

Analysis

December 20, 2021

1 Importing Packages

```
[1]: # Load packages for data wrangling:
import os
import glob
import numpy as np
import pandas as pd

# Load packages for fine-tuning BERT model:
from simpletransformers.classification import ClassificationModel

# Load scikit-learn train_test_split:
from sklearn.model_selection import train_test_split

# Load classification metrics:
from sklearn.metrics import (accuracy_score, recall_score, precision_score,
                             f1_score,
                             classification_report, confusion_matrix)

# Load softmax for converting raw model output to probabilities:
from scipy.special import softmax

# Load packages for data cleaning:
import string
import re
import nltk
nltk.download('stopwords')
nltk.download('wordnet')
nltk.download('punkt')
nltk.download('averaged_perceptron_tagger')
from nltk.corpus import stopwords
from nltk.stem import WordNetLemmatizer
from nltk import pos_tag
from nltk import sent_tokenize, word_tokenize

# Set stopword corpus
stopword = nltk.corpus.stopwords.words('english')
```

```
# Set NLTK lemmatizer
lemmatizer = WordNetLemmatizer()
```

```
[nltk_data] Downloading package stopwords to /home/ucloud/nltk_data...
[nltk_data]   Unzipping corpora/stopwords.zip.
[nltk_data] Downloading package wordnet to /home/ucloud/nltk_data...
[nltk_data]   Unzipping corpora/wordnet.zip.
[nltk_data] Downloading package punkt to /home/ucloud/nltk_data...
[nltk_data]   Unzipping tokenizers/punkt.zip.
[nltk_data] Downloading package averaged_perceptron_tagger to
[nltk_data]   /home/ucloud/nltk_data...
[nltk_data]   Unzipping taggers/averaged_perceptron_tagger.zip.
```

2 Preprocessing of dataset 1

2.1 Data loading and wrangling

```
[2]: # Loading data with fake news as pandas dataframe:
fake_df = pd.read_csv(os.path.join("data", "dataset_1", "Fake.csv"))

# Loading data with true news as pandas dataframe:
true_df = pd.read_csv(os.path.join("data", "dataset_1", "True.csv"))
```

```
[3]: # Inspecting fake data:
fake_df.head(10)
```

```
[3]:
```

	title \		text subject \
0	Donald Trump Sends Out Embarrassing New Year'...		
1	Drunk Bragging Trump Staffer Started Russian ...		
2	Sheriff David Clarke Becomes An Internet Joke...		
3	Trump Is So Obsessed He Even Has Obama's Name...		
4	Pope Francis Just Called Out Donald Trump Dur...		
5	Racist Alabama Cops Brutalize Black Boy While...		
6	Fresh Off The Golf Course, Trump Lashes Out A...		
7	Trump Said Some INSANELY Racist Stuff Inside ...		
8	Former CIA Director Slams Trump Over UN Bully...		
9	WATCH: Brand-New Pro-Trump Ad Features So Muc...		
0	Donald Trump just couldn t wish all Americans ...	News	
1	House Intelligence Committee Chairman Devin Nu...	News	
2	On Friday, it was revealed that former Milwauk...	News	
3	On Christmas day, Donald Trump announced that ...	News	
4	Pope Francis used his annual Christmas Day mes...	News	
5	The number of cases of cops brutalizing and ki...	News	
6	Donald Trump spent a good portion of his day a...	News	

7	In the wake of yet another court decision that...	News
8	Many people have raised the alarm regarding th...	News
9	Just when you might have thought we d get a br...	News

	date
0	December 31, 2017
1	December 31, 2017
2	December 30, 2017
3	December 29, 2017
4	December 25, 2017
5	December 25, 2017
6	December 23, 2017
7	December 23, 2017
8	December 22, 2017
9	December 21, 2017

```
[4]: # Inspecting true data:
true_df.head(10)
```

```
[4]:                                     title \
0  As U.S. budget fight looms, Republicans flip t...
1  U.S. military to accept transgender recruits o...
2  Senior U.S. Republican senator: 'Let Mr. Muell...
3  FBI Russia probe helped by Australian diplomat...
4  Trump wants Postal Service to charge 'much mor...
5  White House, Congress prepare for talks on spe...
6  Trump says Russia probe will be fair, but time...
7  Factbox: Trump on Twitter (Dec 29) - Approval ...
8      Trump on Twitter (Dec 28) - Global Warming
9  Alabama official to certify Senator-elect Jone...
```

	text	subject \
0	WASHINGTON (Reuters) - The head of a conservat...	politicsNews
1	WASHINGTON (Reuters) - Transgender people will...	politicsNews
2	WASHINGTON (Reuters) - The special counsel inv...	politicsNews
3	WASHINGTON (Reuters) - Trump campaign adviser ...	politicsNews
4	SEATTLE/WASHINGTON (Reuters) - President Donal...	politicsNews
5	WEST PALM BEACH, Fla./WASHINGTON (Reuters) - T...	politicsNews
6	WEST PALM BEACH, Fla (Reuters) - President Don...	politicsNews
7	The following statements were posted to the ve...	politicsNews
8	The following statements were posted to the ve...	politicsNews
9	WASHINGTON (Reuters) - Alabama Secretary of St...	politicsNews

	date
0	December 31, 2017
1	December 29, 2017
2	December 31, 2017

```
3 December 30, 2017
4 December 29, 2017
5 December 29, 2017
6 December 29, 2017
7 December 29, 2017
8 December 29, 2017
9 December 28, 2017
```

```
[5]: # Adding category-labels to each dataset:
```

```
fake_df["label"]="fake"
true_df["label"]="true"
```

```
[6]: # Merge fake- and true news into a single dataframe:
```

```
merged_df = pd.concat([true_df, fake_df])
```

```
[7]: # Assessing whether merge was succesful:
```

```
len(true_df) + len(fake_df) == len(merged_df)
```

```
[7]: True
```

2.2 Data cleaning

2.2.1 Removing bad rows

```
[8]: # Remove rows with only whitespace and replace it with NaN:
```

```
merged_df.replace(" ", float("NaN"), inplace=True)
```

```
# Remove NA's:
```

```
merged_df.dropna(subset = ["text"], inplace=True)
```

```
[9]: # Remove duplicate texts:
```

```
merged_df = merged_df.drop_duplicates(subset=['text'])
```

```
[10]: # Reset indices:
```

```
merged_df = merged_df.reset_index()
```

```
[11]: # Selecting only relevant columns:
```

```
merged_df = merged_df[["text", "label"]]
```

2.2.2 Regex

Remove “[city name] Reuters -” from true articles

```
[12]: # Define regex pattern:
```

```
pattern = r".*\(Reuters\) - "
```

```
for i in range(len(merged_df['text'])):
```

```
    merged_df['text'][i] = re.sub(pattern, '', merged_df['text'][i])
```

Remove hashtags

```
[13]: # Define regex pattern:
      pattern = r"#(\S+)"

      for i in range(len(merged_df['text'])):
          merged_df['text'][i] = re.sub(pattern, '', merged_df['text'][i])
```

Remove twitter tags (“@[username]”)

```
[14]: # Define regex pattern:
      pattern = r"@(\S+)"

      for i in range(len(merged_df['text'])):
          merged_df['text'][i] = re.sub(pattern, '', merged_df['text'][i])
```

Remove ‘(CAPSLOCK)’ e.g. from (VIDEO); something which was quite frequent in the fake news dataset

```
[15]: # Define regex pattern:
      pattern = r"\([A-Z]*\)"

      for i in range(len(merged_df['text'])):
          merged_df['text'][i] = re.sub(pattern, '', merged_df['text'][i])
```

Remove systematic patterns:

```
[16]: # Define regex pattern:
      pattern = r"The following statement.*accuracy[.]"

      for i in range(len(merged_df['text'])):
          merged_df['text'][i] = re.sub(pattern, '', merged_df['text'][i])
```

```
[17]: # Define regex pattern:
      pattern = r"pic\.twitter\.com\/.* "

      for i in range(len(merged_df['text'])):
          merged_df['text'][i] = re.sub(pattern, '', merged_df['text'][i])
```

2.2.3 Remove punctuation

```
[18]: # Define function:
      def remove_punctuation(text):
          no_punct=[words for words in text if words not in string.punctuation]
          words_wo_punct=' '.join(no_punct)
          return words_wo_punct
```

```
[19]: merged_df['text']=merged_df['text'].apply(lambda x: remove_punctuation(x))
```

2.2.4 Tokenization + Lower

```
[20]: # Define function:
def tokenize(text):
    split=re.split("\W+",text)
    return split
```

```
[21]: merged_df['tokenized']=merged_df['text'].apply(lambda x: tokenize(x.lower()))
```

2.2.5 Remove stopwords

```
[22]: # Define function:
def remove_stopwords(text):
    text=[words for words in text if words not in stopwords]
    #text=' '.join(text)
    return text
```

```
[23]: merged_df['tokenized'] = merged_df['tokenized'].apply(lambda x:
    ↪remove_stopwords(x))
```

2.2.6 Lemmatize

```
[24]: # Define function:
def penn2morph(penntag):
    morphy_tag = {'NN':'n', 'JJ':'a',
                  'VB':'v', 'RB':'r'}

    try:
        return morphy_tag[penntag[:2]]
    except:
        return 'n'
```

```
[25]: for i in range(len(merged_df['tokenized'])):
    tagged = pos_tag(merged_df['tokenized'][i])
    merged_df['tokenized'][i] = [lemmatizer.lemmatize(word,
    ↪pos=penn2morph(tag)) for word, tag in tagged]
```

2.2.7 Concatenate tokens into sentences

```
[26]: # Define function:
def concat(text):
    text=[words for words in text]
    text=' '.join(text)
    return text
```

```
[27]: merged_df['text'] = merged_df['tokenized'].apply(lambda x: concat(x))
```

2.2.8 Remove newly induced empty columns

```
[28]: merged_df.replace(" ", float("NaN"), inplace=True)

merged_df.dropna(subset = ["text"], inplace=True)
```

```
[29]: merged_df = merged_df.reset_index()
```

2.2.9 Assess whether we have missed anything

```
[30]: true_idx = merged_df[merged_df['label']=="true"].index.tolist()
fake_idx = merged_df[merged_df['label']=="fake"].index.tolist()
```

```
[31]: from collections import Counter
Counter(" ".join(merged_df['text'][true_idx]).split()).most_common(10)
```

```
[31]: [('say', 113426),
      ('trump', 53621),
      ('u', 40552),
      ('state', 36143),
      ('would', 31145),
      ('president', 26582),
      ('republican', 20154),
      ('government', 19171),
      ('year', 18520),
      ('house', 16787)]
```

```
[32]: from collections import Counter
Counter(" ".join(merged_df['text'][fake_idx]).split()).most_common(10)
```

```
[32]: [('trump', 58413),
      ('say', 36515),
      ('people', 19204),
      ('president', 18091),
      ('go', 17802),
      ('would', 17078),
      ('make', 16956),
      ('one', 16919),
      ('state', 16195),
      ('get', 14812)]
```

2.2.10 Remove newly found systematic patterns

```
[33]: # Define regex pattern:
pattern = r"21st century wire say"

for i in range(len(merged_df['text'])):
```

```
merged_df['text'][i] = re.sub(pattern, '', merged_df['text'][i])
```

/opt/conda/lib/python3.7/site-packages/ipykernel_launcher.py:5:

SettingWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy

"""

```
[34]: # Define regex pattern:
pattern = r"21st century wire"

for i in range(len(merged_df['text'])):
    merged_df['text'][i] = re.sub(pattern, '', merged_df['text'][i])
```

/opt/conda/lib/python3.7/site-packages/ipykernel_launcher.py:5:

SettingWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy

"""

```
[35]: # Define regex pattern:
pattern = r"filessupport.*"

for i in range(len(merged_df['text'])):
    merged_df['text'][i] = re.sub(pattern, '', merged_df['text'][i])
```

/opt/conda/lib/python3.7/site-packages/ipykernel_launcher.py:5:

SettingWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy

"""

```
[36]: # Define regex pattern:
pattern = r"21wire"

for i in range(len(merged_df['text'])):
    merged_df['text'][i] = re.sub(pattern, '', merged_df['text'][i])
```

/opt/conda/lib/python3.7/site-packages/ipykernel_launcher.py:5:

SettingWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy

```
[37]: from collections import Counter
      Counter(" ".join(merged_df['text'][fake_idx]).split()).most_common(10)
```

```
[37]: [('trump', 58413),
      ('say', 36151),
      ('people', 19204),
      ('president', 18091),
      ('go', 17802),
      ('would', 17078),
      ('make', 16956),
      ('one', 16919),
      ('state', 16195),
      ('get', 14812)]
```

2.3 Saving and loading cleaned dataset 1

2.3.1 Write dataframe to csv-file

```
[38]: # Selecting only relevant columns
      merged_df = merged_df[["text", "label"]]
```

```
[39]: # Write to csv
      merged_df.to_csv(os.path.join("data", "generated_data", "cleaned_dataset_1.
      ↪csv"), index=False)
```

3 BERT trained- and evaluated on dataset 1

3.0.1 Load cleaned data and prepare for classification

```
[2]: cleaned_dataset_1 = pd.read_csv(os.path.join("data", "generated_data", "
      ↪cleaned_dataset_1.csv"))
```

One row is corrupted when loading CSV and is turned into blank space. This is removed

```
[3]: cleaned_dataset_1.replace(" ", float("NaN"), inplace=True)

      cleaned_dataset_1.dropna(subset = ["text"], inplace=True)
```

Create training-, validation and testing dataset:

```
[4]: # Create train/test split with 20% of all articles in testing data:
      train_1, test_1 = train_test_split(cleaned_dataset_1, test_size=0.2)
```

```
[5]: # Create train/val split with 10% of remaining articles in validation data:
train_1, val_1 = train_test_split(train_1, test_size=0.1)
```

```
[6]: # Assess that split was successful:
len(train_1) + len(val_1) + len(test_1) == len(cleaned_dataset_1)
```

```
[6]: True
```

```
[7]: # Convert label column to binary integer (0 = true, 1 = fake):
train_1["label"] = np.where(train_1["label"] == "true", 0,1)
val_1["label"] = np.where(val_1["label"] == "true", 0,1)
test_1["label"] = np.where(test_1["label"] == "true", 0,1)
```

```
[8]: # Inspecting transformed training data:
train_1.head(10)
```

```
[8]:
```

	text	label
22238	president obama typically refrain outright bla...	1
6692	numerous vote machine heavily democratic detro...	0
22097	spouse tradition g20 summit truth little world...	1
38014	yesterday u mainstream medium outlet cnn threa...	1
36244	news come heel obama release drug offender pri...	1
17666	u president donald trump host singapore prime ...	0
33492	know good think either bill clinton senile try...	1
34700	democrat want spend whop 2 billion zika virus ...	1
35018	unfortunately day age really matter story true...	1
13853	britain parliament debate brexit withdrawal bi...	0

```
[9]: # Inspecting transformed validation data:
val_1.head(10)
```

```
[9]:
```

	text	label
3818	democrat u senate formally request detail thur...	0
14347	u secretary state rex tillerson urge african l...	0
31932	nancy pelosi sink low question answer giggle s...	1
17293	myanmar military launch internal probe conduct...	0
22758	western world see wave nationalistic xenophobi...	1
9843	democratic presidential candidate hillary clin...	0
36350	author clinton cash responds hillary clinton b...	1
18818	russian president vladimir putin speak phone g...	0
16129	britain talk exit european union cannot progre...	0
36440	private jet lot cash presidential suite name u...	1

```
[10]: # Assess that data is roughly balanced across categories:
train_1.groupby('label').count()
```

```
[10]:          text
      label
0       15277
1       12545
```

```
[11]: # Assess that data is roughly balanced across categories:
      val_1.groupby('label').count()
```

```
[11]:          text
      label
0       1727
1       1365
```

```
[12]: # Assess that data is roughly balanced across categories:
      test_1.groupby('label').count()
```

```
[12]:          text
      label
0       4187
1       3542
```

```
[13]: # Define number of unique labels:
      n_labels = len(train_1['label'].unique())
```

```
[14]: # Create list of texts to predict:
      X_dataset_1 = test_1['text'].tolist()
```

```
[15]: # Inspect length
      len(X_dataset_1)
```

```
[15]: 7729
```

3.1 Training

```
[19]: # Initialize the model with the specified hyperparameters:
      FN_model_1 = ClassificationModel('bert', "bert-base-uncased",
                                      num_labels=n_labels, use_cuda=False,
                                      args={'reprocess_input_data': True,
↳ 'overwrite_output_dir': True,
                                      "num_train_epochs": 3, "max_seq_length":
↳ 512, "train_batch_size": 128,
                                      "learning_rate": 1e-5})

      # Fine-tune the model:
      FN_model_1.train_model(train_1)
```

Some weights of the model checkpoint at bert-base-uncased were not used when initializing BertForSequenceClassification: ['cls.predictions.bias', 'cls.predictions.transform.dense.bias', 'cls.predictions.transform.LayerNorm.bias', 'cls.predictions.decoder.weight', 'cls.seq_relationship.bias', 'cls.seq_relationship.weight', 'cls.predictions.transform.LayerNorm.weight', 'cls.predictions.transform.dense.weight']

- This IS expected if you are initializing BertForSequenceClassification from the checkpoint of a model trained on another task or with another architecture (e.g. initializing a BertForSequenceClassification model from a BertForPreTraining model).

- This IS NOT expected if you are initializing BertForSequenceClassification from the checkpoint of a model that you expect to be exactly identical (initializing a BertForSequenceClassification model from a BertForSequenceClassification model).

Some weights of BertForSequenceClassification were not initialized from the model checkpoint at bert-base-uncased and are newly initialized:

['classifier.bias', 'classifier.weight']

You should probably TRAIN this model on a down-stream task to be able to use it for predictions and inference.

/opt/conda/lib/python3.7/site-

packages/simpletransformers/classification/classification_model.py:586:

UserWarning: Dataframe headers not specified. Falling back to using column 0 as text and column 1 as labels.

"Dataframe headers not specified. Falling back to using column 0 as text and column 1 as labels."

0%| | 0/27822 [00:00<?, ?it/s]

Epoch: 0%| | 0/3 [00:00<?, ?it/s]

Running Epoch 0 of 3: 0%| | 0/218 [00:00<?, ?it/s]

Running Epoch 1 of 3: 0%| | 0/218 [00:00<?, ?it/s]

Running Epoch 2 of 3: 0%| | 0/218 [00:00<?, ?it/s]

[19]: (654, 0.06579963080759448)

3.2 Predictions

```
[16]: # Loading trained model, so we don't have to rerun the training each time we
      ↪ restart the kernel:
      FN_model_1 = ClassificationModel("bert", "outputs_dataset_1/",
      ↪ num_labels=n_labels, use_cuda=False)
```

```

↳ -----
HTTPError                                Traceback (most recent call↳
↳ last)

<ipython-input-16-c5b37e71791a> in <module>
    1 # Loading trained model, so we don't have to rerun the training each↳
↳ time we restart the kernel:
    ----> 2 FN_model_1 = ClassificationModel("bert", "outputs_dataset_1/",↳
↳ num_labels=n_labels, use_cuda=False)

/opt/conda/lib/python3.7/site-packages/simpletransformers/classification/
↳ classification_model.py in __init__(self, model_type, model_name,↳
↳ tokenizer_type, tokenizer_name, num_labels, weight, args, use_cuda,↳
↳ cuda_device, onnx_execution_provider, **kwargs)
    338         if num_labels:
    339             self.config = config_class.from_pretrained(
--> 340                 model_name, num_labels=num_labels, **self.args.config
    341             )
    342             self.num_labels = num_labels

/opt/conda/lib/python3.7/site-packages/transformers/configuration_utils.
↳ py in from_pretrained(cls, pretrained_model_name_or_path, **kwargs)
    499
    500         """
--> 501         config_dict, kwargs = cls.
↳ get_config_dict(pretrained_model_name_or_path, **kwargs)
    502         if "model_type" in config_dict and hasattr(cls,↳
↳ "model_type") and config_dict["model_type"] != cls.model_type:
    503             logger.warn(

/opt/conda/lib/python3.7/site-packages/transformers/configuration_utils.
↳ py in get_config_dict(cls, pretrained_model_name_or_path, **kwargs)
    552             revision=revision,
    553             use_auth_token=use_auth_token,
--> 554             local_files_only=local_files_only,
    555         )
    556

```

```

/opt/conda/lib/python3.7/site-packages/transformers/configuration_utils.
↳py in get_configuration_file(path_or_repo, revision, use_auth_token,
↳local_files_only)
    840     # Inspect all files from the repo/folder.
    841     all_files = get_list_of_files(
--> 842         path_or_repo, revision=revision,
↳use_auth_token=use_auth_token, local_files_only=local_files_only
    843     )
    844     configuration_files_map = {}

```

```

/opt/conda/lib/python3.7/site-packages/transformers/file_utils.py in
↳get_list_of_files(path_or_repo, revision, use_auth_token, local_files_only)
    1950     else:
    1951         token = None
-> 1952     return list_repo_files(path_or_repo, revision=revision,
↳token=token)
    1953
    1954

```

```

/opt/conda/lib/python3.7/site-packages/huggingface_hub/hf_api.py in
↳list_repo_files(self, repo_id, revision, repo_type, token, timeout)
    601         if repo_type is None:
    602             info = self.model_info(
--> 603                 repo_id, revision=revision, token=token,
↳timeout=timeout
    604             )
    605             elif repo_type == "dataset":

```

```

/opt/conda/lib/python3.7/site-packages/huggingface_hub/hf_api.py in
↳model_info(self, repo_id, revision, token, timeout)
    584         )
    585         r = requests.get(path, headers=headers, timeout=timeout)
--> 586         r.raise_for_status()
    587         d = r.json()
    588         return ModelInfo(**d)

```

```

/opt/conda/lib/python3.7/site-packages/requests/models.py in
↳raise_for_status(self)
    938
    939         if http_error_msg:
--> 940             raise HTTPError(http_error_msg, response=self)
    941
    942     def close(self):

```

```
HTTPError: 404 Client Error: Not Found for url: https://huggingface.co/
↳api/models/outputs_dataset_1
```

```
[56]: # Use the fine-tuned model to predict the testing labels and save the raw model
↳outputs:
_, raw_pred = FN_model_1.predict(X_dataset_1)
```

```
0%|          | 0/7729 [00:00<?, ?it/s]
```

```
0%|          | 0/967 [00:00<?, ?it/s]
```

```
[57]: # Convert raw model outputs to class probabilities:
probabilities = softmax(raw_pred, axis=1)
```

```
[58]: # Assess probabilities:
probabilities
```

```
[58]: array([[4.79282272e-04, 9.99520718e-01],
           [9.99563380e-01, 4.36620433e-04],
           [4.79260344e-04, 9.99520740e-01],
           ...,
           [3.38563774e-03, 9.96614362e-01],
           [5.35239229e-04, 9.99464761e-01],
           [9.99326263e-01, 6.73736558e-04]])
```

```
[59]: # Binarize probabilities to the most probable class:
binary_preds = [np.argmax(pred) for pred in probabilities]
```

```
[60]: # Inspect length of predictions:
len(binary_preds)
```

```
[60]: 7729
```

3.3 Results

```
[61]: # Print classification report:
print(classification_report(test_1.label, binary_preds))

# Print confusion matrix:
confusion_matrix(test_1.label, binary_preds)
```

```
precision    recall  f1-score   support
```

0	1.00	1.00	1.00	4187
1	1.00	1.00	1.00	3542
accuracy			1.00	7729
macro avg	1.00	1.00	1.00	7729
weighted avg	1.00	1.00	1.00	7729

```
[61]: array([[4182,    5],
           [ 13, 3529]])
```

4 Preprocess dataset 2

4.1 Data loading and wrangling

```
[62]: file_list = glob.glob(os.path.join(os.getcwd(), "data", "dataset_2", "fake", "*.
      ↪txt"))

fake = []

for file_path in file_list:
    with open(file_path, encoding='windows-1252') as f_input:
        encoded_f = f_input.read().replace("\n", " ")
        fake.append(encoded_f)
```

```
[63]: file_list = glob.glob(os.path.join(os.getcwd(), "data", "dataset_2", "real", "*.
      ↪txt"))

real = []

for file_path in file_list:
    with open(file_path, encoding='windows-1252') as f_input:
        encoded_f = f_input.read().replace("\n", " ")
        real.append(encoded_f)
```

4.2 Data cleaning

```
[64]: # Remove \ from the data:
for i in range(len(fake)):
    fake[i] = fake[i].replace("\'", "")
```

```
[65]: # Remove \ from the data:
for i in range(len(real)):
    real[i] = real[i].replace("\'", "")
```



```

[66]: # Convert data to pandas dataframe:
fake_new = pd.DataFrame(fake)

[67]: # Rename column with texts to text:
fake_new = fake_new.rename({0: "text"},axis = 'columns')

[68]: # Add label-column with fake labels:
fake_new['label'] = 'fake'

[69]: # Convert data to pandas dataframe:
real_new = pd.DataFrame(real)

[70]: # Rename column with texts to text:
real_new = real_new.rename({0: "text"},axis = 'columns')

[71]: # Add label-column with fake labels:
real_new['label'] = 'true'

[72]: # Merge fake- and true news into a single dataframe:
merged_new = pd.concat([fake_new, real_new])

[73]: # Reset indeces:
merged_new = merged_new.reset_index()

[74]: # Selecting only relevant columns:
merged_new = merged_new[["text", "label"]]

[75]: # Inspecting:
merged_new

```

```

[75]:
                                     text label
0    The warranty on ‘Make America Great Again’ bas...  fake
1    Calling it a total disaster, president-elect D...  fake
2    WASHINGTON, D.C. - Former presidential inter...  fake
3    President Barack Obama’s legacy might soon be ...  fake
4    atican City - In a final speech to the synod, ...  fake
..
246  WASHINGTON - Republicans are united on repeali...  true
247  President-elect Donald Trump escalated his rhe...  true
248  Congress is preparing to do major battle next ...  true
249  PALM BEACH, Fla. -- President-elect Donald Tru...  true
250  This is my last column until after the electio...  true

[251 rows x 2 columns]

```

4.2.1 Remove punctuation

```
[76]: merged_new['text']=merged_new['text'].apply(lambda x: remove_punctuation(x))
```

4.2.2 Tokenize and lower

```
[77]: merged_new['tokenized']=merged_new['text'].apply(lambda x: tokenize(x.lower()))
```

4.2.3 Remove stopwords

```
[78]: merged_new['tokenized'] = merged_new['tokenized'].apply(lambda x: ↵  
    ↵remove_stopwords(x))
```

4.2.4 Lemma

```
[79]: for i in range(len(merged_new['tokenized'])):  
    tagged = pos_tag(merged_new['tokenized'][i])  
    merged_new['tokenized'][i] = [lemmatizer.lemmatize(word, ↵  
    ↵pos=penn2morphy(tag)) for word, tag in tagged]
```

4.2.5 Concatenate tokens into sentences

```
[80]: merged_new['text'] = merged_new['tokenized'].apply(lambda x: concat(x))
```

4.2.6 Write dataframe to csv-file

```
[81]: # Selecting only relevant columns:  
merged_new = merged_new[["text", "label"]]
```

```
[82]: # Write to csv:  
merged_new.to_csv(os.path.join("data", "generated_data", "cleaned_dataset_2.  
    ↵csv"), index=False)
```

5 BERT trained on dataset 1, evaluated on dataset 2

5.0.1 Load cleaned data

```
[83]: cleaned_dataset_2 = pd.read_csv(os.path.join("data", "generated_data", ↵  
    ↵"cleaned_dataset_2.csv"))
```

```
[84]: # Change labels to binary integers:  
cleaned_dataset_2["label"] = np.where(cleaned_dataset_2["label"] == "true", 0,1)
```

```
[85]: # Define number of unique labels:  
n_labels = len(cleaned_dataset_2['label'].unique())
```

```
[86]: # Create list of texts to predict:
X_dataset_2 = cleaned_dataset_2['text'].tolist()
```

```
[87]: # Use the 1st fine-tuned model to predict dataset 2 save the raw model outputs:
_, raw_pred = FN_model_1.predict(X_dataset_2)
```

```
0%|          | 0/251 [00:00<?, ?it/s]
```

```
0%|          | 0/32 [00:00<?, ?it/s]
```

```
[88]: # Convert raw model outputs to class probabilities:
probabilities = softmax(raw_pred, axis=1)
```

```
[89]: # Binarize probabilities to the most probable class:
binary_preds = [np.argmax(pred) for pred in probabilities]
```

```
[90]: # Inspect length of predictions:
len(binary_preds)
```

```
[90]: 251
```

```
[91]: # Print classification report:
print(classification_report(cleaned_dataset_2.label, binary_preds))

# Print confusion matrix:
confusion_matrix(cleaned_dataset_2.label, binary_preds)
```

	precision	recall	f1-score	support
0	0.59	0.62	0.61	128
1	0.58	0.56	0.57	123
accuracy			0.59	251
macro avg	0.59	0.59	0.59	251
weighted avg	0.59	0.59	0.59	251

```
[91]: array([[79, 49],
           [54, 69]])
```

6 BERT trained- and evaluated dataset 2:

```
[92]: # Create train/test split with 20% of all articles in testing data:
train_2, test_2 = train_test_split(cleaned_dataset_2, test_size=0.2)
```

```
[93]: # Create list of texts to predict:
X_dataset_2 = test_2['text'].tolist()
```

```
[94]: # Define number of unique labels:
n_labels = len(train_2['label'].unique())
```

```
[115]: # Initialize the model with the specified hyperparameters:
FN_model_2 = ClassificationModel('bert', "bert-base-uncased",
                                num_labels=n_labels, use_cuda=False,
                                args={'reprocess_input_data': True,
    ↪ 'overwrite_output_dir': True,
                                "num_train_epochs": 3, "max_seq_length":
    ↪ 512, "train_batch_size": 16,
                                "learning_rate": 1e-5})

# Fine-tune the model:
FN_model_2.train_model(train_2)
```

Downloading: 0%| | 0.00/570 [00:00<?, ?B/s]

Downloading: 0%| | 0.00/420M [00:00<?, ?B/s]

Some weights of the model checkpoint at bert-base-uncased were not used when initializing BertForSequenceClassification: ['cls.seq_relationship.bias', 'cls.predictions.transform.LayerNorm.weight', 'cls.seq_relationship.weight', 'cls.predictions.bias', 'cls.predictions.transform.dense.bias', 'cls.predictions.transform.LayerNorm.bias', 'cls.predictions.transform.dense.weight', 'cls.predictions.decoder.weight']
- This IS expected if you are initializing BertForSequenceClassification from the checkpoint of a model trained on another task or with another architecture (e.g. initializing a BertForSequenceClassification model from a BertForPreTraining model).

- This IS NOT expected if you are initializing BertForSequenceClassification from the checkpoint of a model that you expect to be exactly identical (initializing a BertForSequenceClassification model from a BertForSequenceClassification model).

Some weights of BertForSequenceClassification were not initialized from the model checkpoint at bert-base-uncased and are newly initialized:

['classifier.bias', 'classifier.weight']

You should probably TRAIN this model on a down-stream task to be able to use it for predictions and inference.

Downloading: 0%| | 0.00/226k [00:00<?, ?B/s]

Downloading: 0%| | 0.00/455k [00:00<?, ?B/s]

Downloading: 0%| | 0.00/28.0 [00:00<?, ?B/s]

/opt/conda/lib/python3.7/site-packages/simpletransformers/classification/classification_model.py:586:
UserWarning: Dataframe headers not specified. Falling back to using column 0 as text and column 1 as labels.

"Dataframe headers not specified. Falling back to using column 0 as text and column 1 as labels."

0%| | 0/200 [00:00<?, ?it/s]

Epoch: 0%| | 0/3 [00:00<?, ?it/s]

Running Epoch 0 of 3: 0%| | 0/13 [00:00<?, ?it/s]

Running Epoch 1 of 3: 0%| | 0/13 [00:00<?, ?it/s]

Running Epoch 2 of 3: 0%| | 0/13 [00:00<?, ?it/s]

[115]: (39, 0.6268747601753626)

```
[95]: # Loading trained model, so we don't have to rerun the training each time we
      ↪ restart the kernel:
      FN_model_2 = ClassificationModel("bert", "outputs_dataset_2/",
      ↪ num_labels=n_labels, use_cuda=False)
```

```
[96]: # Use the fine-tuned model to predict the testing labels and save the raw model
      ↪ outputs:
      _, raw_pred = FN_model_2.predict(X_dataset_2)
```

0%| | 0/51 [00:00<?, ?it/s]

0%| | 0/7 [00:00<?, ?it/s]

```
[97]: # Convert raw model outputs to class probabilities:
      probabilities = softmax(raw_pred, axis=1)
```

```
[98]: # Binarize probabilities to the most probable class:
      binary_preds = [np.argmax(pred) for pred in probabilities]
```

```
[99]: # Inspect length of predictions:
      len(binary_preds)
```

[99]: 51

```
[100]: # Print classification report:
print(classification_report(test_2.label, binary_preds))

# Print confusion matrix:
confusion_matrix(test_2.label, binary_preds)
```

	precision	recall	f1-score	support
0	0.86	0.72	0.78	25
1	0.77	0.88	0.82	26
accuracy			0.80	51
macro avg	0.81	0.80	0.80	51
weighted avg	0.81	0.80	0.80	51

[100]: array([[18, 7],
[3, 23]])

7 BERT trained on dataset 2, evaluated on dataset 1

```
[101]: # Use the fine-tuned model to predict the testing labels from dataset 1 and
→save the raw model outputs:
_, raw_pred = FN_model_2.predict(X_dataset_1)
```

0%| | 0/7729 [00:00<?, ?it/s]

0%| | 0/967 [00:00<?, ?it/s]

```
[102]: # Convert raw model outputs to class probabilities:
probabilities = softmax(raw_pred, axis=1)
```

```
[103]: # Binarize probabilities to the most probable class:
binary_preds = [np.argmax(pred) for pred in probabilities]
```

```
[104]: # Inspect length of predictions:
len(binary_preds)
```

[104]: 7729

```
[105]: # Print classification report:
print(classification_report(test_1.label, binary_preds))
```

```
# Print confusion matrix:
confusion_matrix(test_1.label, binary_preds)
```

	precision	recall	f1-score	support
0	0.67	0.30	0.41	4187
1	0.50	0.83	0.62	3542
accuracy			0.54	7729
macro avg	0.58	0.56	0.52	7729
weighted avg	0.59	0.54	0.51	7729

```
[105]: array([[1237, 2950],
              [ 618, 2924]])
```

8 Periods - for temporal word embedding analysis

8.1 Data wrangling

```
[106]: import numpy as np
import regex as re
from datetime import *
```

```
[107]: # Load data:
fake = pd.read_csv(os.path.join("data", "dataset_1", "Fake.csv"))

# NA for wrong entries:
fake["date"] = [re.sub("^.*:.*|^.* .* .* .*|^d.*", "NA", date) for date in fake["date"]]
# All webpages, entries that start with a number and sequences
# of words upon words upon words should be NA

# Drop rows with NAs:
fake = fake[(fake != 'NA').all(1)]

# Streamline dates:
months = ["January", "February", "March", "April", "May", "June", "July",
          "August", "September", "October", "November", "December"]
for i in months:
    fake["date"] = [re.sub(f"^{i}", f"{i[0:3]}", date) for date in fake["date"]]

# Convert to date format:
fake["date"] = [datetime.strptime(date, "%b %d, %Y").date() for date in fake["date"]]
# fake["date"]
```

```
[108]: # Find date range:
date_range = max(fake["date"]) - min(fake["date"])

[109]: # Create categorical variable pertaining to split:
period = []
for date in fake["date"]:
    if date <= min(fake["date"]) + date_range/5:
        period.append(1)
    if date > min(fake["date"]) + date_range/5 and date <= min(fake["date"]) +
↪date_range/5*2:
        period.append(2)
    if date > min(fake["date"]) + date_range/5*2 and date <= min(fake["date"])
↪+ date_range/5*3:
        period.append(3)
    if date > min(fake["date"]) + date_range/5*3 and date <= min(fake["date"])
↪+ date_range/5*4:
        period.append(4)
    if date > min(fake["date"]) + date_range/5*4:
        period.append(5)

[110]: # Create column with periods:
fake["period"] = period

[111]: # Ensure that the unique entries in the period-column is correct:
fake['period'].unique()

[111]: array([5, 4, 3, 2, 1])

[112]: # Write data to csv:
fake.to_csv(os.path.join("data", "generated_data", "fake_periods.csv"),
↪index=False)

[113]: # Load data from csv:
fake = pd.read_csv(os.path.join("data", "generated_data", "fake_periods.csv"))
```

8.2 Data cleaning

```
[114]: # Remove rows with only whitespace and replace it with NA's
fake.replace(" ", float("NaN"), inplace=True)

# Remove NA's
fake.dropna(subset = ["text"], inplace=True)

[115]: # Remove duplicate texts:
fake = fake.drop_duplicates(subset=['text'])
```



```
[116]: # Reset indeces:
fake = fake.reset_index()
```

8.2.1 Remove reuters

```
[117]: # Define regex pattern:
pattern = r".*\(Reuters\) - "

for i in range(len(fake['text'])):
    fake['text'][i] = re.sub(pattern, '', fake['text'][i])
```

/opt/conda/lib/python3.7/site-packages/ipykernel_launcher.py:5:

SettingWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy

"""

8.2.2 Remove hashtags

```
[118]: # Define regex pattern:
pattern = r"#(\S+)"

for i in range(len(fake['text'])):
    fake['text'][i] = re.sub(pattern, '', fake['text'][i])
```

/opt/conda/lib/python3.7/site-packages/ipykernel_launcher.py:5:

SettingWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy

"""

8.2.3 Remove tags

```
[119]: # Define regex pattern:
pattern = r"@(\S+)"

for i in range(len(fake['text'])):
    fake['text'][i] = re.sub(pattern, '', fake['text'][i])
```

/opt/conda/lib/python3.7/site-packages/ipykernel_launcher.py:5:

SettingWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy

```
"""
```

8.2.4 Remove (capslock)

```
[120]: # Define regex pattern:
pattern = r"\([A-Z]*\)"

for i in range(len(fake['text'])):
    fake['text'][i] = re.sub(pattern, '', fake['text'][i])
```

```
/opt/conda/lib/python3.7/site-packages/ipykernel_launcher.py:5:
SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame
```

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy

```
"""
```

8.2.5 Remove systematic patterns

```
[122]: # Define regex pattern:
pattern = r"The following statement.*accuracy[.]"

for i in range(len(fake['text'])):
    fake['text'][i] = re.sub(pattern, '', fake['text'][i])
```

```
/opt/conda/lib/python3.7/site-packages/ipykernel_launcher.py:5:
SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame
```

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy

```
"""
```

```
[123]: # Define regex pattern:
pattern = r"pic\.twitter\.com\/.* "

for i in range(len(fake['text'])):
    fake['text'][i] = re.sub(pattern, '', fake['text'][i])
```

```
/opt/conda/lib/python3.7/site-packages/ipykernel_launcher.py:5:
SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame
```

See the caveats in the documentation: <https://pandas.pydata.org/pandas->

docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy
"""

8.2.6 Remove punctuation:

```
[124]: fake['text']=fake['text'].apply(lambda x: remove_punctuation(x))
```

8.2.7 Tokenize and lower

```
[125]: fake['tokenized']=fake['text'].apply(lambda x: tokenize(x.lower()))
```

8.2.8 Remove stopwords

```
[126]: fake['tokenized'] = fake['tokenized'].apply(lambda x: remove_stopwords(x))
```

8.2.9 Lemmatize

```
[127]: for i in range(len(fake['tokenized'])):
        tagged = pos_tag(fake['tokenized'][i])
        fake['tokenized'][i] = [lemmatizer.lemmatize(word, pos=penn2morphy(tag))
        ↪for word, tag in tagged]
```

/opt/conda/lib/python3.7/site-packages/ipykernel_launcher.py:3:

SettingWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy

This is separate from the ipykernel package so we can avoid doing imports until

8.2.10 Concatenate words

```
[129]: fake['text'] = fake['tokenized'].apply(lambda x: concat(x))
```

8.2.11 Remove newly induced empty columns

```
[130]: fake.replace(" ", float("NaN"), inplace=True)
```

```
fake.dropna(subset = ["text"], inplace=True)
```

```
[131]: fake = fake.reset_index()
```

8.2.12 Remove more systematic patterns

```
[132]: # Define regex pattern:
pattern = r"21st century wire say"

for i in range(len(fake['text'])):
    fake['text'][i] = re.sub(pattern, '', fake['text'][i])
```

/opt/conda/lib/python3.7/site-packages/ipykernel_launcher.py:5:

SettingWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy

"""

```
[133]: # Define regex pattern:
pattern = r"21st century wire"

for i in range(len(fake['text'])):
    fake['text'][i] = re.sub(pattern, '', fake['text'][i])
```

/opt/conda/lib/python3.7/site-packages/ipykernel_launcher.py:5:

SettingWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy

"""

```
[134]: # Define regex pattern:
pattern = r"filessupport.*"

for i in range(len(fake['text'])):
    fake['text'][i] = re.sub(pattern, '', fake['text'][i])
```

/opt/conda/lib/python3.7/site-packages/ipykernel_launcher.py:5:

SettingWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy

"""

```
[135]: # Define regex pattern:
pattern = r"21wire"

for i in range(len(fake['text'])):
```

```
fake['text'][i] = re.sub(pattern, '', fake['text'][i])
```

/opt/conda/lib/python3.7/site-packages/ipykernel_launcher.py:5:

SettingWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy

"""

8.3 Write data for word-embedding analysis

```
[136]: # Concatenate
period_texts = []
for i in range(1, 6):
    period_text = " ".join(fake.loc[fake['period'] == i]["text"])
    period_texts.append(period_text)

# Write as .txt files
for i, n in zip(period_texts, range(1,6)):
    text_file = open(os.path.join("word_embeddings", "output", "texts",
    ↪f"00{n}0.txt"), "w")
    n = text_file.write(i)
    text_file.close()
```