Installing Libraries

%pip install matplotlib Requirement already satisfied: matplotlib in c:\users\aakas\appdata\ local\programs\python\python312\lib\site-packages (3.8.3)Note: you may need to restart the kernel to use updated packages. Requirement already satisfied: contourpy>=1.0.1 in c:\users\aakas\ appdata\local\programs\python\python312\lib\site-packages (from matplotlib) (1.2.0) Requirement already satisfied: cycler>=0.10 in c:\users\aakas\appdata\ local\programs\python\python312\lib\site-packages (from matplotlib) Requirement already satisfied: fonttools>=4.22.0 in c:\users\aakas\ appdata\local\programs\python\python312\lib\site-packages (from matplotlib) (4.49.0) Requirement already satisfied: kiwisolver>=1.3.1 in c:\users\aakas\ appdata\local\programs\python\python312\lib\site-packages (from matplotlib) (1.4.5) Requirement already satisfied: numpy<2,>=1.21 in c:\users\aakas\ appdata\local\programs\python\python312\lib\site-packages (from matplotlib) (1.26.4) Requirement already satisfied: packaging>=20.0 in c:\users\aakas\ appdata\roaming\python\python312\site-packages (from matplotlib) (23.2)Requirement already satisfied: pillow>=8 in c:\users\aakas\appdata\ local\programs\python\python312\lib\site-packages (from matplotlib) Requirement already satisfied: pyparsing>=2.3.1 in c:\users\aakas\ appdata\local\programs\python\python312\lib\site-packages (from matplotlib) (3.1.1) Requirement already satisfied: python-dateutil>=2.7 in c:\users\aakas\

Importing Libraries

matplotlib) (2.8.2)

>matplotlib) (1.16.0)

```
import pandas as pd
import numpy as np

import matplotlib.pyplot as plt

C:\Users\aakas\AppData\Local\Temp\ipykernel_15468\3311980270.py:1:
DeprecationWarning:
Pyarrow will become a required dependency of pandas in the next major release of pandas (pandas 3.0),
```

appdata\local\programs\python\python312\lib\site-packages (from

Requirement already satisfied: six>=1.5 in c:\users\aakas\appdata\roaming\python\python312\site-packages (from python-dateutil>=2.7-

```
(to allow more performant data types, such as the Arrow string type,
and better interoperability with other libraries)
but was not found to be installed on your system.
If this would cause problems for you,
please provide us feedback at
https://github.com/pandas-dev/pandas/issues/54466
  import pandas as pd
virus = pd.read csv("assignment3-part2.csv")
virus.replace("?", np.nan, inplace=True)
virus.isna().sum()
Unnamed: 0
ANTI A
              0
ANTI B
              0
              0
۷1
dtype: int64
Binning
num bin = 5
```

```
label = ["A", "B", "C", "D", "E"]
virus["E-W-Partition"] = pd.cut(virus["V1"], bins=num_bin,
labels=label)
virus["E-F-Partition"] = pd.qcut(virus["V1"], q = num bin,
labels=label)
virus.iloc[:]
                 ANTI A
                                   V1 E-W-Partition E-F-Partition
    Unnamed: 0
                         ANTI B
                   2.77
0
              0
                            98.1
                                   6.0
                                                     Α
                                                                    Α
1
              1
                   9.79
                            16.8
                                   6.7
                                                     Α
                                                                    Α
2
              2
                   9.29
                            13.6
                                   7.5
                                                     Α
                                                                    Α
3
              3
                   3.41
                            1.6
                                   7.6
                                                     Α
                                                                    Α
4
              4
                   9.83
                            55.3
                                   8.4
                                                     Α
                                                                    Α
5
              5
                   5.76
                            94.4
                                   9.3
                                                     Α
                                                                    Α
6
              6
                   2.73
                            67.2
                                   9.4
                                                     Α
                                                                    Α
7
              7
                            41.9
                   0.17
                                   9.6
                                                     Α
                                                                    Α
8
              8
                   7.50
                            22.7
                                  10.6
                                                     Α
                                                                    В
9
              9
                                  11.2
                                                                    В
                   0.21
                            39.6
                                                     Α
10
                                  11.3
                                                                    В
             10
                   0.66
                            61.6
                                                     Α
11
             11
                   2.84
                            78.9
                                 11.5
                                                     Α
                                                                    В
12
             12
                   7.90
                            15.5
                                  11.5
                                                     Α
                                                                    В
                                  11.9
13
             13
                   2.00
                            93.3
                                                     Α
                                                                    В
14
             14
                   5.07
                            87.9
                                  13.0
                                                                    В
                                                     Α
15
             15
                                                                    В
                   2.14
                            85.0 13.8
```

16	16	9.94	16.4	14.1	Α	C
17	17	7.21	20.6	14.9	Α	С
18	18	4.41	4.8	15.2	Α	С
19	19	8.70	48.3	15.8	A	Ċ
20	20	3.05	86.3	16.5	A	Ċ
21	21	9.34	52.6	17.7	Ä	Č
22	22	7.36	13.0	21.4	Ä	Č
23	23	7.79	61.4	55.5	Č	Č
24	24	4.24	53.3	60.0	D	D
25	25	0.63	11.3	61.4	D	D
26	26	1.43	26.2	62.1	D	D
27	27	1.43	32.9	64.9	D	D
28	28	7.34	83.4	64.9	D	D
29		1.74	12.7	69.0		D
	29				D	
30	30	6.12	35.4	70.1	D	D
31	31	7.59	76.7	73.4	D	D
32	32	7.31	58.4	74.9	E	E
33	33	7.17	86.9	76.8	Ē	E
34	34	2.99	20.6	77.0	E	E
35	35	2.84	33.6	77.9	E	E
36	36	0.67	9.2	80.1	E	E
37	37	8.62	78.9	81.2	E	E
38	38	7.44	98.8	83.8	E	E
39	39	7.21	88.0	90.7	Е	E

Dividing Dataset with V1 less than and greater than 40

```
threshold = 40
virus low = virus[virus["V1"] < threshold]</pre>
virus high = virus[virus["V1"] > threshold]
virus_low.describe()
                       ANTI A
                                   ANTI B
                                                   ٧1
       Unnamed: 0
         23.00000
                    23.000000
                                23.000000
                                            23.000000
count
                                            11.952174
         11.00000
                     5.307826
                                48.495652
mean
std
          6.78233
                     3.363449
                                32.804302
                                             3.830840
          0.00000
                     0.170000
                                 1.600000
                                             6.000000
min
25%
          5.50000
                     2.750000
                                16.600000
                                             9.350000
                                48.300000
                                            11.500000
50%
         11.00000
                     5.070000
75%
         16.50000
                     8.300000
                                81.950000
                                            14.500000
         22.00000
                     9.940000
                                98.100000
                                            21.400000
max
virus high.describe()
                                                   ٧1
       Unnamed: 0
                       ANTI A
                                   ANTI B
                    17.000000
                                            17.000000
count
        17.000000
                                17.000000
```

4.857059

2.937230

51.041176

30.416526

71.982353

9.593112

31.000000

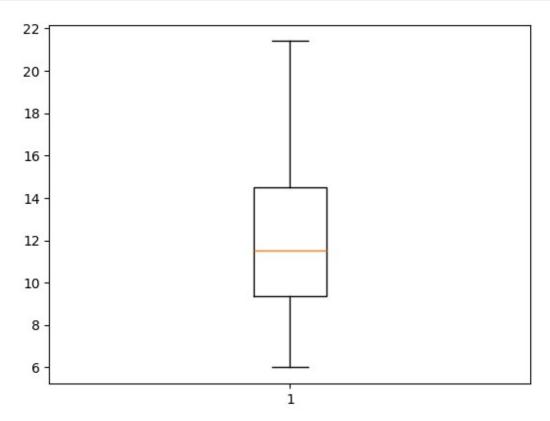
5.049752

mean

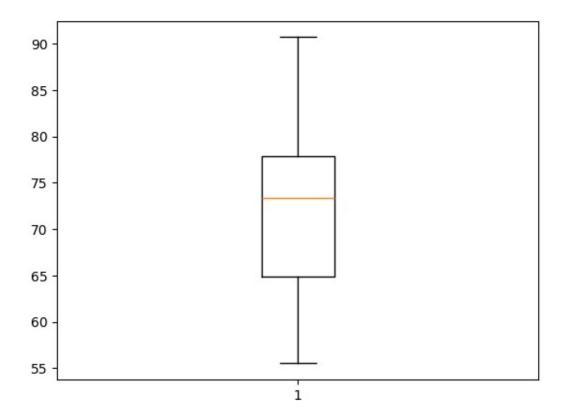
std

```
min
        23.000000
                    0.630000
                                9.200000
                                          55.500000
25%
        27.000000
                    1.740000
                                          64.900000
                               26.200000
50%
        31.000000
                    6.120000
                               53.300000
                                          73.400000
                                          77.900000
75%
        35,000000
                    7.340000
                               78,900000
max
        39,000000
                    8.620000
                               98.800000
                                          90.700000
```

Plotting Boxplot from V1 for both low and high groups



```
plt.boxplot(virus_high["V1"])
```



```
virus_low["ANTI_A"] = (virus_low["ANTI_A"] -
virus_low["ANTI_A"].min())/(virus_low["ANTI_A"].max() -
virus_low["ANTI_A"].min())
virus_low["ANTI_B"] = (virus_low["ANTI_B"] -
virus_low["ANTI_B"].min())/(virus_low["ANTI_B"].max() -
virus_low["ANTI_B"].min())
virus_low

C:\Users\aakas\AppData\Local\Temp\ipykernel_15468\3142245604.py:1:
SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation:
https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#
returning-a-view-versus-a-copy
```

```
virus low["ANTI A"] = (virus low["ANTI A"] -
virus low["ANTI A"].min())/(virus low["ANTI A"].max() -
virus low["ANTI A"].min())
C:\Users\aakas\AppData\Local\Temp\ipykernel 15468\3142245604.py:2:
SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row indexer,col indexer] = value instead
See the caveats in the documentation:
https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#
returning-a-view-versus-a-copy
  virus low["ANTI B"] = (virus low["ANTI B"] -
virus low["ANTI B"].min())/(virus low["ANTI B"].max() -
virus low["ANTI B"].min())
    Unnamed: 0
                            ANTI B
                                     V1 E-W-Partition E-F-Partition
                  ANTI A
0
                0.266121
                          1.000000
                                     6.0
1
                                                                   Α
             1
                0.984647
                          0.157513
                                     6.7
                                                     Α
2
             2
                0.933470
                          0.124352
                                     7.5
                                                     Α
                                                                   Α
3
             3
                0.331627
                          0.000000
                                     7.6
                                                                   Α
                                                     Α
4
             4
                                                                   Α
                0.988741
                          0.556477
                                     8.4
                                                     Α
5
             5
                0.572160 0.961658
                                     9.3
                                                     Α
                                                                   Α
6
             6
                                     9.4
                                                     Α
                                                                   Α
                0.262027
                          0.679793
7
             7
                                                                   Α
                0.000000
                          0.417617
                                     9.6
                                                     Α
8
             8
                0.750256 0.218653
                                    10.6
                                                     Α
                                                                   В
9
             9
                                                                   В
                0.004094 0.393782
                                    11.2
                                                     Α
10
                         0.621762
                                                                   В
            10
                0.050154
                                    11.3
                                                     Α
                                                                   В
11
                                    11.5
                                                     Α
            11
                0.273286
                          0.801036
12
                          0.144041
                                    11.5
                                                                   В
            12
                0.791198
                                                     Α
13
            13
                                    11.9
                                                     Α
                                                                   В
                0.187308 0.950259
14
                                    13.0
                                                                   В
            14
                0.501535
                          0.894301
                                                     Α
15
            15
                0.201638 0.864249
                                    13.8
                                                     Α
                                                                   В
                1.000000 0.153368
                                    14.1
                                                                   C
16
            16
                                                     Α
                                                                   C
17
            17
                0.720573
                          0.196891
                                    14.9
                                                     Α
                                                                   C
18
            18
                0.433982 0.033161
                                    15.2
                                                     Α
                                                                   C
19
            19
                0.873081
                          0.483938
                                    15.8
                                                     Α
                                                                   C
20
            20
                0.294780
                                    16.5
                                                     Α
                         0.877720
                                                                   C
21
                          0.528497
                                    17.7
                                                     Α
            21
                0.938588
22
            22
                0.735926 0.118135
                                    21.4
                                                     Α
                                                                   C
```

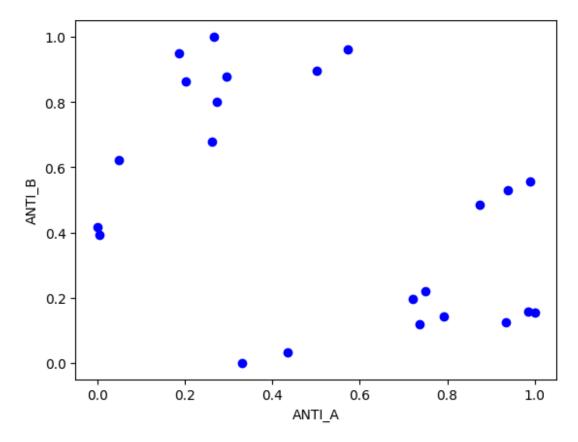
Min-Max Normalization

```
virus_high["ANTI_A"] = (virus_high["ANTI_A"] -
virus_high["ANTI_A"].min())/(virus_high["ANTI_A"].max() -
virus_high["ANTI_A"].min())
virus_high["ANTI_B"] = (virus_high["ANTI_B"] -
virus_high["ANTI_B"].min())/(virus_high["ANTI_B"].max() -
virus_high["ANTI_B"].min())
virus_high
```

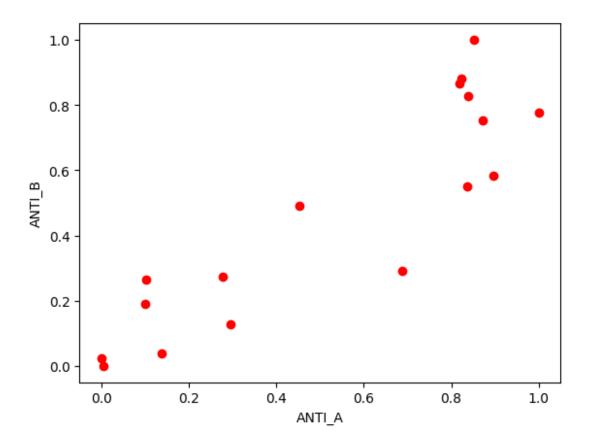
```
C:\Users\aakas\AppData\Local\Temp\ipykernel 15468\34499113.py:1:
SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row indexer,col indexer] = value instead
See the caveats in the documentation:
https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#
returning-a-view-versus-a-copy
  virus_high["ANTI_A"] = (virus_high["ANTI_A"] -
virus high["ANTI_A"].min())/(virus_high["ANTI_A"].max() -
virus high["ANTI A"].min())
C:\Users\aakas\AppData\Local\Temp\ipykernel 15468\34499113.py:2:
SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row indexer,col indexer] = value instead
See the caveats in the documentation:
https://pandas.pydata.org/pandas-docs/stable/user guide/indexing.html#
returning-a-view-versus-a-copy
  virus high["ANTI B"] = (virus high["ANTI B"] -
virus high["ANTI B"].min())/(virus high["ANTI B"].max() -
virus high["ANTI B"].min())
   Unnamed: 0
                  ANTI A
                            ANTI B
                                   V1 E-W-Partition E-F-Partition
23
                0.896120
                         0.582589
            23
                                    55.5
                                                     C
24
                                                     D
            24
                0.451815 0.492187
                                    60.0
                                                                   D
25
            25
                                                                   D
                0.000000 0.023438
                                    61.4
                                                     D
26
            26
                0.100125
                         0.189732
                                    62.1
                                                     D
                                                                   D
27
                                    64.9
                                                     D
                                                                   D
            27
                0.101377
                         0.264509
                                    64.9
28
            28
                0.839800 0.828125
                                                     D
                                                                   D
29
            29
                0.138924 0.039062
                                    69.0
                                                     D
                                                                   D
30
               0.687109 0.292411
                                                     D
                                                                   D
            30
                                    70.1
31
                                                     D
                                                                   D
           31
                0.871089 0.753348
                                    73.4
32
                                    74.9
                                                     Ε
                                                                   Ε
            32
                0.836045
                         0.549107
                                                     Ε
                                                                   Ε
33
            33
                0.818523 0.867188
                                    76.8
                                    77.0
                                                                   Ε
34
                                                     Ε
            34 0.295369 0.127232
                                                     Ε
                                                                   Ε
35
            35
                0.276596 0.272321
                                    77.9
                                                                   Ε
                                                     Ε
36
            36
                0.005006 0.000000
                                    80.1
                                                     Ε
                                                                   Ε
37
            37
                1.000000
                                    81.2
                         0.777902
                                                     Ε
                                                                   Ε
38
            38
                0.852315
                         1.000000
                                    83.8
39
            39
                0.823529
                                                     Е
                                                                   Ε
                         0.879464
                                    90.7
```

Plotting Scatterplot

```
plt.scatter(virus_low["ANTI_A"], virus_low["ANTI_B"], c="blue")
plt.xlabel("ANTI_A")
plt.ylabel("ANTI_B")
plt.show()
```



```
plt.scatter(virus_high["ANTI_A"], virus_high["ANTI_B"], c="red")
plt.xlabel("ANTI_A")
plt.ylabel("ANTI_B")
plt.show()
```



Corelation between ANTI_A and ANTI_B

```
virus_low["ANTI_A"].corr(virus_low["ANTI_B"])
-0.44962747254599306
virus_high["ANTI_A"].corr(virus_high["ANTI_B"])
0.8938242533758364
```

For low values, we must use only one of the medicines i.e ANTI_A or ANTI_B as the corelation is negative but for high values, we can use both ANTI_A and ANTI_B together as the corelation is positive