Praful file2 h2

February 15, 2024

Spam Detection HW

Read complete instructions before starting the HW

1 Installing/Importing Modules

```
[1]: !pip install -U spacy -q
[2]: | python -m spacy download en_core_web_sm
    Collecting en-core-web-sm==3.7.1
      Downloading https://github.com/explosion/spacy-
    models/releases/download/en_core_web_sm-3.7.1/en_core_web_sm-3.7.1-py3-none-
    any.whl (12.8 MB)
                                12.8/12.8 MB
    24.2 MB/s eta 0:00:00
    Requirement already satisfied: spacy<3.8.0,>=3.7.2 in
    /usr/local/lib/python3.10/dist-packages (from en-core-web-sm==3.7.1) (3.7.2)
    Requirement already satisfied: spacy-legacy<3.1.0,>=3.0.11 in
    /usr/local/lib/python3.10/dist-packages (from spacy<3.8.0,>=3.7.2->en-core-web-
    sm==3.7.1) (3.0.12)
    Requirement already satisfied: spacy-loggers<2.0.0,>=1.0.0 in
    /usr/local/lib/python3.10/dist-packages (from spacy<3.8.0,>=3.7.2->en-core-web-
    sm==3.7.1) (1.0.5)
    Requirement already satisfied: murmurhash<1.1.0,>=0.28.0 in
    /usr/local/lib/python3.10/dist-packages (from spacy<3.8.0,>=3.7.2->en-core-web-
    sm==3.7.1) (1.0.10)
    Requirement already satisfied: cymem<2.1.0,>=2.0.2 in
    /usr/local/lib/python3.10/dist-packages (from spacy<3.8.0,>=3.7.2->en-core-web-
    sm==3.7.1) (2.0.8)
    Requirement already satisfied: preshed<3.1.0,>=3.0.2 in
    /usr/local/lib/python3.10/dist-packages (from spacy<3.8.0,>=3.7.2->en-core-web-
    sm==3.7.1) (3.0.9)
    Requirement already satisfied: thinc<8.3.0,>=8.1.8 in
    /usr/local/lib/python3.10/dist-packages (from spacy<3.8.0,>=3.7.2->en-core-web-
    sm==3.7.1) (8.2.3)
    Requirement already satisfied: wasabi<1.2.0,>=0.9.1 in
    /usr/local/lib/python3.10/dist-packages (from spacy<3.8.0,>=3.7.2->en-core-web-
```

```
sm==3.7.1) (1.1.2)
Requirement already satisfied: srsly<3.0.0,>=2.4.3 in
/usr/local/lib/python3.10/dist-packages (from spacy<3.8.0,>=3.7.2->en-core-web-
sm==3.7.1) (2.4.8)
Requirement already satisfied: catalogue<2.1.0,>=2.0.6 in
/usr/local/lib/python3.10/dist-packages (from spacy<3.8.0,>=3.7.2->en-core-web-
sm==3.7.1) (2.0.10)
Requirement already satisfied: weasel<0.4.0,>=0.1.0 in
/usr/local/lib/python3.10/dist-packages (from spacy<3.8.0,>=3.7.2->en-core-web-
sm==3.7.1) (0.3.4)
Requirement already satisfied: typer<0.10.0,>=0.3.0 in
/usr/local/lib/python3.10/dist-packages (from spacy<3.8.0,>=3.7.2->en-core-web-
sm==3.7.1) (0.9.0)
Requirement already satisfied: smart-open<7.0.0,>=5.2.1 in
/usr/local/lib/python3.10/dist-packages (from spacy<3.8.0,>=3.7.2->en-core-web-
sm==3.7.1) (6.4.0)
Requirement already satisfied: tqdm<5.0.0,>=4.38.0 in
/usr/local/lib/python3.10/dist-packages (from spacy<3.8.0,>=3.7.2->en-core-web-
sm==3.7.1) (4.66.1)
Requirement already satisfied: requests<3.0.0,>=2.13.0 in
/usr/local/lib/python3.10/dist-packages (from spacy<3.8.0,>=3.7.2->en-core-web-
sm==3.7.1) (2.31.0)
Requirement already satisfied: pydantic!=1.8,!=1.8.1,<3.0.0,>=1.7.4 in
/usr/local/lib/python3.10/dist-packages (from spacy<3.8.0,>=3.7.2->en-core-web-
sm==3.7.1) (2.6.1)
Requirement already satisfied: jinja2 in /usr/local/lib/python3.10/dist-packages
(from spacy<3.8.0,>=3.7.2->en-core-web-sm==3.7.1) (3.1.3)
Requirement already satisfied: setuptools in /usr/local/lib/python3.10/dist-
packages (from spacy<3.8.0,>=3.7.2->en-core-web-sm==3.7.1) (67.7.2)
Requirement already satisfied: packaging>=20.0 in
/usr/local/lib/python3.10/dist-packages (from spacy<3.8.0,>=3.7.2->en-core-web-
sm==3.7.1) (23.2)
Requirement already satisfied: langcodes<4.0.0,>=3.2.0 in
/usr/local/lib/python3.10/dist-packages (from spacy<3.8.0,>=3.7.2->en-core-web-
sm==3.7.1) (3.3.0)
Requirement already satisfied: numpy>=1.19.0 in /usr/local/lib/python3.10/dist-
packages (from spacy<3.8.0,>=3.7.2->en-core-web-sm==3.7.1) (1.25.2)
Requirement already satisfied: annotated-types>=0.4.0 in
/usr/local/lib/python3.10/dist-packages (from
pydantic!=1.8,!=1.8.1,<3.0.0,>=1.7.4->spacy<3.8.0,>=3.7.2->en-core-web-
sm==3.7.1) (0.6.0)
Requirement already satisfied: pydantic-core==2.16.2 in
/usr/local/lib/python3.10/dist-packages (from
