vectorization and model

May 31, 2020

```
[2]: import pandas as pd
     from sklearn.model_selection import train_test_split
     from sklearn.feature_extraction.text import TfidfVectorizer
     from sklearn.model_selection import train_test_split, cross_val_score
     from sklearn.svm import LinearSVC
     from sklearn.metrics import accuracy_score
     from sklearn.metrics import *
     #Load dataset from CSV file
     df=pd.read_csv('mal_corpus.
     →csv',names=['sentence','ambiguous_word','label','sense'],skiprows=1)
     df x=df["sentence"]
     df_y=df['label']
     #vectorization and data splitting
     cv = TfidfVectorizer(input="content",encoding="utf-8",norm="12")
     x_train,x_test,y_train,y_test=train_test_split(df_x,df_y,test_size=0.
     \rightarrow23,random_state=11)
     xtrain_cv=cv.fit_transform(x_train)
     xtest cv=cv.transform(x test)
     clf = LinearSVC(penalty='12', loss='squared_hinge', dual=True, tol=0.0001, C=1.
     →0, multi_class='ovr', fit_intercept=True, intercept_scaling=1,__
     ⇒class_weight=None, verbose=0, random_state=None, max_iter=1000)
     y_train=y_train.astype('int')
     y_test=y_test.astype('int')
     #Model Training using training set data
     clf.fit(xtrain_cv,y_train)
     #Predicting the label of test set data
     pred=clf.predict(xtest_cv)
     #Calculating Accuracy
     Accuracy=accuracy_score(y_test,pred)
     Accuracy=Accuracy*100
     Accuracy
```

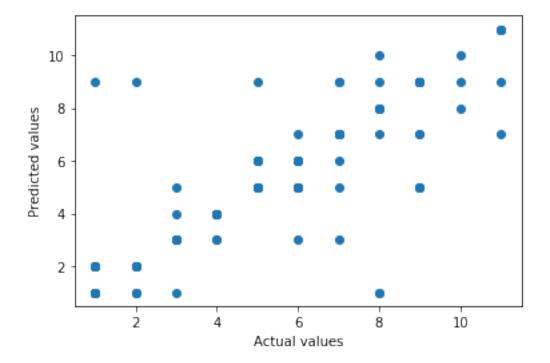
[2]: 73.07692307692307

```
[]: #Plotting the model
```

```
[3]: from matplotlib import pyplot as plt
plt.scatter(y_test, pred)

plt.xlabel("Actual values")
plt.ylabel("Predicted values")
print("\n\nAccuracy Score : ")
print(accuracy_score(y_test,pred ))
```

Accuracy Score: 0.7307692307692307



[]: #Printing Classification Report

[35]: print(classification_report(y_test, pred))

	precision	recall	f1-score	support
1	0.70	0.56	0.62	25
2	0.66	0.83	0.73	23

```
3
                    0.87
                               0.90
                                          0.89
                                                       30
            4
                    0.95
                               0.91
                                          0.93
                                                       22
            5
                    0.70
                               0.89
                                          0.78
                                                       54
            6
                    0.65
                               0.39
                                          0.49
                                                       28
            7
                    0.72
                               0.72
                                          0.72
                                                        18
            8
                               0.44
                                                        9
                    0.80
                                          0.57
            9
                               0.62
                    0.56
                                          0.59
                                                       16
                               0.33
           10
                    0.50
                                          0.40
                                                        3
           11
                     1.00
                               0.67
                                          0.80
                                                        6
                                          0.73
                                                      234
    accuracy
   macro avg
                    0.74
                               0.66
                                          0.68
                                                      234
weighted avg
                    0.73
                               0.73
                                          0.72
                                                      234
```

[]: #Confusion Matrix

[4]: print(confusion_matrix(y_test,pred))

```
[[14 10 0 0
               0
                  0
                     0
                         0
                           1
                                  0]
[ 3 19
                            1
                               0
                                  0]
        0
            0
               0
                  0
                     0
                         0
     0 27
            1
               1
                  0
                         0
                            0
                                  0]
[ 0
     0
        2 20
               0
                  0
                     0
                         0
                            0
                               0
                                  0]
[ 0
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         0
            0 48
                  5
                     0
                         0
                            1
                              0
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[ 0
                            0
                                  0]
         1
            0 15 11
                      1
                         0
                               0
[ 0
     0
         1
            0
               1
                  1 13
                         0
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                               0
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[ 2
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         0
                     1
                                  0]
            0
               0
                  0
                         4
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[ 0
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               4
                  0
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               0
                  0
                         1
                            1
                                  0]
                         0
                            1
                                  4]]
```

[]: