Import Library

```
In []: from torch.utils.data import DataLoader from torch.utils.data import DataLoader import warnings, transformers, logging, torch from transformers import TrainingArguments, Trainer from transformers import AutoModelForSequenceClassification, AutoTokenizer from datasets import load_dataset, Dataset, DatasetDict import datasets import pandas as pd import numpy as np
```

/workspace/miniconda3/envs/n1p/1ib/python3.7/site-packages/tqdm/auto.py:22: TqdmWarnin g: IProgress not found. Please update jupyter and ipywidgets. See https://ipywidgets.re adthedocs.io/en/stable/user_install.html

from .autonotebook import tqdm as notebook_tqdm

```
In [ ]: data = pd. read_csv('/workspace/xxl/train.csv')
    data. head(5)
```

Out[]:		discourse_id	essay_id	discourse_text	discourse_type	discourse_effectiveness
	0	0013cc385424	007ACE74B050	Hi, i'm Isaac, i'm going to be writing about h	Lead	Adequate
	1	9704a709b505	007ACE74B050	On my perspective, I think that the face is a	Position	Adequate
	2	c22adee811b6	007ACE74B050	I think that the face is a natural landform be	Claim	Adequate
	3	a10d361e54e4	007ACE74B050	If life was on Mars, we would know by now. The	Evidence	Adequate
	4	db3e453ec4e2	007ACE74B050	People thought that the face was formed by ali	Counterclaim	Adequate

Load Model

Hugging Face Models: https://huggingface.co/models?sort=downloads&search=deberta

```
In [ ]: tokz = AutoTokenizer.from_pretrained('microsoft/deberta-v3-small')
```

Special tokens have been added in the vocabulary, make sure the associated word embeddings are fine-tuned or trained.

/workspace/miniconda3/envs/nlp/lib/python3.7/site-packages/transformers/convert_slow_to kenizer.py:435: UserWarning: The sentencepiece tokenizer that you are converting to a f ast tokenizer uses the byte fallback option which is not implemented in the fast tokenizers. In practice this means that the fast version of the tokenizer can produce unknown tokens whereas the sentencepiece version would have converted these unknown tokens into a sequence of byte tokens matching the original piece of text.

"The sentencepiece tokenizer that you are converting to a fast tokenizer uses the byt e fallback option"

Special tokens have been added in the vocabulary, make sure the associated word embeddings are fine-tuned or trained.

```
In []: # 查看该模型的句间连接词是什么(这个地方理解可能有误)
sep = tokz. sep_token
sep
'[SEP]'
```

Data Processing

Out[]:

```
In []: # 将议论元素的类型和文本内容拼接在一起 data['inputs']= data. discourse_type + sep +data. discourse_text data. head(5)
```

rs is a natural landform or if there is life on Mars that made it. The story is about h ow NASA took a picture of Mars and a face was seen on the planet. NASA doesn't know if

the landform was created by life on Mars, or if it is just a natural landform.

```
# 这里的 tokz 是前面tokz = AutoTokenizer.from_pretrained('microsoft/deberta-v3-small')
       # 是为了将内容分词, 然后变成序列化的数字
       def tok func(x): return tokz(x["inputs"], truncation=True)
In [ ]: # 举个例子,将 data_1 的第一个数据使用 tokenizer 变成序列化的数字
       tok func (data 1[0])
       {'input_ids': [1, 8380, 2, 2684, 261, 584, 280, 358, 11759, 261, 584, 280, 358, 446, 26
Out[]: { Input_las . [1, 6560, 2, 2661, 261, 661, 262, 266, 1008, 1311, 9930, 289, 337, 343, 4, 282, 898, 314, 361, 291, 812, 277, 7583, 269, 266, 1008, 1311, 9930, 289, 337, 343,
       269, 432, 277, 7583, 272, 412, 278, 260, 279, 697, 269, 314, 361, 7658, 681, 266, 1355,
       265,\ 7583,\ 263,\ 266,\ 812,\ 284,\ 757,\ 277,\ 262,\ 3568,\ 260,\ 7658,\ 702,\ 280,\ 297,\ 391,\ 337,
       262, 1311, 9930, 284, 994, 293, 432, 277, 7583, 261, 289, 337, 278, 269, 348, 266, 100
       # 删除不必要的列,只需要保留经过序列化的 'input_ids' 和 'label'列
       # 'input_ids', 'token_type_ids', 'attention_mask' 是经过 tokenizer 后自动生成的列(具体
       # map 函数就是传入一个函数参数,然后原来的数据集按照这个函数变化
       inps = "discourse_text", "discourse_type"
       data_2 = data_1.map(tok_func, batched=True, remove_columns=inps+('inputs', 'discourse_i
       \mathtt{data}\_2
       100% | 37/37 [00:01<00:00, 19.07ba/s]
Out[]:
          features: ['label', 'input ids', 'token type ids', 'attention mask'],
          num rows: 36765
       })
       Split Dataset
      # 选取随机种子,目的是为了每次跑这个代码时,产生的随机数是相同的,保持复现结果相同
In [ ]:
       # 打乱数据集
       essay_ids = data.essay_id.unique()
       np. random. seed (42)
       np. random. shuffle (essay ids)
       essay_ids[:10]
      array(['B5C606F0A883', 'FA4FE7706A1A', '37A77BEAD718', '0ED28D8A5EC4', 'F25BA634ADDD', '1065173BBE31', '763EF698F56B', '1B976DF43007', '9922F6B9A55B', 'A187E6D70752'], dtype=object)
Out[ ]:
In [ ]: # 确定测试集、训练集比例,制作训练集、测试集的索引
       val\_prop = 0.2
       val_sz = int(len(essay_ids)*val_prop)
       val_essay_ids = essay_ids[:val_sz]
       is_val = np. isin(data. essay_id, val_essay_ids)
       idxs = np. arange(len(data))
       val_idxs = idxs[ is_val]
trn_idxs = idxs[~is_val]
       print(len(val_idxs), len(trn_idxs))
       print(val idxs, trn idxs)
       7181 29584
       [ 53 54
                  55 ... 36762 36763 36764] [ 0
                                                 1
                                                      2 ... 36757 36758 36759]
       Get Finally Dataset
In [ ]: data_3 = DatasetDict({"train":data_2.select(trn_idxs),
                  "test": data_2.select(val_idxs)})
In [ ]: data_3
```

In []: # 定义一个函数,为后面调用 dataset.map

```
Out[]: DatasetDict({
            train: Dataset({
                features: ['label', 'input_ids', 'token_type_ids', 'attention_mask'],
                num rows: 29584
            })
            test: Dataset({
                features: ['label', 'input_ids', 'token_type_ids', 'attention_mask'],
                num rows: 7181
            })
        })
In [ ]: | type(data_3['train'][0]['input_ids'])
        list
Out[]:
        Set A Function
In [ ]: # 上面这些步骤我们可以编写一个函数来统一调用
        def get_data(data, train=True):
            ds = Dataset. from_pandas(data)
            to_remove = ['discourse_text', 'discourse_type', 'inputs', 'discourse_id', 'essay_id']
            data_out = ds.map(tok_func, batched=True, remove_columns=to_remove)
            if train:
                return DatasetDict({"train":data_out.select(trn_idxs), "test": data_out.select
                return data_out
In []: # 使用函数会和上面结果是一样的
        data_out = get_data(data)
        data out
```

```
100% | 37/37 [00:01<00:00, 20.89ba/s]
DatasetDict({
```

```
Out[]:
              train: Dataset({
                  features: ['label', 'input_ids', 'token_type_ids', 'attention_mask'],
                  num_rows: 29584
             })
             test: Dataset({
    features: ['label', 'input_ids', 'token_type_ids', 'attention_mask'],
                  num_rows: 7181
             })
         })
```