Red Wine Quality Data Analytics using Numpy

This project analyzes the **Red and White Wine Quality datasets** using **NumPy**. The goal is to perform exploratory data analysis and demonstrate NumPy operations such as aggregation, filtering, reshaping, sorting, and combining datasets.

Importing modules for numpy

```
In [1]: import numpy as np
        np.genfromtxt() - numpy func used to load data from text files like CSV,TSV, TXT into array
In [2]: # Loading the csv into array
        wines = np.genfromtxt("winequality-red.csv", delimiter=",", skip_header=1)
In [3]:
        # size
        wines.shape
Out[3]: (1599, 12)
In [4]: # dimension
        wines.ndim
Out[4]: 2
In [5]: # rows and column
        row, col = wines.shape
        print(row)
        print(col)
         1599
        12
In [6]:
        # data type
        wines.dtype
Out[6]: dtype('float64')
In [7]: # type of wines
        type(wines)
Out[7]: numpy.ndarray
```

```
In [8]:
         # top 5 rows
         wines[0:5]
 Out[8]: array([[7.400e+00, 7.000e-01, 0.000e+00, 1.900e+00, 7.600e-02, 1.100e+01,
                  3.400e+01, 9.978e-01, 3.510e+00, 5.600e-01, 9.400e+00, 5.000e+00],
                 [7.800e+00, 8.800e-01, 0.000e+00, 2.600e+00, 9.800e-02, 2.500e+01,
                  6.700e+01, 9.968e-01, 3.200e+00, 6.800e-01, 9.800e+00, 5.000e+00],
                 [7.800e+00, 7.600e-01, 4.000e-02, 2.300e+00, 9.200e-02, 1.500e+01,
                  5.400e+01, 9.970e-01, 3.260e+00, 6.500e-01, 9.800e+00, 5.000e+00],
                 [1.120e+01, 2.800e-01, 5.600e-01, 1.900e+00, 7.500e-02, 1.700e+01,
                  6.000e+01, 9.980e-01, 3.160e+00, 5.800e-01, 9.800e+00, 6.000e+00],
                 [7.400e+00, 7.000e-01, 0.000e+00, 1.900e+00, 7.600e-02, 1.100e+01,
                  3.400e+01, 9.978e-01, 3.510e+00, 5.600e-01, 9.400e+00, 5.000e+0
         0]])
         # value at 3rd row 4th col
 In [9]:
         wines[2,3]
Out[9]: 2.3
In [10]: # first 3 items in 4th col
         wines[0:3,3]
Out[10]: array([1.9, 2.6, 2.3])
In [12]: # first column
         wines[:,0]
Out[12]: array([7.4, 7.8, 7.8, ..., 6.3, 5.9, 6.])
         # second row
In [13]:
         wines[1,:]
Out[13]: array([ 7.8
                           0.88 ,
                                    0.
                                             2.6
                                                      0.098, 25.
                                                                      , 67.
                  0.9968,
                           3.2
                                             9.8
                                                       5.
                                                             1)
                                    0.68
In [15]: | # items from rows 1 to 3 and 5th column
         wines[1:4,4]
Out[15]: array([0.098, 0.092, 0.075])
In [16]:
         # entire array
         print(wines)
         [[ 7.4
                   0.7
                           0.
                                      0.56
                                             9.4
                                                     5.
          [ 7.8
                   0.88
                                      0.68
                                             9.8
                                                     5.
                                                          ]
                           0.
          [ 7.8
                   0.76
                           0.04
                                      0.65
                                             9.8
                                                     5.
          [ 6.3
                   0.51
                           0.13
                                      0.75
                                            11.
                                                     6.
          [ 5.9
                                 . . .
                                      0.71
                                            10.2
                                                     5.
                                                          1
                   0.645
                           0.12
          [ 6.
                   0.31
                           0.47 ...
                                      0.66
                                           11.
                                                          ]]
                                                     6.
```

```
In [18]: # changing 1st value into 100
    print(wines[0,0])
    wines[0,0] = 100
    print(wines[0,0])

7.4
    100.0

In [19]: # changing back to 7.4
    wines[0,0] = 7.4
    print(wines[0,0])

7.4
```

