

ADL HW2 Report

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1. Model

- Architecture:

It inputs with text and also output the text. It is a seq2seq model. It is multilingual transformer that supports many different languages and have also been pretrained on mC4 datasets. It is a transformers that contained both encode and decode stages. The decode strategies also affects the result of the prediction.

mT5 is the model based on the T5. It improves upon T5 by using GeGLU nonlinearities, scaling both dmodel and dff. It doesn't do the dropout. One of its great benefit is that mT5 can perform zero-shot learning.

- Text Summarization:

Text summarization needs to condense the long piece text into a shorter but also meaningful text to represent the whole content. The model needs to understand and know the difference between the input words and generate a coherent summary. The mT5 has been pretrained on multi languages so it can easily know some of the common phrases or words. Then I finetuned with the model to make it have a better fit on the news article from udn.com.

2. Preprocess

- Tokenization:

I use the t5 pretrained tokenizer. Its split character into groups and transform the group into index id based on the t5 tokenizer library. Since we have set the max_source_length and max_target_length, the tokenizer will also pad the encoded sequence with meaningless index to fill up the length or just truncate the exceeded length.

- Data Cleaning:

I have removed other unrelated columns in the datasets. Only leaves the "maintext" and "title". Other columns are removed in the tokenization.

3. Hyperparameter

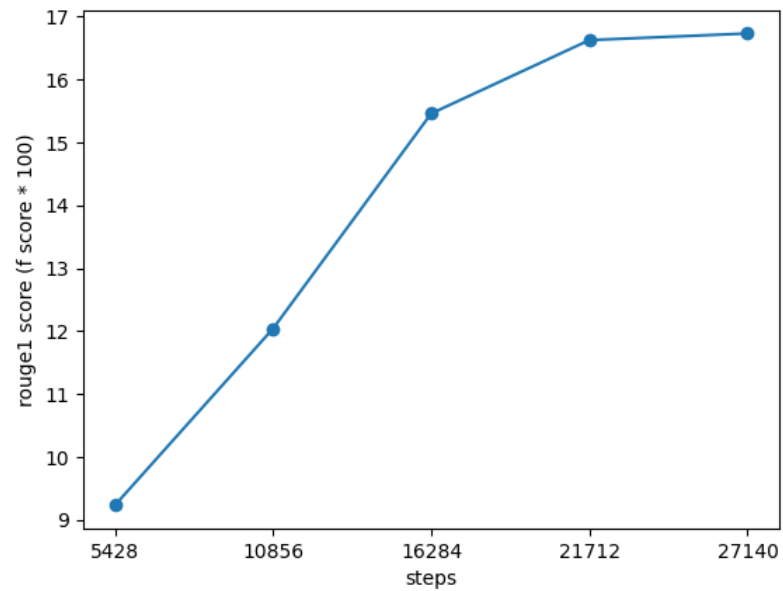
- batch size: 2*2
- max_source_length: 1024
- max_target_length: 128
- epochs: 5
- warmup steps: 300
- learning rate: 5e-4
- optimizer: AdamW
- lr_scheduler: linear

The prediction evaluation on public.jsonl file is {rouge1: 0.25517, rouge2: 0.10395, rougeL: 0.23007}

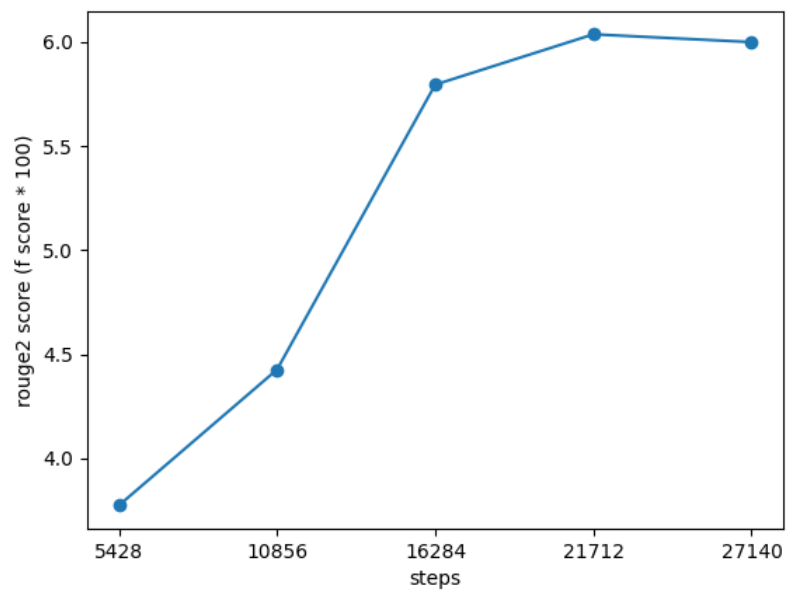
4. Learning Curve

I plotted these rouge scores based on the `evaluate("rouge")` metric score which is quite different with the `tw_rouge` score.

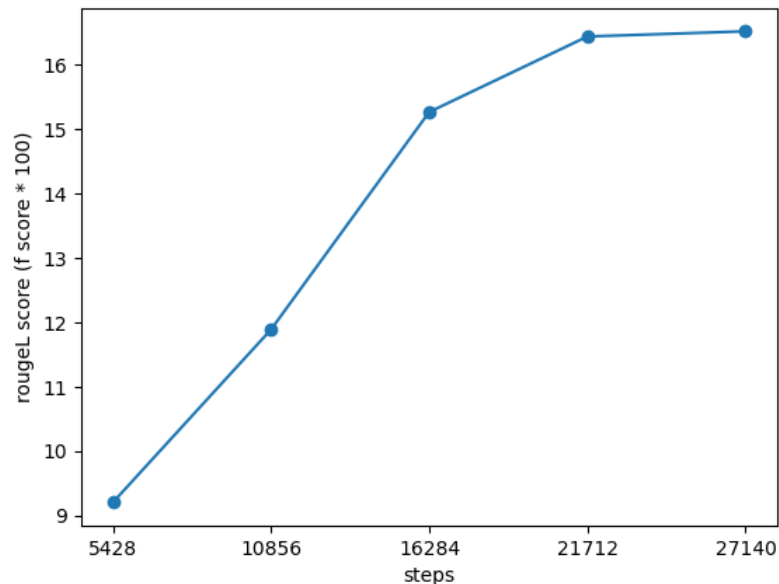
Rouge 1 score curve:



Rouge 2 score curve:



Rouge 3 score curve:



5. Strategies

- Greedy

Greedy strategy selected **one** word with the **highest probability** at each step.

- Beam Search (beam size: n)

Beam search strategy keeps the **top n candidates** with the highest probability in each step. After all the steps, it chooses the sequence with the **highest probability**. The probability is calculated by multiplying the probability of each word in the sequence. So the beam search results in high probability of short length output.

- Top-K (size: k)

Top-K also selects one word in each step. However, it chooses the word by **sampling** from the **top k words** with the high probability.

- Top-P (probability: p)

Similar to the top k strategy. Top-P also **samples** one word in each step. The difference is that it samples from the top n words that have the **probability higher than p**.

- Temperature

It is used with the sampling strategies. It makes the sampling task not so randomly. It **increases the probability** of the higher potential one and decreases the probability of the lower potential one.

6. Generation Strategy Hyperparameters

The text summarization is more based on the original text. Also, it is actually already determined with lower fraction of creativity. That means, the sampling strategy can't help much with our prediction.

In this case, I choose the beam search decode strategy. Trying with different beam size, I found out that **num_beam=5** is the best for my model.

These results are predicted with the public.jsonl file and the model with $5e-5$ learning rate.

	rouge1	rouge2	rougeL
greedy	24.532	9.538	22.085
beam (n=3)	24.642	9.766	22.103
beam (n=5)	24.508	9.721	21.958
beam (n=7)	23.043	8.333	20.625
top k (k=10)	20.281	6.333	17.913
top k (k=20)	19.099	5.888	16.875
top p (p=0.8)	19.589	6.269	17.393
top p (p=0.9)	18.978	6.059	15.815
top p (p=0.8) + tempertature (0.5)	22.549	8.076	20.179
top p (p=0.8) + tempertature (1.0)	19.843	6.506	17.715