Question 1: Perform elementary mathematical operations in Octave/MATLAB or python like addition, multiplication, division and exponentiation.

```
In [1]: val1=2
        val2=3
        #using addition operator
        res1=val1+val2
        print("First value-",val1,"\nSecond value-",val2)
        print("Addition-",res1)
        #using multiplication operator
        res2=val1*val2
        print("Multiplication-", res2)
        #using division operator
        res3=val1/val2
        print("Division-", res3)
        #using exponential operator
        res4=val1 ** val2
        print("Exponent-", res4)
      First value- 2
      Second value- 3
      Addition- 5
      Multiplication- 6
      Exponent- 8
```

Question 2: Perform elementary logical operations in Octave/MATLAB or python (like OR, AND, Checking for Equality, NOT, XOR).

```
In [2]: #Logical and operator
        a=10
        h=5
        c=-4
        if a>0 and b>0:
            print("The numbers are greater than 0")
        if a>0 and b>0 and c>0:
            print("The numbers are greater than 0")
        else:
            print("Atleast one number is not greater than 0")
        #Logical or operator
        a=10
        b=-10
        c=0
        if a>0 or b>0:
            print("Either of the number is greater than 0")
        else:
            print("No number is greater than 0")
        if b>0 or c>0:
            print("Either of the number is greater than 0")
            print("No number is greater than 0")
```

```
#Logical not operator
a=10
if not a:
    print("Boolean value of a is true")
if not(a%3==0 or a%6==0):
    print("10 is not divisible by either 3 and 6")
else:
    print("10 is divisible by either 3 or 6")
#Logical xor operator
a=6
h=3
c=a^b
print("XOR of a=6, b=3 is",c)
#Logical equality operator
a=5
h=3
c=5
if a==b:
    print("Equal")
else:
    print("Not Equal")
if a==c:
    print("Equal")
else:
    print("Not Equal")
```

The numbers are greater than 0
Atleast one number is not greater than 0
Either of the number is greater than 0
No number is greater than 0
10 is not divisible by either 3 and 6
XOR of a=6, b=3 is 5
Not Equal
Equal

Question 3: Create, initialize and display simple variables and simple strings and use simple formatting for variable.

Question 4: Create/Define single dimension / multi-dimension arrays, and arrays with specific values like array of all ones, all zeros, array with random values within a range, or a diagonal matrix.

```
In [6]: arr = py.array([1,2,3,4,5,6])
            print('dimension of array is :',arr.ndim)
          dimension of array is : 1
   Out[6]: array([1, 2, 3, 4, 5, 6])
   In [7]: arr1 = arr.reshape((2,3))
            arr1
   Out[7]: array([[1, 2, 3],
                   [4, 5, 6]])
   In [8]: arr=py.array([1,2,3,4,5,6])
            print('dimension of array is ',arr.ndim)
          dimension of array is 1
   Out[8]: array([1, 2, 3, 4, 5, 6])
   In [9]: arr1=arr.reshape((2,3))
  In [10]: print('dimension of array is ',arr1.ndim)
          dimension of array is 2
  In [11]: arr2=py.zeros((2,3))
            arr2
  Out[11]: array([[0., 0., 0.],
                   [0., 0., 0.]])
  In [12]: arr3=py.ones((3,3))
            arr3
  Out[12]: array([[1., 1., 1.],
                   [1., 1., 1.],
                   [1., 1., 1.]])
arr2 = py.zeros like((arr3)) arr2
  In [13]: py.eye(4)
  Out[13]: array([[1., 0., 0., 0.],
                   [0., 1., 0., 0.],
                   [0., 0., 1., 0.],
                   [0., 0., 0., 1.]])
  In [14]: t = py.random.rand(12)
  In [15]: py.array(t).reshape(6,2)
```

Question 5: Use command to compute the size of a matrix, size/length of a particular row/column, load data from a text file, store matrix data to a text file, finding out variables and their features in the current scope.

```
In [16]: mat=py.array([1,2,3,4,5,6]).reshape(2,3)
         mat
Out[16]: array([[1, 2, 3],
                [4, 5, 6]]
In [17]: mat.size
         mat
Out[17]: array([[1, 2, 3],
                [4, 5, 6]])
In [18]: mat[0].size
Out[18]: 3
In [19]: import pandas as pd
In [20]: df=pd.read_csv('pro-5-text.txt',sep=" ",header=None)
         df
Out[20]:
            0 1 2
         0 a b c
         1 d e f
         2 g h i
In [21]: df.to_csv('text_file.txt')
In [22]: f= open('text_file.txt')
In [23]: print(f.read())
        ,0,1,2
        0,a,b,c
        1,d,e,f
        2,g,h,i
```

Question 6: Perform basic operations on matrices (like addition, subtraction, multiplication) and display specific rows or columns of the matrix.

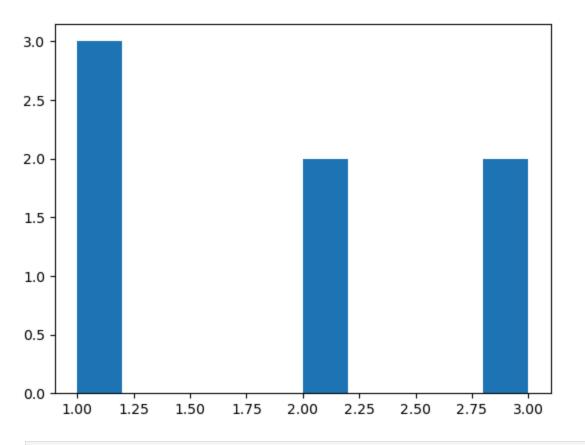
```
In [28]: a=py.array([[11,12],
                      [13,14],
                      [15,16],
                      [17,18]
         b=py.array([[1,2],
                      [3,4],
                      [5,6],
                      [7,8]])
         print('addition of two arrays \n',a+b)
        addition of two arrays
         [[12 14]
         [16 18]
         [20 22]
         [24 26]]
In [25]: print('Subtraction of two arrays \n',a-b)
        Subtraction of two arrays
         [[10 10]
         [10 10]
         [10 10]
         [10 10]]
In [26]: a=py.array([[1,2,5],
                      [3,4,6],
                      [4,5,6]])
         b=py.array([[4,5,1],
                      [4,5,7],
                      [2,3,2]])
         print('multiplication of two arrays \n',py.dot(a,b))
        multiplication of two arrays
         [[22 30 25]
         [40 53 43]
         [48 63 51]]
```

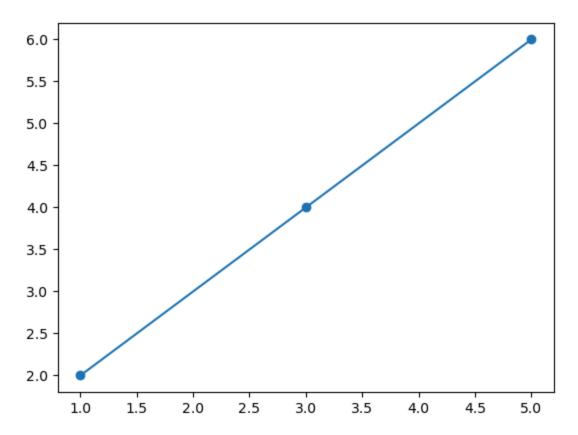
Question 7: Perform other matrix operations like converting matrix data to absolute values, taking the negative of matrix values, adting/removing rows/columns from a matrix, finding the maximum or minimum values in a matrix or in a row/column, and finding the sum of some/all elements in a matrix.

```
In [30]: print('Adding row')
          py.insert(a,2,[[88,88]],axis=0)
        Adding row
Out[30]: array([[11, 12],
                 [13, 14],
                 [88, 88],
                 [15, 16],
                 [17, 18]])
In [31]: print('maximum of matrix',py.max(a))
        maximum of matrix 18
In [32]: print('minimum of matrix',py.min(a))
        minimum of matrix 11
In [33]: print('Sum of matrix',py.sum(a))
        Sum of matrix 116
In [34]: print("Sum of first row of matrix",py.sum(a[0]))
        Sum of first row of matrix 23
In [35]: print("Absolute values of matrix")
          py.absolute(a)
        Absolute values of matrix
Out[35]: array([[11, 12],
                 [13, 14],
                 [15, 16],
                 [17, 18]])
In [36]: print("Negative of matrix",py.negative(a))
        Negative of matrix [[-11 -12]
         [-13 -14]
         [-15 -16]
         [-17 -18]]
          Question 8: Create various type of plots/charts like histograms, plot based on
          sine/cosine function based on data from a matrix. Further label different axes in a plot
          and data in a plot.
```

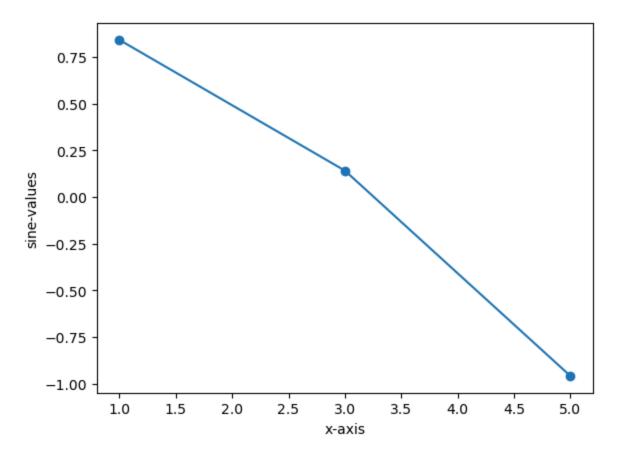
```
In [37]: import matplotlib.pyplot as plt
import math

In [38]: plt.hist([1,1,1,2,2,3,3])
   plt.show()
```





```
In [41]: sin_value=[math.sin(i) for i in arr[:,0]]
In [42]: plt.plot(arr[:,0],sin_value,marker='o')
    plt.xlabel('x-axis')
    plt.ylabel('sine-values')
    plt.show()
```

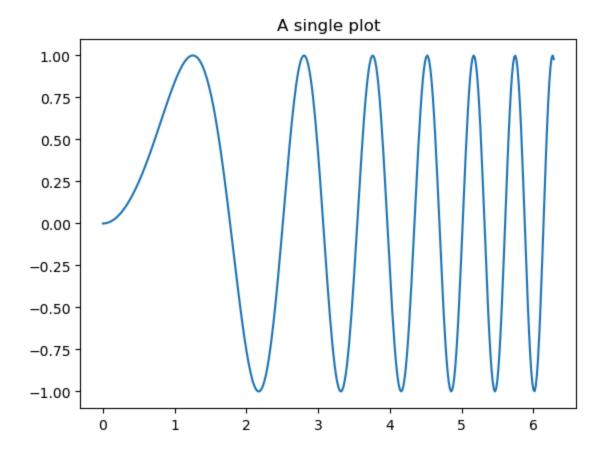


Question 9: Generate different subplots from a given plot and colour plot data.

```
In [44]: import numpy as np
    x=np.linspace(0,2*np.pi,400)
    y=np.sin(x**2)

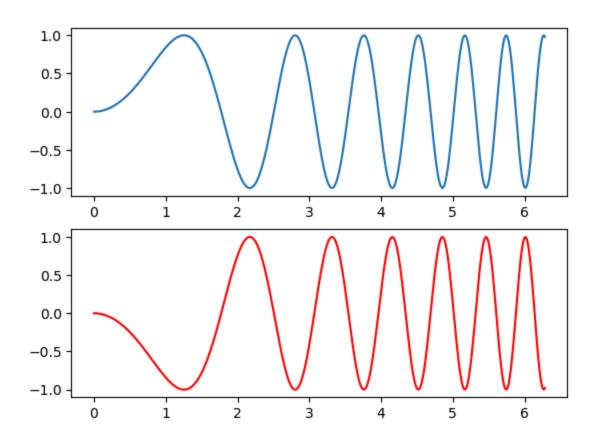
In [45]: fig,ax=plt.subplots()
    ax.plot(x,y)
    ax.set_title("A single plot")
```

Out[45]: Text(0.5, 1.0, 'A single plot')



```
In [46]: fig,axis=plt.subplots(2)
    fig.suptitle("Vertically stacked subplots")
    axis[0].plot(x,y)
    axis[1].plot(x,-y,'red')
    plt.show()
```

Vertically stacked subplots



Question 10: Use conditional statements and different type of loops based on simple example/s.

```
In [47]: fruits =["apple","banana","cherry"]
         for x in fruits:
              print(x)
              if x== "banana":
                  break
        apple
        banana
In [48]: i=1
         while i<6:
              print(i)
              if i ==3:
                  break
              i+=1
        1
        2
        3
```

Question 11. Perform vectorized implementation of simple matrix operation like finding the transpose of a matrix, adding, subtracting or multiplying two matrices.

```
In [91]: #import numpy as np
         # Function to find the transpose of a matrix
         def transpose(matrix):
             return np.transpose(matrix)
         # Function to add two matrices
         def add matrices(matrix1, matrix2):
             return np.add(matrix1, matrix2)
         # Function to subtract two matrices
         def subtract_matrices(matrix1, matrix2):
             return np.subtract(matrix1, matrix2)
         # Function to multiply two matrices
         def multiply_matrices(matrix1, matrix2):
             return np.dot(matrix1, matrix2)
         # Example usage:
         # Define two matrices
         matrix_a = np.array([[1, 2, 3],
                               [4, 5, 6]])
         matrix_b = np.array([[7, 8, 9],
                               [10, 11, 12]])
         # Find the transpose of matrix a
         transposed_matrix_a = transpose(matrix_a)
         print("Transpose of matrix_a:")
         print(transposed_matrix_a)
         # Add matrix a and matrix b
         added matrices = add matrices(matrix a, matrix b)
         print("\nAddition of matrix_a and matrix_b:")
         print(added_matrices)
         # Subtract matrix_b from matrix_a
         subtracted_matrices = subtract_matrices(matrix_a, matrix_b)
         print("\nSubtraction of matrix b from matrix a:")
         print(subtracted_matrices)
         # Multiply matrix_a and matrix_b
         multiplied_matrices = multiply_matrices(matrix_a, matrix_b.transpose()) # Multiply
         print("\nMultiplication of matrix_a and the transpose of matrix_b:")
         print(multiplied matrices)
```

```
Transpose of matrix_a:
       [[1 4]
       [2 5]
       [3 6]]
       Addition of matrix_a and matrix_b:
       [[ 8 10 12]
        [14 16 18]]
       Subtraction of matrix_b from matrix_a:
       [[-6 -6 -6]
       [-6 -6 -6]]
       Multiplication of matrix_a and the transpose of matrix_b:
       [[ 50 68]
        [122 167]]
In [1]: from sklearn.linear_model import LinearRegression
        from sklearn.model selection import train test split
        from sklearn.metrics import mean_squared_error, r2_score
```

Question 12: Implement Linear Regression problem. For example, based on a dataset comprising of existing set of prices and area/size of the houses, predict the estimated price of a given house.

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
df=pd.read_csv('lindata.csv')
df.columns=['X','Y']
df
```

```
      Out[10]:
      X
      Y

      0
      5.5277
      9.1302

      1
      8.5186
      13.6620

      2
      7.0032
      11.8540

      3
      5.8598
      6.8233

      4
      8.3829
      11.8860
```

```
In [12]: def leastSquareRegression(indep_val, dep_val, predict_indep_val):
    mean_dep_val = dep_val.mean()
    mean_indep_val = indep_val.mean()
    diff_dep_val = dep_val - mean_dep_val
    diff_indep_val = indep_val - mean_indep_val

numerator = (diff_indep_val * diff_dep_val).sum()
    denominator = (diff_indep_val ** 2).sum()

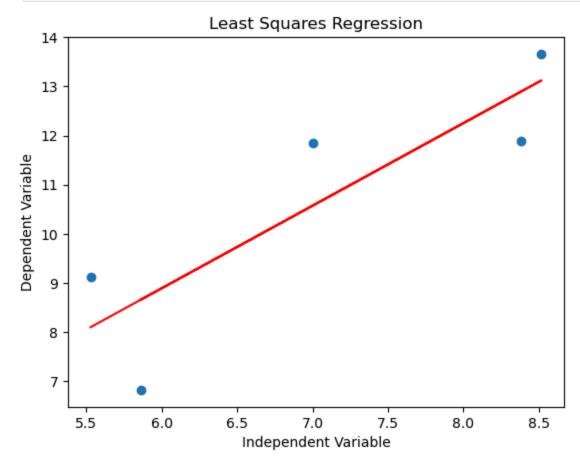
beta1 = numerator / denominator
    beta0 = mean_dep_val - (beta1 * mean_indep_val)
```

```
predict_dep_val = beta0 + beta1 * predict_indep_val

plt.scatter(indep_val, dep_val)
plt.plot(indep_val, beta0 + beta1 * indep_val, 'red')
plt.xlabel('Independent Variable')
plt.ylabel('Dependent Variable')
plt.title('Least Squares Regression')
plt.show()

return {'beta0': beta0, 'beta1': beta1, 'predict_val': predict_dep_val}

result = leastSquareRegression(df.X, df.Y, 15.67)
print(f'predicted value for x = 15.67 is : {result["predict_val"]}')
```

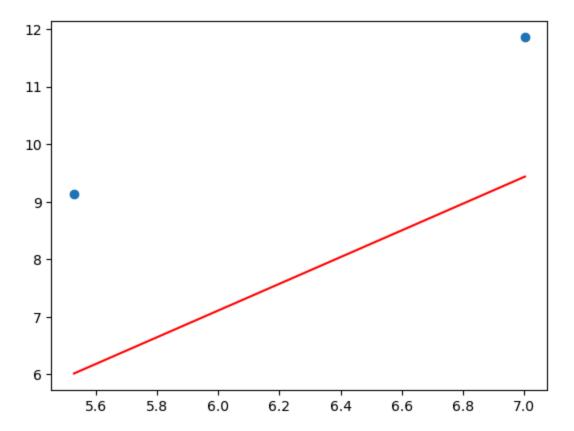


predicted value for x = 15.67 is : 25.135710655002136

Question 13: Based on multiple features/variables perform Linear Regression. For example, based on several additional features like number of bedrooms, servant room, number of balconies, number of houses of years a house has been built – predict the price of a house.

```
In [13]: df=pd.read_csv('lindata.csv')
    df.columns=['X','Y']
    df
```

```
Out[13]:
                X
                         Υ
          0 5.5277
                     9.1302
          1 8.5186 13.6620
          2 7.0032 11.8540
          3 5.8598
                    6.8233
          4 8.3829 11.8860
In [14]: X=np.array(df['X']).reshape(-1,1)
         Y=np.array(df['Y']).reshape(-1,1)
In [15]: x_train, x_test, y_train, y_test = train_test_split(X,Y,test_size = 0.25)
In [16]: lr=LinearRegression()
         lr.fit(x_train,y_train)
Out[16]:
         ▼ LinearRegression
         LinearRegression()
         Y_pred=lr.predict(x_test)
In [17]:
In [24]: print('mean square error :',mean_squared_error(y_test,Y_pred))
         print(r2_score(y_test,Y_pred))
        mean square error : 7.774748341791334
        -3.191755092580558
In [25]: lr.score(x_test, y_test)
Out[25]: -3.191755092580558
In [20]: print('slope:',lr.coef_)
         print('Intercept:',lr.intercept_)
        slope: [[2.31888009]]
        Intercept: [-6.80314178]
In [21]: y=lr.predict([[15.67]])
In [22]: y[0][0]
Out[22]: 29.53370919583682
In [23]: plt.scatter(x_test,y_test)
         plt.plot(x_test,Y_pred,'red')
          plt.show()
```



Question 14: Implement a classification/ logistic regression problem. For example based on different features of students data, classify, whether a student is suitable for a particular activity. Based on the available dataset, a student can also implement another classification problem like checking whether an email is spam or not.

```
In [51]: import seaborn as sns
#from sklearn.model_selection import train_test_split
from sklearn.linear_model import LogisticRegression
In [31]: df=pd.read_csv('diabetes.csv')
df
```

Out[31]:		Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	ВМІ	DiabetesPedigreeFu
	0	6	148	72	35	0	33.6	
	1	1	85	66	29	0	26.6	
	2	8	183	64	0	0	23.3	
	3	1	89	66	23	94	28.1	
	4	0	137	40	35	168	43.1	
	•••							
	763	10	101	76	48	180	32.9	
	764	2	122	70	27	0	36.8	
	765	5	121	72	23	112	26.2	
	766	1	126	60	0	0	30.1	
	767	1	93	70	31	0	30.4	

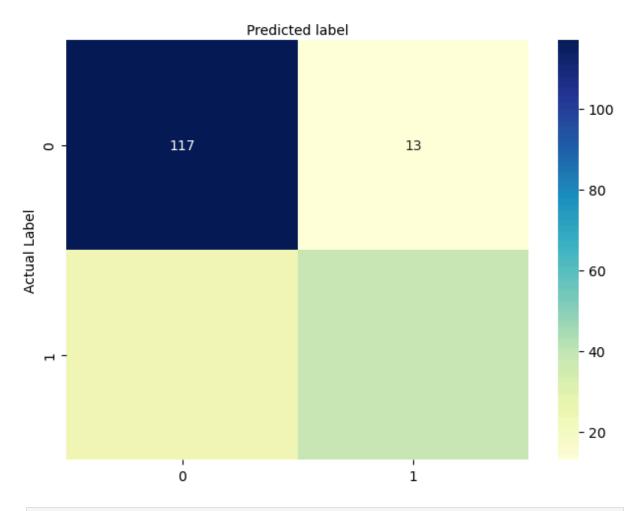
768 rows × 9 columns

```
In [45]: | feature_cols = ['Pregnancies','Insulin','BMI','Age','Glucose','BloodPressure','Diab
         x=df[feature_cols]
         y=df.Outcome
In [46]: x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.25,random_state=0)
In [47]: logreg=LogisticRegression()
         logreg.fit(x_train,y_train)
         y_pred=logreg.predict(x_test)
        C:\Users\lamot\anaconda3\Lib\site-packages\sklearn\linear_model\_logistic.py:458: Co
        nvergenceWarning: lbfgs failed to converge (status=1):
        STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.
        Increase the number of iterations (max_iter) or scale the data as shown in:
            https://scikit-learn.org/stable/modules/preprocessing.html
        Please also refer to the documentation for alternative solver options:
            https://scikit-learn.org/stable/modules/linear_model.html#logistic-regression
          n_iter_i = _check_optimize_result(
In [48]: from sklearn import metrics
         cnf_matrix=metrics.confusion_matrix(y_test,y_pred)
         cnf_matrix
Out[48]: array([[117, 13],
                 [ 24, 38]], dtype=int64)
In [49]: class_names=[0,1]
         fig,ax=plt.subplots()
```

```
tick_marks=np.arange(len(class_names))
plt.xticks(tick_marks,class_names)
plt.yticks(tick_marks,class_names)
#create heatmap
sns.heatmap(pd.DataFrame(cnf_matrix),annot=True,cmap="YlGnBu",fmt='g')
ax.xaxis.set_label_position("top")
plt.tight_layout()
plt.title('Confusion matrix',y=1.1)
plt.ylabel('Actual Label')
plt.xlabel('Predicted label')
```

Out[49]: Text(0.5, 427.95555555555, 'Predicted label')

Confusion matrix



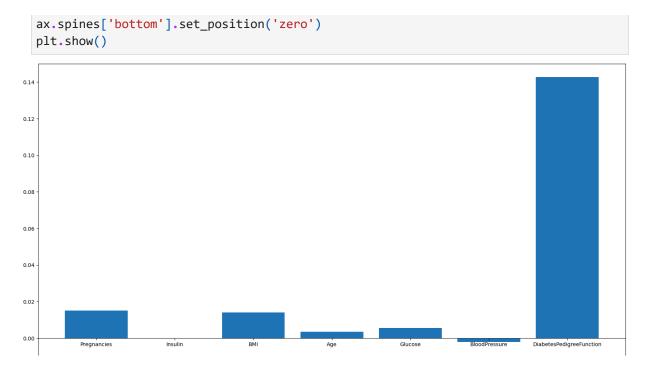
```
In [50]: print("Accuracy:",metrics.accuracy_score(y_test,y_pred))
    print("Precision:",metrics.precision_score(y_test,y_pred))
    print("Recall:",metrics.recall_score(y_test,y_pred))
```

Question 15: Use some function for regularization of dataset based on problem 14.

```
In [53]: #import pandas as pd
#import numpy as py
```

```
#import matplotlib.ppyplot as plt
          from sklearn import datasets
          #from sklearn.linear model import LinearRegression
          #from sklearn.model_selection import train_test_split
In [56]:
         data_url = "http://lib.stat.cmu.edu/datasets/boston"
          raw_df = pd.read_csv(data_url, sep="\s+", skiprows=22, header=None)
          data = np.hstack([raw_df.values[::2, :], raw_df.values[1::2, :2]])
          target = raw_df.values[1::2, 2]
In [59]:
          boston_dataset=raw_df
          boston_dataset
Out[59]:
                              1
                                    2
                                          3
                                                 4
                                                       5
                                                             6
                                                                    7
                                                                          8
                                                                                 9
                                                                                     10
             0
                  0.00632
                           18.00
                                  2.31
                                         0.0
                                             0.538
                                                    6.575
                                                           65.2
                                                                4.0900
                                                                         1.0
                                                                             296.0
                                                                                    15.3
                396.90000
                            4.98
                                 24.00
                                        NaN
                                                          NaN
                                              NaN
                                                     NaN
                                                                  NaN
                                                                       NaN
                                                                              NaN
                                                                                    NaN
             2
                  0.02731
                            0.00
                                  7.07
                                         0.0
                                             0.469
                                                    6.421
                                                           78.9
                                                                4.9671
                                                                         2.0
                                                                             242.0
                                                                                    17.8
                396.90000
                                 21.60
                                        NaN
                            9.14
                                              NaN
                                                     NaN
                                                          NaN
                                                                  NaN
                                                                       NaN
                                                                              NaN
                                                                                    NaN
                  0.02729
                            0.00
                                  7.07
                                         0.0
                                             0.469
                                                    7.185
                                                           61.1
                                                                4.9671
                                                                         2.0
                                                                             242.0
                                                                                    17.8
          1007
               396.90000
                                 23.90
                                       NaN
                            5.64
                                              NaN
                                                     NaN
                                                          NaN
                                                                  NaN
                                                                       NaN
                                                                              NaN
                                                                                   NaN
          1008
                  0.10959
                            0.00
                                11.93
                                         0.0
                                             0.573
                                                    6.794
                                                           89.3
                                                                2.3889
                                                                         1.0
                                                                                    21.0
                                                                             273.0
               393.45000
                                 22.00
          1009
                            6.48
                                       NaN
                                              NaN
                                                     NaN
                                                          NaN
                                                                  NaN
                                                                       NaN
                                                                              NaN
                                                                                   NaN
                                                                                    21.0
          1010
                            0.00
                                11.93
                                             0.573
                                                    6.030
                                                          8.08
                                                                2.5050
                                                                             273.0
                  0.04741
                                         0.0
                                                                         1.0
          1011 396.90000
                            7.88 11.90 NaN
                                              NaN
                                                    NaN
                                                          NaN
                                                                  NaN NaN
                                                                              NaN NaN
         1012 rows × 11 columns
         x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.25,random_state=0)
In [60]:
In [61]: lreg=LinearRegression()
          lreg.fit(x_train,y_train)
Out[61]:
          ▼ LinearRegression
          LinearRegression()
In [62]: y_pred=lreg.predict(x_test)
In [67]:
         lreg_coff=pd.DataFrame()
          lreg_coff['Columns']=x_train.columns
          lreg_coff['Cofficient']=pd.Series(lreg.coef_)
```

```
lreg_coff
In [68]:
Out[68]:
                            Columns Cofficient
                          Pregnancies
          0
                                       0.015115
          1
                              Insulin
                                      -0.000119
          2
                                 BMI
                                       0.014168
          3
                                       0.003623
                                Age
                                       0.005652
          4
                             Glucose
          5
                        BloodPressure
                                      -0.002039
          6 DiabetesPedigreeFunction
                                       0.145175
In [83]:
          fig,ax=plt.subplots(figsize=(20,10))
          ax.bar(['Pregnancies','Insulin','BMI','Age','Glucose','BloodPressure','DiabetesPedi
          ax.spines['bottom'].set_position('zero')
          plt.show()
        0.14
        0.12
        0.10
        0.06
        0.04
        0.02
                 Pregnancies
                              Insulin
In [77]: from sklearn.linear_model import Ridge
          ri=Ridge()
          ri.fit(x_train,y_train)
          ri_y_pred=ri.predict(x_test)
In [78]:
          ri_coff=pd.DataFrame()
          ri_coff['Columns']=x_train.columns
          ri_coff['Cofficient']=pd.Series(ri.coef_)
In [82]: fig,ax=plt.subplots(figsize=(20,10))
          ax.bar(['Pregnancies','Insulin','BMI','Age','Glucose','BloodPressure','DiabetesPedi
```



Question 16: Use some function for neural networks, like Stochastic Gradient Descent or backpropagation - algorithm to predict the value of a variable based on the dataset of problem 14.

```
In [84]: #import pandas as pd
    #from sklearn.model_selection import train_test_split
    #numpy as np
    from sklearn.linear_model import SGDClassifier
    #from sklearn import metrics
In [85]: df=pd.read_csv('diabetes.csv')
df
```

Out[85]:		Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	вмі	DiabetesPedigreeFu
	0	6	148	72	35	0	33.6	
	1	1	85	66	29	0	26.6	
	2	8	183	64	0	0	23.3	
	3	1	89	66	23	94	28.1	
	4	0	137	40	35	168	43.1	
	•••		•••					
	763	10	101	76	48	180	32.9	
	764	2	122	70	27	0	36.8	
	765	5	121	72	23	112	26.2	
	766	1	126	60	0	0	30.1	
	767	1	93	70	31	0	30.4	
	768 rd	ows × 9 colum	ins					

```
In [86]: | feature_cols=['Pregnancies','Insulin','BMI','Age','Glucose','BloodPressure','Diabet
         x=df[feature_cols]
         y=df.Outcome
In [87]: x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.25,random_state=0)
In [88]: sgdc=SGDClassifier(max_iter=100,)
         sgdc.fit(x_train,y_train)
         y_pred=sgdc.predict(x_test)
In [89]: print("Accuracy:",metrics.accuracy_score(y_test,y_pred))
         print("Precision:", metrics.precision_score(y_test,y_pred))
         print("Recall:",metrics.recall_score(y_test,y_pred))
         print("F1-score", metrics.accuracy_score(y_test,y_pred))
        Accuracy: 0.359375
        Precision: 0.33513513513513515
        Recall: 1.0
        F1-score 0.359375
In [90]: cnf_matrix=metrics.confusion_matrix(y_test,y_pred)
         cnf_matrix
Out[90]: array([[ 7, 123],
                 [ 0, 62]], dtype=int64)
```