


```
i': 22, 'the': 73, 'servic': 68, 'never': 53, 'use': 81, 'thi': 75, 'again': 2, 'lov
e': 46, 'work': 92, 'worst': 93, 'custom': 28, 'care': 16, 'angri': 4, 'brilliant': 1
2, 'effort': 24, 'guy': 35, 'you': 94, 'point': 68, 'one': 56, 'finger': 29, 'millio
n': 58, 'are': 5, 'right': 66, 'back': 6, 'jewishsupremacist': 38, 'word': 91, 'free':
31, 'how': 37, 'that': 72, 'can': 15, 'cost': 19, 'verbal': 83, 'abus': 6, 'adult': 1,
'teen': 71, 'right': 49, 'libard': 44, 'sjw': 78, 'liber': 43, 'polit': 61, 'pic': 5
8, 'say': 67, 'otherwis': 57, 'for': 38, 'young': 95, 'girl': 33, 'confin': 18, 'kitch
en': 40, 'void': 84, 'mean': 48, 'beyond': 9, 'cheap': 17, 'public': 63, 'topoli': 78,
'night': 54, 'faith': 28, 'ever': 25, 'vaitacacommafiasdv': 82, 'when': 87, 'block': 1
1, 'troll': 88, 'becaus': 7, 'promis': 62, 'blacklivesmatt': 10, 'amp': 3, 'let': 42,
'hi': 36, 'nonsens': 65, 'rant': 64, 'dinner': 21, 'sister': 69, 'who': 88, 'els': 25,
'plan': 59, 'watch': 86, 'tomorrow': 77)
['abus', 'adult', 'again', 'amp', 'angri', 'are', 'back', 'becaus', 'best', 'beyond',
'blacklivesmatt', 'block', 'brilliant', 'buddi', 'but', 'can', 'care', 'cheap', 'confi
n', 'cost', 'custom', 'dinner', 'dissatisfi', 'eat', 'effort', 'els', 'even', 'expec
t', 'faith', 'finger', 'for', 'free', 'futar', 'girl', 'good', 'guy', 'hi', 'how', 'je
wishsupremacist', 'job', 'kitchen', 'ksleg', 'let', 'liber', 'libtard', 'lot', 'love',
'lunch', 'mean', 'night', 'million', 'money', 'more', 'never', 'night', 'nonsens', 'on
e', 'otherwis', 'pic', 'plan', 'point', 'polit', 'promis', 'public', 'rant', 'reflec
t', 'right', 'say', 'servic', 'sister', 'sjw', 'teen', 'that', 'the', 'they', 'thi',
'time', 'tomorrow', 'topoli', 'total', 'troll', 'use', 'vaitacacommafiasdv', 'verbal',
'void', 'want', 'watch', 'when', 'who', 'will', 'with', 'word', 'work', 'worst', 'yo
u', 'young']
```

```
[[0 0 0 ... 0 0 0]
 [0 0 0 ... 0 0 0]
 [0 0 0 ... 0 0 0]

 [0 0 0 ... 0 2 0]
 [0 0 0 ... 0 0 0]
 [0 0 0 ... 0 0 0]]
```

In [32]: # Already have frature extraction for both test and train data in bag of words

```
In [33]: X_train = bag_of_words_all_data[:10]
Y_train = allData['Polarity'][:10]

X_test = bag_of_words_all_data[10:]
Y_test = allData['Polarity'][10:]
```

```
In [34]: decisionTree_bag_of_words = DecisionTreeClassifier(criterion = "entropy" , random_sta
decisionTree_bag_of_words.fit(X_train , Y_train)
Y_pred = decisionTree_bag_of_words.predict(X_test)
print('Predicted values: ', end= ' ')
print(Y_pred)

print('Accuracy: ', end= ' ')
print(accuracy_score(Y_test , Y_pred))
```

Predicted values: [0 1 0 1 1]
Accuracy: 1.0

Decision Tree through TF-IDF

```
In [35]: tfidf = TfidfVectorizer()

tfidf_matrix = tfidf.fit_transform(allData['Processed_Comment'])

allData_tfidf = pd.DataFrame(tfidf_matrix.todense())

display(allData_tfidf)
```

	0	1	2	3	4	5	6	7	8	9	...
0	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.422559	0.000000	...
1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	...
2	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	...
3	0.000000	0.000000	0.314278	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	...
4	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	...
5	0.000000	0.000000	0.000000	0.000000	0.458638	0.000000	0.000000	0.000000	0.000000	0.000000	...
6	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	...
7	0.000000	0.000000	0.000000	0.000000	0.000000	0.246500	0.283877	0.000000	0.000000	0.000000	...
8	0.292656	0.292656	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	...
9	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	...
10	0.000000	0.000000	0.000000	0.000000	0.212639	0.000000	0.000000	0.000000	0.000000	0.244882	...
11	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	...
12	0.000000	0.000000	0.000000	0.281105	0.000000	0.000000	0.000000	0.281105	0.000000	0.000000	...
13	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	...
14	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	...

15 rows × 96 columns

```
In [36]: X_train = allData_tfidf[:10]
Y_train = allData['Polarity'][:10]

X_test = allData_tfidf[10:]
Y_test = allData['Polarity'][10:]
```

```
In [37]: decisionTree_tfidf = DecisionTreeClassifier(criterion = "entropy" , random_state = 100)
decisionTree_tfidf.fit(X_train , Y_train)
Y_pred = decisionTree_tfidf.predict(X_test)
print('Predicted values: ', end= ' ')
print(Y_pred)

print('Accuracy: ', end= ' ')
print(accuracy_score(Y_test , Y_pred))
```

Predicted values: [1 1 0 1 1]
Accuracy: 0.8

Application Phase

Full Decision Tree Modal

```
In [38]: full_decisionTree = DecisionTreeClassifier(criterion = "entropy" , random_state = 100)
full_decisionTree.fit(bag_of_words_all_data[:], allData['Polarity'][:])
```

Out[38]: DecisionTreeClassifier(criterion='entropy' , random_state=100)

Saving in Pickle File

```
In [39]: filename = 'finalized_model.sav'
pickle.dump(full_decisionTree, open(filename, 'wb'))
```

Loading Pickle Model

```
In [40]: loaded_model = pickle.load(open(filename, 'rb'))
```

Input from User

```
In [41]: userTweet = input('Enter A Tweet: ')

Enter A Tweet: they want to kill this sad is it not @someone #eating
```

```
In [42]: userTweet = remove_pattern(userTweet , "@{w}")

r = re.findall("[a-zA-Z]" , userTweet)
text = ''
for i in r:
    text = re.sub(i," ",userTweet)

userTweet = text

userTweet = userTweet.lower()

userTweet = [x for x in userTweet.split() if len(x) >= 3]
```

```
In [43]: ps = PorterStemmer()

for x in range(len(userTweet)):
    userTweet[x] = ps.stem(userTweet[x])

userTweet
```

Out[43]: ['they', 'want', 'kill', 'thi', 'sad', 'not', 'eat']

```
In [44]: # All of Our Feature names are in cv.get_feature_names()

lables = cv.get_feature_names()

userTweetArr = []

for x in lables:
    total = 0
    for y in userTweet:
        if x == y:
            total = total + 1
    userTweetArr.append(total)

predictArr = [userTweetArr]
```

```
In [45]: predict = loaded_model.predict(predictArr)
```

```
In [46]: if predict == 1:
    print('The Tweet Polarity is Happy')
else:
    print('The Tweet Polarity is Sad')
```

The Tweet Polarity is Happy

In []: