Wine Quality Prediction

March 11, 2024

```
[1]: import pandas as pd
     import matplotlib.pyplot as plt
     import seaborn as sns
     import numpy as np
[2]: data = pd.read_excel("WineQT.xlsx")
[3]: data.head()
[3]:
        fixed acidity volatile acidity
                                         citric acid residual sugar
                                                                        chlorides \
     0
                  7.4
                                    0.70
                                                 0.00
                                                                   1.9
                                                                            0.076
                  7.8
     1
                                    0.88
                                                 0.00
                                                                   2.6
                                                                            0.098
     2
                  7.8
                                                                   2.3
                                    0.76
                                                 0.04
                                                                            0.092
     3
                 11.2
                                    0.28
                                                 0.56
                                                                   1.9
                                                                            0.075
                  7.4
                                    0.70
     4
                                                 0.00
                                                                   1.9
                                                                            0.076
        free sulfur dioxide total sulfur dioxide density
                                                                pH sulphates
     0
                       11.0
                                              34.0
                                                     0.9978 3.51
                                                                         0.56
     1
                       25.0
                                              67.0
                                                     0.9968 3.20
                                                                         0.68
                                                     0.9970
     2
                       15.0
                                                                         0.65
                                              54.0
                                                              3.26
     3
                       17.0
                                              60.0
                                                     0.9980
                                                              3.16
                                                                         0.58
     4
                       11.0
                                              34.0
                                                     0.9978 3.51
                                                                         0.56
        alcohol quality
                          Ιd
     0
            9.4
                           0
            9.8
     1
                           1
            9.8
                       5
     2
                           2
            9.8
     3
                       6
                           3
            9.4
                       5
    Data Preprocessing
[4]: data.info()
    <class 'pandas.core.frame.DataFrame'>
    RangeIndex: 1143 entries, 0 to 1142
    Data columns (total 13 columns):
         Column
                                Non-Null Count Dtype
```

```
0
          fixed acidity
                                1143 non-null
                                                float64
          volatile acidity
                                1143 non-null
                                                float64
      1
      2
          citric acid
                                1143 non-null
                                                float64
      3
          residual sugar
                                1143 non-null
                                                float64
      4
          chlorides
                                1143 non-null
                                                float64
          free sulfur dioxide
      5
                                1143 non-null
                                                float64
         total sulfur dioxide 1143 non-null
                                               float64
                                                float64
      7
          density
                                1143 non-null
      8
                                1143 non-null float64
          Нq
          sulphates
                                1143 non-null
                                                float64
      9
      10 alcohol
                                1143 non-null
                                                float64
                                1143 non-null
                                                int64
      11
         quality
                                1143 non-null
                                                int64
      12 Id
     dtypes: float64(11), int64(2)
     memory usage: 116.2 KB
 [6]: data.isnull().sum()
 [6]: fixed acidity
                              0
     volatile acidity
                              0
      citric acid
                              0
     residual sugar
                              0
      chlorides
                              0
     free sulfur dioxide
                              0
     total sulfur dioxide
                              0
     density
                              0
                              0
     рΗ
                              0
     sulphates
                              0
     alcohol
      quality
                              0
      Ιd
                              0
      dtype: int64
     Feature Selection
 [7]: from sklearn.model_selection import train_test_split
      from sklearn.preprocessing import StandardScaler
 [8]: # Split the dataset into features (X) and target variable (y)
      X = data.drop(["quality", "Id"], axis=1)
      y = data["quality"]
[10]: # Split the dataset into training and testing sets
      X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.
       →2,random_state=42)
[11]: # Feature Scaling
      scaler = StandardScaler()
```

```
X_train_scaled = scaler.fit_transform(X_train)
      X_test_scaled = scaler.transform(X_test)
     MODEL BUILDING
[12]: from sklearn.ensemble import RandomForestClassifier
      from sklearn.linear model import SGDClassifier
      from sklearn.svm import SVC
[13]: # Initialize classifier models
      rf_model = RandomForestClassifier()
      sgd_model = SGDClassifier()
      svc_model = SVC()
[14]: # Train the random forest model
      rf_model.fit(X_train_scaled, y_train)
[14]: RandomForestClassifier()
[15]: # Train the sqd model
      sgd_model.fit(X_train_scaled, y_train)
[15]: SGDClassifier()
[16]: # Train the svc model
      svc_model.fit(X_train_scaled, y_train)
[16]: SVC()
     MODEL EVALUATION
[17]: from sklearn.metrics import classification_report
[18]: # Evaluate Random Forest model
      rf_predictions = rf_model.predict(X_test_scaled)
      print("\033[1mRandom Forest Classifier:\033[0m")
      print(classification_report(y_test, rf_predictions))
     Random Forest Classifier:
                   precision
                                recall f1-score
                                                   support
                4
                        0.00
                                  0.00
                                            0.00
                                                         6
                5
                        0.69
                                  0.78
                                            0.74
                                                        96
                6
                        0.64
                                  0.63
                                            0.63
                                                        99
                7
                        0.67
                                  0.62
                                            0.64
                                                        26
                        0.00
                                  0.00
                                            0.00
                8
                                                         2
                                                       229
         accuracy
                                            0.67
```

0.40

229

macro avg

0.40

0.40

weighted avg 0.64 0.67 0.65 229

C:\ProgramData\anaconda3\Lib\site-

packages\sklearn\metrics_classification.py:1344: UndefinedMetricWarning: Precision and F-score are 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, msg_start, len(result))

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_warn_prf(average, modifier, msg_start, len(result))

[19]: # Evaluate SGD model

sgd_predictions = sgd_model.predict(X_test_scaled)
print("\033[1mStochastic Gradient Descent Classifier:\033[0m")
print(classification_report(y_test, sgd_predictions))

Stochastic Gradient Descent Classifier:

	precision	recall	f1-score	support
3	0.00	0.00	0.00	0
4	0.00	0.00	0.00	6
5	0.67	0.55	0.61	96
6	0.54	0.67	0.60	99
7	0.28	0.27	0.27	26
8	0.00	0.00	0.00	2
accuracy			0.55	229
macro avg	0.25	0.25	0.25	229
weighted avg	0.55	0.55	0.54	229

C:\ProgramData\anaconda3\Lib\site-

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_warn_prf(average, modifier, msg_start, len(result))

[20]: # Evaluate SVC model

svc_predictions = svc_model.predict(X_test_scaled)
print("\033[1mSupport Vector Classifier:\033[0m")
print(classification_report(y_test, svc_predictions))

Support Vector Classifier:

	precision	recall	f1-score	support
4	0.00	0.00	0.00	6
5	0.70	0.74	0.72	96
6	0.59	0.69	0.64	99
7	0.54	0.27	0.36	26
8	0.00	0.00	0.00	2
accuracy			0.64	229
macro avg	0.37	0.34	0.34	229
weighted avg	0.61	0.64	0.62	229

C:\ProgramData\anaconda3\Lib\site-

packages\sklearn\metrics_classification.py:1344: UndefinedMetricWarning: Precision and F-score are ill-defined and being set to 0.0 in labels with no predicted samples. Use `zero_division` parameter to control this behavior.

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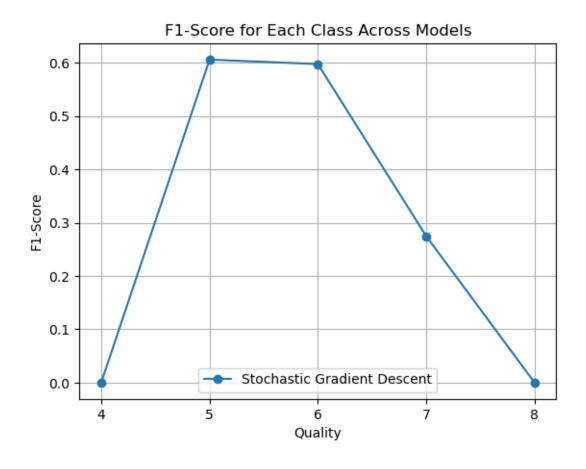
```
_warn_prf(average, modifier, msg_start, len(result))
     C:\ProgramData\anaconda3\Lib\site-
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     predicted samples. Use `zero division` parameter to control this behavior.
       _warn_prf(average, modifier, msg_start, len(result))
[21]: # Define classes and models
      classes = np.unique(y test)
      models = ['Random Forest', 'Stochastic Gradient Descent', 'Support Vector⊔

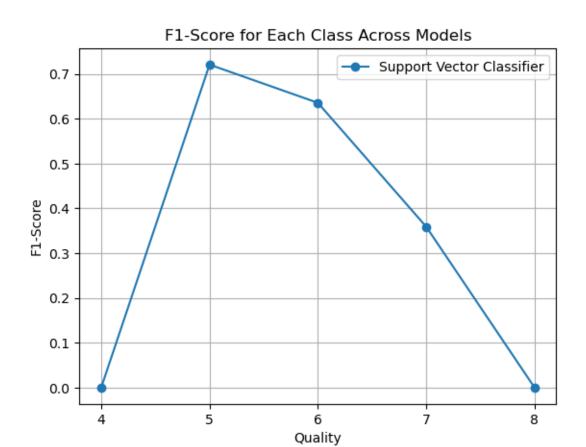
Glassifier']

[22]: # Initialize empty lists to store F1-scores for each class and model
      f1 scores = {model: [] for model in models}
[23]: # Calculate F1-score for each model
      for model_name, predictions in zip(models, [rf_predictions,_
       →sgd_predictions,svc_predictions]):
          report = classification_report(y_test, predictions, output_dict=True)
          for class label in classes:
              f1_scores[model_name].append(report[str(class_label)]['f1-score'])
     C:\ProgramData\anaconda3\Lib\site-
     packages\sklearn\metrics\_classification.py:1344: UndefinedMetricWarning:
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     C:\ProgramData\anaconda3\Lib\site-
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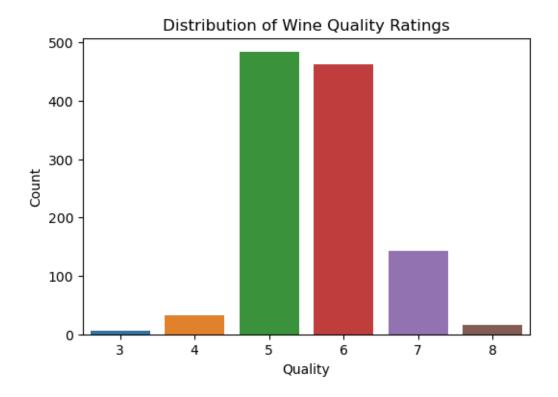
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     predicted samples. Use `zero_division` parameter to control this behavior.
       _warn_prf(average, modifier, msg_start, len(result))
[24]: # Plot F1-scores for each class across models
      plt.figure(figsize=(6, 4))
      for model_name in models:
          plt.plot(classes, f1 scores[model name], marker='o', label=model name)
          plt.title('F1-Score for Each Class Across Models')
          plt.xlabel('Quality')
          plt.ylabel('F1-Score')
          plt.legend()
          plt.xticks(classes)
          plt.grid(True)
          plt.show()
```



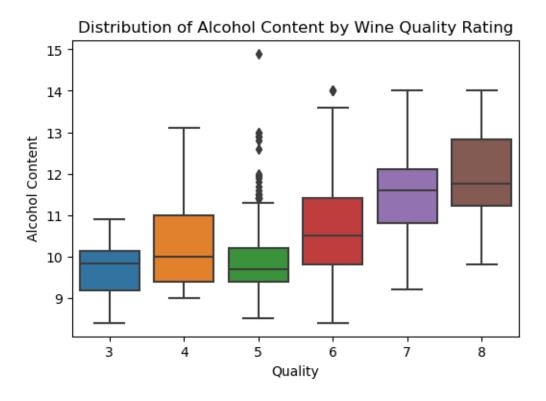




```
[25]: # Visualize the distribution of wine quality ratings
plt.figure(figsize=(6, 4))
sns.countplot(x='quality', data=data)
plt.title('Distribution of Wine Quality Ratings')
plt.xlabel('Quality')
plt.ylabel('Count')
plt.show()
```



```
[26]: # Boxplot to visualize distribution of features by quality rating
plt.figure(figsize=(6, 4))
sns.boxplot(x='quality', y='alcohol', data=data)
plt.title('Distribution of Alcohol Content by Wine Quality Rating')
plt.xlabel('Quality')
plt.ylabel('Alcohol Content')
plt.show()
```

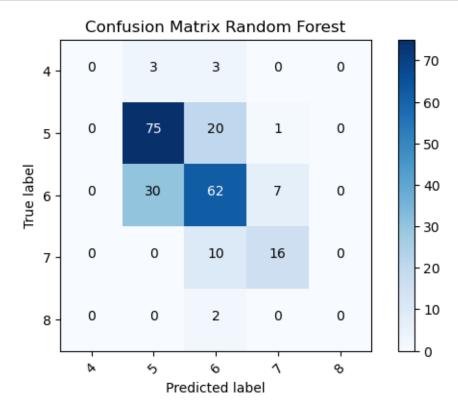


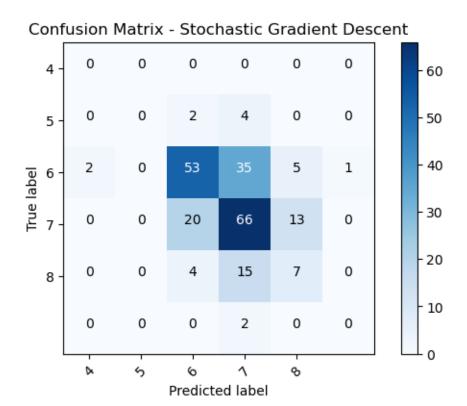
```
[27]: from sklearn.metrics import confusion_matrix
      import itertools
      # Function to plot confusion matrix
      def plot confusion matrix(cm, classes, title='Confusion matrix', cmap=plt.cm.
       →Blues):
          plt.imshow(cm, interpolation='nearest', cmap=cmap)
          plt.title(title)
          plt.colorbar()
          tick_marks = np.arange(len(classes))
          plt.xticks(tick_marks, classes, rotation=45)
          plt.yticks(tick_marks, classes)
          fmt = 'd'
          thresh = cm.max() / 2.
          for i, j in itertools.product(range(cm.shape[0]), range(cm.shape[1])):
              plt.text(j, i, format(cm[i, j], __
       ofmt), horizontalalignment="center", color="white" if cm[i, j] > thresh else⊔

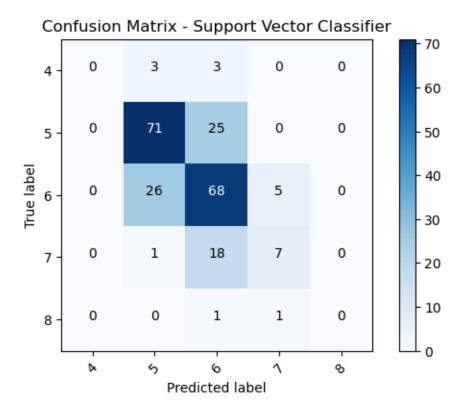
¬"black")

          plt.tight_layout()
          plt.ylabel('True label')
          plt.xlabel('Predicted label')
```

```
[28]: # Calculate confusion matrix for Random Forest model
rf_cm = confusion_matrix(y_test, rf_predictions)
```







```
Gacid', 'residual sugar', 'chlorides', 'free sulfur dioxide', 'total sulfur
 ⇔dioxide', 'density', 'pH', 'sulphates', 'alcohol', 'quality'], hue='quality')
plt.title('Pairplot of Wine Features')
plt.show()
C:\ProgramData\anaconda3\Lib\site-packages\seaborn\_oldcore.py:1119:
FutureWarning: use_inf_as_na option is deprecated and will be removed in a
future version. Convert inf values to NaN before operating instead.
  with pd.option_context('mode.use_inf_as_na', True):
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```

sns.pairplot(data=data, vars=['fixed acidity', 'volatile acidity', 'citric

Pairplot to visualize relationships between features

[35]:

```
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