This document explains in more details about how functions work, how some of the parameters are configured, etc.

PREPROCESSING:

```
\PiThe code commented out(df_.text = 'summarize: ' + df_.text) is
required only for t5, because t5 has a unified model structure for
various downstream tasks and the user must specify the need by
adding a keyword.
2train test split(df ,test size=0.1,shuffle=True, random state=42)
""" Split dataset into training set and test set."""
Args:
    *arrays:
    dataset, vaild for lists, numpy arrays, spcipy-sparse matrices or pandas
    dataframes.
     test size:
    a ratio that indicates the proportion of test set.
     shuffle:
     a Boolean value, True by default.
     random state:
     an option whether preserves the same shuffle at each run. When a number is
    given, then the results become reproducible.
Returns:
     List containing train-test split of inputs.
MODELS:
1 tokenizer.batch encode plus(train['text'], return_tensors='pt', max
_length=max_length, pad_to_max_length=True, truncation=True):
""" Encode the input sequence"""
Args:
    text:
    a list of sequence to be encoded.
    return tensors:
    If set, will return tensors. 'pt' represents torch. Tensor objects.
    max length:
```

If not set, it will use the predefined max_length of model. Here it is still set in order to make it clear - for bart max_length=1024, for t5 max_length=512 and for gpt-2 it is set to be 1024 although there's no restriction.

pad to max length:

A Boolean value, if set True, then the input sequence will be padded to max length.

truncation:

A Boolean value, if set \mbox{True} , the input sequence will be truncated to \mbox{max} length.

Returns:

A dictionary of input ids, attention masks, token type ids.

2def train(dataset, lr, eps, batch_size, epochs, model, tokenizer, d evice):

""" Train GPT2 model and print out training details."""

Args:

dataset:

training data, it's a 2-dim tensor, i.e. [4500,1024] represents 4500 samples in total, and each of them has 1024 dims.

lr:

learning rate, set as a rather small number 1e-05 to avoid large change on the weights, default value is 1e-03.

eps:

term added to the denominator to improve numerical stability, set by default 1e-08.

batch_size:

number of samples in a batch for one backpropagation, set to be 2 but it can be larger if GPU allows.

epochs:

number of iterations over the whole input data, set by 5 due to GPU limits, but a larger number would be better.

model:

the model for training, it's loaded as a pre-trained model from the hugging face's libraries.

tokenizer:

the tokenizer for generating tokens, it's also directly loaded from the huggingface's libraries.

device:

an option for GPU.

Returns:

a list containing average training loss for each epoch.

[*Note: those hyperparameters on training() are set the same way among different language models for simplicity, but it's worth exploring different optimal settings for each of the models.]

3 model.generate(input_ids, attention_mask, decoder_start_token_id,
max_length, min_length, num_beams, no_repeat_ngram_size):
 """ Generates sequence for models with a LM head."""

Args:

input ids, attention mask, decoder start token id:

input_ids, attention_mask and decoder_start_token_id of the input tokens, directly obtained from the tokenizing step. for gpt-2, it only has input_ids, but for encoder-decoder structure like t5 and bart, all three are required.

max_length, min_length:

max and min length of the generated sequence, but from my experiments, these constraints are not accurate, thus for a generated summary of length around 100, different max_length and min_length are set for different models and they are all chosen by the observations of limited experiments. num beams:

Number of beams for beam search. 1 means no beam search. Default to 1. It's set to be 3 for all models when apply beam search. no repeat ngram size:

the number defines that the ngram can only appear once. For gpt-2 and bart, the value is set to be 2 but for t5 it's 3. The reason is that when it's 2, t5 outputs bad summaries and only with 3 it performs in a normal way.

Returns:

`torch.LongTensor` of shape `(batch_size * num_return_sequences,
sequence length)`

[*Note: Only greedy search and beam search are discussed and compared in the project, thus many other settings are absent, but it would be interesting to also include method like top-k sampling. Besides, there are also parameters like repetition_penalty, length_penalty to penalize on word repetitions or sequence length which can make an difference on the results.]

[source code from

https://github.com/huggingface/transformers/blob/c4d4e8bdbd25d9463d41de6398940329c89b7fb6/src/transformers/generation_utils.py#L101

more explanations please refer to https://huggingface.co/blog/how-to-generate]

EVALUATIONS:

```
1 datasets.load metric('rouge'):
""" Provide evaluation metrics for NLP tasks."""
Args:
     metric name:
     name of the metric chosen from bertscore, bleu, bleurt, coval, gleu, glue, meteor, rouge, sacrebleu, sequal, squad, squad_v2, xnli.
Returns:
    the metric object.
metric.compute(rouge types=['rouge1',"rouge2","rouge3",'rougeL']):
""" Compute the metric value."""
Args:
     rouge types:
     a list of the rouge types that to be returned.
Returns:
     A dictionary containing all specified touge types. To access the rouge-1,
     use results["rouge1"], to access the recall of rouge-1, use
     results["rouge1"].mid.recall
```