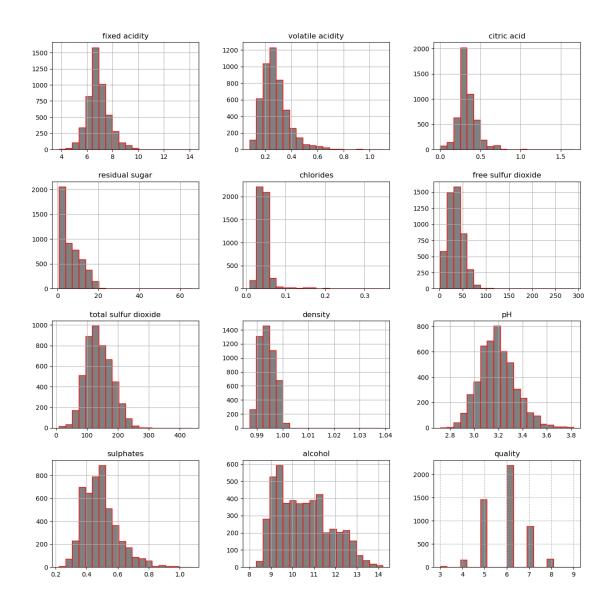
# wine-quality-prediction

### February 3, 2024

```
[1]: import pandas as pd
     import numpy as np
     import seaborn as sns
     import matplotlib.pyplot as plt
[2]: df = pd.read_csv('wine-quality.csv')
     df.head()
[2]:
        fixed acidity volatile acidity
                                          citric acid residual sugar
                                                                         chlorides
                  7.0
                                    0.27
                                                  0.36
                                                                  20.7
                                                                             0.045
     1
                  6.3
                                    0.30
                                                  0.34
                                                                   1.6
                                                                             0.049
                  8.1
                                                                   6.9
     2
                                    0.28
                                                  0.40
                                                                             0.050
                                                  0.32
     3
                  7.2
                                    0.23
                                                                   8.5
                                                                             0.058
     4
                  7.2
                                    0.23
                                                  0.32
                                                                   8.5
                                                                             0.058
        free sulfur dioxide total sulfur dioxide density
                                                                   sulphates
                                                                рΗ
     0
                       45.0
                                              170.0
                                                      1.0010
                                                              3.00
                                                                          0.45
                       14.0
     1
                                              132.0
                                                      0.9940
                                                              3.30
                                                                          0.49
     2
                       30.0
                                              97.0
                                                      0.9951
                                                              3.26
                                                                          0.44
     3
                       47.0
                                              186.0
                                                      0.9956
                                                                          0.40
                                                              3.19
     4
                       47.0
                                              186.0
                                                      0.9956 3.19
                                                                          0.40
        alcohol
                 quality
     0
            8.8
                       6
            9.5
                       6
     1
     2
           10.1
                       6
     3
            9.9
                       6
     4
            9.9
                       6
    df.shape
[3]: (4898, 12)
[4]:
    df.columns
[4]: Index(['fixed acidity', 'volatile acidity', 'citric acid', 'residual sugar',
            'chlorides', 'free sulfur dioxide', 'total sulfur dioxide', 'density',
```

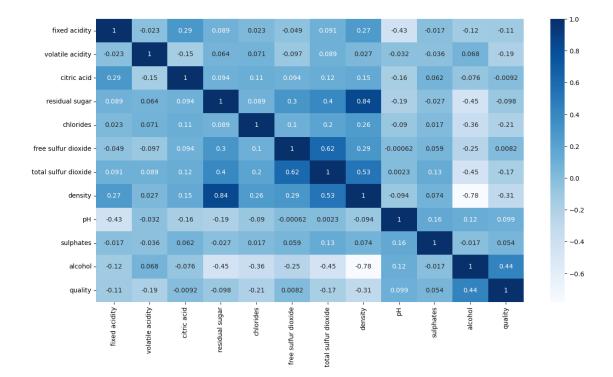
```
dtype='object')
[4]: df.isnull().sum()
[4]: fixed acidity
                             0
    volatile acidity
                             0
                             0
     citric acid
                             0
     residual sugar
     chlorides
                             0
     free sulfur dioxide
                             0
     total sulfur dioxide
                             0
     density
                             0
                             0
    рΗ
                             0
     sulphates
     alcohol
                             0
     quality
                             0
     dtype: int64
[5]: df.info()
    <class 'pandas.core.frame.DataFrame'>
    RangeIndex: 4898 entries, 0 to 4897
    Data columns (total 12 columns):
     #
         Column
                               Non-Null Count Dtype
         -----
                               -----
     0
         fixed acidity
                               4898 non-null
                                               float64
         volatile acidity
     1
                               4898 non-null
                                               float64
