Decoding Strategies

Decoding Hyperparameters & Practical considerations

Learning goals

- Learn how to use the generate() function of the Transformers library
- Learn how to choose from the different decoding strategies with the choice of hyperparameters
- See how hyperparameters affect the output of the generate() function

GENERATING TEXT WITH LLMs

In order to generate text with a generative language model you have to use their built in generate() method:

- You need a tokenized input prompt
- This will be the input to the model
- Then you specify the hyperparameters of the models' generate() method to control the output length and to choose the desired decoding strategy

GREEDY SEARCH

The default decoding strategy is **greedy search**, if all the default hyperparameters are used

- model.generate(tokenized_prompt)
- Output: Once upon a time, the world was a place of great beauty and great danger. The world was a place of great danger, and the world was a place of great danger. The world was a place of great danger, and the world was a place of great danger.

BEAM SEARCH

In order to use **beam search** you simply have to add the num_beams argument to the generate() method and set it to a value > 1

- model.generate(tokenized_prompt, num_beams=5)
- Once upon a time, it was said, there would be a time when the world would be a better place. It was a time when the world would be a better place. It was a time when the world would be a better place. It was a time when the world would be a better place. It was a time when the world would be a better place.

SAMPLING WITH TEMPERATURE (1)

In order to activate **sampling** you have to set do_sample=True. To additionally use **temperature** you have to set the temperature. Also top_k has to be set to 0, as 50 is the default value

- model.generate(tokenized_prompt, do_sample=True, temperature=0.7, top_k=0)
- Output: Once upon a time, we could have seen ourselves as a
 modern day, modern version of ourselves. Ken, who was
 assembled with great courage and bravery to fight for the cause of
 women's reproductive rights, has been doing so for more than 15
 years, and his career is full of inspiring stories. He is a poet, a civil
 rights leader, a hard-nosed activist and a true American hero.

SAMPLING WITH TEMPERATURE (2)

If we set temperature to a super high value, the output distribution will approximate a uniform distribution. For temperature \rightarrow 0 the output distribution will have all the probability mass in the most probable token and generate() will be equivalent to **greedy search**

Prompt: "Once upon a time"

- model.generate(tokenized_prompt, do_sample=True, temperature=10000000.0, top_k=0)
- Output: Once upon a time indicators Gran slew extrater rockedIA moaningJohn Chin knocking transgender Button nin forgiving submarine traveller efforts Pascal despairaways Hem Melvin Toryunn Amb velvet

This output is as expected since with such a high temperature every token in the vocabulary will have the same probability

TOP-K SAMPLING

In order to activate **top-k sampling** you have to set do_sample=True and set top_k to a integer > 0. If you use top_k=1 it will again be equivalent to **greedy search** as only the most probable token is chosen

Prompt: "Once upon a time"

- model.generate(tokenized_prompt, do_sample=True, top_k=10)
- Output: Once upon a time, he did not think he had a chance and turned to her and said:
 - "'My dear, what do you mean by that? I know your father, but he is dead.'

"So she said:

'I knew it. I know your father, but I do not know his whereabouts. You know my mother.'

TOP-P SAMPLING

Set top_p to some float and do_sample=True to use **top-p sampling**. Using a very small value is equal to using **greedy search**. Using top_p=1.0 will consider all tokens

- model.generate(tokenized_prompt, do_sample=True, top_p=0.7, top_k=0)
- Output: Once upon a time they arrived at the well of water, they saw that their creature was burning in flames. After they had brought their clothes back to the dry land, they were found that they had died in the burning body.

CONTRASTIVE SEARCH

For **contrastive search** you have to set penalty_alpha to a float and set top_k as previously mentioned. Remember we are **not** sampling here. For penalty_alpha=0 we only maximize the model confidence term and it becomes **greedy search** again. Using penalty_alpha=1 we only care about the degeneration penalty term

- model.generate(tokenized_prompt, do_sample=True, penalty_alpha=0.6, top_k=20)
- Output: Once upon a time, the man who had been the object of the most intense and most intense hatred, the most intense and most intense hatred, was the man who had been the object of the most intense and most intense hatred, and the man who had been the object of the most intense and most intense hatred.

CONCLUSION

- In practice you wouldn't use one decoding strategy in isolation
- You would have to play around with the different hyperparameters and decoding strategies until you get a good result
- But how do we determine if a generated text is good or not?
- For that there are several evaluation metrics → next chapter