```
DATA MANIPULATION
         provides features to create random data
 In [3]: a=np.random.randint(0,9,size=[10,10])
 Out[3]: array([[5, 5, 1, 6, 0, 7, 8, 7, 7, 1],
                [3, 8, 7, 7, 7, 8, 5, 3, 2, 3],
                [3, 7, 5, 3, 2, 1, 4, 6, 6, 4],
                [6, 3, 4, 7, 6, 4, 1, 8, 6, 2],
                [7, 1, 8, 2, 7, 5, 6, 0, 3, 7],
                [3, 1, 7, 4, 4, 6, 7, 7, 6, 3],
                [6, 4, 8, 8, 7, 6, 4, 5, 1, 7],
                [5, 1, 5, 0, 8, 5, 5, 1, 7, 0],
                [3, 0, 4, 5, 1, 8, 8, 7, 8, 0],
                [8, 5, 2, 0, 4, 1, 7, 7, 4, 8]])
In [18]: a.shape
Out[18]: (10, 10)
         reshaping data
 In [4]: a.reshape(20,5)
 Out[4]: array([[5, 5, 1, 6, 0],
                [7, 8, 7, 7, 1],
                [3, 8, 7, 7, 7],
                [8, 5, 3, 2, 3],
                [3, 7, 5, 3, 2],
                [1, 4, 6, 6, 4],
                [6, 3, 4, 7, 6],
                [4, 1, 8, 6, 2],
                [7, 1, 8, 2, 7],
                [5, 6, 0, 3, 7],
                [3, 1, 7, 4, 4],
                [6, 7, 7, 6, 3],
                [6, 4, 8, 8, 7],
                [6, 4, 5, 1, 7],
                [5, 1, 5, 0, 8],
                [5, 5, 1, 7, 0],
                [3, 0, 4, 5, 1],
                [8, 8, 7, 8, 0],
                [8, 5, 2, 0, 4],
                [1, 7, 7, 4, 8]])
         index slicing:provides required part's view from the array
In [16]: a[2:6,4:8]
Out[16]: array([[2, 1, 4, 6],
                [6, 4, 1, 8],
                [7, 5, 6, 0],
                [4, 6, 7, 7]])
         vectorized computation without any loops
 In [5]: print('a+a: \n',a+a)
         print('\na*a:\n',a*a)
        a+a:
         [[10 10 2 12 0 14 16 14 14 2]
         [ 6 16 14 14 14 16 10 6 4 6]
         [61410642812128]
         [12 6 8 14 12 8 2 16 12 4]
         [14 2 16 4 14 10 12 0 6 14]
         [ 6 2 14 8 8 12 14 14 12 6]
         [12 8 16 16 14 12 8 10 2 14]
         [10 2 10 0 16 10 10 2 14 0]
         [ 6 0 8 10 2 16 16 14 16 0]
         [16 10 4 0 8 2 14 14 8 16]]
         [[25 25 1 36 0 49 64 49 49 1]
         [ 9 64 49 49 49 64 25 9 4 9]
         [ 9 49 25 9 4 1 16 36 36 16]
         [36 9 16 49 36 16 1 64 36 4]
         [49 1 64 4 49 25 36 0 9 49]
         [ 9 1 49 16 16 36 49 49 36 9]
         [36 16 64 64 49 36 16 25 1 49]
         [25  1  25  0  64  25  25  1  49  0]
         [ 9 0 16 25 1 64 64 49 64 0]
         [64 25 4 0 16 1 49 49 16 64]]
 In [6]: a1=np.arange(1,11)
         a2=np.arange(11,21)
         print('a1:',a1)
         print('a2:',a2)
         print('a1+a2:',a1+a2)
         print('a1*a2:',a1*a2)
         print('a2-a1:',a2-a1)
         print('a2/a1:',a2/a1)
        a1: [ 1 2 3 4 5 6 7 8 9 10]
        a2: [11 12 13 14 15 16 17 18 19 20]
        a1+a2: [12 14 16 18 20 22 24 26 28 30]
        a1*a2: [ 11 24 39 56 75 96 119 144 171 200]
        a2-a1: [10 10 10 10 10 10 10 10 10]
        a2/a1: [11.
                                       4.33333333 3.5
                                                                           2.66666667
                         6.
         2.42857143 2.25
                                 2.11111111 2.
         linear algerbra
 In [7]: x = np.array([[1., 2., 3.], [4., 5., 6.]])
         y = np.array([[6., 23.], [-1, 7], [8, 9]])
         print('x:\n',x,'\ny:\n',y)
        х:
         [[1. 2. 3.]
         [4. 5. 6.]]
        у:
         [[ 6. 23.]
         [-1. 7.]
         [ 8. 9.]]
 In [8]: np.dot(x,y)
 Out[8]: array([[ 28., 64.],
                [ 67., 181.]])
 In [9]: x.dot(y)
 Out[9]: array([[ 28., 64.],
                [ 67., 181.]])
In [10]: y.dot(x)
Out[10]: array([[ 98., 127., 156.],
                  27., 33., 39.],
                [ 44., 61., 78.]])
         DATA AGGREGATION
         data aggregation(row wise,column wise,wholedata)
In [30]: print('row wise mean:',a.mean(axis=1))
         print('column wise mean:',a.mean(axis=0))
         print('row wise median:',np.median(a,axis=1))
         print('column wise median:', np.median(a, axis=0))
         print('row wise sum:',a.sum(axis=1))
         print('column wise sum:',a.sum(axis=0))
         print('whole mean:',a.mean())
         print('whole sum:',a.sum())
         print('whole median:',np.median(a))
         print('row standard deviation:',np.std(a,axis=1))
         print('row standard deviation:', np.std(a, axis=0))
         print('whole standard deviation:', np.std(a))
        row wise mean: [4.7 5.3 4.1 4.7 4.6 4.8 5.6 3.7 4.4 4.6]
        column wise mean: [4.9 3.5 5.1 4.2 4.6 5.1 5.5 5.1 5. 3.5]
        row wise median: [5.5 6. 4. 5. 5.5 5. 6. 5. 4.5 4.5]
        column wise median: [5. 3.5 5. 4.5 5. 5.5 5.5 6.5 6. 3. ]
        row wise sum: [47 53 41 47 46 48 56 37 44 46]
        column wise sum: [49 35 51 42 46 51 55 51 50 35]
        whole mean: 4.65
        whole sum: 465
        whole median: 5.0
        row standard deviation: [2.79463772 2.23830293 1.81383571 2.14709106 2.72763634 1.98997487
         2.05912603 2.79463772 3.13687743 2.76405499]
        row standard deviation: [1.75783958 2.61725047 2.3
                                                                2.74954542 2.69072481 2.38537209
         2.06155281 2.66270539 2.23606798 2.80178515]
        whole standard deviation: 2.535251466817444
         ANALYSING DATA
         co-variance and correlation matrix of a given data
In [13]: np.cov(a)
Out[13]: array([[ 8.67777778, -1.23333333, 0.81111111, 0.67777778, -5.02222222,
                  1.93333333, -3.8
                                     , -0.87777778, 7.46666667, -0.13333333],
                [-1.23333333, 5.56666667, -1.36666667, -1.01111111, 0.13333333,
                 -0.82222222, 2.24444444, -0.01111111, -1.46666667, -4.64444444],
                [ 0.81111111, -1.36666667, 3.65555556, -0.07777778, -2.84444444,
                         , -2.17777778, -1.74444444, -0.48888889, 1.71111111],
                [ 0.67777778, -1.01111111, -0.07777778, 5.12222222, -2.8
                  0.4888889, 0.31111111, 0.23333333, 1.35555556, -1.57777778],
                [-5.02222222, 0.13333333, -2.84444444, -2.8 , 8.26666667,
                  0.46666667, 2.6 , 4.2
                                                 , -2.26666667, 0.93333333],
                                                  , 0.48888889, 0.46666667,
                [ 1.93333333, -0.82222222, -0.2
                  4.4 , -0.53333333, 2.15555556, 5.64444444, -1.08888889],
                           , 2.24444444, -2.17777778, 0.31111111, 2.6
                 -0.53333333, 4.71111111, -1.68888889, -2.93333333, -1.84444444],
                [-0.87777778, -0.01111111, -1.74444444, 0.233333333, 4.2
                  2.15555556, -1.68888889, 8.67777778, 2.57777778, -0.91111111],
                [ 7.46666667, -1.46666667, -0.48888889, 1.35555556, -2.26666667,
                  5.64444444, -2.93333333, 2.57777778, 10.93333333, -2.37777778],
                [-0.13333333, -4.64444444, 1.71111111, -1.57777778, 0.93333333,
                 -1.08888889, -1.84444444, -0.91111111, -2.37777778, 8.48888889]])
In [14]: np.corrcoef(a)
                            , -0.17745106, 0.14401224, 0.10166088, -0.59296144,
Out[14]: array([[ 1.
                  0.3128788 , -0.59431633, -0.10115237, 0.76656009, -0.01553492],
                                     , -0.3029621 , -0.18935295, 0.01965513,
                [-0.17745106, 1.
                 -0.16613658, 0.43827791, -0.00159866, -0.18799986, -0.67563282],
                                                   , -0.0179742 , -0.5174347 ,
                [ 0.14401224, -0.3029621 , 1.
                 -0.04986858, -0.52477749, -0.30972495, -0.07733163, 0.30716809],
                [ 0.10166088, -0.18935295, -0.0179742 , 1.
                                                               , -0.43029234,
                  0.10298041, 0.06333221, 0.03499801, 0.18113894, -0.23927163],
                [-0.59296144, 0.01965513, -0.5174347 , -0.43029234, 1.
                  0.0773776 , 0.41662609, 0.49588368, -0.2384219 , 0.11141563],
                 [ 0.3128788 , -0.16613658, -0.04986858, 0.10298041, 0.0773776 ,
                  1. , -0.11714148, 0.34884188, 0.81380166, -0.17816886],
                [-0.59431633, 0.43827791, -0.52477749, 0.06333221, 0.41662609,
                 -0.11714148, 1. , -0.26414059, -0.4087177 , -0.29166108],
                [-0.10115237, -0.00159866, -0.30972495, 0.03499801, 0.49588368,
                  0.34884188, -0.26414059, 1. , 0.26464575, -0.10615529],
                [ 0.76656009, -0.18799986, -0.07733163, 0.18113894, -0.2384219 ,
                  0.81380166, -0.4087177 , 0.26464575, 1. , -0.24681386],
                [-0.01553492, -0.67563282, 0.30716809, -0.23927163, 0.11141563,
                 -0.17816886, -0.29166108, -0.10615529, -0.24681386, 1.
```

the functions cov(co-varaiance matrix) and corr(corrlation matrix) reduce the need for coding to calulation covariance or correlation

here the data contains 10 rows and 10 columns, consider a data having some 10,000 rows and some 1000 columns in genral format using for loops we need size of the data to iterate through the data and perform any required opertaion making it 3 to 5 lines of code this is where numpy enters the field completing the computation with a single line of code(vectorized computation)

In [2]: import numpy as np