Text Classification Using Transformer Networks (BERT)

Sofía Cantú Talamantes A01571120

Some initialization:

```
In [1]:
        import random
        import torch
        import numpy as np
        import pandas as pd
        from tqdm.notebook import tqdm
        # enable tqdm in pandas
        tqdm.pandas()
        # set to True to use the gpu (if there is one available)
        use_gpu = True
        # select device
        device = torch.device('cuda' if use_gpu and torch.cuda.is_available() else
        print(f'device: {device.type}')
        # random seed
        seed = 1122
        # set random seed
        if seed is not None:
            print(f'random seed: {seed}')
            random.seed(seed)
            np.random.seed(seed)
            torch.manual_seed(seed)
```

device: cuda
random seed: 1122

Read the train/dev/test datasets and create a HuggingFace Dataset object:

```
In [2]: def read_data(filename):
    # read csv file
    df = pd.read_csv(filename, header=None)
    # add column names
    df.columns = ['label', 'title', 'description']
    # make labels zero-based
    df['label'] -= 1
    # concatenate title and description, and remove backslashes
    df['text'] = df['title'] + " " + df['description']
    df['text'] = df['text'].str.replace('\\', ' ', regex=False)
```

return df

In [3]: #labels = open('data/ag_news_csv/classes.txt').read().splitlines()
 #train_df = read_data('data/ag_news_csv/train.csv')
 #test_df = read_data('data/ag_news_csv/test.csv')
 labels = open('/kaggle/input/ag-news-dataset/data/ag_news_csv/classes.txt').
 train_df = read_data('/kaggle/input/ag-news-dataset/data/ag_news_csv/train.c
 test_df = read_data('/kaggle/input/ag-news-dataset/data/ag_news_csv/test.csv
 train_df

Out[3]:	label	title	description	text
	0 2	Wall St. Bears Claw Back Into the Black (Reuters)	Reuters - Short-sellers, Wall Street's dwindli	Wall St. Bears Claw Back Into the Black (Reute
	1 2	Carlyle Looks Toward Commercial Aerospace (Reu	Reuters - Private investment firm Carlyle Grou	Carlyle Looks Toward Commercial Aerospace (Reu
	2 2	Oil and Economy Cloud Stocks' Outlook (Reuters)	Reuters - Soaring crude prices plus worries\ab	Oil and Economy Cloud Stocks' Outlook (Reuters
	3 2	Iraq Halts Oil Exports from Main Southern Pipe	Reuters - Authorities have halted oil export\f	Iraq Halts Oil Exports from Main Southern Pipe
	4 2	Oil prices soar to all- time record, posing new	AFP - Tearaway world oil prices, toppling reco	Oil prices soar to all-time record, posing new
11999	5 0	Pakistan's Musharraf Says Won't Quit as Army C	KARACHI (Reuters) - Pakistani President Perve	Pakistan's Musharraf Says Won't Quit as Army C
11999	6 1	Renteria signing a top-shelf deal	Red Sox general manager Theo Epstein acknowled	Renteria signing a top- shelf deal Red Sox gene
11999	7 1	Saban not going to Dolphins yet	The Miami Dolphins will put their courtship of	Saban not going to Dolphins yet The Miami Dolp
11999	8 1	Today's NFL games	PITTSBURGH at NY GIANTS Time: 1:30 p.m. Line:	Today's NFL games PITTSBURGH at NY GIANTS Time
11999	9 1	Nets get Carter from Raptors	INDIANAPOLIS All- Star Vince Carter was trad	Nets get Carter from Raptors INDIANAPOLIS A

120000 rows × 4 columns

```
In [4]: from sklearn.model_selection import train_test_split
    train_df, eval_df = train_test_split(train_df, train_size=0.9)
```

```
train_df.reset_index(inplace=True, drop=True)
        eval_df.reset_index(inplace=True, drop=True)
        print(f'train rows: {len(train_df.index):,}')
        print(f'eval rows: {len(eval_df.index):,}')
        print(f'test rows: {len(test_df.index):,}')
       train rows: 108,000
       eval rows: 12,000
       test rows: 7,600
In [5]: from datasets import Dataset, DatasetDict
        ds = DatasetDict()
        ds['train'] = Dataset.from_pandas(train_df)
        ds['validation'] = Dataset.from_pandas(eval_df)
        ds['test'] = Dataset.from_pandas(test_df)
        ds
Out[5]: DatasetDict({
             train: Dataset({
                 features: ['label', 'title', 'description', 'text'],
                 num_rows: 108000
             })
             validation: Dataset({
                 features: ['label', 'title', 'description', 'text'],
                 num_rows: 12000
             })
             test: Dataset({
                 features: ['label', 'title', 'description', 'text'],
                 num rows: 7600
             })
         })
        Tokenize the texts:
```

```
In [6]: !pip install ipywidgets
```

```
Requirement already satisfied: ipywidgets in /opt/conda/lib/python3.10/site-
packages (7.7.1)
Requirement already satisfied: ipykernel>=4.5.1 in /opt/conda/lib/python3.1
0/site-packages (from ipywidgets) (6.29.4)
Requirement already satisfied: ipython-genutils~=0.2.0 in /opt/conda/lib/pyt
hon3.10/site-packages (from ipywidgets) (0.2.0)
Requirement already satisfied: traitlets>=4.3.1 in /opt/conda/lib/python3.1
0/site-packages (from ipywidgets) (5.14.3)
Requirement already satisfied: widgetsnbextension~=3.6.0 in /opt/conda/lib/p
ython3.10/site-packages (from ipywidgets) (3.6.9)
Requirement already satisfied: ipython>=4.0.0 in /opt/conda/lib/python3.10/s
ite-packages (from ipywidgets) (8.21.0)
Requirement already satisfied: jupyterlab-widgets>=1.0.0 in /opt/conda/lib/p
ython3.10/site-packages (from ipywidgets) (3.0.11)
Requirement already satisfied: comm>=0.1.1 in /opt/conda/lib/python3.10/sit
e-packages (from ipykernel>=4.5.1->ipywidgets) (0.2.2)
Requirement already satisfied: debugpy>=1.6.5 in /opt/conda/lib/python3.10/s
ite-packages (from ipykernel>=4.5.1->ipywidgets) (1.8.1)
Requirement already satisfied: jupyter-client>=6.1.12 in /opt/conda/lib/pyth
on3.10/site-packages (from ipykernel>=4.5.1->ipywidgets) (7.4.9)
Requirement already satisfied: jupyter-core!=5.0.*,>=4.12 in /opt/conda/lib/
python3.10/site-packages (from ipykernel>=4.5.1->ipywidgets) (5.7.2)
Requirement already satisfied: matplotlib-inline>=0.1 in /opt/conda/lib/pyth
on3.10/site-packages (from ipykernel>=4.5.1->ipywidgets) (0.1.7)
Requirement already satisfied: nest-asyncio in /opt/conda/lib/python3.10/sit
e-packages (from ipykernel>=4.5.1->ipywidgets) (1.6.0)
Requirement already satisfied: packaging in /opt/conda/lib/python3.10/site-p
ackages (from ipykernel>=4.5.1->ipywidgets) (21.3)
Requirement already satisfied: psutil in /opt/conda/lib/python3.10/site-pack
ages (from ipykernel>=4.5.1->ipywidgets) (5.9.3)
Requirement already satisfied: pyzmq>=24 in /opt/conda/lib/python3.10/site-p
ackages (from ipykernel>=4.5.1->ipywidgets) (26.0.3)
Requirement already satisfied: tornado>=6.1 in /opt/conda/lib/python3.10/sit
e-packages (from ipykernel>=4.5.1->ipywidgets) (6.4.1)
Requirement already satisfied: decorator in /opt/conda/lib/python3.10/site-p
ackages (from ipython>=4.0.0->ipywidgets) (5.1.1)
Requirement already satisfied: jedi>=0.16 in /opt/conda/lib/python3.10/site-
packages (from ipython>=4.0.0->ipywidgets) (0.19.1)
Requirement already satisfied: prompt-toolkit<3.1.0,>=3.0.41 in /opt/conda/l
ib/python3.10/site-packages (from ipython>=4.0.0->ipywidgets) (3.0.47)
Requirement already satisfied: pygments>=2.4.0 in /opt/conda/lib/python3.10/
site-packages (from ipython>=4.0.0->ipywidgets) (2.18.0)
Requirement already satisfied: stack-data in /opt/conda/lib/python3.10/site-
packages (from ipython>=4.0.0->ipywidgets) (0.6.2)
Requirement already satisfied: exceptiongroup in /opt/conda/lib/python3.10/s
ite-packages (from ipython>=4.0.0->ipywidgets) (1.2.0)
Requirement already satisfied: pexpect>4.3 in /opt/conda/lib/python3.10/sit
e-packages (from ipython>=4.0.0->ipywidgets) (4.9.0)
Requirement already satisfied: notebook>=4.4.1 in /opt/conda/lib/python3.10/
site-packages (from widgetsnbextension~=3.6.0->ipywidgets) (6.5.7)
Requirement already satisfied: parso<0.9.0,>=0.8.3 in /opt/conda/lib/python
3.10/site-packages (from jedi>=0.16->ipython>=4.0.0->ipywidgets) (0.8.4)
Requirement already satisfied: entrypoints in /opt/conda/lib/python3.10/sit
e-packages (from jupyter-client>=6.1.12->ipykernel>=4.5.1->ipywidgets) (0.4)
Requirement already satisfied: python-dateutil>=2.8.2 in /opt/conda/lib/pyth
on3.10/site-packages (from jupyter-client>=6.1.12->ipykernel>=4.5.1->ipywidg
```

```
ets) (2.9.0.post0)
```

(1.16.0)

Requirement already satisfied: platformdirs>=2.5 in /opt/conda/lib/python3.1 0/site-packages (from jupyter-core!=5.0.*,>=4.12->ipykernel>=4.5.1->ipywidge ts) (3.11.0)

Requirement already satisfied: jinja2 in /opt/conda/lib/python3.10/site-pack ages (from notebook>=4.4.1->widgetsnbextension~=3.6.0->ipywidgets) (3.1.4) Requirement already satisfied: argon2-cffi in /opt/conda/lib/python3.10/sit e-packages (from notebook>=4.4.1->widgetsnbextension~=3.6.0->ipywidgets) (2 3.1.0)

Requirement already satisfied: nbformat in /opt/conda/lib/python3.10/site-pa ckages (from notebook>=4.4.1->widgetsnbextension~=3.6.0->ipywidgets) (5.1 0.4)

Requirement already satisfied: nbconvert>=5 in /opt/conda/lib/python3.10/sit e-packages (from notebook>=4.4.1->widgetsnbextension~=3.6.0->ipywidgets) (6.4.5)

Requirement already satisfied: Send2Trash>=1.8.0 in /opt/conda/lib/python3.1 0/site-packages (from notebook>=4.4.1->widgetsnbextension~=3.6.0->ipywidget s) (1.8.3)

Requirement already satisfied: terminado>=0.8.3 in /opt/conda/lib/python3.1 0/site-packages (from notebook>=4.4.1->widgetsnbextension~=3.6.0->ipywidget s) (0.18.1)

Requirement already satisfied: prometheus-client in /opt/conda/lib/python3.1 0/site-packages (from notebook>=4.4.1->widgetsnbextension~=3.6.0->ipywidget s) (0.20.0)

Requirement already satisfied: nbclassic>=0.4.7 in /opt/conda/lib/python3.1 0/site-packages (from $notebook>=4.4.1->widgetsnbextension\sim=3.6.0->ipywidget s) (1.1.0)$

Requirement already satisfied: ptyprocess>=0.5 in /opt/conda/lib/python3.10/site-packages (from pexpect>4.3->ipython>=4.0.0->ipywidgets) (0.7.0)
Requirement already satisfied: wcwidth in /opt/conda/lib/python3.10/site-packages (from prompt-toolkit<3.1.0,>=3.0.41->ipython>=4.0.0->ipywidgets) (0.2.13)

Requirement already satisfied: pyparsing!=3.0.5,>=2.0.2 in /opt/conda/lib/py thon3.10/site-packages (from packaging->ipykernel>=4.5.1->ipywidgets) (3.1.2)

Requirement already satisfied: executing>=1.2.0 in /opt/conda/lib/python3.1 0/site-packages (from stack-data->ipython>=4.0.0->ipywidgets) (2.0.1)
Requirement already satisfied: asttokens>=2.1.0 in /opt/conda/lib/python3.1 0/site-packages (from stack-data->ipython>=4.0.0->ipywidgets) (2.4.1)
Requirement already satisfied: pure-eval in /opt/conda/lib/python3.10/site-packages (from stack-data->ipython>=4.0.0->ipywidgets) (0.2.2)
Requirement already satisfied: six>=1.12.0 in /opt/conda/lib/python3.10/site-packages (from asttokens>=2.1.0->stack-data->ipython>=4.0.0->ipywidgets)

Requirement already satisfied: notebook-shim>=0.2.3 in /opt/conda/lib/python 3.10/site-packages (from nbclassic>=0.4.7->notebook>=4.4.1->widgetsnbextensi on \sim =3.6.0->ipywidgets) (0.2.4)

Requirement already satisfied: mistune<2,>=0.8.1 in /opt/conda/lib/python3.1 0/site-packages (from nbconvert>=5->notebook>=4.4.1->widgetsnbextension~= 3.6.0->ipywidgets) (0.8.4)

Requirement already satisfied: jupyterlab-pygments in /opt/conda/lib/python 3.10/site-packages (from nbconvert>=5->notebook>=4.4.1->widgetsnbextension~= 3.6.0->ipywidgets) (0.3.0)

Requirement already satisfied: bleach in /opt/conda/lib/python3.10/site-pack ages (from nbconvert>=5->notebook>=4.4.1->widgetsnbextension~=3.6.0->ipywidg ets) (6.1.0)

Requirement already satisfied: pandocfilters>=1.4.1 in /opt/conda/lib/python 3.10/site-packages (from nbconvert>=5->notebook>=4.4.1->widgetsnbextension~= 3.6.0->ipywidgets) (1.5.0)

Requirement already satisfied: testpath in /opt/conda/lib/python3.10/site-pa ckages (from nbconvert>=5->notebook>=4.4.1->widgetsnbextension~=3.6.0->ipywidgets) (0.6.0)

Requirement already satisfied: defusedxml in /opt/conda/lib/python3.10/site-packages (from nbconvert>=5->notebook>=4.4.1->widgetsnbextension \sim =3.6.0->ipy widgets) (0.7.1)

Requirement already satisfied: beautifulsoup4 in /opt/conda/lib/python3.10/s ite-packages (from nbconvert>=5->notebook>=4.4.1->widgetsnbextension~=3.6.0->ipywidgets) (4.12.3)

Requirement already satisfied: nbclient<0.6.0,>=0.5.0 in /opt/conda/lib/pyth on3.10/site-packages (from $nbconvert>=5->notebook>=4.4.1->widgetsnbextension <math>\sim=3.6.0->ipywidgets$) (0.5.13)

Requirement already satisfied: MarkupSafe>=2.0 in /opt/conda/lib/python3.10/site-packages (from nbconvert>=5->notebook>=4.4.1->widgetsnbextension~= 3.6.0->ipywidgets) (2.1.5)

Requirement already satisfied: fastjsonschema>=2.15 in /opt/conda/lib/python 3.10/site-packages (from nbformat->notebook>=4.4.1->widgetsnbextension~= 3.6.0->ipywidgets) (2.19.1)

Requirement already satisfied: jsonschema>=2.6 in /opt/conda/lib/python3.10/ site-packages (from nbformat->notebook>=4.4.1->widgetsnbextension \sim =3.6.0->ip ywidgets) (4.22.0)

Requirement already satisfied: argon2-cffi-bindings in /opt/conda/lib/python 3.10/site-packages (from argon2-cffi->notebook>=4.4.1->widgetsnbextension~= 3.6.0->ipywidgets) (21.2.0)

Requirement already satisfied: attrs>=22.2.0 in /opt/conda/lib/python3.10/si te-packages (from jsonschema>=2.6->nbformat->notebook>=4.4.1->widgetsnbexten sion~=3.6.0->ipywidgets) (23.2.0)

Requirement already satisfied: jsonschema-specifications>=2023.03.6 in /opt/conda/lib/python3.10/site-packages (from jsonschema>=2.6->nbformat->notebook >=4.4.1->widgetsnbextension~=3.6.0->ipywidgets) (2023.12.1)

Requirement already satisfied: referencing>=0.28.4 in /opt/conda/lib/python 3.10/site-packages (from jsonschema>=2.6->nbformat->notebook>=4.4.1->widgets nbextension~=3.6.0->ipywidgets) (0.35.1)

Requirement already satisfied: rpds-py>=0.7.1 in /opt/conda/lib/python3.10/s ite-packages (from jsonschema>=2.6->nbformat->notebook>=4.4.1->widgetsnbexte nsion~=3.6.0->ipywidgets) (0.18.1)

Requirement already satisfied: jupyter-server<3,>=1.8 in /opt/conda/lib/pyth on3.10/site-packages (from notebook-shim>=0.2.3->nbclassic>=0.4.7->notebook>=4.4.1->widgetsnbextension~=3.6.0->ipywidgets) (2.12.5)

Requirement already satisfied: cffi>=1.0.1 in /opt/conda/lib/python3.10/sit e-packages (from argon2-cffi-bindings->argon2-cffi->notebook>=4.4.1->widgets nbextension~=3.6.0->ipywidgets) (1.16.0)

Requirement already satisfied: soupsieve>1.2 in /opt/conda/lib/python3.10/si te-packages (from beautifulsoup4->nbconvert>=5->notebook>=4.4.1->widgetsnbex tension~=3.6.0->ipywidgets) (2.5)

Requirement already satisfied: webencodings in /opt/conda/lib/python3.10/sit e-packages (from bleach->nbconvert>=5->notebook>=4.4.1->widgetsnbextension~= 3.6.0->ipywidgets) (0.5.1)

Requirement already satisfied: pycparser in /opt/conda/lib/python3.10/site-p ackages (from cffi>=1.0.1->argon2-cffi-bindings->argon2-cffi->notebook>= 4.4.1->widgetsnbextension~=3.6.0->ipywidgets) (2.22)

Requirement already satisfied: anyio>=3.1.0 in /opt/conda/lib/python3.10/sit e-packages (from jupyter-server<3,>=1.8->notebook-shim>=0.2.3->nbclassic>=

```
0.4.7->notebook>=4.4.1->widgetsnbextension\sim=3.6.0->ipywidgets) (4.4.0)
Requirement already satisfied: jupyter-events>=0.9.0 in /opt/conda/lib/pytho
n3.10/site-packages (from jupyter-server<3,>=1.8->notebook-shim>=0.2.3->nbcl
assic>=0.4.7->notebook>=4.4.1->widgetsnbextension~=3.6.0->ipywidgets) (0.1
0.0)
Requirement already satisfied: jupyter-server-terminals in /opt/conda/lib/py
thon3.10/site-packages (from jupyter-server<3,>=1.8->notebook-shim>=0.2.3->n
bclassic>=0.4.7->notebook>=4.4.1->widgetsnbextension~=3.6.0->ipywidgets)
Requirement already satisfied: overrides in /opt/conda/lib/python3.10/site-p
ackages (from jupyter-server<3,>=1.8->notebook-shim>=0.2.3->nbclassic>=
0.4.7--notebook>=4.4.1--widgetsnbextension\sim=3.6.0--ipywidgets) (7.7.0)
Requirement already satisfied: websocket-client in /opt/conda/lib/python3.1
0/site-packages (from jupyter-server<3,>=1.8->notebook-shim>=0.2.3->nbclassi
c = 0.4.7 - \text{notebook} = 4.4.1 - \text{widgetsnbextension} = 3.6.0 - \text{pywidgets}) (1.8.0)
Requirement already satisfied: idna>=2.8 in /opt/conda/lib/python3.10/site-p
ackages (from anyio>=3.1.0->jupyter-server<3,>=1.8->notebook-shim>=0.2.3->nb
classic>=0.4.7->notebook>=4.4.1->widgetsnbextension~=3.6.0->ipywidgets)
Requirement already satisfied: sniffio>=1.1 in /opt/conda/lib/python3.10/sit
e-packages (from anyio>=3.1.0->jupyter-server<3,>=1.8->notebook-shim>=0.2.3-
>nbclassic>=0.4.7->notebook>=4.4.1->widgetsnbextension~=3.6.0->ipywidgets)
(1.3.1)
Requirement already satisfied: typing-extensions>=4.1 in /opt/conda/lib/pyth
on3.10/site-packages (from anyio>=3.1.0->jupyter-server<3,>=1.8->notebook-sh
im > 0.2.3 - nbclassic > 0.4.7 - notebook > 4.4.1 - widgetsnbextension <math>\sim 3.6.0 - ipy
widgets) (4.12.2)
Requirement already satisfied: python-json-logger>=2.0.4 in /opt/conda/lib/p
ython3.10/site-packages (from jupyter-events>=0.9.0->jupyter-server<3,>=1.8-
>notebook-shim>=0.2.3->nbclassic>=0.4.7->notebook>=4.4.1->widgetsnbextension
\sim=3.6.0->ipywidgets) (2.0.7)
Requirement already satisfied: pyyaml>=5.3 in /opt/conda/lib/python3.10/sit
e-packages (from jupyter-events>=0.9.0->jupyter-server<3,>=1.8->notebook-shi
m>=0.2.3->nbclassic>=0.4.7->notebook>=4.4.1->widgetsnbextension~=3.6.0->ipyw
idgets) (6.0.2)
Requirement already satisfied: rfc3339-validator in /opt/conda/lib/python3.1
0/site-packages (from jupyter-events>=0.9.0->jupyter-server<3,>=1.8->noteboo
k-shim >= 0.2.3 - nbclassic >= 0.4.7 - notebook >= 4.4.1 - widgetsnbextension \sim= 3.6.0 - notebook >= 4.4.1 - notebook >= 4.4
>ipywidgets) (0.1.4)
Requirement already satisfied: rfc3986-validator>=0.1.1 in /opt/conda/lib/py
thon3.10/site-packages (from jupyter-events>=0.9.0->jupyter-server<3,>=1.8->
notebook-shim>=0.2.3->nbclassic>=0.4.7->notebook>=4.4.1->widgetsnbextension~
=3.6.0->ipywidgets) (0.1.1)
Requirement already satisfied: fqdn in /opt/conda/lib/python3.10/site-packaq
es (from jsonschema[format-nongpl]>=4.18.0->jupyter-events>=0.9.0->jupyter-s
erver<3,>=1.8->notebook-shim>=0.2.3->nbclassic>=0.4.7->notebook>=4.4.1->widg
etsnbextension~=3.6.0->ipywidgets) (1.5.1)
Requirement already satisfied: isoduration in /opt/conda/lib/python3.10/sit
e-packages (from jsonschema[format-nongpl]>=4.18.0->jupyter-events>=0.9.0->j
upyter-server<3,>=1.8->notebook-shim>=0.2.3->nbclassic>=0.4.7->notebook>=
```

7 of 15

Requirement already satisfied: jsonpointer>1.13 in /opt/conda/lib/python3.1 0/site-packages (from jsonschema[format-nongpl]>=4.18.0->jupyter-events>= 0.9.0->jupyter-server<3,>=1.8->notebook-shim>=0.2.3->nbclassic>=0.4.7->noteb

Requirement already satisfied: uri-template in /opt/conda/lib/python3.10/sit

4.4.1->widgetsnbextension~=3.6.0->ipywidgets) (20.11.0)

ook>=4.4.1->widgetsnbextension~=3.6.0->ipywidgets) (2.4)

Map:

0%|

```
e-packages (from jsonschema[format-nongpl]>=4.18.0->jupyter-events>=0.9.0->j
       upyter-server<3,>=1.8->notebook-shim>=0.2.3->nbclassic>=0.4.7->notebook>=
       4.4.1->widgetsnbextension~=3.6.0->ipywidgets) (1.3.0)
       Requirement already satisfied: webcolors>=1.11 in /opt/conda/lib/python3.10/
       site-packages (from jsonschema[format-nongpl]>=4.18.0->jupyter-events>=
       0.9.0->jupyter-server<3,>=1.8->notebook-shim>=0.2.3->nbclassic>=0.4.7->noteb
       ook>=4.4.1->widgetsnbextension~=3.6.0->ipywidgets) (24.6.0)
       Requirement already satisfied: arrow>=0.15.0 in /opt/conda/lib/python3.10/si
       te-packages (from isoduration->jsonschema[format-nongpl]>=4.18.0->jupyter-ev
       ents>=0.9.0->jupyter-server<3,>=1.8->notebook-shim>=0.2.3->nbclassic>=0.4.7-
       >notebook>=4.4.1->widgetsnbextension~=3.6.0->ipywidgets) (1.3.0)
       Requirement already satisfied: types-python-dateutil>=2.8.10 in /opt/conda/l
       ib/python3.10/site-packages (from arrow>=0.15.0->isoduration->jsonschema[for
       mat-nongpl]>=4.18.0->jupyter-events>=0.9.0->jupyter-server<3,>=1.8->noteboo
       k-shim>=0.2.3->nbclassic>=0.4.7->notebook>=4.4.1->widgetsnbextension\sim=3.6.0-
       >ipywidgets) (2.9.0.20240316)
In [7]: from transformers import AutoTokenizer
        transformer_name = 'bert-base-cased'
        #tokenizer = AutoTokenizer.from_pretrained(transformer_name)
        tokenizer = AutoTokenizer.from_pretrained(
            transformer_name, clean_up_tokenization_spaces=True, quiet=True
                                             | 0.00/49.0 [00:00<?, ?B/s]
       tokenizer_config.json:
                                0%|
                                   | 0.00/570 [00:00<?, ?B/s]
       config.json:
                      0%|
                    0%|
                                 | 0.00/213k [00:00<?, ?B/s]
       vocab.txt:
       tokenizer.json:
                                      | 0.00/436k [00:00<?, ?B/s]
                         0%|
        import logging
In [9]:
        logging.disable(logging.WARNING) # Suppress progress bar and warnings
        def tokenize(examples):
            return tokenizer(examples['text'], truncation=True)
        train_ds = ds['train'].map(
            tokenize,
            batched=True,
            remove_columns=['title', 'description', 'text']
        eval_ds = ds['validation'].map(
            tokenize,
            batched=True,
            remove_columns=['title', 'description', 'text']
        )
        train_ds.to_pandas()
                           | 0/108000 [00:00<?, ? examples/s]
       Map:
              0%|
```

8 of 15 22/11/24, 9:53 PM

| 0/12000 [00:00<?, ? examples/s]

Out[9]:		label	input_ids	token_type_ids	attention_mask
	0	2	[101, 16752, 13335, 1186, 2101, 6690, 9717, 11	[0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0	[1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1
	1	1	[101, 145, 11680, 17308, 9741, 2428, 150, 1469	[0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0	[1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1
	2	2	[101, 1418, 14099, 27086, 1494, 1114, 4031, 11	[0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0	[1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1
	3	1	[101, 2404, 117, 6734, 1996, 118, 1565, 5465,	[0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0	[1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1
	4	3	[101, 142, 10044, 27302, 4317, 1584, 3273, 111	[0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0	[1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1
	•••				
	107995	1	[101, 4922, 2274, 1654, 1112, 10503, 1505, 112	[0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0	[1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1
	107996	3	[101, 10605, 24632, 11252, 21285, 10221, 118,	[0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0	[1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1
	107997	2	[101, 13832, 3484, 11300, 4060, 5058, 112, 188	[0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0	[1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1
	107998	3	[101, 142, 13675, 3756, 5795, 2445, 1104, 109,	[0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0	[1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1
	107999	2	[101, 157, 16450, 1658, 5302, 185, 7776, 11006	[0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0	[1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1

108000 rows × 4 columns

Create the transformer model:

```
In [10]: from torch import nn
    from transformers.modeling_outputs import SequenceClassifierOutput
    from transformers.models.bert.modeling_bert import BertModel, BertPreTrained

# https://github.com/huggingface/transformers/blob/65659a29cf5a079842e61a63a

class BertForSequenceClassification(BertPreTrainedModel):
    def __init__(self, config):
        super().__init__(config)
        self.num_labels = config.num_labels
```

```
self.bert = BertModel(config)
    self.dropout = nn.Dropout(config.hidden_dropout_prob)
    self.classifier = nn.Linear(config.hidden size, config.num labels)
    self.init_weights()
def forward(self, input_ids=None, attention_mask=None, token_type_ids=No
    outputs = self.bert(
        input_ids,
        attention mask=attention mask,
        token_type_ids=token_type_ids,
        **kwargs,
    )
    cls_outputs = outputs.last_hidden_state[:, 0, :]
    cls_outputs = self.dropout(cls_outputs)
    logits = self.classifier(cls outputs)
    loss = None
    if labels is not None:
        loss_fn = nn.CrossEntropyLoss()
        loss = loss_fn(logits, labels)
    return SequenceClassifierOutput(
        loss=loss,
        logits=logits,
        hidden_states=outputs.hidden_states,
        attentions=outputs.attentions,
```

| 0.00/436M [00:00<?, ?B/s]

Create the trainer object and train:

0%|

model.safetensors:

```
In [13]: from transformers import TrainingArguments
         num_epochs = 2
         batch size = 24
         weight decay = 0.01
         model_name = f'{transformer_name}-sequence-classification'
         training_args = TrainingArguments(
             output_dir='./results',
                                            # Directory for model outputs
                                            # Do not save checkpoints
             save_strategy="no",
                                            # Number of epochs
             num_train_epochs=3,
             per_device_train_batch_size=8, # Batch size for training
             per_device_eval_batch_size=8, # Batch size for evaluation
                                             # Weight decay
             weight_decay=0.01,
```

```
logging_dir='./logs',
                                                # Directory for logs
In [14]: from sklearn.metrics import accuracy_score
         def compute_metrics(eval_pred):
             y_true = eval_pred.label_ids
             y_pred = np.argmax(eval_pred.predictions, axis=-1)
             return {'accuracy': accuracy_score(y_true, y_pred)}
In [16]: from transformers import Trainer
         import os
         # Ensure wandb is completely disabled
         os.environ["WANDB DISABLED"] = "true"
         os.environ["WANDB_MODE"] = "disabled"
         # Define training arguments
         training_args = TrainingArguments(
             output_dir="./results",
             report_to="none", # Disable all reporting integrations
             num_train_epochs=3,
             per_device_train_batch_size=16,
             per device eval batch size=16,
             evaluation_strategy="epoch",
             save_strategy="epoch",
             weight_decay=0.01,
         # Create Trainer
         trainer = Trainer(
             model=model,
             args=training_args,
             compute_metrics=compute_metrics,
             train_dataset=train_ds,
             eval_dataset=eval_ds,
             tokenizer=tokenizer,
         )
        /opt/conda/lib/python3.10/site-packages/transformers/training_args.py:1545:
        FutureWarning: `evaluation_strategy` is deprecated and will be removed in ve
        rsion 4.46 of 🕮 Transformers. Use `eval_strategy` instead
          warnings.warn(
In [17]: trainer.train()
                                             [20250/20250 1:04:17, Epoch 3/3]
        Epoch Training Loss Validation Loss Accuracy
            1
                  0.198000
                                 0.197395 0.938167
            2
                  0.146300
                                 0.201212 0.943667
```

11 of 15 22/11/24, 9:53 PM

0.251510 0.946000

3

0.070600

Evaluate on the test partition:

```
In [18]: test_ds = ds['test'].map(
          tokenize,
          batched=True,
          remove_columns=['title', 'description', 'text'],
)
test_ds.to_pandas()
```

test	test_us.to_pandas()						
Map:	0%						
Out[18]:	label	input_ids	token_type_ids	attention_mask			
	0 2	[101, 11284, 1116, 1111, 157, 151, 12966, 1170	[0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0	[1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1			
	1 3	[101, 1109, 6398, 1110, 1212, 131, 2307, 7219,	[0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0	[1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1,			
:	2 3	[101, 148, 1183, 119, 1881, 16387, 1116, 4468,	[0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0	[1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1,			
:	3 3	[101, 11689, 15906, 6115, 12056, 1116, 1370, 2	[0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0	[1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1,			
	4 3	[101, 11917, 8914, 119, 19294, 4206, 1106, 215	[0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0	[1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1,			
•	·•						
759	5 0	[101, 5596, 1103, 1362, 5284, 5200, 3234, 1384	[0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0	[1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1,			
759	6 1	[101, 159, 7874, 1110, 2709, 1114, 13875, 1556	[0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0	[1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1,			
759	7 1	[101, 16247, 2972, 9178, 2409, 4271, 140, 1418	[0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0	[1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1			
759	8 2	[101, 126, 1104, 1893, 8167, 10721, 4420, 1107	[0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0	[1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1,			
759	9 2	[101, 142, 2064, 4164, 3370, 1154, 13519, 1116	[0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0	[1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1,			

7600 rows × 4 columns

```
In [19]: output = trainer.predict(test_ds)
         output
Out[19]: PredictionOutput(predictions=array([[ 0.19795685, -4.619936 ,
                [-1.4081583, -2.9251745, -3.5345733, 6.778545],
                [-1.3091689, -3.3134522, -3.2881854, 6.4918365],
                [-2.041592 , 8.425348 , -2.4603012 , -3.4557693 ],
                [-0.81975 , -3.7158365 , 6.2912555 , -3.2601109 ],
                [-2.8590794, -4.609074, 4.9792194, 0.66535884]],
               dtype=float32), label_ids=array([2, 3, 3, ..., 1, 2, 2]), metrics={'t
         est_loss': 0.2598552703857422, 'test_accuracy': 0.9438157894736842, 'test_r
         untime': 24.6787, 'test_samples_per_second': 307.958, 'test_steps_per_secon
         d': 19.247})
In [20]: from sklearn.metrics import classification_report
         y_true = output.label_ids
         y_pred = np.argmax(output.predictions, axis=-1)
         target_names = labels
         print(classification_report(y_true, y_pred, target_names=target_names))
                      precision
                                  recall f1-score
                                                     support
               World
                          0.96
                                    0.95
                                              0.96
                                                        1900
              Sports
                          0.98
                                    0.99
                                              0.99
                                                        1900
            Business
                          0.92
                                    0.91
                                              0.91
                                                        1900
            Sci/Tech
                          0.91
                                    0.93
                                              0.92
                                                        1900
                                              0.94
                                                        7600
            accuracy
                          0.94
                                    0.94
                                              0.94
                                                        7600
           macro avg
```

Descripción de la estructura del pipeline del código del notebook

0.94

7600

Inicialización y Configuración:

0.94

weighted avg

• Importación de librerías necesarias (torch, transformers, pandas, etc.).

0.94

 Configuración del dispositivo de cómputo (CPU o GPU) y la semilla para garantizar reproducibilidad.

Carga y Preprocesamiento de Datos:

- Lectura de los datasets (entrenamiento, validación y prueba) desde archivos CSV.
- Creación de una columna combinada text que concatena el título y descripción de los textos.
- Normalización de datos, como la eliminación de caracteres especiales y ajuste de

- las etiquetas (label) para que comiencen desde 0.
- Dividir los datos de entrenamiento en subconjuntos de entrenamiento y validación (90%-10%).

Conversión de Datos a Objetos de HuggingFace:

 Transformación de los DataFrames a objetos del tipo Dataset y DatasetDict para ser utilizados por el modelo.

Tokenización:

- Uso de un tokenizador BERT preentrenado (bert-base-cased) para convertir los textos en secuencias de tokens compatibles con el modelo.
- Aplicación de la tokenización a los datasets mediante mapeo batched y eliminación de columnas innecesarias.

Definición del Modelo:

- Construcción del modelo de clasificación de secuencias (BertForSequenceClassification) basado en BERT, añadiendo una capa lineal para clasificar los datos en categorías específicas.
- Configuración de hiperparámetros del modelo, como el tamaño de la capa oculta y el número de etiquetas.

Entrenamiento del Modelo:

- Configuración de los argumentos de entrenamiento, como el número de épocas (3), el tamaño del batch, la estrategia de evaluación, y el peso de decaimiento.
- Creación de un objeto Trainer de HuggingFace que gestiona el entrenamiento, validación y evaluación.
- Entrenamiento del modelo utilizando los datos tokenizados y supervisando la pérdida y precisión durante las épocas.

Evaluación del Modelo:

- Evaluación del modelo entrenado en el conjunto de prueba mediante métricas de clasificación (precisión, recall, F1-score) y cálculo de la pérdida en prueba.
- Generación de un reporte detallado de clasificación usando classification_report de sklearn.

Predicción y Análisis:

- Uso del modelo entrenado para realizar predicciones en el conjunto de prueba.
- Análisis de las métricas y resultados obtenidos, como precisión global y métricas

específicas para cada clase.