## EDA with Red Wine Data Quality - Fazal Rehman

```
In [ ]:
         import warnings
         warnings.filterwarnings("ignore")
         import pandas as pd
         import matplotlib.pyplot as plt
         import seaborn as sns
         df=pd.read csv("C:/Users/fazal/OneDrive/Desktop/winequality-red.csv",sep=";")
In [4]:
In [5]:
         df.head()
Out[5]:
                                                              free
                                                                       total
                     volatile citric residual
              fixed
                                                chlorides
                                                            sulfur
                                                                      sulfur
                                                                             density
                                                                                        pH sulphates a
             acidity
                      acidity
                                acid
                                        sugar
                                                           dioxide dioxide
         0
                 7.4
                                0.00
                                                    0.076
                                                                        34.0
                                                                                                  0.56
                         0.70
                                           1.9
                                                              11.0
                                                                              0.9978
                                                                                       3.51
          1
                 7.8
                         0.88
                                0.00
                                           2.6
                                                   0.098
                                                              25.0
                                                                        67.0
                                                                              0.9968
                                                                                       3.20
                                                                                                  0.68
         2
                 7.8
                                0.04
                                           2.3
                                                                                                  0.65
                         0.76
                                                   0.092
                                                              15.0
                                                                        54.0
                                                                              0.9970
                                                                                       3.26
          3
                11.2
                         0.28
                                           1.9
                                                   0.075
                                                                        60.0
                                                                              0.9980
                                                                                                  0.58
                                0.56
                                                              17.0
                                                                                       3.16
          4
                 7.4
                         0.70
                                0.00
                                           1.9
                                                    0.076
                                                              11.0
                                                                        34.0
                                                                              0.9978 3.51
                                                                                                  0.56
         df.describe() #for descriptive statistics
In [6]:
Out[6]:
                        fixed
                                    volatile
                                                               residual
                                                                                        free sulfur
                                                                                                     tota
                                               citric acid
                                                                           chlorides
                                                                                           dioxide
                       acidity
                                    acidity
                                                                 sugar
          count 1599.000000
                               1599.000000
                                             1599.000000
                                                           1599.000000
                                                                         1599.000000
                                                                                       1599.000000
                                                                                                    1599
          mean
                     8.319637
                                  0.527821
                                                0.270976
                                                              2.538806
                                                                            0.087467
                                                                                         15.874922
                                                                                                      46
            std
                     1.741096
                                  0.179060
                                                0.194801
                                                              1.409928
                                                                            0.047065
                                                                                         10.460157
                                                                                                      32
           min
                     4.600000
                                  0.120000
                                                0.000000
                                                              0.900000
                                                                            0.012000
                                                                                          1.000000
                                                                                                       6
           25%
                     7.100000
                                  0.390000
                                                0.090000
                                                              1.900000
                                                                            0.070000
                                                                                          7.000000
                                                                                                      22
           50%
                     7.900000
                                  0.520000
                                                0.260000
                                                              2.200000
                                                                            0.079000
                                                                                         14.000000
                                                                                                      38
           75%
                     9.200000
                                  0.640000
                                                0.420000
                                                              2.600000
                                                                            0.090000
                                                                                         21.000000
                                                                                                      62
                    15.900000
                                   1.580000
                                                1.000000
                                                             15.500000
                                                                            0.611000
                                                                                         72.000000
                                                                                                     289
           max
                                                                                                      In [7]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
       RangeIndex: 1599 entries, 0 to 1598
       Data columns (total 12 columns):
            Column
                                 Non-Null Count Dtype
        --- -----
                                 _____
           fixed acidity
        0
                                 1599 non-null float64
        1
            volatile acidity
                                 1599 non-null float64
            citric acid
                                 1599 non-null float64
        2
        3
            residual sugar
                                 1599 non-null float64
        4
           chlorides
                                 1599 non-null float64
        5
           free sulfur dioxide 1599 non-null float64
           total sulfur dioxide 1599 non-null float64
        6
        7
            density
                                 1599 non-null float64
                                 1599 non-null float64
            рΗ
        9
            sulphates
                                 1599 non-null float64
        10 alcohol
                                 1599 non-null float64
        11 quality
                                 1599 non-null int64
       dtypes: float64(11), int64(1)
       memory usage: 150.0 KB
 In [9]: df.shape
 Out[9]: (1599, 12)
In [11]: df.columns
Out[11]: Index(['fixed acidity', 'volatile acidity', 'citric acid', 'residual sugar',
                'chlorides', 'free sulfur dioxide', 'total sulfur dioxide', 'density',
                'pH', 'sulphates', 'alcohol', 'quality'],
               dtype='object')
In [13]: df["quality"].unique()
Out[13]: array([5, 6, 7, 4, 8, 3], dtype=int64)
In [15]: df["quality"].value_counts()
Out[15]: quality
         5
              681
              638
         6
         7
              199
         4
               53
         8
               18
         3
               10
         Name: count, dtype: int64
In [17]: df.isnull().sum() #missing values
```

```
Out[17]: fixed acidity
         volatile acidity
         citric acid
                                0
         residual sugar
                                0
         chlorides
                                0
         free sulfur dioxide
         total sulfur dioxide
         density
                                0
                                0
         рΗ
         sulphates
                                0
         alcohol
                                0
                                0
         quality
         dtype: int64
In [18]: df.duplicated() #to check duplicacy
Out[18]: 0
                 False
         1
                 False
         2
                 False
         3
                False
                True
                . . .
         1594 False
         1595 False
         1596
                True
         1597
               False
         1598
               False
         Length: 1599, dtype: bool
In [19]: df[df.duplicated()]
```

Out[19]:

		fixed acidity	volatile acidity	citric acid	residual sugar	chlorides	sulfur dioxide	total sulfur dioxide	density	рН	sulphate
	4	7.4	0.700	0.00	1.90	0.076	11.0	34.0	0.99780	3.51	0.50
	11	7.5	0.500	0.36	6.10	0.071	17.0	102.0	0.99780	3.35	0.80
	27	7.9	0.430	0.21	1.60	0.106	10.0	37.0	0.99660	3.17	0.9
	40	7.3	0.450	0.36	5.90	0.074	12.0	87.0	0.99780	3.33	0.83
	65	7.2	0.725	0.05	4.65	0.086	4.0	11.0	0.99620	3.41	0.39
	•••										
	1563	7.2	0.695	0.13	2.00	0.076	12.0	20.0	0.99546	3.29	0.54
	1564	7.2	0.695	0.13	2.00	0.076	12.0	20.0	0.99546	3.29	0.54
	1567	7.2	0.695	0.13	2.00	0.076	12.0	20.0	0.99546	3.29	0.54
	1581	6.2	0.560	0.09	1.70	0.053	24.0	32.0	0.99402	3.54	0.60
	1596	6.3	0.510	0.13	2.30	0.076	29.0	40.0	0.99574	3.42	0.7!

240 rows × 12 columns

In [20]: df.drop\_duplicates(inplace=True) #remove duplicate

In [21]: df.shape

Out[21]: (1359, 12)

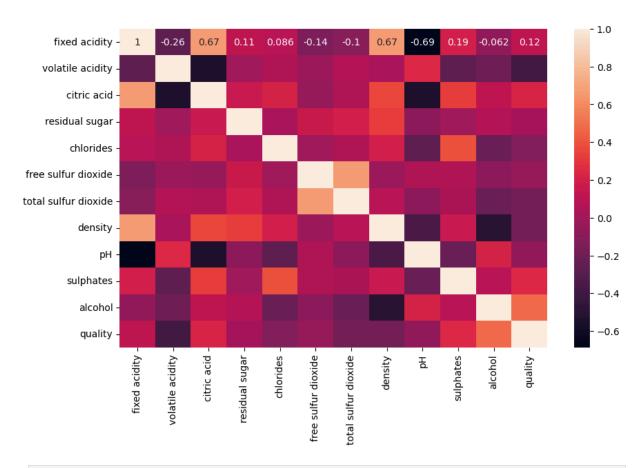
In [22]: df.corr()

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$\cup$	I L	44	

	fixed acidity	volatile acidity	citric acid	residual sugar	chlorides	free sulfur dioxide	total sulfur dioxide	deı
fixed acidity	1.000000	-0.255124	0.667437	0.111025	0.085886	-0.140580	-0.103777	0.67
volatile acidity	-0.255124	1.000000	-0.551248	-0.002449	0.055154	-0.020945	0.071701	0.02
citric acid	0.667437	-0.551248	1.000000	0.143892	0.210195	-0.048004	0.047358	0.35
residual sugar	0.111025	-0.002449	0.143892	1.000000	0.026656	0.160527	0.201038	0.32
chlorides	0.085886	0.055154	0.210195	0.026656	1.000000	0.000749	0.045773	0.19
free sulfur dioxide	-0.140580	-0.020945	-0.048004	0.160527	0.000749	1.000000	0.667246	-0.01
total sulfur dioxide	-0.103777	0.071701	0.047358	0.201038	0.045773	0.667246	1.000000	0.07
density	0.670195	0.023943	0.357962	0.324522	0.193592	-0.018071	0.078141	1.00
рН	-0.686685	0.247111	-0.550310	-0.083143	-0.270893	0.056631	-0.079257	-0.35
sulphates	0.190269	-0.256948	0.326062	-0.011837	0.394557	0.054126	0.035291	0.14
alcohol	-0.061596	-0.197812	0.105108	0.063281	-0.223824	-0.080125	-0.217829	-0.50
quality	0.119024	-0.395214	0.228057	0.013640	-0.130988	-0.050463	-0.177855	-0.18
4								

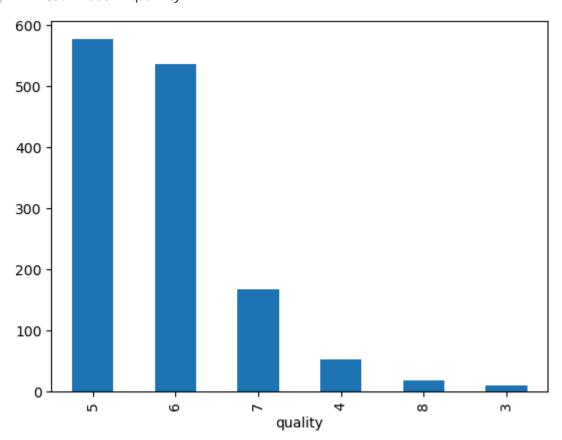
In [24]: plt.figure(figsize=(10,6))
 sns.heatmap(df.corr(),annot=True)

Out[24]: <Axes: >



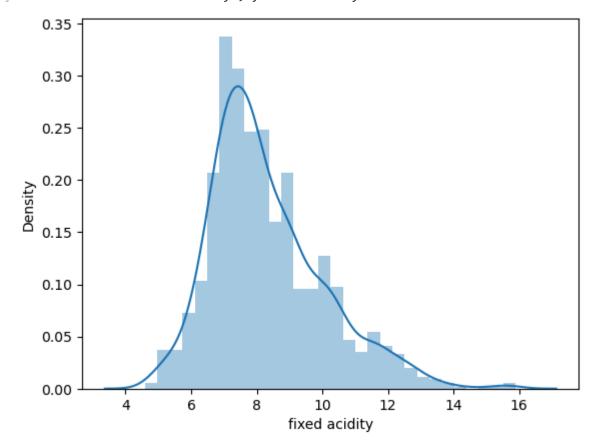
In [25]: df.quality.value\_counts().plot(kind="bar")

Out[25]: <Axes: xlabel='quality'>



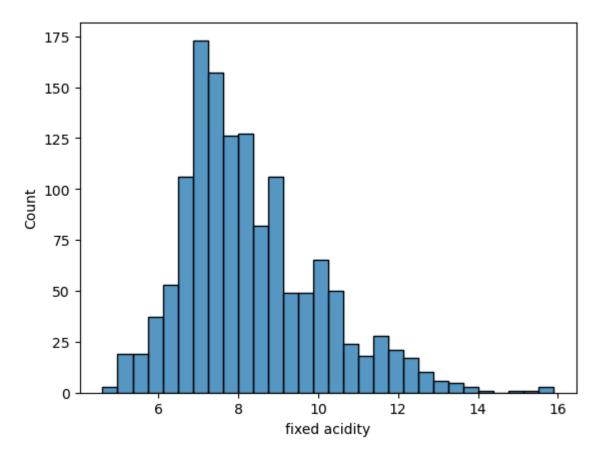
```
In [29]: sns.distplot(df["fixed acidity"])
```

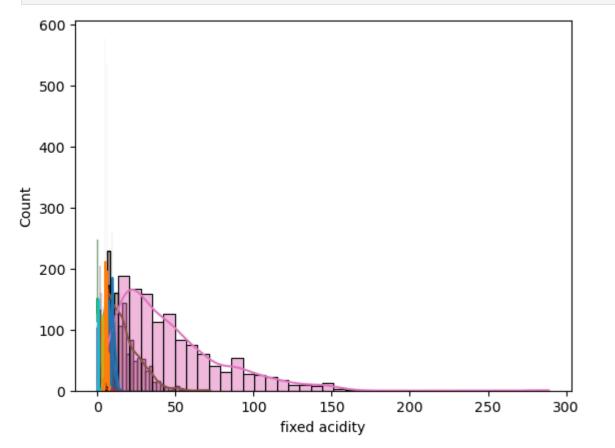
Out[29]: <Axes: xlabel='fixed acidity', ylabel='Density'>



In [30]: sns.histplot(df["fixed acidity"])

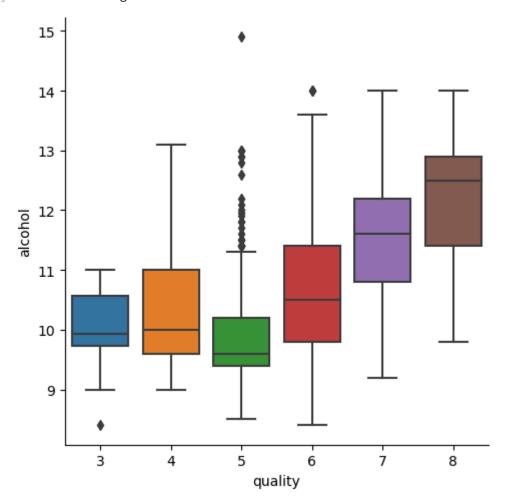
Out[30]: <Axes: xlabel='fixed acidity', ylabel='Count'>





In [34]: sns.catplot(x="quality",y="alcohol",data=df,kind="box") #categorical plot

Out[34]: <seaborn.axisgrid.FacetGrid at 0x1d7254b7550>



In [36]: sns.scatterplot(x="alcohol",y="pH",hue="quality",data=df)

Out[36]: <Axes: xlabel='alcohol', ylabel='pH'>