pydantic!=1.8,!=1.8.1,<3.0.0,>=1.7.4->spacy<3.8.0,>=3.7.2->en-core-web-
sm==3.7.1) (2.16.2)
Requirement already satisfied: typing-extensions>=4.6.1 in
/usr/local/lib/python3.10/dist-packages (from
pydantic!=1.8,!=1.8.1,<3.0.0,>=1.7.4->spacy<3.8.0,>=3.7.2->en-core-web-
```

```
sm==3.7.1) (4.9.0)
    Requirement already satisfied: charset-normalizer<4,>=2 in
    /usr/local/lib/python3.10/dist-packages (from
    requests<3.0.0,>=2.13.0->spacy<3.8.0,>=3.7.2->en-core-web-sm==3.7.1) (3.3.2)
    Requirement already satisfied: idna<4,>=2.5 in /usr/local/lib/python3.10/dist-
    packages (from requests<3.0.0,>=2.13.0->spacy<3.8.0,>=3.7.2->en-core-web-
    sm==3.7.1) (3.6)
    Requirement already satisfied: urllib3<3,>=1.21.1 in
    /usr/local/lib/python3.10/dist-packages (from
    requests<3.0.0,>=2.13.0->spacy<3.8.0,>=3.7.2->en-core-web-sm==3.7.1) (2.0.7)
    Requirement already satisfied: certifi>=2017.4.17 in
    /usr/local/lib/python3.10/dist-packages (from
    requests<3.0.0,>=2.13.0->spacy<3.8.0,>=3.7.2->en-core-web-sm==3.7.1) (2024.2.2)
    Requirement already satisfied: blis<0.8.0,>=0.7.8 in
    /usr/local/lib/python3.10/dist-packages (from
    thinc\{8.3.0, >=8.1.8-\} spacy\{3.8.0, >=3.7.2-\} en-core-web-sm==3.7.1) (0.7.11)
    Requirement already satisfied: confection<1.0.0,>=0.0.1 in
    /usr/local/lib/python3.10/dist-packages (from
    thinc<8.3.0,>=8.1.8->spacy<3.8.0,>=3.7.2->en-core-web-sm==3.7.1) (0.1.4)
    Requirement already satisfied: click<9.0.0,>=7.1.1 in
    /usr/local/lib/python3.10/dist-packages (from
    typer<0.10.0,>=0.3.0->spacy<3.8.0,>=3.7.2->en-core-web-sm==3.7.1) (8.1.7)
    Requirement already satisfied: cloudpathlib<0.17.0,>=0.7.0 in
    /usr/local/lib/python3.10/dist-packages (from
    weasel<0.4.0,>=0.1.0->spacy<3.8.0,>=3.7.2->en-core-web-sm==3.7.1) (0.16.0)
    Requirement already satisfied: MarkupSafe>=2.0 in
    /usr/local/lib/python3.10/dist-packages (from jinja2->spacy<3.8.0,>=3.7.2->en-
    core-web-sm==3.7.1) (2.1.5)
     Download and installation successful
    You can now load the package via spacy.load('en_core_web_sm')
[3]: !pip install pyspellchecker
    Collecting pyspellchecker
      Downloading pyspellchecker-0.8.1-py3-none-any.whl (6.8 MB)
                                6.8/6.8 MB
    19.1 MB/s eta 0:00:00
    Installing collected packages: pyspellchecker
    Successfully installed pyspellchecker-0.8.1
[4]: import spacy
     from spacy.matcher import Matcher
     from spacy.tokens import Token
     import pandas as pd
     import numpy as np
     from nltk.stem.porter import PorterStemmer
     import os
```

```
import sys
     from pathlib import Path
     from typing import List
     from sklearn.pipeline import Pipeline, FeatureUnion
     from sklearn.feature_extraction.text import TfidfVectorizer
     from sklearn.model_selection import train_test_split
     from collections import Counter
     from xgboost import XGBClassifier
     from sklearn.metrics import fbeta score, make scorer
     from sklearn.model selection import RandomizedSearchCV
     from sklearn.base import BaseEstimator, TransformerMixin
     from bs4 import BeautifulSoup
     import re
     from spellchecker import SpellChecker
     import warnings
     warnings.filterwarnings('ignore')
[5]: from google.colab import drive
     drive.mount('/content/drive')
    Mounted at /content/drive
[6]: base_folder = Path('/content/drive/MyDrive/NLP_HW/')
     data_folder = base_folder/'HW_2'
[7]: file = data_folder/'spam.csv'
     df = pd.read_csv(file, encoding = 'ISO-8859-1')
     df = df[['v1','v2']]
     df = df.rename(columns ={'v1': 'label', 'v2': 'message'})
[8]: df.head()
[8]:
      label
                                                        message
     0
        ham Go until jurong point, crazy.. Available only ...
     1
                                  Ok lar... Joking wif u oni...
     2 spam Free entry in 2 a wkly comp to win FA Cup fina...
       ham U dun say so early hor... U c already then say...
         ham Nah I don't think he goes to usf, he lives aro...
```

2 Taking a subset of the dataset and splitting it to create train and test datasets

```
[9]: df_large = df.sample(frac = 0.4, random_state = 0)
df_large
```

```
[9]:
           label
                                                             message
      4456
             ham
                 Aight should I just plan to come up later toni...
      690
                                                  Was the farm open?
             ham
      944
                  I sent my scores to sophas and i had to do sec...
             ham
                  Was gr8 to see that message. So when r u leavi...
      3768
      1189
                  In that case I guess I'll see you at campus lodge
      2672
             ham
                                         Super msg da:)nalla timing.
                  Guy, no flash me now. If you go call me, call ...
      5076
             ham
      3302
             ham
                         It'll be tough, but I'll do what I have to
      4225
                  Ok thats cool. Its , just off either raglan rd...
             ham
      5343
             ham No go. No openings for that room 'til after th...
      [2229 rows x 2 columns]
[10]: df_large['label'] = df_large['label'].map({'spam':1, 'ham':0}).astype(int)
[11]: X = df large['message'].values
      y = df_large['label'].values
      X_train, X_test, y_train, y_test = train_test_split(X, y, random_state=0)
[12]: f2score = make_scorer(fbeta_score, beta=2)
[13]: counter = Counter(y)
      estimate = counter[0] / counter[1]
```

2.0.1 Feature Engineering

We will use the featurizer class that was provided in the lecture along with few additions to count exclamations and misspelled words. We will also include the definition for the Custom Preprocessor class since this is used in the featurizer class.

Spacy Preprocessor

```
[14]: class SpacyPreprocessor(BaseEstimator, TransformerMixin):

"""

A text preprocessor that utilizes spaCy for efficient and flexible NLP.

Designed as a part of a scikit-learn

pipeline, it provides a wide range of text cleaning and preprocessing

functionalities.

Attributes:

model (str): The spaCy language model to be used for tokenization and

other NLP tasks.

batch_size (int): The number of documents to process at once during

spaCy's pipeline processing.
```

lemmatize (bool): If True, lemmatize tokens.

lower (bool): If True, convert all characters to lowercase.

remove_stop (bool): If True, remove stopwords.

remove_punct (bool): If True, remove punctuation.

remove_email (bool): If True, remove email addresses.

remove_url (bool): If True, remove URLs.

remove_num (bool): If True, remove numbers.

stemming (bool): If True, apply stemming to tokens (mutually exclusive ω with lemmatization).

 $add_user_mention_prefix~(bool):~If~True,~add~'@'~as~a~separate~token_{\sqcup}\\ \hookrightarrow (useful~for~user~mentions~in~social$

media data).

basic_clean_only (bool): If True, perform only basic cleaning (HTML $_{\sqcup}$ +tags removal, line breaks, etc.)

and ignore other preprocessing steps.

Methods:

basic clean(text: str) -> str:

Performs basic cleaning of the text such as removing HTML tags and $_{\!\!\!\!\perp}$ +excessive whitespace.

spacy_preprocessor(texts: list) -> list:

Processes a list of texts through the spaCy pipeline with specified $_{\!\!\!\perp}$ -preprocessing options.

fit(X, y=None) -> 'SpacyPreprocessor':

change the state of the object.

transform(X, y=None) -> list:

Transforms the provided data using the defined preprocessing \neg pipeline. Performs basic cleaning,

and if `basic_clean_only` is False, it applies advanced \textit{spaCy}_{\sqcup} $_{\hookrightarrow} \textit{preprocessing steps}.$

Raises:

ValueError: If both 'lemmatize' and 'stemming' are set to True.

 $\label{lem:valueError:} \textit{ValueError: If 'basic_clean_only' is True but other processing options} _{\hookrightarrow} \textit{are also set to True.}$

TypeError: If the input X is not a list or a numpy array.

```
def __init__(self, model, *, batch_size = 64, lemmatize=True, lower=True, __
→remove_stop=True,
               remove_punct=True, remove_email=True, remove_url=True,_
→remove num=False, stemming = False,
               add_user_mention_prefix=True, remove_hashtag_prefix=False,__
⇔basic_clean_only=False):
      self.model = model
      self.batch_size = batch_size
      self.remove_stop = remove_stop
      self.remove_punct = remove_punct
      self.remove_num = remove_num
      self.remove_url = remove_url
      self.remove_email = remove_email
      self.lower = lower
      self.add_user_mention_prefix = add_user_mention_prefix
      self.remove_hashtag_prefix = remove_hashtag_prefix
      self.basic_clean_only = basic_clean_only
      if lemmatize and stemming:
           raise ValueError("Only one of 'lemmatize' and 'stemming' can be ⊔
Grue.")
       # Validate basic_clean_only option
      if self.basic_clean_only and (lemmatize or lower or remove_stop or_u
→remove_punct or remove_num or stemming or
                                     add_user_mention_prefix or_
→remove_hashtag_prefix):
           raise ValueError("If 'basic_clean_only' is set to True, other_
⇔processing options must be set to False.")
       # Assign lemmatize and stemming
      self.lemmatize = lemmatize
      self.stemming = stemming
  def basic_clean(self, text):
      soup = BeautifulSoup(text, "html.parser")
      text = soup.get_text()
      text = re.sub(r'[\n\r]', '', text)
      return text.strip()
  def get_cores(self):
       Get the number of CPU cores to use in parallel processing.
```

```
# Get the number of CPU cores available on the system.
      num_cores = os.cpu_count()
       if num_cores < 3:</pre>
           use_cores = 1
       else:
           use_cores = num_cores // 2 + 1
      return use_cores
  def spacy_preprocessor(self, texts):
      final result = []
      nlp = spacy.load(self.model)
       # Disable unnecessary pipelines in spaCy model
      if self.lemmatize:
           # Disable parser and named entity recognition
           disabled_pipes = ['parser', 'ner']
       else:
           # Disable tagger, parser, attribute ruler, lemmatizer and named_
⇔entity recognition
           disabled_pipes = ['tok2vec', 'tagger', 'parser', 'attribute_ruler',_
with nlp.select_pipes(disable=disabled_pipes):
         # Modify tokenizer behavior based on user_mention_prefix and_
⇔hashtaq_prefix settings
         if self.add_user_mention_prefix or self.remove_hashtag_prefix:
             prefixes = list(nlp.Defaults.prefixes)
             if self.add_user_mention_prefix:
                prefixes += ['0'] # Treat '0' as a separate token
             if self.remove_hashtag_prefix:
                prefixes.remove(r'#') # Don't separate '#' from the
\hookrightarrow following text
             prefix_regex = spacy.util.compile_prefix_regex(prefixes)
             nlp.tokenizer.prefix_search = prefix_regex.search
         # Process text data in parallel using spaCy's nlp.pipe()
         for doc in nlp.pipe(texts, batch_size=self.batch_size, n_process=self.
→get_cores()):
             filtered tokens = []
             for token in doc:
                 # Check if token should be removed based on specified filters
                 if self.remove_stop and token.is_stop:
                     continue
                 if self.remove_punct and token.is_punct:
                     continue
                 if self.remove_num and token.like_num:
                     continue
```

```
if self.remove_url and token.like_url:
                     continue
                 if self.remove_email and token.like_email:
                     continue
                 # Append the token's text, lemma, or stemmed form to the
\rightarrow filtered_tokens list
                 if self.lemmatize:
                     filtered_tokens.append(token.lemma_)
                 elif self.stemming:
                     filtered_tokens.append(PorterStemmer().stem(token.text))
                 else:
                     filtered_tokens.append(token.text)
             # Join the tokens and apply lowercasing if specified
             text = ' '.join(filtered_tokens)
             if self.lower:
                 text = text.lower()
             final_result.append(text.strip())
      return final result
  def fit(self, X, y=None):
      return self
  def transform(self, X, y=None):
      try:
           if not isinstance(X, (list, np.ndarray)):
               raise TypeError(f'Expected list or numpy array, got {type(X)}')
          x_clean = [self.basic_clean(text).encode('utf-8', 'ignore').
⇒decode() for text in X]
           # Check if only basic cleaning is required
           if self.basic_clean_only:
               return x_clean # Return the list of basic-cleaned texts
           x_clean_final = self.spacy_preprocessor(x_clean)
           return x_clean_final
       except Exception as error:
           print(f'An exception occurred: {repr(error)}')
```

####Featurizer

```
[15]: class ManualFeatures (TransformerMixin, BaseEstimator):
          """A transformer class for extracting manual features from text data.
          This class is designed to be used in a scikit-learn pipeline. It uses the \Box
       \hookrightarrow spaCy
          library to extract a variety of manual features from text data, such as
          part-of-speech (POS) features, named entity recognition (NER) features,
          and count-based features.
          11 11 11
          def __init__(self, spacy_model='en_core_web_sm', batch_size = 64,_
       pos_features = True, ner_features = True, count_features = True):
              Initialize the feature extractor.
              Parameters
               _____
              spacy_model : str
                   The name of the spaCy model to use for feature extraction.
              pos_features : bool, optional (default=True)
                   Whether to extract part-of-speech (POS) features from the text data.
              ner_features : bool, optional (default=True)
                   Whether to extract named entity recognition (NER) features from the \Box
       \rightarrow text data.
              count_features : bool, optional (default=True)
                   Whether to extract count-based features from the text data.
               11 11 11
              self.spacy_model = spacy_model
              self.batch_size = batch_size
              self.pos_features = pos_features
              self.ner_features = ner_features
              self.count_features = count_features
          def get_cores(self):
              Get the number of CPU cores to use in parallel processing.
              # Get the number of CPU cores available on the system.
              num_cores = os.cpu_count()
              if num_cores < 3:</pre>
                  use_cores = 1
              else:
```

```
use_cores = num_cores // 2 + 1
      return num_cores
  def get_pos_features(self, cleaned_text):
      nlp = spacy.load(self.spacy_model)
      noun count = []
      aux_count = []
      verb count = []
      adj_count =[]
      # Disable the lemmatizer and NER pipelines for improved performance
      disabled_pipes = ['lemmatizer', 'ner']
      with nlp.select_pipes(disable=disabled_pipes):
          n_process = self.get_cores()
          for doc in nlp.pipe(cleaned_text, batch_size=self.batch_size,_
→n_process=n_process):
              # Extract nouns, auxiliaries, verbs, and adjectives from the
\rightarrow document
              nouns = [token.text for token in doc if token.pos_ in_
auxs = [token.text for token in doc if token.pos_ in ["AUX"]]
              verbs = [token.text for token in doc if token.pos in ["VERB"]]
              adjectives = [token.text for token in doc if token.pos_ in_
⇔["ADJ"]]
              # Store the count of each type of word in separate lists
              noun_count.append(len(nouns))
              aux_count.append(len(auxs))
              verb_count.append(len(verbs))
              adj_count.append(len(adjectives))
      # Stack the count lists vertically to form a 2D numpy array
      return np.transpose(np.vstack((noun_count, aux_count, verb_count, u
→adj count)))
  def get_ner_features(self, cleaned_text):
      nlp = spacy.load(self.spacy_model)
      count_ner = []
       # Disable the tok2vec, tagger, parser, attribute ruler, and lemmatizer
⇒pipelines for improved performance
      disabled_pipes = ['tok2vec', 'tagger', 'parser', 'attribute_ruler', |
```

```
with nlp.select_pipes(disable=disabled_pipes):
          n_process = self.get_cores()
          for doc in nlp.pipe(cleaned text, batch size=self.batch size, __
→n_process=n_process):
             ners = [ent.label_ for ent in doc.ents]
             count ner.append(len(ners))
      # Convert the list of NER counts to a 2D numpy array
      return np.array(count_ner).reshape(-1, 1)
  def get_count_features(self, cleaned_text):
      list_count_words = []
      list_count_characters = []
      list_count_characters_no_space = []
      list_avg_word_length = []
      list count digits = []
      list_count_numbers = []
      list count misspell=[]
      list_count_sentences = []
      nlp = spacy.load(self.spacy model)
      with nlp.select_pipes(disable=disabled_pipes):
          if not nlp.has_pipe('sentencizer'):
             nlp.add pipe('sentencizer')
          n process = self.get cores()
         for doc in nlp.pipe(cleaned_text, batch_size=self.batch_size,_
→n process=n process):
             count_word = len([token for token in doc if not token.is_punct])
             count_char = len(doc.text)
             count_char_no_space = len(doc.text_with_ws.replace(' ', ''))
             avg_word_length = count_char_no_space / (count_word + 1)
             count_numbers = len([token for token in doc if token.is_digit])
             count_sentences = len(list(doc.sents))
             list count words.append(count word)
             list_count_characters.append(count_char)
             list count characters no space append(count char no space)
             list_avg_word_length.append(avg_word_length)
             list count numbers.append(count numbers)
             list_count_sentences.append(count_sentences)
      count_features = np.vstack((list_count_words, list_count_characters,_
space, list_avg_word_length,
                                list_count_numbers, list_count_sentences))
```

```
return np.transpose(count_features)
  def fit(self, X, y=None):
       Fit the feature extractor to the input data.
       This method does not actually do any fitting, as the feature extractor u
\hookrightarrow is stateless.
       It simply returns the instance of the class.
       Parameters:
       X (list or numpy.ndarray): The input data.
       y (list or numpy.ndarray, optional): The target labels. Not used in \sqcup
⇔this implementation.
       Returns:
       FeatureExtractor: The instance of the class.
       return self
  def transform(self, X, y=None):
       Transform the input data into a set of features.
       Parameters:
       X (list or numpy.ndarray): The input data.
       y (list or numpy.ndarray, optional): The target labels. Not used in_{\sqcup}
\hookrightarrow this implementation.
       Returns:
       tuple: A tuple containing a 2D numpy array with shape (len(X), \bot)
\negnum_features) where num_features is the number of features extracted and a_{\sqcup}
\hookrightarrow list of feature names.
       Raises:
       TypeError: If the input data is not a list or numpy array.
       Exception: If an error occurs while transforming the data into features.
       11 11 11
       try:
           # Check if the input data is a list or numpy array
           if not isinstance(X, (list, np.ndarray)):
               raise TypeError(f"Expected list or numpy array, got {type(X)}")
           preprocessor1 = SpacyPreprocessor(model='en_core_web_sm',__
⇒batch_size=64, lemmatize=False, lower=False,
```

```
remove_stop=False, remove_email=True,
                                   remove_url=True, remove_num=False,__
⇔stemming=False,
                                   add_user_mention_prefix=True,_
→remove_hashtag_prefix=False, basic_clean_only=False)
          preprocessor2 = SpacyPreprocessor(model='en_core_web_sm',__
⇔batch_size=64, lemmatize=False, lower=False,
                                   remove_stop=False, remove_punct=False,_
→remove_email=True,
                                  remove_url=True, remove_num=False,_
⇔stemming=False,
                                   add_user_mention_prefix=True,_
→remove_hashtag_prefix=False, basic_clean_only=False)
          feature names = []
          if self.pos_features or self.ner_features:
               cleaned_x_count_ner_pos = preprocessor2.fit_transform(X)
          if self.count_features:
               cleaned_x_count_features = preprocessor1.fit_transform(X)
              count_features = self.
→get_count_features(cleaned_x_count_features)
               feature_names.extend(['count_words', 'count_characters',
                                     'count_characters_no_space', __
'count_numbers', 'count_sentences'])
          else:
              count_features = np.empty(shape=(0, 0))
          if self.pos_features:
              pos_features = self.get_pos_features(cleaned_x_count_ner_pos)
              feature_names.extend(['noun_count', 'aux_count', 'verb_count', |
else:
              pos_features = np.empty(shape=(0, 0))
          if self.ner_features:
              ner_features = self.get_ner_features(cleaned_x_count_ner_pos)
              feature_names.extend(['ner'])
          else:
              ner_features = np.empty(shape=(0, 0))
          # Stack the feature arrays horizontally to form a single 2D numpy
\hookrightarrow array
          if ner_features.shape == (0, 0) and pos_features.shape == (0, 0):
            return np.hstack((count_features))
```

```
elif pos_features.shape == (0, 0):
    return np.hstack((count_features, ner_features))
elif ner_features.shape == (0, 0):
    return np.hstack((count_features, pos_features))
else:
    return np.hstack((count_features, ner_features, pos_features))
except Exception as error:
    print(f'An exception occured: {repr(error)}')
```

3 Chosen Pipeline -> Sparse Embeddings + Feature Engineering

We are using the hyperparameters as per our results from the cross validation

```
[16]: feature_extract = FeatureUnion([("sparse_embed",__
       Garage TfidfVectorizer(max_features=100, max_df=0.75)), ('feature_eng', □
       →ManualFeatures(pos_features=True, ner_features = False))])
[17]: final_pipe = Pipeline([('fe', feature_extract), ('classifier', ___
       →XGBClassifier(scale_pos_weight=estimate))])
[18]: final_pipe.fit(X_train, y_train)
[18]: Pipeline(steps=[('fe',
                       FeatureUnion(transformer list=[('sparse embed',
                                                        TfidfVectorizer(max df=0.75,
      max_features=100)),
                                                       ('feature_eng',
      ManualFeatures(ner_features=False))])),
                      ('classifier',
                       XGBClassifier(base_score=None, booster=None, callbacks=None,
                                      colsample_bylevel=None, colsample_bynode=None,
                                      colsample_bytree=None, device=None,
                                      early_stopping_rounds=None,
                                      feature_types=None, gamma=None, grow_policy=None,
                                      importance_type=None,
                                      interaction_constraints=None, learning_rate=None,
                                      max_bin=None, max_cat_threshold=None,
                                      max cat to onehot=None, max delta step=None,
                                      max_depth=None, max_leaves=None,
                                      min_child_weight=None, missing=nan,
                                      monotone_constraints=None, multi_strategy=None,
                                      n_estimators=None, n_jobs=None,
                                      num_parallel_tree=None, random_state=None,
      ...))])
```

The test score of the final pipeline is 0.8815426997245178

Running the final pipeline on the large dataset (40% of the dataset), we get a test score of 0.8815426997245178

[]: