model_selection_training

April 26, 2024

[]: import pandas as pd

```
from sklearn.model_selection import train_test_split
     from sklearn.ensemble import RandomForestClassifier, GradientBoostingClassifier
     from sklearn.svm import SVC
     from sklearn.neighbors import KNeighborsClassifier
     from sklearn.neural network import MLPClassifier
     from sklearn.tree import DecisionTreeClassifier
     from sklearn.metrics import accuracy score, classification report
[]: data = pd.read_csv('./final_data/final_data.csv')
[]: | # Split the dataset into features (X) and target variable (y)
     X = data.drop(columns=['compound'])
     y = data['compound']
[]: # Encode categorical variables using one-hot encoding
     X_encoded = pd.get_dummies(X)
[]: #Split the dataset into training and testing sets (80% train, 20% test)
     X_train, X_test, y_train, y_test = train_test_split(X_encoded, y, test_size=0.
      →2, random_state=42)
[]: # Make predictions on new data
     # new data is the first row of the original data
     new data = data.iloc[[0]]
     new_data = new_data.drop(columns=["compound"])
     # Encode new data using the same encoding as the training data
     new_data_encoded = pd.get_dummies(new_data)
    Random Forest Classifier
[]: # Choose a machine learning algorithm
     model = RandomForestClassifier(n estimators=100, random state=42)
     # Train the model
     model.fit(X_train, y_train)
```

```
[]: RandomForestClassifier(random_state=42)
```

```
[]: # Evaluate the model
    y_pred = model.predict(X_test)
    accuracy = accuracy_score(y_test, y_pred)
    print(f'Accuracy: {accuracy}')

# Print classification report
    print(classification_report(y_test, y_pred))
```

Accuracy: 0.7356521739130435

	precision	recall	f1-score	support
HARD	0.77	0.80	0.78	271
HYPERSOFT	0.71	0.80	0.75	15
INTERMEDIATE	0.91	0.97	0.94	64
MEDIUM	0.73	0.68	0.70	391
SOFT	0.70	0.75	0.72	293
SUPERSOFT	0.71	0.64	0.67	53
ULTRASOFT	0.60	0.57	0.58	49
WET	0.90	0.64	0.75	14
accuracy			0.74	1150
macro avg	0.75	0.73	0.74	1150
weighted avg	0.74	0.74	0.73	1150

```
[]: # Make predictions
predictions = model.predict(new_data_encoded)
print(f'Predicted tire compound: {predictions}')
```

Predicted tire compound: ['ULTRASOFT']

Gradient Boosting Classifier

```
[]: # Choose a machine learning algorithm
model = GradientBoostingClassifier(n_estimators=100, random_state=42)

# Train the model
model.fit(X_train, y_train)
```

[]: GradientBoostingClassifier(random_state=42)

```
[]: # Evaluate the model
y_pred = model.predict(X_test)
accuracy = accuracy_score(y_test, y_pred)
print(f'Accuracy: {accuracy}')
# Print classification report
```

```
Accuracy: 0.7365217391304347
                  precision
                                recall f1-score
                                                   support
            HARD.
                        0.75
                                  0.77
                                            0.76
                                                        271
       HYPERSOFT
                        0.71
                                  0.67
                                            0.69
                                                         15
    INTERMEDIATE
                        0.93
                                  0.97
                                            0.95
                                                         64
          MEDIUM
                        0.70
                                  0.71
                                            0.70
                                                        391
                        0.73
                                  0.74
                                            0.74
                                                        293
            SOFT
       SUPERSOFT
                        0.73
                                  0.68
                                            0.71
                                                         53
       ULTRASOFT
                                  0.59
                                            0.62
                        0.66
                                                         49
             WET
                        0.90
                                  0.64
                                            0.75
                                                         14
                                            0.74
                                                       1150
        accuracy
       macro avg
                        0.76
                                  0.72
                                            0.74
                                                       1150
                        0.74
                                  0.74
                                            0.74
    weighted avg
                                                       1150
[]: # Make predictions
     predictions = model.predict(new_data_encoded)
     print(f'Predicted tire compound: {predictions}')
    Predicted tire compound: ['ULTRASOFT']
    Support Vector Machines with RBF kernel
[]: # Choose a machine learning algorithm
     model = SVC(kernel='rbf', random_state=42)
     # Train the model
     model.fit(X_train, y_train)
[]: SVC(random_state=42)
[]: # Evaluate the model
     y_pred = model.predict(X_test)
     accuracy = accuracy_score(y_test, y_pred)
     print(f'Accuracy: {accuracy}')
     # Print classification report
     print(classification_report(y_test, y_pred))
    Accuracy: 0.3391304347826087
                  precision
                                recall f1-score
                                                   support
                                  0.00
                                            0.00
            HARD
                        0.00
                                                        271
       HYPERSOFT
                        0.00
                                  0.00
                                            0.00
                                                         15
    INTERMEDIATE
                        0.00
                                  0.00
                                            0.00
                                                         64
```

print(classification_report(y_test, y_pred))

MEDIUM	0.34	1.00	0.51	391
SOFT	0.00	0.00	0.00	293
SUPERSOFT	0.00	0.00	0.00	53
ULTRASOFT	0.00	0.00	0.00	49
UNKNOWN	0.00	0.00	0.00	0
WET	0.00	0.00	0.00	14
accuracy			0.34	1150
macro avg	0.04	0.11	0.06	1150
weighted avg	0.12	0.34	0.17	1150

/Users/neahabijo/anaconda3/envs/tyre_pred/lib/python3.12/site-packages/sklearn/metrics/_classification.py:1509: UndefinedMetricWarning: Precision is ill-defined and being set to 0.0 in labels with no predicted samples. Use `zero_division` parameter to control this behavior.

_warn_prf(average, modifier, f"{metric.capitalize()} is", len(result))
/Users/neahabijo/anaconda3/envs/tyre_pred/lib/python3.12/sitepackages/sklearn/metrics/_classification.py:1509: UndefinedMetricWarning: Recall
is ill-defined and being set to 0.0 in labels with no true samples. Use
`zero_division` parameter to control this behavior.

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is ill-defined and being set to 0.0 in labels with no true samples. Use
`zero_division` parameter to control this behavior.

_warn_prf(average, modifier, f"{metric.capitalize()} is", len(result))

```
[]: # Make predictions
predictions = model.predict(new_data_encoded)
print(f'Predicted tire compound: {predictions}')
```

Predicted tire compound: ['MEDIUM']

K Nearest Neighbours Classifiers

```
[]: model = KNeighborsClassifier(n_neighbors=5)

# Train the model
model.fit(X_train, y_train)
```

[]: KNeighborsClassifier()

```
[]: # Evaluate the model
y_pred = model.predict(X_test)
accuracy = accuracy_score(y_test, y_pred)
print(f'Accuracy: {accuracy}')

# Print classification report
print(classification_report(y_test, y_pred))
```

Accuracy: 0.3243478260869565

	precision	recall	f1-score	${ t support}$
HARD	0.29	0.39	0.33	271
HYPERSOFT	0.30	0.20	0.24	15
INTERMEDIATE	0.42	0.55	0.47	64
MEDIUM	0.34	0.34	0.34	391
SOFT	0.35	0.29	0.32	293
SUPERSOFT	0.33	0.11	0.17	53
ULTRASOFT	0.12	0.06	0.08	49
UNKNOWN	0.00	0.00	0.00	0
WET	1.00	0.07	0.13	14
accuracy			0.32	1150
macro avg	0.35	0.22	0.23	1150
weighted avg	0.33	0.32	0.32	1150

/Users/neahabijo/anaconda3/envs/tyre_pred/lib/python3.12/site-packages/sklearn/metrics/_classification.py:1509: UndefinedMetricWarning: Recall is ill-defined and being set to 0.0 in labels with no true samples. Use `zero_division` parameter to control this behavior.

_warn_prf(average, modifier, f"{metric.capitalize()} is", len(result))
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/Users/neahabijo/anaconda3/envs/tyre_pred/lib/python3.12/sitepackages/sklearn/metrics/_classification.py:1509: UndefinedMetricWarning: Recall
is ill-defined and being set to 0.0 in labels with no true samples. Use
`zero_division` parameter to control this behavior.

```
_warn_prf(average, modifier, f"{metric.capitalize()} is", len(result))
```

```
[]: # Make predictions
predictions = model.predict(new_data_encoded)
print(f'Predicted tire compound: {predictions}')
```

Predicted tire compound: ['HARD']

Multi-Layer Perceptron Classifier

[]: MLPClassifier(hidden_layer_sizes=(100, 50), max_iter=500, random_state=42)

```
[]: # Evaluate the model
y_pred = model.predict(X_test)
accuracy = accuracy_score(y_test, y_pred)
print(f'Accuracy: {accuracy}')

# Print classification report
print(classification_report(y_test, y_pred))
```

Accuracy: 0.23565217391304347

	precision	recall	f1-score	support
HARD	0.24	1.00	0.38	271
HYPERSOFT	0.00	0.00	0.00	15
INTERMEDIATE	0.00	0.00	0.00	64
MEDIUM	0.00	0.00	0.00	391
SOFT	0.00	0.00	0.00	293
SUPERSOFT	0.00	0.00	0.00	53
ULTRASOFT	0.00	0.00	0.00	49
WET	0.00	0.00	0.00	14
accuracy			0.24	1150
macro avg	0.03	0.12	0.05	1150
weighted avg	0.06	0.24	0.09	1150

/Users/neahabijo/anaconda3/envs/tyre_pred/lib/python3.12/site-packages/sklearn/metrics/_classification.py:1509: UndefinedMetricWarning: Precision is ill-defined and being set to 0.0 in labels with no predicted samples. Use `zero_division` parameter to control this behavior.

_warn_prf(average, modifier, f"{metric.capitalize()} is", len(result))
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Precision is ill-defined and being set to 0.0 in labels with no predicted samples. Use `zero_division` parameter to control this behavior.

_warn_prf(average, modifier, f"{metric.capitalize()} is", len(result))

```
[]: # Make predictions
predictions = model.predict(new_data_encoded)
print(f'Predicted tire compound: {predictions}')
```

Predicted tire compound: ['HARD']

Decision Tree Classifier

```
[]: model = DecisionTreeClassifier(random_state=42)

# Train the model
model.fit(X_train, y_train)
```

[]: DecisionTreeClassifier(random_state=42)

```
[]: # Evaluate the model
y_pred = model.predict(X_test)
accuracy = accuracy_score(y_test, y_pred)
print(f'Accuracy: {accuracy}')

# Print classification report
print(classification_report(y_test, y_pred))
```

Accuracy: 0.6843478260869565

	precision	recall	f1-score	support
HARD	0.74	0.70	0.72	271
HYPERSOFT	0.65	0.87	0.74	15
INTERMEDIATE	0.91	0.94	0.92	64
MEDIUM	0.64	0.63	0.63	391
SOFT	0.67	0.71	0.69	293
SUPERSOFT	0.60	0.64	0.62	53
ULTRASOFT	0.56	0.47	0.51	49
WET	0.91	0.71	0.80	14
accuracy			0.68	1150
macro avg	0.71	0.71	0.71	1150
weighted avg	0.68	0.68	0.68	1150

```
[]: # Make predictions
predictions = model.predict(new_data_encoded)
print(f'Predicted tire compound: {predictions}')
```

Predicted tire compound: ['ULTRASOFT']