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Session - 19



This session deals with

Linear Algebra

System of Equations

Advantages of Numpy

Exercises on Numpy and Linear algebra

In a multiple regression problem we seek a function that can map input data points to outcome values..

Each data point is a *feature vector* $(x_1, x_2, ..., x_m)$ composed of two or more data values that capture various features of the input.

To represent all of the input data along with the vector of output values we set up a input matrix X and an output vector y:



$$X = \begin{bmatrix} 1 & x_{1,1} & x_{1,2} & \cdots & x_{1,m} \\ 1 & x_{2,1} & x_{2,2} & \cdots & x_{2,m} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & x_{n,1} & x_{n,2} & \cdots & x_{n,m} \end{bmatrix} \qquad \mathbf{y} = \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{bmatrix}$$

$$\mathbf{y} = \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{bmatrix}$$

we can compute a predicted outcome value

$$\hat{\mathbf{y}} = \mathbf{x} \cdot \mathbf{\beta} = \beta_0 + \beta_1 \ x_1 + \beta_2 \ x_2 + \dots + \beta_m \ x_m$$



To compute the β vector as follows:

$$\beta = (X^T X)^{-1} X^T y$$

will use numpy to construct the appropriate matrices and vectors and solve for the \beta vector.

Once we have solved for \(\beta \) we will use it to make predictions for some test data points that we initially left out of our input data set.

constructed the input matrix X and the outcomes vector y in numpy, the following code will compute the β vector:

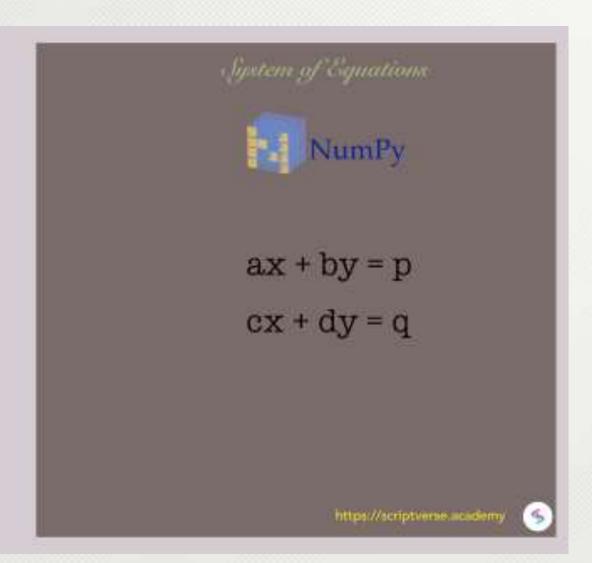
```
Xt = np.transpose(X)
XtX = np.dot(Xt,X)
Xty = np.dot(Xt,y)
beta = np.linalg.solve(XtX,Xty)
```



SONET Solving System equations









SONET Solving System equations DATA SCIENCE



$$3x + 7y = 27$$
$$5x + 2y = 16$$

This can be put in the matrix dot product form as

$$\left[egin{array}{cc} 3 & 7 \ 5 & 2 \end{array}
ight] \cdot \left[egin{array}{c} x \ y \end{array}
ight] = \left[egin{array}{c} 27 \ 16 \end{array}
ight]$$

If A represents the matrix of coefficients, x the column vector of variables and B the column vector of solutions, the above equation can be shortened to

$$Ax = B$$





```
import numpy as np
a = np.array([[3,7], [5,2]])
b = np.array([27,16])
x = np.linalg.solve(a, b)
print(x)
```

Output

[2. 3.]

SONET Solving System equations



As our practice, we will proceed with an example, first writing the matrix model and then using Numpy for a solution.

$$2a+b+c = 4$$

$$a+3b+2c = 5$$

$$a = 6$$

Now, we can formalize the problem with matrices:

$$A = \begin{bmatrix} 2 & 1 & 1 \\ 1 & 3 & 2 \\ 1 & 0 & 0 \end{bmatrix}, x = \begin{bmatrix} a \\ b \\ c \end{bmatrix}, B = \begin{bmatrix} 4 \\ 5 \\ 6 \end{bmatrix}$$

Then, the linear equations can be written this way:

$$Ax = B$$

to solve for the vector x, we must take the inverse of matrix A and the equation is written as follows:

$$x = A^{-1}B$$





```
import numpy as np
# define matrix A using Numpy arrays
A = np.array([[2, 1, 1],
    [1, 3, 2],
    [1, 0, 0]])
#define matrix B
B = np.array([4, 5, 6])
print("Solutions:\n",np.linalg.solve(A, B ))
```

Output

```
Solutions:
[ 6. 15. -23.]
```



Advantages of Numpy



Numpy is the core library for scientific computing in Python.

It provides a high-performance multidimensional array object.

The Python core library provided Lists.

A list is the Python equivalent of an array, but is resizeable and can contain elements of different types.



Advantages of Numpy



Numpy data structures perform better in:

Size - Numpy data structures take up less space

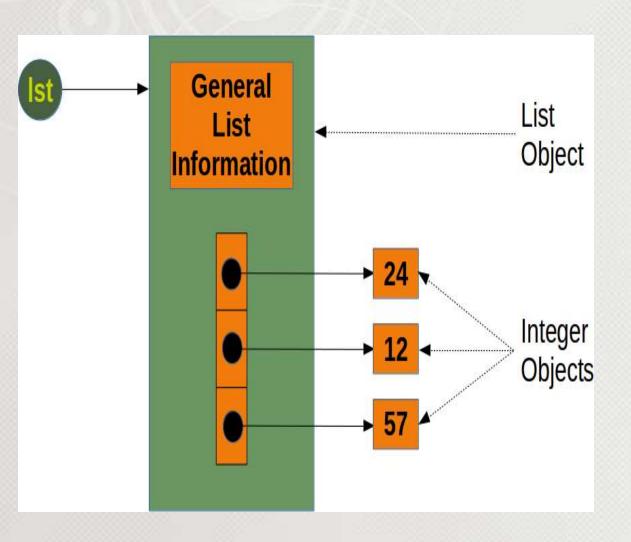
Performance - they have a need for speed and are faster than lists

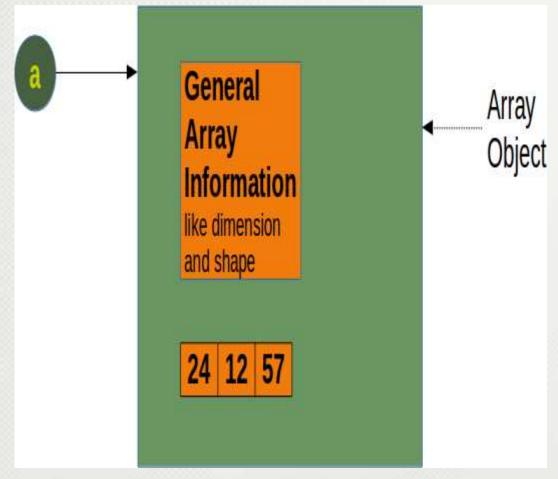
Functionality - SciPy and NumPy have optimized functions such as linear algebra operations built in.



Memory









In [1]: l1=range(1000)

Speed



Numpy arrays are faster than Lists, We can calculate run time speed of list using timeit module

```
In [2]: timeit sum(l1)
28.7 μs ± 1.1 μs per loop (mean ± std. dev. of 7 runs, 10000 loops each)
In [1]: import numpy as np
    ...: y=np.array(range(1000))
In [2]: timeit np.sum(y)
4.45 μs ± 258 ns per loop (mean ± std. dev. of 7 runs, 100000 loops each)
```



Storage Space



Numpy arrays will take less space than Lists

```
In [1]: import sys
In [2]: x=range(1000)
In [3]: sys.getsizeof(1)*len(x)
Out[3]: 28000
In [1]: import numpy as np
In [2]: y=np.array(range(1000))
In [3]: y.itemsize*y.size
Out[3]: 4000
```



Exercise-1



The table below provides the population of daily order volumes for a recent week. Calculate the mean, variance, and standard deviation of this population.

Day	Order Volume
1	16
2	10
3	15
4	12
5	11



Solution



```
import numpy as np
[11,12,15,10,16,10] ordervolume=
arr=np.array(ordervolume)
print(arr)
print("mean = ",np.mean(arr))
print("variance = ",np.var(arr))
print("standard deviation = ",np.std(arr))
```



Exercise-2



A well-known manufacturer of sugarless food products has invested a great deal of time and money in developing the formula for a new kind of sweetener. Although costly to develop, this sweetener is significantly less expensive to produce than the sweeteners the manufacturer had been using. The manufacturer would like to know if the new sweetener is as good as the traditional product. The manufacturer knows that when consumers are asked to indicate their level of satisfaction with the traditional sweeteners, they respond that on average their level of satisfaction is 5.5. The manufacturer conducts market research to determine the level of acceptance of this new product. Consumer taste acceptance data are collected from 25 consumers of sugarless products. The data collected can be seen in the table below.



Exercise-2



Consumers		Satisfaction
1	5.00	
2	6.00	
3	7.00	
4	5.00	
5	6.00	
6	5.00	
7	7.00	
8	4.00	
9	5.00	
10	5.00	
11	6.00	
12	6.00	
13	7.00	
14	5.00	

Cons	sumers
15	5.00
16	7.00
17	5.00
18	6.00
19	6.00
20	7.00
21	7.00
22	7.00
23	6.00
24	5.00
25	6.00

Satisfaction



Solution



```
import numpy as np
import statistics as ss
list sati=[5,6,7,5,6,5,7,4,5,5,6,6,7,5,5,7,5,6,6,7,7,7,7,6,5,6]
print(len(list sati))
list satisifaction=np.array(list sati,dtype=float)
print("Mean = ",np.mean(list satisifaction))
print("Median = ",np.median(list satisifaction))
print("Variance = ",np.var(list satisifaction))
print("Standard deviation = ",np.std(list satisifaction))
```



Exercise-3



The table below identifies the number of consultants of a financial consulting firm for each of its five largest U.S. offices. For this data set, calculate the mode, median, mean, variance, and standard deviation.

City	Number of Consultants
New York	75
Chicago	60
San Francisco	50
Los Angeles	50
Atlanta	70



Solution



```
import numpy as np
import statistics as ss
no of consultants=[75,60,50,50,70]
print("Mean",np.mean(no of consultants))
print("Mode",ss.mode(no of consultants))
print("Median", np.median(no of consultants))
print("Variance",np.var(no of consultants))
print("Standard deviation", np.std(no of consultants))
```

Output

Mean 61.0 Mode 50 Median 60.0 Variance 104.0 Standard deviation 10.198039027185569



Exercise-4



Write a NumPy program to create a vector with values ranging from 15 to 55

- a. print all values except the first and last.
- b. Insert a number 55 at 11th position
- c. Find out the minimum element from the Vector
- d. Find out the maximum element from the Vector



Solution



```
import numpy as np
v = np.arange(15,55)
print("Original vector:")
print(v)
print("All values except the first and last of the said vector:")
print(v[1:-1])
print("after insertion")
print(np.insert(v,11,55))
print(np.min(v))
print(np.max(v))
```







You are aware of

Numpy

Matrix and Linear algebra

We will proceed with

Pandas





