

clean\_words = [word for word in words if (word not in stopwords\_list or word in whitelist) and len(word) > 2]
return " ...ioin(clean words) # list -> string

45

46

words = input\_text.split()

```
48
 50 def expandShortsForms(input text):
           #print("in expandShortsForms\n",input_text)
           return input_text.replace("can't", "can not").replace("won't", "will not")
 52
 def lemmatize(input_text):
55
          """ Return the base or dictionary form of a word, lemma """ #print("in Lemmatize\n", input_text) outtext= ""
 57
58
 59
60
           # Lemmatize
           from nltk.stem import WordNetLemmatizer
from nltk import pos_tag, word_tokenize, wordnet
 61
62
          from nltk.corpus.reader.wordnet import WordNetError
lemmatizer = WordNetLemmatizer()
 63
64
 65
66
           input_text = input_text.split()
           for word in input_text:
    # Get the single character pos constant from pos_tag like this:
 67
68
 69
70
                 pos_label = (pos_tag(word_tokenize(word))[0][1][0]).lower()
                71
72
73
74
75
76
77
                if pos_label == 'j': pos_label = 'a'  # 'j' <--> 'a' reassignment
 78
79
                if pos_label in ['r']: # For adverbs it's a bit different
                     80
81
                                                              ' + (wordnet.wordnet.synset(word+'.r.1').lemmas()[0].pertainyms()[0].name())
                                outtext = outtext + '
 83
                           pass
                pass
outtext = outtext + ' ' + word # To keep the word in the list
elif pos_label in ['a', 's', 'v']: # For adjectives and verbs
outtext = outtext +' ' + (lemmatizer.lemmatize(word, pos=pos_label))
else: # For nouns and everything else as it is the default kwarg
outtext = outtext +' ' + (lemmatizer.lemmatize(word))
 84
85
 86
87
 88
89
 90
91
           return outtext
93 def execute_funcs(input_text, *args):
94 funcs = list(args)
95 for func in funcs:
100 fext = func(input_text)
                input_text = func(input_text)
          return input_text
 98
def apply_funcs(input_text):
    clean_X = execute_funcs(input_text, to_lower, remove_punctuation, remove_digits,
                                          remove_stopwords,
expandShortsForms,
102
104 #
                                             Lemmatize
106
          return clean X
Wall time: 0 ns
```

## Pipeline & ColumnTransformer

```
In [22]: H
                       import warnings
                      warnings.filterwarnings('ignore')
                      df["clean_text"] = df["text"].apply(func=apply_funcs)
df["length"] = df["clean_text"].apply(func=len)
                   9 feature_columns = ['cool', 'useful', 'funny', 'clean_text', 'length']
                  11 X = df[feature_columns]
                  12 y = df["stars"]
                  14 from sklearn.model selection import train test split
                  16 X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=0, stratify=y)
                  18 from sklearn.preprocessing import LabelEncoder
                  enc = LabelEncoder()
y_train = enc.fit_transform(y=y_train)
y_test = enc.transform(y=y_test)
                  24 from sklearn.compose import ColumnTransformer
                      from sklearn.feature_extraction.text import TfidfVectorizer
from sklearn.preprocessing import OneHotEncoder
                  27 from sklearn.preprocessing import StandardScaler, RobustScaler, Normalizer
                  # Whenever the transformer expects a 1D array as input, the columns were specified as a string ("xxx").
# For the transformers which expects 2D data, we need to specify the column as a list of strings (["xxx"]).
                  ct = ColumnTransformer(transformers=[
    ("TfidfVectorizer", TfidfVectorizer(), ("clean_text")),
    ("OneHotEncoder", OneHotEncoder(handle_unknown='ignore'), (["cool", "useful", "funny"])),
    ('Normalizer', Normalizer(), (["length"])),
                           n_jobs=-1)
                  38 from sklearn.pipeline import Pipeline
                  39 from sklearn.naive_bayes import MultinomialNB
                  46 clf.fit(X_train, y_train)
```

```
48 from sklearn.metrics import contusion_matrix, accuracy_score, classification_report 49
               50
51
                  predictions = clf.predict(X_test)
print(classification_report(y_test, predictions))
print(accuracy_score(y_test, predictions))
                   print(confusion_matrix(y_test, predictions))
                             precision
                                           recall f1-score support
                                   0.00
                                             0.00
                                                         0.00
                          0
                                   0.00
                                             0.00
                                                         0.00
                                                                     185
                                   0.00
                                              0.00
                                                         0.00
                                                                    292
                                   0.38
                                             0.87
                                                         0.53
                                                                     705
                                   0.62
                                              0.36
                                                         0.45
                                                                     668
                   accuracy
                                                         0.43
                                                                    2000
                                   0.20
                                             0.25
                                                                    2000
                 macro avg
                                                         0.20
              weighted avg
                                   0.34
                                             0.43
                                                         0.34
                                                                    2000
              0.4265
                      0
                         0 125 25]
              11
                  0
                  0
                      0
                           0 174 11]
                           0 271 21]
                  a
                      a
                           0 613 921
                           0 428 240]]
              Wall time: 10.9 s
In [23]: ▶
               1 from sklearn import set_config
                   set_config(display="diagram")
                4 clf
    Out[23]:
                                                  Pipeline
                                   ColumnTransformer: ColumnTransformer
                 TfidfVectorizer
                                                  OneHotEncoder
                                                                                 Normalizer
                 TfidfVectorizer
                                                  OneHotEncoder
                                                                                 Normalizer
                TfidfVectorizer() | OneHotEncoder(handle_unknown='ignore') | Normalizer()
                                                MultinomialNB
                                              MultinomialNB()
```

## make\_pipeline & make\_column\_transformer

```
In [24]: № 1 %%time
                      import warnings
                      warnings.filterwarnings('ignore')
                     df["clean_text"] = df["text"].apply(func=apply_funcs)
df["length"] = df["clean_text"].apply(func=len)
                      feature_columns = ['cool', 'useful', 'funny', 'clean_text', 'length']
                  11 X = df[feature_columns]
                 12 y = df["stars"]
                  14 from sklearn.model_selection import train_test_split
                  16 X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=0, stratify=y)
                  18 from sklearn.preprocessing import LabelEncoder
                  20 enc = LabelEncoder()
                 y_train = enc.fit_transform(y=y_train)
y_test = enc.transform(y=y_test)
                  24 from sklearn.compose import make_column_transformer
                     from sklearn.feature_extraction.text import TfidfVectorizer, CountVectorizer
from sklearn.preprocessing import OneHotEncoder
                  27 from sklearn.preprocessing import StandardScaler, RobustScaler, Normalizer
                  29 # This is a shorthand for the ColumnTransformer constructor; it does not require, and does not permit, naming the
                      # transformers
                  31 ct = make column transformer(
                           - mode_cotion_collection |
(CountVectorizer(), ("clean_text")),
(OneHotEncoder(handle_unknown='ignore'), (["cool", "useful", "funny"])),
(StandardScaler(), (["length"])),
                           n jobs=-1)
                 from sklearn.pipeline import make_pipeline
from lightgbm import LGBMClassifier
                  39 from sklearn.naive bayes import MultinomialNB
                 # This is a shorthand for the Pipeline constructor; it does not require, and does not permit, naming the estimators.

# Instead, their names will be set to the lowercase of their types automatically.

clf = make_pipeline(ct, LGBMClassifier())
                  45 clf.fit(X train, y train)
                 47 from sklearn.metrics import confusion matrix, accuracy score, classification report
                 49 predictions = clf.predict(X test)
                     print(classification_report(y_test, predictions))
print(accuracy_score(y_test, predictions))
print(confusion_matrix(y_test, predictions))
                                   precision
                                                  recall f1-score support
                               0
                                         0.62
                                                      0.49
                                                                   0.54
                                         0.46
                                                      0.26
                                                                   0.34
                                                                                 185
                                         0.39
                                                      0.22
                                                                   0.28
                                                                                 292
                                         0 49
                                                      0.60
                                                                   0.54
                                                                                 705
                                         0.59
                                                                   0.62
                                                                                 668
                                                      0.65
                                                                   0.52
                                                                                2000
                      accuracy
                macro avg
weighted avg
                                         0.51
                                                      0.45
                                                                   0.46
                                                                                2000
                                                                   0.51
                                                                                2000
                                         0.52
                                                      0.52
```

```
[[ 73 24 14 21 18]
[ 27 49 37 55 17]
[ 5 20 64 157 46]
[ 7 11 42 425 220]
[ 6 2 9 214 437]]
Wall time: 11.2 s
4 clf
   Out[25]:
                                              Pipeline
                                 columntransformer: ColumnTransformer
                                                                        standardscaler
               countvectorizer
                                            onehotencoder
               CountVectorizer
                                            OneHotEncoder
                                                                      StandardScaler
              CountVectorizer() | OneHotEncoder(handle_unknown='ignore') | StandardScaler()
                                           LGBMClassifier
                                           LGBMClassifier()
In [26]: ► I from sklearn.metrics import plot_confusion_matrix
              display = plot_confusion_matrix(estimator=clf, X=X_test, y_true=y_test, cmap="Blues", values_format='.3g')
                                                400
                                                 350
                                         17
                   27
                                   55
                                                 300
                                                250
                             64
                                  157
                                        46
                        20
                                                200
                                                - 150
                        11
                             42
                                                - 100
                                   214
                                                - 50
In [27]: ► display.confusion_matrix
```