     2
         citric acid
                               4898 non-null
                                               float64
     3
        residual sugar
                               4898 non-null
                                               float64
     4
         chlorides
                               4898 non-null
                                               float64
         free sulfur dioxide
                               4898 non-null
                                               float64
         total sulfur dioxide 4898 non-null
                                               float64
     7
         density
                               4898 non-null
                                               float64
     8
                               4898 non-null
                                               float64
         Нq
     9
         sulphates
                               4898 non-null
                                                float64
     10 alcohol
                               4898 non-null
                                                float64
     11 quality
                               4898 non-null
                                                int64
    dtypes: float64(11), int64(1)
    memory usage: 459.3 KB
[6]: # plt.figure(figsize=(15,15))
     df.hist(bins=20,figsize=(15,15), color='grey', edgecolor='red')
     plt.grid(linestyle='--')
     plt.show()
```

'pH', 'sulphates', 'alcohol', 'quality'],



```
[7]: plt.figure(figsize=(15,8))
sns.heatmap(df.corr(), annot=True, cmap='Blues')
```

[7]: <Axes: >



```
[8]: x = df.drop('quality', axis=1)
     y = df['quality']
[9]: x.head()
[9]:
        fixed acidity volatile acidity citric acid residual sugar
                                                                         chlorides \
     0
                  7.0
                                    0.27
                                                  0.36
                                                                   20.7
                                                                             0.045
                  6.3
                                                  0.34
                                    0.30
                                                                    1.6
                                                                             0.049
     1
                  8.1
                                                  0.40
                                                                    6.9
     2
                                    0.28
                                                                             0.050
     3
                  7.2
                                    0.23
                                                  0.32
                                                                    8.5
                                                                             0.058
     4
                  7.2
                                    0.23
                                                  0.32
                                                                    8.5
                                                                             0.058
        free sulfur dioxide total sulfur dioxide density
                                                                pH sulphates
     0
                       45.0
                                              170.0
                                                      1.0010 3.00
                                                                          0.45
                        14.0
     1
                                              132.0
                                                      0.9940 3.30
                                                                          0.49
     2
                        30.0
                                                                          0.44
                                               97.0
                                                      0.9951 3.26
     3
                       47.0
                                              186.0
                                                      0.9956 3.19
                                                                          0.40
     4
                       47.0
                                              186.0
                                                      0.9956 3.19
                                                                          0.40
        alcohol
            8.8
     0
     1
            9.5
     2
           10.1
     3
            9.9
```

```
4
            9.9
[10]: y.head()
[10]: 0
           6
      1
      2
           6
      3
           6
      Name: quality, dtype: int64
     0.1 splitting the data
[11]: from sklearn.model_selection import train_test_split
      x_train, x_test, y_train, y_test = train_test_split(x,y,test_size=0.2,_
       →random_state=101)
      print('x_train: ', x_train.shape)
      print('x_test: ', x_test.shape)
      print('y_train: ', y_train.shape)
      print('y_test: ', y_test.shape)
     x_train: (3918, 11)
     x test: (980, 11)
     y_train: (3918,)
     y_test: (980,)
     0.2 preprocessing the data
[12]: from sklearn.preprocessing import StandardScaler
      scaler = StandardScaler()
      x_train_scaled = scaler.fit_transform(x_train)
      x_test_scaled = scaler.fit_transform(x_test)
[13]: x_train_scaled,x_test_scaled
[13]: (array([[-1.59549189, 1.43374398, -2.01364076, ..., 0.94099026,
                0.18077608, 0.70930052],
              [-0.88939471, -1.17914458, 0.05692741, ..., 0.8748854,
                0.79651933, -0.30440681],
              [-0.18329752, -0.47567458, -0.02589531, ..., 0.8748854,
               -0.78682047, 0.222721 ],
              [0.05206821, -0.87765744, 0.55386377, ..., -0.71163105,
                1.67615255, 0.79039711],
              [-1.00707757, -1.48063172, -0.44000895, ..., 2.65971642,
              -0.5229305 , 0.06052783],
              [-0.18329752, 0.22779541, 1.29926831, ..., -0.44721164,
```

```
-0.34700386, -0.50714827]]),
      array([[-1.16034599, 0.17156545, -0.39719192, ..., -0.0048365,
               -0.96229349, 0.68094974],
              [-0.18415621, 1.68586113, -1.29898824, ..., -0.7391665,
              -0.4441829 , 1.74737515],
              [0.91405729, -1.24808675, 0.5046044, ..., -0.33862286,
                2.31907355, -0.38547568],
              [1.15810474, -0.01772151, -0.72511786, ..., -1.40673924,
              -1.04864525, 2.07550605],
              [-1.28236971, 0.31353067, -0.56115489, ..., 1.19679442,
               -0.78958996, 1.17314608],
              [-0.06213249, -0.11236499, -0.97106231, ..., -0.80592378,
                0.33298298, -1.04173747]]))
[14]: from sklearn.linear_model import LogisticRegression
      from sklearn.tree import DecisionTreeClassifier
      from sklearn.ensemble import RandomForestClassifier, GradientBoostingClassifier
      from sklearn.svm import SVC
      from sklearn.neighbors import KNeighborsClassifier
      from sklearn.metrics import accuracy_score, classification_report,_
       [15]: models = {
          ('Logistic Regression', Logistic Regression()),
          ('Decision Tree', DecisionTreeClassifier()),
          ('Random Forest', RandomForestClassifier()),
          ('Gradient Boosting', GradientBoostingClassifier()),
          ('SVM',SVC()),
          ('KNeighborsClassifier', KNeighborsClassifier())
      }
[16]: models
[16]: {('Decision Tree', DecisionTreeClassifier()),
       ('Gradient Boosting', GradientBoostingClassifier()),
       ('KNeighborsClassifier', KNeighborsClassifier()),
       ('Logistic Regression', LogisticRegression()),
       ('Random Forest', RandomForestClassifier()),
       ('SVM', SVC())}
[17]: result = pd.DataFrame(columns=['Model', 'Accuracy score'])
[18]: result
```

```
[18]: Empty DataFrame
      Columns: [Model, Accuracy_score]
      Index: []
[19]: for model_name , model in models:
          try: #exception
              model.fit(x train scaled, y train)
              prediction = model.predict(x_test_scaled)
              accuracy = accuracy_score(y_test, prediction)
              result = result.append({'Model':model_name,
                                      'Accuracy_score':accuracy,
                                      },ignore_index=True)
              print(f'\nModel: {model_name}')
              print('Classification reports\n',classification_report(y_test,__
       →prediction))
              print('Confusion Matrix\n',confusion_matrix(y_test, prediction))
          except Exception as e:
              print(f'Error occurred while processing {model_name}: {str(e)}')
      print(result)
     C:\Users\Admin\anaconda3\Lib\site-
     packages\sklearn\linear_model\_logistic.py:460: ConvergenceWarning: lbfgs failed
     to converge (status=1):
     STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.
     Increase the number of iterations (max_iter) or scale the data as shown in:
         https://scikit-learn.org/stable/modules/preprocessing.html
     Please also refer to the documentation for alternative solver options:
         https://scikit-learn.org/stable/modules/linear_model.html#logistic-
     regression
       n_iter_i = _check_optimize_result(
     C:\Users\Admin\AppData\Local\Temp\ipykernel_10844\164645047.py:8: FutureWarning:
     The frame.append method is deprecated and will be removed from pandas in a
     future version. Use pandas.concat instead.
       result = result.append({'Model':model_name,
     C:\Users\Admin\anaconda3\Lib\site-
     packages\sklearn\metrics\_classification.py:1471: 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))
     C:\Users\Admin\anaconda3\Lib\site-
     packages\sklearn\metrics\_classification.py:1471: 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))

C:\Users\Admin\anaconda3\Lib\site-

packages\sklearn\metrics\\_classification.py:1471: 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))

C:\Users\Admin\AppData\Local\Temp\ipykernel\_10844\164645047.py:8: FutureWarning: The frame.append method is deprecated and will be removed from pandas in a future version. Use pandas.concat instead.

result = result.append({'Model':model\_name,

Model: Logistic Regression Classification reports

	precision	recall	f1-score	support
3	0.00	0.00	0.00	5
4	0.75	0.07	0.13	41
5	0.59	0.54	0.56	306
6	0.53	0.74	0.62	433
7	0.42	0.27	0.32	158
8	0.00	0.00	0.00	37
accuracy			0.54	980
macro avg	0.38	0.27	0.27	980
weighted avg	0.52	0.54	0.51	980

#### Confusion Matrix

LL	0	U	) ;	31	1	0]
[	0	3	25	13	0	0]
[	0	1	164	137	4	0]
[	0	0	71	319	43	0]
[	0	0	12	104	42	0]
[	0	0	2	24	11	0]]

## ${\tt Model: KNeighborsClassifier}$

Classification reports

	precision	recall	f1-score	support
3	0.00	0.00	0.00	5
4	0.26	0.12	0.17	41
5	0.53	0.59	0.56	306
6	0.59	0.61	0.60	433
7	0.48	0.44	0.46	158
8	0.24	0.11	0.15	37
accuracy			0.54	980

macro	avg	0.35	0.31	0.32	980
weighted	avg	0.52	0.54	0.53	980

#### Confusion Matrix

[[	0	(	) 3	3 1	1	0]
[	1	5	25	8	2	0]
[	1	10	182	96	14	3]
[	0	3	115	266	42	7]
[	0	0	16	69	70	3]
[	1	1	4	9	18	4]]

C:\Users\Admin\AppData\Local\Temp\ipykernel\_10844\164645047.py:8: FutureWarning: The frame.append method is deprecated and will be removed from pandas in a future version. Use pandas.concat instead.

result = result.append({'Model':model\_name,

C:\Users\Admin\anaconda3\Lib\site-

packages\sklearn\metrics\\_classification.py:1471: 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))

C:\Users\Admin\anaconda3\Lib\site-

packages\sklearn\metrics\\_classification.py:1471: 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))

C:\Users\Admin\anaconda3\Lib\site-

packages\sklearn\metrics\\_classification.py:1471: 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))

Model: Random Forest Classification reports

	precision	recall	f1-score	support
3	0.00	0.00	0.00	5
4	0.75	0.15	0.24	41
5	0.70	0.69	0.69	306
6	0.66	0.78	0.71	433
7 8	0.65 0.81	0.60 0.35	0.62 0.49	158 37
accuracy			0.67	980
macro avg weighted avg	0.59 0.68	0.43 0.67	0.46 0.66	980 980

Confusion Matrix

 $[[0 \ 0 \ 3 \ 2 \ 0 \ 0]$ 

```
2 210 89
                        01
                    5
 Γ
       0 59 336
                   36
                        21
 ΓΟ
           5 57
                        1]
       0
                   95
 Γ
   0
                  10 13]]
        0
            0
               14
C:\Users\Admin\AppData\Local\Temp\ipykernel_10844\164645047.py:8: FutureWarning:
The frame.append method is deprecated and will be removed from pandas in a
future version. Use pandas.concat instead.
  result = result.append({'Model':model name,
C:\Users\Admin\anaconda3\Lib\site-
packages\sklearn\metrics\_classification.py:1471: 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))
C:\Users\Admin\anaconda3\Lib\site-
packages\sklearn\metrics\_classification.py:1471: 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))
C:\Users\Admin\anaconda3\Lib\site-
packages\sklearn\metrics\_classification.py:1471: 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))
C:\Users\Admin\AppData\Local\Temp\ipykernel_10844\164645047.py:8: FutureWarning:
The frame.append method is deprecated and will be removed from pandas in a
future version. Use pandas.concat instead.
  result = result.append({'Model':model_name,
C:\Users\Admin\anaconda3\Lib\site-
packages\sklearn\metrics\_classification.py:1471: 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))
C:\Users\Admin\anaconda3\Lib\site-
packages\sklearn\metrics\_classification.py:1471: UndefinedMetricWarning: Recall
and F-score are 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, msg_start, len(result))
C:\Users\Admin\anaconda3\Lib\site-
packages\sklearn\metrics\_classification.py:1471: 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))
C:\Users\Admin\anaconda3\Lib\site-
packages\sklearn\metrics\_classification.py:1471: UndefinedMetricWarning: Recall
and F-score are ill-defined and being set to 0.0 in labels with no true samples.
Use `zero_division` parameter to control this behavior.
```

6 25 10

01

\_warn\_prf(average, modifier, msg\_start, len(result))

C:\Users\Admin\anaconda3\Lib\site-

packages\sklearn\metrics\\_classification.py:1471: 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))

C:\Users\Admin\anaconda3\Lib\site-

packages\sklearn\metrics\\_classification.py:1471: UndefinedMetricWarning: Recall and F-score are 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, msg\_start, len(result))

Model: SVM Classification reports

		precision	recall	f1-score	support
	3	0.00	0.00	0.00	5
	4	1.00	0.07	0.14	41
	5	0.63	0.57	0.60	306
	6	0.55	0.78	0.65	433
	7	0.49	0.26	0.34	158
	8	0.00	0.00	0.00	37
accur	cacy			0.57	980
macro	avg	0.45	0.28	0.29	980
weighted	avg	0.56	0.57	0.53	980

### Confusion Matrix

LL	O	(	) ;	3 2	0	0]
[	0	3	25	13	0	0]
[	0	0	175	130	1	0]
[	0	0	67	339	27	0]
[	0	0	6	111	41	0]
[	0	0	0	23	14	0]]

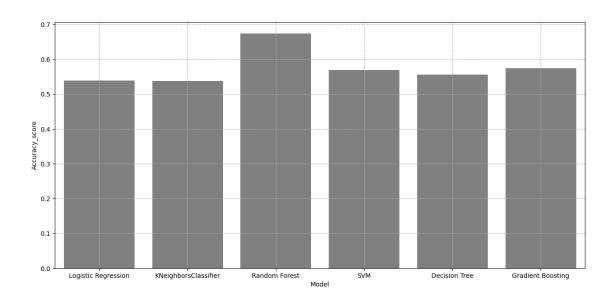
Model: Decision Tree Classification reports

	precision	recall	f1-score	support
3	0.00	0.00	0.00	5
4	0.25	0.22	0.23	41
5	0.62	0.57	0.60	306
6	0.60	0.60	0.60	433
7	0.47	0.56	0.51	158
8	0.33	0.41	0.37	37
9	0.00	0.00	0.00	0

accuracy macro avg weighted avg	0.33 0.56	0.34 0.56	0.56 0.33 0.56	980 980 980
Confusion Matrix [[ 0 0 3 2 [ 0 9 18 11 [ 0 14 175 98 [ 0 13 73 258 [ 0 0 11 52 [ 0 0 1 10 [ 0 0 0 0	0 0 2 1 17 1 68 21 88 7 11 15 0 0	0] 0] 1] 0] 0] 0]		
Model: Gradient Boo	_			
Classification repo	cision	recall	f1-score	support
3 4 5 6 7 8 9	0.17 0.60 0.63 0.57 0.47 0.50	0.20 0.15 0.58 0.73 0.36 0.14 0.00	0.18 0.24 0.61 0.64 0.41 0.21	5 41 306 433 158 37 0
accuracy macro avg weighted avg	0.42 0.57	0.31 0.57	0.57 0.33 0.56	980 980 980
Confusion Matrix [[ 1 0 3 1	42 2 57 2 15 5 0 0	0] 0] 1] 0] 1] 0] 0]] curacy_s	core	
O Logistic Regres  KNeighborsClass  Random Fo  Decision  Gradient Boos	ssion ifier orest SVM Tree	0.53 0.53 0.67 0.56 0.55	8776 7755 3469 9388 6122	

C:\Users\Admin\AppData\Local\Temp\ipykernel\_10844\164645047.py:8: FutureWarning: The frame.append method is deprecated and will be removed from pandas in a future version. Use pandas.concat instead.

```
result = result.append({'Model':model_name,
     C:\Users\Admin\anaconda3\Lib\site-
     packages\sklearn\metrics\_classification.py:1471: UndefinedMetricWarning: Recall
     and F-score are 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, msg_start, len(result))
     C:\Users\Admin\anaconda3\Lib\site-
     packages\sklearn\metrics\_classification.py:1471: UndefinedMetricWarning: Recall
     and F-score are 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, msg_start, len(result))
     C:\Users\Admin\anaconda3\Lib\site-
     packages\sklearn\metrics\_classification.py:1471: UndefinedMetricWarning: Recall
     and F-score are 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, msg_start, len(result))
[20]: data= pd.DataFrame(result)
[21]: data
[21]:
                        Model
                              Accuracy_score
      0
          Logistic Regression
                                     0.538776
       KNeighborsClassifier
      1
                                     0.537755
                Random Forest
                                     0.673469
      3
                          SVM
                                     0.569388
      4
                Decision Tree
                                     0.556122
      5
            Gradient Boosting
                                     0.574490
[26]: plt.figure(figsize=(15,7))
      sns.barplot(data=data, x='Model', y='Accuracy_score', color='grey')
      plt.grid(linestyle='--')
```



[]: