

README

November 15, 2021

System Info: Tested on Ubuntu 20.04 or Ubuntu 18.04

1 Step 1: Set up environment

Note: The commands below in step 1 need to be run in a terminal. Then in the created anaconda environment, start jupyter notebook and open this README.ipynb, then the commands in the following steps can be directly run in this notebook

1.1 create an Anaconda environment, with a name e.g. memsum

Note: the symbol “!” is used to run command in jupyter notebook. In a terminal, “!” is not needed.

Note: Without further notification, the following commands need to be run in the working directory where this jupyter notebook is located.

```
[ ]: !conda create -n memsum python=3.7
```

1.2 activate this environment

```
[ ]: !source activate memsum
```

1.3 In the created anaconda environment, install jupyter notebook and ipython to run this code

```
[ ]: !conda install ipython  
!conda install -c anaconda jupyter
```

```
[ ]:
```

2 Setp 2: Install dependencies, download word embeddings and load them to pretrained model

2.1 Install dependencies via pip

```
[1]: !pip install -r requirements.txt
```

2.2 Install pytorch (GPU version)

You need to specify your CUDA version correctly before installing pytorch. If CUDA version is 11.3 then we can use the following command to install pytorch

```
[2]: !conda install pytorch cudatoolkit=11.3 -c pytorch -y
```

We provide a trained MemSum model on PubMed dataset. In order to use this model, we need to download the pretrained GLOVE word embedding from the official website and add them to MemSum using the following script.

This command takes time, as we need to first download and the unzip GloVe embeddings.

```
[3]: !python download_and_load_word_embedding.py
```

```
--2021-11-15 03:30:18-- https://nlp.stanford.edu/data/glove.6B.zip
Resolving proxy.ethz.ch (proxy.ethz.ch)... 129.132.202.155
Connecting to proxy.ethz.ch (proxy.ethz.ch)|129.132.202.155|:3128... connected.
Proxy request sent, awaiting response... 301 Moved Permanently
Location: http://downloads.cs.stanford.edu/nlp/data/glove.6B.zip [following]
--2021-11-15 03:30:19-- http://downloads.cs.stanford.edu/nlp/data/glove.6B.zip
Connecting to proxy.ethz.ch (proxy.ethz.ch)|129.132.202.155|:3128... connected.
Proxy request sent, awaiting response... 200 OK
Length: 862182613 (822M) [application/zip]
Saving to: 'glove.6B.zip'
```

```
glove.6B.zip      100%[=====>] 822.24M  5.10MB/s   in 2m 50s
```

```
2021-11-15 03:33:09 (4.84 MB/s) - 'glove.6B.zip' saved [862182613/862182613]
```

```
Archive:  glove.6B.zip
  inflating: glove.6B.50d.txt
  inflating: glove.6B.100d.txt
  inflating: glove.6B.200d.txt
  inflating: glove.6B.300d.txt
All model loaded!
```

3 Step 3: Testing trained model on a given dataset

For example, the following command test the performance of the full MemSum model, on the Pubmed's test set. The model is evaluated by ROUGE 1/2/L's precision, recall and F1 scores.

```
[4]: !python my_test.py -model_type MemSum_Final -summarizer_model_path model/
↳MemSum_Final/pubmed_full/200dim/best/model.pt -vocabulary_path model/glove/
↳vocabulary_200dim.pkl -corpus_path data/pubmed_full/test_PUBMED.jsonl -gpu 0
↳-max_extracted_sentences_per_document 7 -p_stop_thres 0.6 -output_file
↳results/MemSum_Final/pubmed_full/200dim/test_results.txt -max_doc_len 500
↳-max_seq_len 100
```

Start Computation ...

100it [00:09, 10.68it/s]

p_stop_thres: 0.6000, avg. # sentences: 6.24 ± 1.10, avg. extraction time: 69.72 ± 6.16 ms, R-1 (p,r,f1): 0.4878, 0.5433, 0.4969 R-2: 0.2296, 0.2500, 0.2323

R-L: 0.4418, 0.4907, 0.4493

Due to the size limit of the appendix, we only provided the trained MemSum on the PubMed dataset, and we only provide the first 100 training/validation/testing examples for the PubMed, arXiv and GovReport datasets.

In the future, we will release all the trained models and full datasets used in our experiments.

4 Step 4: Use the pretrained summarizer as a module

4.1 load the full MemSum model

```
[5]: from my_summarizers import ExtractiveSummarizer_MemSum_Final

memsum_model = ExtractiveSummarizer_MemSum_Final(
    "model/MemSum_Final/pubmed_full/200dim/best/model.pt",
    "model/glove/vocabulary_200dim.pkl",
    gpu = 0,
    embed_dim = 200,
    max_doc_len = 500,
    max_seq_len = 100
)
```

4.2 Get a document to be summarized

The format of the document to be summarized is a list of sentences

```
[6]: import json

database = [ json.loads(line) for line in open( "data/pubmed_full/test_PUBMED.
→jsonl" ).readlines() ]
pos = 6
document = database[pos]["text"]
gold_summary = database[pos]["summary"]
```

```
[7]: print(document[:5])
```

```
['the family is the cornerstone of human social support network and its presence
is essential in everyone s life . changes inevitably occur in families with
illness and hospitalization of a family member . in other words , among the
sources of stress for families are accidents leading to hospitalization
particularly intensive care unit ( icu ) .', 'statistics show that 8% of
hospital beds in the united states are occupied by the intensive care units .',
```


patients in the neurosurgery intensive care unit of al - zahra hospital in 2010 .', 'in this study , it was shown that the use of nursing interventions based on family needs (confidence , support , information , proximity and convenience) had significant impact on the family satisfaction of the patient hospitalized in intensive care unit .', 'the statistical research community was the families of hospitalized patients in neurosurgery intensive care unit of al - zahra (sa) hospital , isfahan , iran from may to september 2010 .', 'the aim of this study was to analyze the satisfaction of the families of icu patients .', 'the results of the present study showed that the nursing interventions based on the family needs increased the patient s family satisfaction in the neurosurgery intensive care unit of al - zahra hospital .', 'comparison of mean satisfaction score (100 *) of participants in the intervention and control groups the mean of satisfaction score changes of the studied subjects in the intervention and control groups after intervention', 'the mean satisfaction score in the intervention group after the intervention was significantly higher than before the intervention (p < 0.001) .']

4.4 Evaluate the extracted summary via ROUGE scores

```
[11]: from rouge_score import rouge_scorer
rouge_cal = rouge_scorer.RougeScorer(['rouge1', 'rouge2', 'rougeLsum'],
→use_stemmer=True)
print(rouge_cal.score( "\n".join(gold_summary), "\n".join(extracted_summary) ))
```

```
{'rouge1': Score(precision=0.6517412935323383, recall=0.4833948339483395,
fmeasure=0.5550847457627119), 'rouge2': Score(precision=0.36,
recall=0.26666666666666666, fmeasure=0.30638297872340425), 'rougeLsum':
Score(precision=0.6069651741293532, recall=0.45018450184501846,
fmeasure=0.5169491525423728)}
```

5 Step 5: Training model from script

For example, if we want to train the full MemSum model on the PubMed dataset, we first change working directory to “src/MemSum_Final/”, then run the python script “train.py”. The train.py takes one parameter: config_file_path, which is the path to the training configuration file.

In the configuration file there are detailed list of key-value pairs that configure the training procedure. For example, the number of GPU devices, batch size, learning rate, etc.

```
[ ]: !cd src/MemSum_Final/; python train.py -config_file_path config/pubmed_full/
→200dim/run0/training.config
```

```
[ ]:
```