STANDARDIZATION AND NORMALIZATION

June 30, 2022

```
[]: # NORMALIZATION FORMULA = X-Xmin/Xmax-Xmin
     # LOOK THROUGH THE NOTES
[4]: import numpy as np
     import pandas as pd
     import seaborn as sns
[]: | #to understand normalization first we should look at the data in the csv file
[5]: df = pd.read_csv('wine.csv')
     print(df)
         Wine
               Alcohol
                         Malic.acid
                                      Ash
                                             Acl
                                                       Phenols
                                                                Flavanoids \
                                                   Mg
                  14.23
                               1.71 2.43
    0
            1
                                            15.6
                                                  127
                                                          2.80
                                                                       3.06
                               1.78 2.14
    1
            1
                  13.20
                                            11.2
                                                  100
                                                          2.65
                                                                       2.76
    2
                               2.36 2.67
                                            18.6
                                                          2.80
            1
                  13.16
                                                  101
                                                                       3.24
    3
            1
                  14.37
                               1.95 2.50 16.8
                                                  113
                                                          3.85
                                                                       3.49
    4
            1
                  13.24
                               2.59 2.87
                                            21.0
                                                  118
                                                          2.80
                                                                       2.69
    173
            3
                  13.71
                               5.65 2.45
                                            20.5
                                                   95
                                                          1.68
                                                                       0.61
    174
                  13.40
                               3.91 2.48
                                            23.0
                                                          1.80
                                                                       0.75
            3
                                                 102
                               4.28 2.26
                                                          1.59
    175
            3
                  13.27
                                            20.0
                                                  120
                                                                       0.69
            3
                  13.17
                               2.59 2.37
                                                          1.65
    176
                                            20.0
                                                  120
                                                                       0.68
    177
            3
                  14.13
                               4.10 2.74
                                            24.5
                                                   96
                                                          2.05
                                                                       0.76
         Nonflavanoid.phenols
                               Proanth Color.int
                                                      Hue
                                                             OD
                                                                 Proline
    0
                          0.28
                                   2.29
                                               5.64
                                                     1.04
                                                           3.92
                                                                     1065
                          0.26
                                   1.28
                                               4.38
                                                     1.05
                                                           3.40
    1
                                                                     1050
    2
                          0.30
                                   2.81
                                               5.68
                                                     1.03
                                                           3.17
                                                                     1185
    3
                          0.24
                                   2.18
                                               7.80
                                                     0.86
                                                           3.45
                                                                     1480
    4
                          0.39
                                   1.82
                                               4.32 1.04
                                                           2.93
                                                                      735
    . .
                           •••
    173
                          0.52
                                   1.06
                                               7.70 0.64
                                                           1.74
                                                                      740
    174
                          0.43
                                   1.41
                                               7.30 0.70
                                                           1.56
                                                                      750
                          0.43
                                   1.35
                                              10.20 0.59
    175
                                                           1.56
                                                                      835
                          0.53
                                   1.46
                                               9.30
                                                     0.60
                                                           1.62
    176
                                                                      840
    177
                          0.56
                                   1.35
                                               9.20 0.61
                                                           1.60
                                                                      560
```

```
[6]: df.head()
 [6]:
         Wine
              Alcohol Malic.acid
                                     Ash
                                                     Phenols Flavanoids \
                                           Acl
                                                 Mg
            1
                 14.23
                              1.71
                                    2.43
                                          15.6
                                                         2.80
      0
                                                 127
                                                                     3.06
      1
            1
                 13.20
                              1.78 2.14
                                          11.2
                                                 100
                                                         2.65
                                                                     2.76
      2
            1
                 13.16
                              2.36
                                    2.67
                                           18.6
                                                101
                                                         2.80
                                                                     3.24
      3
            1
                 14.37
                              1.95
                                    2.50
                                          16.8
                                                                     3.49
                                                113
                                                         3.85
      4
            1
                 13.24
                              2.59
                                    2.87
                                          21.0
                                                118
                                                         2.80
                                                                     2.69
         Nonflavanoid.phenols Proanth Color.int
                                                     Hue
                                                            OD
                                                                Proline
      0
                         0.28
                                  2.29
                                              5.64 1.04 3.92
                                                                   1065
                         0.26
                                  1.28
      1
                                              4.38 1.05 3.40
                                                                   1050
      2
                         0.30
                                  2.81
                                              5.68 1.03 3.17
                                                                   1185
      3
                         0.24
                                  2.18
                                              7.80
                                                    0.86 3.45
                                                                   1480
      4
                         0.39
                                  1.82
                                              4.32
                                                   1.04 2.93
                                                                    735
[21]: #intially lets just read the first 3 columns and rename it
      df = pd.read_csv('wine.csv',usecols = [0,1,2])
      df.head()
         Wine Alcohol Malic.acid
[21]:
      0
            1
                 14.23
                              1.71
      1
            1
                 13.20
                              1.78
      2
                 13.16
                              2.36
            1
      3
            1
                 14.37
                              1.95
      4
            1
                 13.24
                              2.59
[10]: !pip install scikit-learn
     Requirement already satisfied: scikit-learn in
     c:\users\91936\anaconda3\lib\site-packages (0.24.1)
     Requirement already satisfied: joblib>=0.11 in
     c:\users\91936\anaconda3\lib\site-packages (from scikit-learn) (1.0.1)
     Requirement already satisfied: threadpoolctl>=2.0.0 in
     c:\users\91936\anaconda3\lib\site-packages (from scikit-learn) (2.1.0)
     Requirement already satisfied: scipy>=0.19.1 in
     c:\users\91936\anaconda3\lib\site-packages (from scikit-learn) (1.6.2)
     Requirement already satisfied: numpy>=1.13.3 in
     c:\users\91936\anaconda3\lib\site-packages (from scikit-learn) (1.20.1)
[22]: import sklearn as sl
[23]: #to perform normalization we have to import few things
      from sklearn.preprocessing import MinMaxScaler
```

```
[24]: scaler = MinMaxScaler()
      scaler.fit_transform(df[['Alcohol','Malic.acid']])
      #therefore it has scaled down the feature between the values of 0-1
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, 0.22727273],

[0.75]

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            [0.81578947, 0.66403162]])
[]: #to understand standardization.
                                       STANDARDIZATION(Z-SCORE NORMALIZATION)
[]: #here all the features will be transformed in such a way that it will have mean
      \rightarrow value = 0, std dev = 1
```

[0.39473684, 0.94268775], [0.64736842, 0.56324111],

```
\#formua = x - mean / std dev
[25]: from sklearn.preprocessing import StandardScaler
[26]: scaling = StandardScaler()
      scaling.fit_transform(df[['Alcohol', 'Malic.acid']])
[26]: array([[ 1.51861254, -0.5622498 ],
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[0.55510587,
              1.22409986],
[-0.22311106,
               0.92787102],
[ 0.71569031,
               0.21871714],
[ 0.49334262,
               2.03199669],
[-0.98897534,
               0.62266555],
[-0.28487431,
               0.04816114],
[ 1.43214399,
               0.15588072],
[ 0.87627476,
               2.974543 ],
[ 0.49334262,
              1.41260912],
[ 0.33275817,
               1.74474449],
               0.22769377],
[ 0.20923168,
[ 1.39508604,
               1.58316512]])
```

[]: #after this go through the theory notes to know when to use standardization and \rightarrow normalization