```
In [62]: x=5
          y = 10
          z=x+\lambda
Out[62]: 15
In [63]: x=5
          y = 10
          z=x-\lambda
In [64]: z
Out[64]: -5
In [65]: x=5
          y = 10
          z=x*y
Out[65]: 50
In [66]: x=5
          y = 10
          z=x**y
Out[66]: 9765625
```

```
In [ ]: x=10

In [ ]: y=10

In [72]: x = 5
    y = 8

    print("x == y:", x == y)
    print("x != y:", x != y)
    print("x < y:", x < y)
    print("x > y:", x > y)
    print("x <= y:", x <= y)
    print("x >= y:", x <= y)
    print("x >= y:", x >= y)

x == y: False
x != y: True
```

```
x < y: True
x > y: False
x <= y: True
x >= y: False
```

```
In [81]: x='abc'
In [82]: x
Out[82]: 'abc'
```

```
In [84]: import numpy as np
         array = np.arange(20)
         array
Out[84]: array([ 0, 1, 2,
                            3,
                                4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14, 15, 16,
                17, 18, 19])
In [85]: array = np.arange(27).reshape(3,3,3)
         array
Out[85]: array([[[ 0, 1,
                           2],
                 [ 3, 4,
                          5],
                 [6,7,8]],
                [[ 9, 10, 11],
                [12, 13, 14],
                [15, 16, 17]],
                [[18, 19, 20],
                 [21, 22, 23],
                 [24, 25, 26]]])
In [86]: np.ones(5)
Out[86]: array([1., 1., 1., 1., 1.])
In [87]: np.zeros(5)
Out[87]: array([0., 0., 0., 0., 0.])
In [89]: x = np.arange(9).reshape((3,3))
Out[89]: array([[0, 1, 2],
                [3, 4, 5],
                [6, 7, 8]])
```

```
In [91]: | np.diag(x)
Out[91]: array([0, 4, 8])
```

## practical 5

```
In [95]: a=[[1,5,6,8],[1,2,5,9],[7,5,6,2]]
In [96]: a
Out[96]: [[1, 5, 6, 8], [1, 2, 5, 9], [7, 5, 6, 2]]
In [98]: np.shape(a)
Out[98]: (3, 4)
```

## **Practical 6**

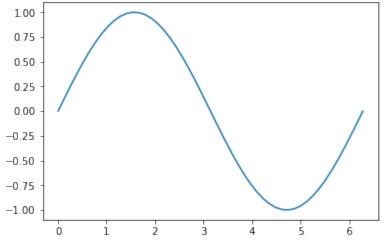
```
In [99]: A = np.array([[2, 4], [5, -6]])
          B = np.array([[9, -3], [3, 6]])
          C = A + B
                     # element wise addition
          print(C)
          [[11 \quad 1]
          [8 0]]
In [101]: A = np.array([[3, 6, 7], [5, -3, 0]])
          B = np.array([[1, 1], [2, 1], [3, -3]])
          C = A.dot(B)
          print(C)
          [[ 36 -12]
          [-1 2]
In [102]: A = np.array([[2, 4], [5, -6]])
          B = np.array([[9, -3], [3, 6]])
          C = A - B # element wise addition
          print(C)
          [[ -7 7]
          [ 2 -12]]
```

### **Practical 7**

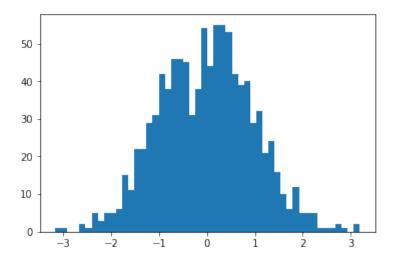
Sum of arr : 279

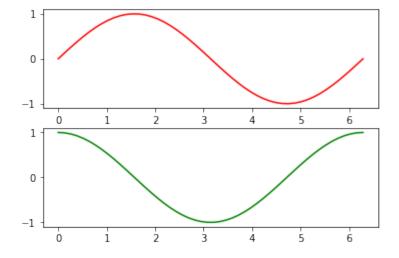
```
In [9]: import matplotlib.pyplot as plt
import numpy as np

In [10]: x = np.linspace(0, 2 * np.pi, 50)
plt.plot(x, np.sin(x))
plt.show()
```



```
In [11]: x = np.random.randn(1000)
plt.hist(x, 50)
plt.show()
```





## **Practical 10**

```
In [13]: for i in [12, 16, 17, 24, 29, 30]:
    if i % 2 == 1:  # if the number is odd
        continue  # don't process it
    print(i)
    print("done")
```

12

16

24

30

```
In [ ]:
        # input two matrices
        mat1 = ([1, 6, 5], [3, 4, 8], [2, 12, 3])
        mat2 = ([3, 4, 6], [5, 6, 7], [6, 56, 7])
        res = np.dot(mat1, mat2)
        print(res)
In [ ]: X = [[12,7,3],
            [4 ,5,6],
            [7,8,9]]
        Y = [[5,8,1],
            [6,7,3],
            [4,5,9]]
        result = [[0,0,0],
                 [0,0,0],
                 [0,0,0]]
        for i in range(len(X)):
           for j in range(len(X[0])):
               result[i][j] = X[i][j] + Y[i][j]
        for r in result:
           print(r)
```

```
and Practical 13
In [16]: import pandas as pd
In [17]: data_train=pd.read_csv("/home/desktop/Downloads/house.csv")
In [18]: data_train.head()
Out[18]:
                                  date
                                          price bedrooms bathrooms sqft_living sqft_lot floors waterl
           0 7129300520 20141013T000000 221900.0
                                                       3
                                                               1.00
                                                                         1180
                                                                                5650
                                                                                        1.0
           1 6414100192 20141209T000000 538000.0
                                                       3
                                                               2.25
                                                                         2570
                                                                                7242
                                                                                        2.0
           2 5631500400 20150225T000000 180000.0
                                                               1.00
                                                                         770
                                                                               10000
                                                       2
                                                                                        1.0
           3 2487200875 20141209T000000 604000.0
                                                       4
                                                               3.00
                                                                         1960
                                                                                5000
                                                                                        1.0
           4 1954400510 20150218T000000 510000.0
                                                               2.00
                                                                         1680
                                                                                8080
                                                                                        1.0
          5 rows x 21 columns
In [20]: data_train.rename(columns = {'price': 'SalePrice'}, inplace = True)
In [21]: data_train.head()
Out[21]:
```

	id	date	SalePrice	bedrooms	bathrooms	sqft_living	sqft_lot	floors	water
0	7129300520	20141013T000000	221900.0	3	1.00	1180	5650	1.0	
1	6414100192	20141209T000000	538000.0	3	2.25	2570	7242	2.0	
2	5631500400	20150225T000000	180000.0	2	1.00	770	10000	1.0	
3	2487200875	20141209T000000	604000.0	4	3.00	1960	5000	1.0	
4	1954400510	20150218T000000	510000.0	3	2.00	1680	8080	1.0	

5 rows x 21 columns

id **SalePrice** bedrooms bathrooms sqft\_living sqft\_lot f **count** 2.161300e+04 2.161300e+04 21613.000000 21613.000000 21613.000000 2.161300e+04 21613.00 mean 4.580302e+09 5.400881e+05 3.370842 2.114757 2079.899736 1.510697e+04 1.49 std 2.876566e+09 3.671272e+05 0.930062 0.770163 918.440897 4.142051e+04 0.53 min 1.000102e+06 7.500000e+04 0.000000 0.000000 290.000000 5.200000e+02 1.00 2.123049e+09 3.219500e+05 25% 3.000000 1427.000000 5.040000e+03 1.750000 1.00 **50%** 3.904930e+09 4.500000e+05 3.000000 2.250000 1910.000000 7.618000e+03 1.50 75% 7.308900e+09 6.450000e+05 4.000000 2.500000 2550.000000 1.068800e+04 2.00 max 9.900000e+09 7.700000e+06 33.000000 8.000000 13540.000000 1.651359e+06 3.50

#### **Practical 14**

```
In [32]: df = pd.read_csv("/home/desktop/Downloads/diabetes.csv")
In [33]: df.head()
```

Out[33]:

	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	ВМІ	DiabetesPedigreeFunction	Ag€
0	6	148	72	35	0	33.6	0.627	50
1	1	85	66	29	0	26.6	0.351	31
2	8	183	64	0	0	23.3	0.672	32
3	1	89	66	23	94	28.1	0.167	21

In [35]: df2 = pd.DataFrame(data=df)
 df2.head()

Out[35]:

	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	ВМІ	DiabetesPedigreeFunction	Ag€
0	6	148	72	35	0	33.6	0.627	50
1	1	85	66	29	0	26.6	0.351	31
2	8	183	64	0	0	23.3	0.672	32
3	1	89	66	23	94	28.1	0.167	21
4	0	137	40	35	168	43.1	2.288	33

In [37]: from sklearn.linear\_model import LogisticRegression

In [39]: logmodel = LogisticRegression()

In [40]: logmodel.fit(X\_train,y\_train)

/home/desktop/anaconda3/lib/python3.7/site-packages/sklearn/linear\_model/
lo gistic.py:433: FutureWarning: Default solver will be changed to 'lbfgs'
in.22. Specify a solver to silence this warning.
FutureWarning)

intercept\_scaling=1, max\_iter=100, multi\_class='warn',
n\_jobs=None, penalty='l2', random\_state=None, solver='warn',
tol=0.0001, verbose=0, warm\_start=False)

In [41]: predictions = logmodel.predict(X\_test)

In [43]: from sklearn.metrics import classification\_report, confusion\_matrix
 print(classification\_report(y\_test,predictions))

	precision	recall	il-score	support
0	0.79	0.91	0.84	150
1	0.76	0.56	0.64	81

micro	avg	0.78	0.78	0.78	231
macro	avg	0.78	0.73	0.74	231
weighted	avg	0.78	0.78	0.77	231

```
In [46]: from sklearn.linear_model import (LinearRegression, Ridge, Lasso, Randomiz edLasso)
In [54]: lasso = Lasso(alpha=0.2, normalize=True)
In [57]: lasso.fit(X_train,y_train)
Out[57]: Lasso(alpha=0.2, copy_X=True, fit_intercept=True, max_iter=1000, normalize=True, positive=False, precompute=False, random_state=None, selection='cyclic', tol=0.0001, warm_start=False)
In [58]: lasso_coef = lasso.coef_print(lasso_coef)
      [0. 0. 0. 0. 0. 0. 0.]
In [59]: ridge = Ridge(normalize=True)
In [60]: ridge.fit(X_train,y_train)
Out[60]: Ridge(alpha=1.0, copy_X=True, fit_intercept=True, max_iter=None, normalize=True, random_state=None, solver='auto', tol=0.001)
In [ ]:
```