problems and solutions did you find in the literature?
- What are the specific dataset characteristics (biases/ limitations)?
- What is your baseline model?
- What is your evaluation metric?

BREAKOUT ROOMS

Room 1

- Klassifikation von Antwort-Mails hinsichtlich Höflichkeit und ggf. hinsichtlich von fachlichen Kriterien (Chris und Sabrina)
- Sentiment-Analyse & Themen-Tagging von Nachrichtenartikeln (Leon, Jule, Jonas, Laura, Awais)

Room 2

- Vorhersage der Beantwortungsschwierigkeiten von Aufgaben (Karo, Sina)
- SHU-T: Generierung von Antworten auf Hass-Artikel (Martin)
- Paraphrasing Texts (Peyman)

HUGGING FACE COURSE

- Token classification
- Fine-tuning a masked language model
- Translation
- Summarization
- Training a causal language model from scratch
- Question answering

EVALUATION METRICS FOR LANGUAGE MODELS

BLUE Score

ROUGE Score

(Log-) Perplexity

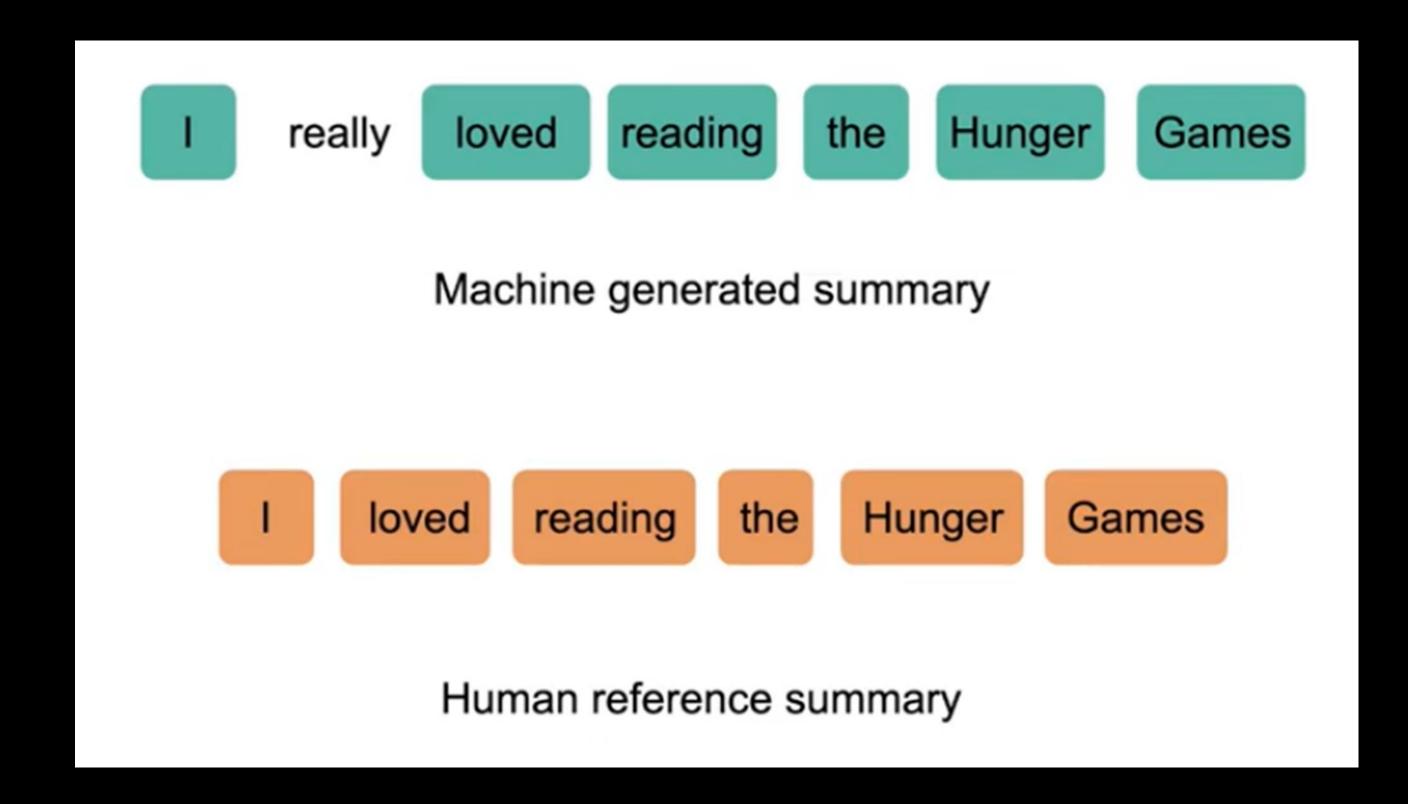
the BLEU metrica





with Lewis

ROUGE-1 SCORE



$$\frac{\text{ROUGE-1}}{\text{recall}} = \frac{\text{Num word matches}}{\text{Num words in reference}} = \frac{6}{6}$$

$$\frac{\text{ROUGE-1}}{\text{precision}} = \frac{\text{Num word matches}}{\text{Num words in summary}} = \frac{6}{7}$$

$$\frac{\text{ROUGE-1}}{\text{F1-score}} = 2\left(\frac{\text{precision} \cdot \text{recall}}{\text{precision} + \text{recall}}\right)$$

ROUGE-2 SCORE

I really

really loved

loved reading

reading the

the Hunger

Hunger Games

loved

loved reading

reading the

the Hunger

Hunger Games

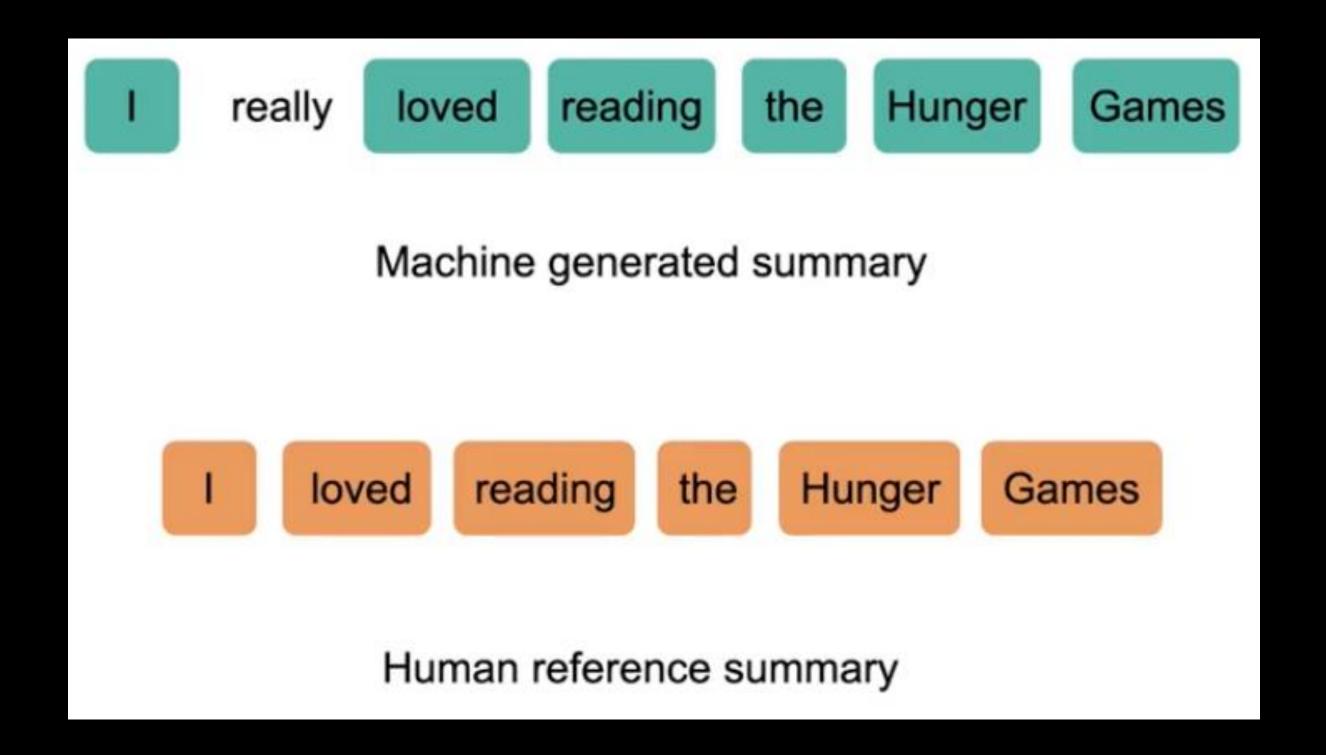
Generated summary bigrams

Reference summary bigrams

ROUGE-2 = Num bigram matches recall =
$$\frac{1}{100}$$
 Num bigrams in reference = $\frac{4}{5}$

ROUGE-2 = Num bigram matches precision =
$$\frac{1}{6}$$

ROUGE-L SCORE



ROUGE-L recall =
$$\frac{LCS(gen, ref)}{Num \text{ words in reference}} = \frac{6}{6}$$

ROUGE-L recision = $\frac{LCS(gen, ref)}{Num \text{ words in summary}} = \frac{6}{7}$

LIKELIHOOD OF A SEQUENCE

$$P(X) = \prod_{i=0}^{t} p(x_i \mid x_{< i})$$

Hugging Face is a startup based in New York City and Paris

p(word|context)

CROSS-ENTROPY

$$CE(X) = -\frac{1}{t}\log P(X)$$

LOG-PERPLEXITY

$$PPL(X) = e^{CE(X)} = e^{-\frac{1}{t} \sum_{i=0}^{t} \log p(x_i | x_{< i})}$$

BREAKOUT DISKUSSION

 Angenommen, Ihr designt einen Prompt für ein GPT-Modell, um nicht gegenderte Texte in gegenderte Text zu überführen.

Wie überprüft Ihr die Qualität der generierten Texte?

TODOS BIS ZUM NÄCHSTEN TERMIN

- Video 3 und Video 4 der Rasa Serie zu Attention und Transformers
- Weiter am Projekt arbeiten.
 Versucht insbesondere unterschiedliche Evaluationsmetriken zu betrachten!