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65070501037
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Paweekorn Soratyathorn
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```
1 import pandas as pd
 2 import numpy as np
 3 import matplotlib.pyplot as plt
 1 df = pd.read_csv('IMDB Dataset.csv')
 2 df.head()
₹
                                             review sentiment

    One of the other reviewers has mentioned that

                                                        positive
          A wonderful little production. <br /><br />The...
                                                        positive
     2 I thought this was a wonderful way to spend ti...
                                                        positive
            Basically there's a family where a little boy ...
                                                       negative
          Petter Mattei's "Love in the Time of Money" is...
                                                        positive
Next steps:
              Generate code with df
                                        View recommended plots
                                                                        New interactive sheet
 1 pd.crosstab(df.sentiment, columns="Count").sort_values(by='Count', ascending=False)
\rightarrow
          col_0 Count
                          sentiment
                           ıl.
      negative
                 25000
       positive
                 25000
```

# Data cleansing

```
1 import nltk
2 from nltk.corpus import stopwords
3 import re
4
5 nltk.download('stopwords')
7 REPLACE_BY_SPLACE_RE = re.compile(r'[/(){}\[\]\\[@,;]')
8 BAD_SYMBOLS_RE = re.compile(r'[^0-9a-z #+_]')
9 STOPWORDS = set(stopwords.words('english'))
   [nltk_data] Downloading package stopwords to /root/nltk_data...
   [nltk_data] Unzipping corpora/stopwords.zip.
1 def clean_text(text):
   text = text.lower()
   text = REPLACE_BY_SPLACE_RE.sub(' ', text)
   text = BAD_SYMBOLS_RE.sub('', text)
   text = text.replace('x', '')
   text = ' '.join(word for word in text.split() if word not in STOPWORDS)
6
   return text
1 df['review'] = df['review'].apply(clean_text)
2 df['review'] = df['review'].str.replace(r'\d+','')
1 def print plot(index):
   example = df[df.index == index][['review', 'sentiment']].values[0]
3
   if len(example) > 0:
4
     print(example[0])
     print('Review: ',example[1])
5
7 print_plot(10)
   phil alien one quirky films humour based around oddness everything rather actual punchlinesbr br first odd pretty funny movie progre
```

Review: negative

## Preprocessing

#### **Train Tokenizer**

```
1 from tensorflow.keras.preprocessing.text import Tokenizer
 3 MAX_NB_WORDS = 50000
 5 MAX_SEQUENCE_LENGTH = 250
 7 EMBEDDING_DIM = 100
8
9 tokenizer = Tokenizer(num_words = MAX_NB_WORDS,
                            filters = '!"#$%&()*+,-./:;<=>?@[\]^_`{|}~',
10
11
                            lower=True)
12
13 tokenizer.fit_on_texts(df['review'].values)
14 word index = tokenizer.word index
15 print('Found %s unique tokens.'%len(word_index))
Found 165306 unique tokens.
Pad sequence
 1 from tensorflow.keras.preprocessing.sequence import pad_sequences
 3 X = tokenizer.texts_to_sequences(df['review'].values)[:2500]
 4 X = pad_sequences(X, maxlen = MAX_SEQUENCE_LENGTH)
 5 print('Shape of data tensor:', X.shape)
Shape of data tensor: (2500, 250)
 1 print(df['review'].values[0], '\n')
🚁 one reviewers mentioned watching 1 oz episode youll hooked right eactly happened mebr br first thing struck oz brutality unflinchinį
    array([
               0,
                       0,
                              0,
                                     0,
                                            0,
                                                   0,
                                                                 0,
                                                                        0,
               0,
                                                   0,
                       0,
                              0,
                                     0,
                                            0,
                                                          0,
                                                                 0,
                                                                        0,
               0,
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                             0.
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                                                         0.
                                                                 0.
                                                                        0.
               0,
                      0,
                             0,
                                     0,
                                           4,
                                               1834,
                                                        950,
                                                                57,
                                                                      233,
             3193,
                    288,
                           353,
                                 3079,
                                         110,
                                                 492,
                                                        480,
                                                              2107,
                                                                        1,
              20,
                     58,
                           3138,
                                 3193,
                                         5451, 15271,
                                                         51,
                                                               461,
                                                                      182,
                    560,
                            53,
                                 1585,
                                          42,
                                               8154,
             110,
                                                       5657,
                                                            11761,
                           5452,
             2394,
                                 1337,
                   5953,
                                          264,
                                                       3267,
                                                                      239,
            23365,
                      1,
                           364,
                                 3193, 11400,
                                                 237, 15905,
                                                              6763,
             947,
                   2521,
                           1257, 25213,
                                         425,
                                                4483,
                                                       2409,
                                                              1081,
                                                                     6991,
             2860, 12894,
                           302, 17412,
                                                4942,
                                                       3559,
                                         214,
                                                               425,
                                                                      241,
             8294, 40799, 15272,
                                 5061,
                                                      18295,
                                         7725,
                                                2315.
                                                               224,
                                                                     9040.
            7356, 13122,
                          8621, 34707,
                                          35.
                                                 128.
                                                       5513.
                                                                 1.
                                                                       8.
                                                               159,
                                                                     439,
              47,
                    171,
                          1191,
                                   42,
                                         557,
                                                  95,
                                                        163,
             2874,
                    706,
                            86,
                                 1150,
                                         4228,
                                                2379,
                                                        984,
                                                               706,
                                                                     1295,
             706,
                     60,
                           869,
                                   89,
                                          20,
                                                 288,
                                                         44,
                                                               106,
                                                                     3138,
             1463,
                   2090,
                           293,
                                    47,
                                         1437,
                                                 178,
                                                       1356,
                                                              1138,
                                                                     3193,
              92, 10168,
                                 1973,
                                         1976,
                                                              7856,
                                                                     6992,
                           214,
                                                 461,
             4842, 14080,
                           2861, 32394,
                                         6934,
                                              14080,
                                                        384,
                                                               515,
                                                                       15,
                     14,
                           9815,
                                  639,
                                         703,
                                                6934,
                                                        551,
                                                              1081, 20459,
             557,
                    440,
                                 1880,
                                         1081,
                                                 448,
                                                         57,
                                                              3193,
                           814,
                                                                      102,
             308.
                   3653.
                           3161.
                                   15.
                                         1090.
                                                3906,
                                                        394], dtype=int32)
    4
 1 Y = pd.get_dummies(df['sentiment']).values[:2500]
 2 print('Shape of label tensor:', Y.shape)

→ Shape of label tensor: (2500, 2)
 1 Y
→ array([[False,
                    Truel.
            [False,
                    True],
```

```
[False, True],
...,
[False, True],
[ True, False],
[False, True]])
```

# Training Model

```
1 from sklearn.model_selection import train_test_split
 3 X_train, X_test, Y_train, Y_test = train_test_split(X, Y, test_size=0.3, random_state=42)
 5 print(X_train.shape, Y_train.shape)
 6 print(X_test.shape, Y_test.shape)
   (1750, 250) (1750, 2)
    (750, 250) (750, 2)
 1 from keras import Input
 2 from keras.models import Sequential
 3 from keras.layers import Embedding, SpatialDropout1D, LSTM, Dense
 4 from keras.callbacks import EarlyStopping
 6 model = Sequential()
 7 model.add(Input(shape=(X.shape[1], )))
 8 model.add(Embedding(MAX_NB_WORDS, EMBEDDING_DIM))
9 model.add(SpatialDropout1D(0.2))
10 model.add(LSTM(100, dropout=0.2, recurrent_dropout=0.2))
11 model.add(Dense(2, activation='softmax'))
12 model.compile(loss='categorical_crossentropy', optimizer='adam', metrics=['accuracy'])
13
14 epochs=3
15 batch_size=64
16
17 history = model.fit(X_train, Y_train,
                       epochs=epochs,
19
                       batch_size=batch_size,
20
                       validation_split=0.1,
                       callbacks=[EarlyStopping(monitor='val_loss', patience=3, min_delta=0.0001)])
→ Epoch 1/3
    25/25 -
                          — 16s 504ms/step - accuracy: 0.5439 - loss: 0.6913 - val_accuracy: 0.5943 - val_loss: 0.6696
    Epoch 2/3
    25/25
                           — 25s 682ms/step - accuracy: 0.7983 - loss: 0.5829 - val_accuracy: 0.7429 - val_loss: 0.5268
    Epoch 3/3
    25/25
                           — 16s 480ms/step - accuracy: 0.9142 - loss: 0.2584 - val_accuracy: 0.8057 - val_loss: 0.4707
 1 model.summary()
```

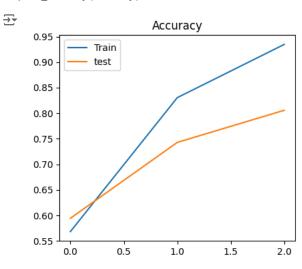
## → Model: "sequential\_1"

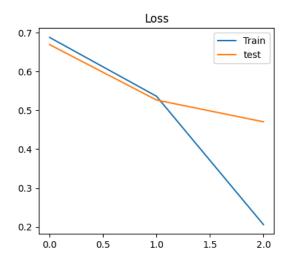
Layer (type)	Output Shape	Param #
embedding_1 (Embedding)	(None, 250, 100)	5,000,000
spatial_dropout1d_1 (SpatialDropout1D)	(None, 250, 100)	0
lstm_1 (LSTM)	(None, 100)	80,400
dense_1 (Dense)	(None, 2)	202

Total params: 15,241,808 (58.14 MB)
Trainable params: 5,080,602 (19.38 MB)
Non-trainable params: 0 (0.00 B)
Ontimiza params: 10 161,306 (38.76 MB)

### Evaluation

```
1 def plot_history(history):
    plt.figure(figsize=(10, 4))
 4
    plt.subplot(1, 2, 1)
    plt.title('Accuracy')
    plt.plot(history.history['accuracy'], label='Train')
 7
    plt.plot(history.history['val_accuracy'], label='test')
 8
    plt.legend()
 9
10
    plt.subplot(1, 2, 2)
    plt.title('Loss')
11
    plt.plot(history.history['loss'], label='Train')
12
    plt.plot(history.history['val_loss'], label='test')
14
    plt.legend()
15
16
17 plot_history(history)
```





#### **Confusion matrix**

```
1 labels = pd.get_dummies(df['sentiment']).columns
3 from sklearn.metrics import confusion_matrix
4 y_pred = model.predict(X_test)
6 pd.DataFrame(confusion_matrix(Y_test.argmax(axis=1),
                                  y_pred.argmax(axis=1)),
8
                index=labels, columns=labels)
  24/24
                            - 6s 224ms/step
            negative positive
                                \blacksquare
    negative
                           82
                                 П.
    positive
                  67
                          306
```

### **Classification Report**

 24/24		<b>– 25</b> 8/ms/step		
	precision	recall	f1-score	support
0	0.81	0.78	0.80	377
1	0.79	0.82	0.80	373
accuracy			0.80	750
macro avg	0.80	0.80	0.80	750
weighted avg	0.80	0.80	0.80	750

## ➤ Test using new review

```
1 new_review = ['''Visually beautiful but unfortunately the dialogues are poorly written and superficial,
2 Galedriel who has a lot of screen time and massive importance unfortunately is portrayed very badly makes me questic
3 how could they allow such performance, not to insult the actress at all but the performance was horrible almost
4 everyone in the show was better performing than her. Overall fun to watch but very superficial and Shallow.
5 The plot is yet to reveal itself but so far nothing interesting. Good to mention the good portrayal of elrond.
6 elendil. Music is also good and appropriate. The dark theme is a bit overlooked.''']

1 seq = tokenizer.texts_to_sequences(new_review)
2 padded = pad_sequences(seq, maxlen=MAX_SEQUENCE_LENGTH)
3 pred = model.predict(padded)
4 print(pred, labels[np.argmax(pred)])

1/1 _______ 0s 151ms/step
[[0.6472341 0.3527659]] negative

1 Start coding or generate with AI.
```