

driving_behavior_XGBoost_binary_v2

September 1, 2022

```
[ ]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
```

```
[ ]: df_training = pd.read_csv("../data_mod/train_motion_data.csv")
df_test = pd.read_csv("../data_mod/test_motion_data.csv")

df_training
```

```
[ ]:      AccX      AccY      GyroZ      Class      DiffAccX      DiffAccY      VelX \
0      0.000000      0.000000      0.101938      NORMAL      0.000000      0.000000      0.000000
1     -1.624864     -1.082492      0.135536      NORMAL     -1.624864     -1.082492     -0.812432
2     -0.594660     -0.122410      0.087888      NORMAL      1.030204      0.960082     -0.297330
3      0.738478     -0.228456      0.054902      NORMAL      1.333138     -0.106046      0.369239
4      0.101741      0.777568      0.054902      NORMAL     -0.636737      1.006023      0.050871
...      ...      ...      ...      ...      ...      ...
3639      0.915688     -2.017489     -1.236468        SLOW      2.374675     -1.824629      0.457844
3640     -1.934203      0.914925     -0.477162        SLOW     -2.849891      2.932414     -0.967102
3641     -0.222845      0.747304      0.054291        SLOW      1.711359     -0.167621     -0.111422
3642     -0.349423      0.067261     -0.004963        SLOW     -0.126579     -0.680043     -0.174712
3643     -0.402428      0.406218      0.001145        SLOW     -0.053005      0.338957     -0.201214
```

```
      VelY
0      0.000000
1     -0.541246
2     -0.061205
3     -0.114228
4      0.388784
...      ...
3639     -1.008745
3640      0.457462
3641      0.373652
3642      0.033630
3643      0.203109
```

```
[3644 rows x 8 columns]
```

```
[ ]: df_training.isna().sum()
```

```
[ ]: AccX      0
      AccY      0
      GyroZ     0
      Class     0
      DiffAccX  0
      DiffAccY  0
      VelX      0
      VelY      0
      dtype: int64
```

0.1 Change categories to numbers

```
[ ]: df_training = df_training.replace(
      {"Class": {"NORMAL": 0, "AGGRESSIVE": 1, "SLOW": 2}})
df_test = df_test.replace(
      {"Class": {"NORMAL": 0, "AGGRESSIVE": 1, "SLOW": 2}})
df_training
```

```
[ ]:      AccX      AccY      GyroZ  Class  DiffAccX  DiffAccY      VelX  \
0      0.000000  0.000000  0.101938      0  0.000000  0.000000  0.000000
1     -1.624864 -1.082492  0.135536      0 -1.624864 -1.082492 -0.812432
2     -0.594660 -0.122410  0.087888      0  1.030204  0.960082 -0.297330
3      0.738478 -0.228456  0.054902      0  1.333138 -0.106046  0.369239
4      0.101741  0.777568  0.054902      0 -0.636737  1.006023  0.050871
...      ...      ...      ...      ...      ...      ...
3639  0.915688 -2.017489 -1.236468      2  2.374675 -1.824629  0.457844
3640 -1.934203  0.914925 -0.477162      2 -2.849891  2.932414 -0.967102
3641 -0.222845  0.747304  0.054291      2  1.711359 -0.167621 -0.111422
3642 -0.349423  0.067261 -0.004963      2 -0.126579 -0.680043 -0.174712
3643 -0.402428  0.406218  0.001145      2 -0.053005  0.338957 -0.201214
```

```
      VelY
0      0.000000
1     -0.541246
2     -0.061205
3     -0.114228
4      0.388784
...      ...
3639 -1.008745
3640  0.457462
3641  0.373652
3642  0.033630
3643  0.203109
```

```
[3644 rows x 8 columns]
```

0.1.1 Only select normal and aggressive values

```
[ ]: df_training = df_training.loc[df_training['Class'] != 1]
df_test = df_test.loc[df_test['Class'] != 1]

df_training
```

```
[ ]:      AccX      AccY      GyroZ  Class  DiffAccX  DiffAccY      VelX  \
0      0.000000  0.000000  0.101938      0  0.000000  0.000000  0.000000
1     -1.624864 -1.082492  0.135536      0 -1.624864 -1.082492 -0.812432
2     -0.594660 -0.122410  0.087888      0  1.030204  0.960082 -0.297330
3      0.738478 -0.228456  0.054902      0  1.333138 -0.106046  0.369239
4      0.101741  0.777568  0.054902      0 -0.636737  1.006023  0.050871
...
3639  0.915688 -2.017489 -1.236468      2  2.374675 -1.824629  0.457844
3640 -1.934203  0.914925 -0.477162      2 -2.849891  2.932414 -0.967102
3641 -0.222845  0.747304  0.054291      2  1.711359 -0.167621 -0.111422
3642 -0.349423  0.067261 -0.004963      2 -0.126579 -0.680043 -0.174712
3643 -0.402428  0.406218  0.001145      2 -0.053005  0.338957 -0.201214

      VelY
0      0.000000
1     -0.541246
2     -0.061205
3     -0.114228
4      0.388784
...
3639 -1.008745
3640  0.457462
3641  0.373652
3642  0.033630
3643  0.203109

[2531 rows x 8 columns]
```

0.2 Normalize the data

```
[ ]: X_training = df_training.drop(columns=["Class"])
X_training = (X_training - X_training.mean()) / X_training.std() * 100

X_training["Class"] = df_training["Class"]
X_training
```

```
[ ]:      AccX      AccY      GyroZ  DiffAccX  DiffAccY      VelX  \
0     -1.855230   3.971188   88.116927    0.012569   -0.067264   -1.855230
1    -190.162298 -135.853745  119.158011  -160.827145 -106.518393 -190.162298
2     -70.770948  -11.840434   75.136116  101.988935   94.346206  -70.770948
```

3	83.727731	-25.538301	44.659418	131.975345	-10.495686	83.727731
4	9.935643	104.409213	44.659418	-63.015859	98.864017	9.935643
...
3639	104.264697	-256.626925	-1148.446773	235.073473	-179.499382	104.264697
3640	-226.011955	122.151549	-446.918404	-282.088379	288.303311	-226.011955
3641	-27.680909	100.500014	44.095035	169.414117	-16.550950	-27.680909
3642	-42.350223	12.659225	-10.650142	-12.517010	-66.941969	-42.350223
3643	-48.492982	56.442174	-5.006309	-5.234178	33.265449	-48.492982

	VelY	Class
0	3.971188	0
1	-135.853745	0
2	-11.840434	0
3	-25.538301	0
4	104.409213	0
...
3639	-256.626925	2
3640	122.151549	2
3641	100.500014	2
3642	12.659225	2
3643	56.442174	2

[2531 rows x 8 columns]

```
[ ]: X_testing = df_test.drop(columns="Class")
X_testing = (X_testing - X_testing.mean()) / X_testing.std() * 100

X_testing["Class"] = df_test["Class"]
X_testing
```

	AccX	AccY	GyroZ	DiffAccX	DiffAccY	VelX \
814	79.340838	21.963793	38.859198	4.511762	78.994793	79.340838
815	132.192943	569.257650	-11.298662	43.314646	415.888708	132.192943
816	-33.998774	10.843662	-4.956864	-136.351589	-424.416667	-33.998774
817	38.437952	57.366915	1.961463	59.378533	35.317808	38.437952
818	-21.053767	3.156469	-25.711843	-48.833088	-41.236835	-21.053767
...
3079	-95.464081	-76.862781	479.325902	-79.996459	-1.185708	-95.464081
3080	180.478081	51.148641	-645.478582	226.299722	97.246415	180.478081
3081	151.492731	-217.785674	-450.612340	-23.810860	-204.420646	151.492731
3082	105.929590	84.125864	374.398012	-37.408456	229.405378	105.929590
3083	174.027088	32.946129	63.073341	55.819369	-38.933584	174.027088