1-Dimensional numpy arrays

```
In [20]: # 4th row all columns
        wines[3,:]
Out[20]: array([11.2 , 0.28 , 0.56 , 1.9 , 0.075, 17. , 60. ,
                                                                    0.998,
                3.16, 0.58, 9.8, 6.
                                           ])
In [21]: # 2nd value in 4th row
        wines[3,1]
Out[21]: 0.28
In [22]: # converting wine data into integer values
        wines.astype(int)
Out[22]: array([[ 7, 0, 0, ...,
                                0,
                                    9,
                                        5],
               [7,0,
                                0,
                        0, ...,
                                        5],
               [ 7,
                    0, 0, ...,
                                        5],
                                0,
                                    9,
               [6,0,
                                        6],
                        0, ..., 0, 11,
               [5, 0, 0, ..., 0, 10,
                                        5],
               [ 6, 0, 0, ..., 0, 11,
                                        6]])
```

Vectorization operations

```
In [23]: # increasing wine quality score by 10
print(wines[:,11])

[5. 5. 5. ... 6. 5. 6.]

In [24]: wines[:,11]+10

Out[24]: array([15., 15., 15., ..., 16., 15., 16.])
```

```
In [25]: # multiplying alcohol of all wine data by 3
print(wines[:,10])
       [ 9.4  9.8  9.8 ... 11. 10.2 11. ]
In [26]: wines[:,10]*3
Out[26]: array([28.2, 29.4, 29.4, ..., 33. , 30.6, 33. ])
```

Broadcasting

```
In [27]: # adding every row of wines data with random array of values
In [28]: |num_cols = col
     random_arr = np.random.rand(num_cols)
     result = wines + random_arr
In [29]: print(random_arr)
      [0.48344637 0.99978372 0.84358806 0.46444581 0.21458066 0.7993041
      0.31131623 0.70924103 0.34359435 0.30167519 0.5198527 0.63798177
In [31]: |print(result)
      5.63798177]
      5.63798177
      [ 8.28344637 1.75978372 0.88358806 ... 0.95167519 10.3198527
       5.63798177]
      6.63798177]
      1.01167519 10.7198527
       5.63798177
      6.63798177]]
```

Numpy Aggregation methods

```
In [32]: # sum of all residual sugar values
         np.sum(wines[:,3])
Out[32]: 4059.55
In [33]: # sum of every feature(column) value
         np.sum(wines, axis=0)
Out[33]: array([13303.1
                                                                        139.859
                                843.985
                                             433.29
                                                         4059.55
                25384.
                            , 74302.
                                            1593.79794,
                                                         5294.47
                                                                       1052.38
                16666.35
                               9012.
                                         1)
```

```
In [34]: # sum of every row
         np.sum(wines, axis=1)
Out[34]: array([74.5438, 123.0548, 99.699, ..., 100.48174, 105.21547,
                 92.49249])
In [35]: # maximum residual sugar value in integer
         np.max(wines[:,3]).astype(int)
Out[35]: 15
In [36]: # minimum residual sugar value in integer
         np.min(wines[:,3]).astype(int)
Out[36]: 0
In [37]: # average residual sugar value
         np.mean(wines[:,3])
Out[37]: 2.53880550343965
In [38]: # 25 percentile of residual sugar
         np.percentile(wines[:,3],25)
Out[38]: 1.9
In [39]: |# 25 percentile of residual sugar
         np.percentile(wines[:,3],75)
Out[39]: 2.6
In [40]: # average of each feature value
         np.mean(wines, axis=0)
Out[40]: array([ 8.31963727, 0.52782051, 0.27097561, 2.5388055 ,
                                                                    0.08746654,
                15.87492183, 46.46779237, 0.99674668, 3.3111132,
                                                                    0.65814884,
                10.42298311, 5.63602251])
```

Numpy Array comparisons

```
In [41]: # wines with quality >5
    wines[:,11]>5

Out[41]: array([False, False, False, ..., True, False, True])

In [42]: # wines with quality >7
    wines[:,11]>7

Out[42]: array([False, False, False, ..., False, False, False])
```

```
In [46]:
         # wines with lot of alcohol >10 and wine quality >7
         wines[(wines[:,10]>10)&(wines[:,11]>7)]
Out[46]: array([[7.9000e+00, 3.5000e-01, 4.6000e-01, 3.6000e+00, 7.8000e-02,
                 1.5000e+01, 3.7000e+01, 9.9730e-01, 3.3500e+00, 8.6000e-01,
                 1.2800e+01, 8.0000e+00],
                [1.0300e+01, 3.2000e-01, 4.5000e-01, 6.4000e+00, 7.3000e-02,
                 5.0000e+00, 1.3000e+01, 9.9760e-01, 3.2300e+00, 8.2000e-01,
                 1.2600e+01, 8.0000e+00],
                [5.6000e+00, 8.5000e-01, 5.0000e-02, 1.4000e+00, 4.5000e-02,
                 1.2000e+01, 8.8000e+01, 9.9240e-01, 3.5600e+00, 8.2000e-01,
                 1.2900e+01, 8.0000e+00],
                [1.1300e+01, 6.2000e-01, 6.7000e-01, 5.2000e+00, 8.6000e-02,
                 6.0000e+00, 1.9000e+01, 9.9880e-01, 3.2200e+00, 6.9000e-01,
                 1.3400e+01, 8.0000e+00],
                [9.4000e+00, 3.0000e-01, 5.6000e-01, 2.8000e+00, 8.0000e-02,
                 6.0000e+00, 1.7000e+01, 9.9640e-01, 3.1500e+00, 9.2000e-01,
                 1.1700e+01, 8.0000e+00],
                [1.0700e+01, 3.5000e-01, 5.3000e-01, 2.6000e+00, 7.0000e-02,
                 5.0000e+00, 1.6000e+01, 9.9720e-01, 3.1500e+00, 6.5000e-01,
                 1.1000e+01, 8.0000e+00],
                [1.0700e+01, 3.5000e-01, 5.3000e-01, 2.6000e+00, 7.0000e-02,
                 5.0000e+00, 1.6000e+01, 9.9720e-01, 3.1500e+00, 6.5000e-01,
                 1.1000e+01, 8.0000e+00],
                [5.0000e+00, 4.2000e-01, 2.4000e-01, 2.0000e+00, 6.0000e-02,
                 1.9000e+01, 5.0000e+01, 9.9170e-01, 3.7200e+00, 7.4000e-01,
                 1.4000e+01, 8.0000e+00],
                [7.8000e+00, 5.7000e-01, 9.0000e-02, 2.3000e+00, 6.5000e-02,
                 3.4000e+01, 4.5000e+01, 9.9417e-01, 3.4600e+00, 7.4000e-01,
                 1.2700e+01, 8.0000e+00],
                [9.1000e+00, 4.0000e-01, 5.0000e-01, 1.8000e+00, 7.1000e-02,
                 7.0000e+00, 1.6000e+01, 9.9462e-01, 3.2100e+00, 6.9000e-01,
                 1.2500e+01, 8.0000e+00],
                [1.0000e+01, 2.6000e-01, 5.4000e-01, 1.9000e+00, 8.3000e-02,
                 4.2000e+01, 7.4000e+01, 9.9451e-01, 2.9800e+00, 6.3000e-01,
                 1.1800e+01, 8.0000e+00],
                [7.9000e+00, 5.4000e-01, 3.4000e-01, 2.5000e+00, 7.6000e-02,
                 8.0000e+00, 1.7000e+01, 9.9235e-01, 3.2000e+00, 7.2000e-01,
                 1.3100e+01, 8.0000e+00],
                [8.6000e+00, 4.2000e-01, 3.9000e-01, 1.8000e+00, 6.8000e-02,
                 6.0000e+00, 1.2000e+01, 9.9516e-01, 3.3500e+00, 6.9000e-01,
                 1.1700e+01, 8.0000e+00],
                [5.5000e+00, 4.9000e-01, 3.0000e-02, 1.8000e+00, 4.4000e-02,
                 2.8000e+01, 8.7000e+01, 9.9080e-01, 3.5000e+00, 8.2000e-01,
                 1.4000e+01, 8.0000e+00],
                [7.2000e+00, 3.8000e-01, 3.1000e-01, 2.0000e+00, 5.6000e-02,
                 1.5000e+01, 2.9000e+01, 9.9472e-01, 3.2300e+00, 7.6000e-01,
                 1.1300e+01, 8.0000e+00],
                [7.4000e+00, 3.6000e-01, 3.0000e-01, 1.8000e+00, 7.4000e-02,
                 1.7000e+01, 2.4000e+01, 9.9419e-01, 3.2400e+00, 7.0000e-01,
                 1.1400e+01, 8.0000e+00]])
```

```
# only alcohol and quality columns from above
         wines[(wines[:,10]>10)&(wines[:,11]>7)][:,[10,11]]
Out[47]: array([[12.8,
                       8.],
                       8. ],
               [12.6,
                       8.],
               [12.9,
                       8.],
               [13.4,
               [11.7,
                       8.],
               [11.,
                       8.],
               [11.,
                       8.],
               [14.,
                       8. ],
               [12.7,
                       8.
               [12.5,
                      8.],
               [11.8, 8.],
               [13.1, 8.],
               [11.7, 8.],
               [14., 8.],
               [11.3, 8.],
               [11.4, 8.]])
```

Combining Numpy arrays

```
In [48]: # combine red wine and white wine data
In [52]: white_wines = np.genfromtxt("winequality-white.csv", delimiter =";", skip_he
In [53]: white_wines
Out[53]: array([[ 7. , 0.27, 0.36, ...,
                                         0.45, 8.8,
                                                          ],
               [ 6.3 , 0.3 , 0.34, ...,
                                         0.49,
                                               9.5,
                                                          ],
               [8.1, 0.28, 0.4, ...,
                                         0.44, 10.1,
                                                          ],
                              0.19, ...,
               [ 6.5 , 0.24,
                                         0.46, 9.4,
               [ 5.5 , 0.29, 0.3 , ..., 0.38, 12.8 , 7.
               [ 6. , 0.21, 0.38, ..., 0.32, 11.8 , 6.
In [55]: # combine wines and white wines using vstack
        all_wines = np.vstack((wines,white_wines))
        all_wines
Out[55]: array([[ 7.4 , 0.7 , 0. , ...,
                                         0.56, 9.4,
                                                          ],
                                               9.8,
               [ 7.8 , 0.88, 0. , ...,
                                         0.68,
                                                      5.
                                                          ٦,
               [ 7.8 , 0.76, 0.04, ...,
                                               9.8,
                                         0.65,
                                                          1,
               [ 6.5 , 0.24, 0.19, ...,
                                         0.46, 9.4,
               [5.5, 0.29, 0.3, ...,
                                         0.38, 12.8, 7.
               [ 6. , 0.21, 0.38, ...,
                                         0.32, 11.8, 6.
```

Matrix operations and reshape

```
In [57]: # transpose of wines and its size
        wines.T.shape
Out[57]: (12, 1599)
In [58]: # wines data into 1D array
        wines.flatten()
Out[58]: array([ 7.4 , 0.7 , 0. , ..., 0.66, 11. , 6. ])
In [59]: wines.size
Out[59]: 19188
In [60]: # reshaping 2nd row of wines into 2D with 2 rows and 6 columns
         wines[1,:].reshape(2,6)
                                         , 2.6
                                                  , 0.098 , 25.
Out[60]: array([[ 7.8
                       , 0.88 , 0.
                                                                   ],
                [67.
                       , 0.9968, 3.2 , 0.68 , 9.8 , 5.
                                                                   11)
In [62]: # sort alcohol column in ascending order
         sort alcohol = wines[:,10][np.argsort(wines[:,10])]
         sort alcohol
Out[62]: array([ 8.4, 8.4, 8.5, ..., 14., 14., 14.9])
In [64]: |sort_alcohol[:10]
Out[64]: array([8.4, 8.4, 8.5, 8.7, 8.7, 8.8, 8.8, 9. , 9. , 9. ])
In [65]: # sort alcohol column in descending order
         sort_alcohol_des = wines[:,10][np.argsort(wines[:,10])][::-1]
         sort_alcohol_des
Out[65]: array([14.9, 14., 14., ..., 8.5, 8.4, 8.4])
In [ ]:
```