	VelY	Class
814	21.963793	0
815	569.257650	0
816	10.843662	0

```

817    57.366915    0
818     3.156469    0
...
3079  -76.862781    2
3080   51.148641    2
3081 -217.785674    2
3082   84.125864    2
3083   32.946129    2

```

```
[2270 rows x 8 columns]
```

0.3 Train model

```
[ ]: X_train = X_training.drop(columns="Class")
     y_train = X_training.Class

     X_test = X_testing.drop(columns="Class")
     y_test = X_testing.Class

```

```
[ ]: from sklearn.ensemble import GradientBoostingClassifier
     from sklearn.model_selection import RandomizedSearchCV
     from sklearn.metrics import confusion_matrix, ConfusionMatrixDisplay

```

```
[ ]: xgb = GradientBoostingClassifier()

     param_grid = {'n_estimators': np.arange(20, 80, 2), 'learning_rate': np.
     ↪ linspace(0.2, 1, 20), 'max_depth': np.arange(1, 10), 'max_features':
     ↪ ['sqrt', None], 'max_leaf_nodes': np.arange(2, 30)}

     xgb_gscv = RandomizedSearchCV(xgb, param_grid, n_iter=100, cv=5, verbose=10,
     ↪ n_jobs=10, random_state=0)
     xgb_gscv.fit(X_train, y_train)

```

```
[ ]: best_params = xgb_gscv.best_params_
     best_params

```

```
[ ]: {'n_estimators': 34,
     'max_leaf_nodes': 11,
     'max_features': None,
     'max_depth': 4,
     'learning_rate': 0.4105263157894737}

```

```
[ ]: xgb_gscv.best_score_

```

```
[ ]: 0.5282448877766603

```

0.3.1 Check for overfitting

```
[ ]: xgb_gscv.score(X_train, y_train)
```

```
[ ]: 0.7890161991307784
```

```
[ ]: xgb_gscv.score(X_test, y_test)
```

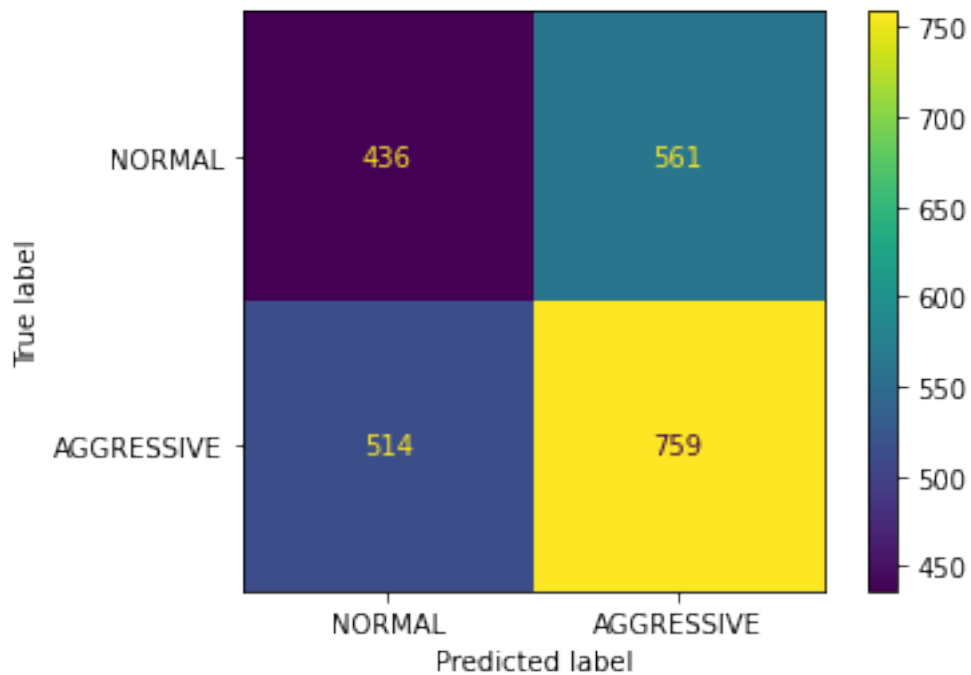
```
[ ]: 0.526431718061674
```

```
[ ]: classes = ["NORMAL", "AGGRESSIVE"]
```

```
[ ]: y_pred = xgb_gscv.predict(X_test)
```

```
CM = confusion_matrix(y_test, y_pred)
display = ConfusionMatrixDisplay(confusion_matrix=CM,
                                display_labels=classes)
display.plot()
```

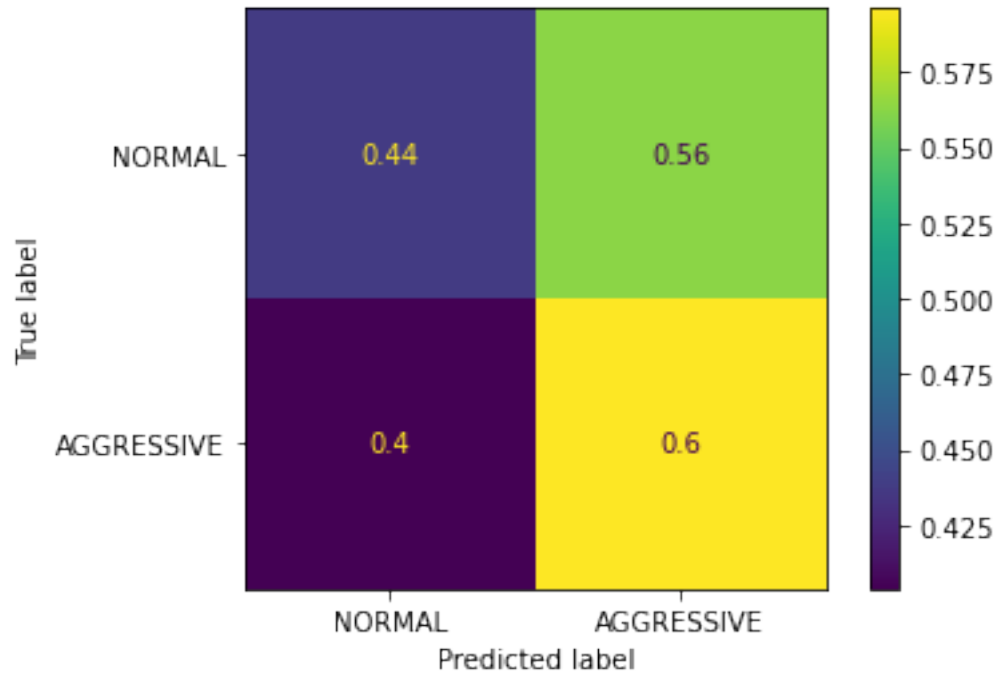
```
[ ]: <sklearn.metrics._plot.confusion_matrix.ConfusionMatrixDisplay at
0x7ff074af2220>
```



```
[ ]: CM_norm = confusion_matrix(y_test, y_pred, normalize="true")
display = ConfusionMatrixDisplay(confusion_matrix=CM_norm,
                                display_labels=classes)
```

```
display.plot()
```

```
[ ]: <sklearn.metrics._plot.confusion_matrix.ConfusionMatrixDisplay at  
0x7ff074c89ac0>
```



```
[ ]: def evaluate(model, test_features, test_labels):  
    accuracy = model.score(test_features, test_labels)  
    print('Model Performance')  
    print('Accuracy = {:.3f}%'.format(accuracy))  
  
    return accuracy  
  
base_model = GradientBoostingClassifier(n_estimators=100, learning_rate=1.0,  
    ↪max_depth=1, random_state=0)  
base_model.fit(X_train, y_train)  
base_accuracy = evaluate(base_model, X_test, y_test)  
  
best_random = xgb_gscv.best_estimator_  
random_accuracy = evaluate(best_random, X_test, y_test)  
  
print(f'Improvement of {100 * (random_accuracy - base_accuracy) / base_accuracy:  
    ↪.3f}%')
```

Model Performance
Accuracy = 0.522%.

Model Performance
Accuracy = 0.526%.
Improvement of 0.929%.

[]: