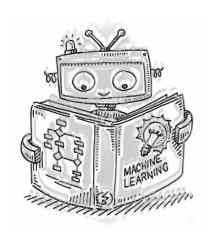


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Machine Learning

Practical File

Submitted By -

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Perform elementary mathematical operations in Octave/MATLAB/R/Python like addition, multiplication, division and exponentiation.

```
In [1]: # Addition
    5 + 7

Out[1]: 12

In [2]: # Subtraction
    5 - 7

Out[2]: -2

In [3]: # Multiplication
    5 * 7

Out[3]: 35

In [4]: # Division
    5 / 7

Out[4]: 0.7142857142857143

In [5]: # Exponentiation
    5 ^ 7

Out[5]: 2
```

2

Perform elementary logical operations in Octave/MATLAB/R/Python (like OR, AND, Checking for Equality, NOT, XOR).

```
In [6]: # OR
    True | False

Out[6]: True

In [7]: # AND
    True & False

Out[7]: False

In [8]: # Check for inequality
    True == False
```

```
Out[8]: False

In [9]: # NOT  
~True

Out[9]: -2

In [10]: # XOR  
   print(True ^ True)  
   print(True ^ False)

False  
True
```

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

```
In [11]: # Creating and initializing a variable
    a = 5
    x1 = 10
    x2 = 20
    name = "Anshul"

In [12]: # Displaying variables
    print(a)
    print(x1)
    print(x2)
    print(name)

5
    10
    20
    Anshul
```

4

Create/Define single dimension / multidimension 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 [13]: import numpy as np
In [14]: # Creating single-dimension arrays
    x = np.array([1, 2, 3, 4, 5])
    print('x = ', x)
```

```
y = np.array([[1], [2], [3]])
           print('y = \n', y)
          x = [1 2 3 4 5]
          y =
           [[1]
           [2]
           [3]]
In [15]:
          # Creating multi-dimension arrays
           z = np.array([[1, 2, 3], [6, 7, 8]])
           print('z = \n', z)
           z1 = np.matrix('1 2 3; 6 7 8')
           print('z1 = \n', z1)
          z =
          [[1 2 3]
           [6 7 8]]
          z1 =
           [[1 2 3]
           [6 7 8]]
In [16]: # Matrix with all ones
          A = np.ones((4, 4))
           Α
Out[16]: array([[1., 1., 1., 1.],
                 [1., 1., 1., 1.],
                 [1., 1., 1., 1.],
                 [1., 1., 1., 1.]])
          # Matrix with all zeros
In [17]:
           B = np.zeros((4, 4))
           В
Out[17]: array([[0., 0., 0., 0.],
                 [0., 0., 0., 0.]
                 [0., 0., 0., 0.]
                 [0., 0., 0., 0.]
In [18]:
          # Matrix with random values within a range
           C = np.random.randint(20, 50, (4,5))
           print("C = \n", C)
           # Matrix with range
           C1 = np.arange(12).reshape((3, 4))
           print("C1 = \n", C1)
           [[30 39 40 29 32]
           [23 47 36 32 22]
           [41 27 26 20 49]
           [35 39 46 30 44]]
          C1 =
           [[0 1 2 3]
           [4567]
           [ 8 9 10 11]]
          # Diagonal matrix
In [19]:
           D = np.diag([1, 2, 3, 4, 5])
           print('D = \n', D)
          D =
```

```
[[1 0 0 0 0]
[0 2 0 0 0]
[0 0 3 0 0]
[0 0 0 4 0]
[0 0 0 0 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 [20]: import numpy
          A = np.arange(12).reshape((3, 4))
Out[20]: array([[ 0, 1, 2, 3],
                 [4, 5, 6, 7],
                 [ 8, 9, 10, 11]])
          # Size of matrix
In [21]:
          # A.size
          np.size(A)
Out[21]: 12
In [22]:
          # Shape of matrix
          A. shape
Out[22]: (3, 4)
          # Length of 1st row
In [23]:
          len(A[0])
Out[23]: 4
In [24]: # Length of first column
          np.size(A, 0)
Out[24]: 3
In [25]:
          # Loading data from text file
          #np.loadtxt("input5.txt", delimiter=',', dtype=int)
          import pandas as pd
          df = pd.read_csv('input5.txt')
          df
Out[25]:
          Sell "List" "Living" "Rooms" "Beds" "Baths" "Age" "Acres" "Taxes"
          0 142
                   160
                           28
                                    10
                                            5
                                                    3
                                                         60
                                                                0.28
                                                                       3167
```

		Sell	"List"	"Living"	"Rooms"	"Beds"	"Baths"	"Age"	"Acres"	"Taxes"
	1	175	180	18	8	4	1	12	0.43	4033
	2	129	132	13	6	3	1	41	0.33	1471
	3	138	140	17	7	3	1	22	0.46	3204
	4	232	240	25	8	4	3	5	2.05	3613
	5	135	140	18	7	4	3	9	0.57	3028
	6	150	160	20	8	4	3	18	4.00	3131
	7	207	225	22	8	4	2	16	2.22	5158
	8	271	285	30	10	5	2	30	0.53	5702
n [26]:	<pre>df.drop(df.tail(5).index,inplace=True) df</pre>									
ut[26]:		Sell	"List"	"Living"	"Rooms"	"Beds"	"Baths"	"Age"	"Acres"	"Taxes"
	0	142	160	28	10	5	3	60	0.28	3167
	1	175	180	18	8	4	1	12	0.43	4033
	2	129	132	13	6	3	1	41	0.33	1471

0.46

3204

6

3 138

140

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

```
In [29]: import numpy as np
A = np.array([[3, 6, 9], [5, -10, 15], [-7, 14, 21]])
B = np.array([[9, -18, 27], [11, 22, 33], [13, -26, 39]])
print("A = \n", A, "\nB = \n", B)

A =
[[ 3  6  9]
[ 5 -10  15]
[ -7  14  21]]
B =
[[ 9 -18  27]
[ 11  22  33]
[ 13 -26  39]]
```

```
In [30]: # Matrix Addition
          C = A + B
          print('C = A + B = \n', C)
          C = A + B =
           [[ 12 -12 36]
           [ 16 12 48]
           [ 6 -12 60]]
In [31]: # Matrix Subtraction
          C = A - B
          print('C = A - B = \n', C)
          C = A - B =
           [[ -6 24 -18]
           [ -6 -32 -18]
           [-20 40 -18]]
In [32]: # Matrix Multiplication
          C = A.dot(B)
          print('C = A * B = \n', C)
          C = A * B =
          [[ 210 -156 630]
           [ 130 -700 390]
           [ 364 -112 1092]]
In [33]: # Print 2nd row of Matrix A
          print(A[1:2])
          [[ 5 -10 15]]
In [34]: # Print 1st row of Matrix B
          print(B[:1])
          [[ 9 -18 27]]
In [35]: # Print 2nd column of Matrix A
          print(A[:,1:2])
          [[ 6]
           [-10]
           [ 14]]
In [36]: # Print 3rd column of Matrix B
          print(B[:,2:3])
          [[27]
           [33]
          [39]]
```

Perform other matrix operations like converting matrix data to absolute values, taking the negative of matrix values, additing/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.

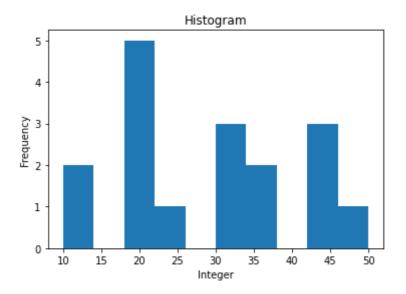
```
In [37]:
          import numpy as np
          A = np.array([[3, 6, 9], [5, -10, 15], [-7, 14, 21]])
          B = np.array([[9, -18, 27], [11, 22, 33], [13, -26, 39]])
          print("A = \n", A, "\nB = \n", B)
          A =
           [[ 3 6 9]
           [ 5 -10 15]
[ -7 14 21]]
          B =
           [[ 9 -18 27]
           [ 11 22 33]
           [ 13 -26 39]]
In [38]:
          # Converting matrix A data to its absolute values
          np.absolute(A)
Out[38]: array([[ 3, 6, 9],
                [ 5, 10, 15],
                 [7, 14, 21]])
In [39]: # Converting matrix B data to its negative values
          np.negative(B)
Out[39]: array([[ -9, 18, -27],
                 [-11, -22, -33],
                 [-13, 26, -39]])
In [40]: | # Deleting a row from Matrix A
          np.delete(A, 1, 0)
Out[40]: array([[ 3, 6, 9],
                [-7, 14, 21]])
          # Deleting a column from Matrix B
In [41]:
          np.delete(B, 0, 1)
Out[41]: array([[-18, 27],
                 [ 22, 33],
                 [-26, 39]])
In [42]:
          # Adding a row to Matrix A
          np.append(A, np.array([[23, -45, 56]]), axis=0)
Out[42]: array([[ 3, 6,
                 [ 5, -10, 15],
                 [ -7, 14, 21],
                 [ 23, -45, 56]])
In [43]: # Adding a column to Matrix B
          np.append(B, [[23], [-45], [56]], axis=1)
Out[43]: array([[ 9, -18, 27, 23],
                 [ 11, 22, 33, -45],
                 [ 13, -26, 39, 56]])
In [44]: # Maximum of 2nd row of Matrix A
          np.max(A, 0)[1]
```

```
Out[44]: 14
In [45]:
          # Minimum of 2nd row of Matrix A
          np.min(A, 0)[1]
Out[45]: -10
          # Maximum of 3rd column of Matrix B
In [46]:
          np.max(B, 1)[2]
Out[46]: 39
In [47]:
          # Minimum of 3rd column of Matrix B
          np.min(B, 1)[2]
Out[47]: -26
In [48]: # Sum of some elements of array
          np.sum(A[1:, 1:])
Out[48]: 40
          # Sum of all elements of array
In [49]:
          sumA = np.sum(A)
          sumB = np.sum(B)
          print('sumA = ', sumA, ', sumB = ', sumB)
          sumA = 56, sumB = 110
```

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.

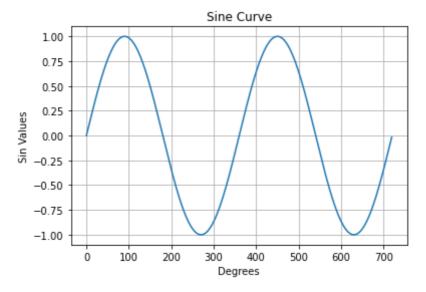
```
import numpy as np
import matplotlib.pyplot as plt

# Histogram
list = np.array([20, 45, 45, 35, 30, 10, 30, 20, 20, 50, 30, 20, 20, 10, 45
plt.hist(list)
plt.xlabel('Integer')
plt.ylabel('Frequency')
plt.title('Histogram')
plt.show()
```

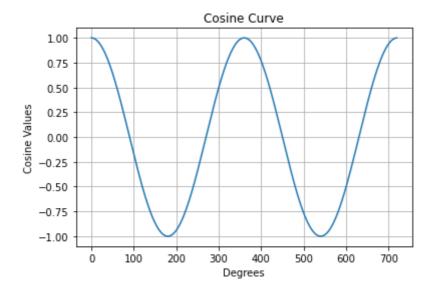


```
import math

# Sine curve
degrees = range(0 , 720)
sinValues = [math.sin(math.radians(i)) for i in degrees]
plt.plot(sinValues)
plt.xlabel('Degrees')
plt.ylabel('Sin Values')
plt.title('Sine Curve')
plt.grid()
plt.show()
```



```
In [52]: # Cosine curve
    degrees = range(0 , 720)
    sinValues = [math.cos(math.radians(i)) for i in degrees]
    plt.plot(sinValues)
    plt.xlabel('Degrees')
    plt.ylabel('Cosine Values')
    plt.title('Cosine Curve')
    plt.grid()
    plt.show()
```



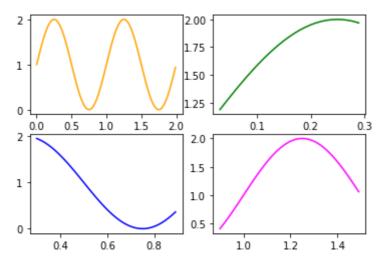
Generate different subplots from a given plot and color plot data.

```
import matplotlib.pyplot as plt
import numpy as np

# Data for plotting
x = np.arange(0.0, 2.0, 0.01)
y = 1 + np.sin(2 * np.pi * x)

# Creating 6 subplots and unpacking the output array immediately
fig, ((ax1, ax2), (ax3, ax4)) = plt.subplots(2, 2)
ax1.plot(x, y, color="orange")
ax2.plot(x[3:30], y[3:30], color="green")
ax3.plot(x[30:90], y[30:90], color="blue")
ax4.plot(x[90:150], y[90:150], color="magenta")
```

Out[53]: [<matplotlib.lines.Line2D at 0x1d7944a4ca0>]



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

```
In [54]: #if - elif - else
           grade = None
           marks = 90
           if marks >= 95:
              grade = 'A+'
           elif marks >= 90:
              grade = 'A'
           elif marks >= 80:
              grade = 'B'
           elif marks >= 75:
              grade = 'C'
           elif marks >= 65:
              grade = 'D'
           else:
              grade = 'F'
           grade
Out[54]: 'A'
In [55]:
          # while loop
           i = 1
           while i < 6:
            print(i)
            if i == 3:
              break
             i += 1
          2
In [56]:
          # for Loop
           fruits = ["apple", "cherry", "banana"]
           for x in fruits:
            print(x)
             if x == "banana":
               break
          apple
          cherry
          banana
```

11

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

```
import numpy as np
A = np.array([[3, 6, 9], [5, -10, 15], [-7, 14, 21]])
B = np.array([[9, -18, 27], [11, 22, 33], [13, -26, 39]])
print("A = \n", A, "\nB = \n", B)
```

```
A =
         [[ 3 6 9]
[ 5 -10 15]
          [ -7 14 21]]
          [[ 9 -18 27]
          [ 11 22 33]
          [ 13 -26 39]]
In [58]: # Addition
          A + B
Out[58]: array([[ 12, -12,
                          36],
               [ 16, 12, 48],
[ 6, -12, 60]])
In [59]: # Subtraction
          A - B
# Multiplication
In [60]:
          A @ B
Out[60]: array([[ 210, -156, 630],
               [ 130, -700, 390],
                [ 364, -112, 1092]])
In [61]:
          # Transpose
          A' =
         [[ 3 5 -7]
[ 6 -10 14]
[ 9 15 21]]
B' =
          [[ 9 11 13]
          [-18 22 -26]
[ 27 33 39]]
```

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.

```
#Required imports
In [1]:
          import pandas as pd
          import numpy as np
          from sklearn import linear_model
          import matplotlib.pyplot as plt
In [2]:
          # Reading csv file to dataframe
          df = pd.read_csv('houseprices.csv')
          df.head()
Out[2]:
                   price
            area
         0 2600 550000
            2800
                 556000
            3000
                 565000
           3200 610000
         4 3400 640000
In [3]:
         # Scatter plot for the dataset
          %matplotlib inline
          plt.xlabel('area')
          plt.ylabel('price')
          plt.scatter(df.area,df.price,color='red',marker='+')
Out[3]: <matplotlib.collections.PathCollection at 0x1c0d22ae3d0>
            800000
            750000
            700000
            650000
            600000
            550000
                 2500
                                   3500
                                                      4500
                                                               5000
                          3000
                                             4000
                                         area
```

Preparing data for training

```
x_df = df.drop('price',axis='columns')
In [4]:
         x_df.head()
Out[4]:
            area
         0 2600
         1 2800
         2 3000
         3 3200
         4 3400
In [5]:
         price = df.price
         price
              550000
Out[5]: 0
         1
              556000
         2
              565000
         3
             610000
         4
             640000
         5
             680000
         6
             720000
         7
             725000
         8
             760000
         9
             779000
         10
              800000
        Name: price, dtype: int64
        Applying Linear Regression
        # Create linear regression object
In [6]:
         reg = linear_model.LinearRegression()
         reg.fit(x_df,price)
Out[6]: LinearRegression()
In [7]: m = reg.coef_
         c = reg.intercept
         print('Coefficient, m = ', m)
         print('Intercept, c = ', c)
         Coefficient, m = [114.12402428]
         Intercept, c = 250142.23764093674
        Predictions
         ans1 = reg.predict([[3300]])
In [8]:
         print('(1) Price of a house with area = 3300 sqr ft: ', ans1)
         (1) Price of a house with area = 3300 sqr ft: [626751.51777971]
         y = m*3300 + c
In [9]:
         print('y = m*x + c =', y)
```

```
y = m*x + c = [626751.51777971]
Here, we can see that y = ans1 = 626751.51777971
```

Another prediction

```
In [10]: ans2 = reg.predict([[6000]])
    print('(2) Price of a house with area = 6000 sqr ft: ', ans2)

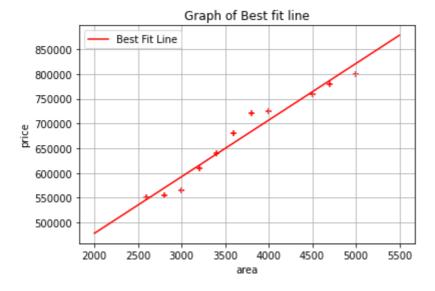
(2) Price of a house with area = 6000 sqr ft: [934886.38334779]

In [11]: y = m*6000 + c
    print('y = m*x + c =', y)

y = m*x + c = [934886.38334779]
```

Visualising Best Fit Line

```
In [12]: x = np.linspace(2000,5500)
y = m*x+c
plt.plot(x, y, '-r', label='Best Fit Line')
plt.legend(loc='upper left')
plt.title('Graph of Best fit line')
plt.xlabel('area')
plt.ylabel('price')
plt.scatter(df.area,df.price,color='red',marker='+')
plt.grid()
plt.show()
```



13.

Based on multiple features/variables perform Linear Regression. For example, based on a number of 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.

```
#Required imports
In [13]:
           import pandas as pd
           import numpy as np
           from sklearn import linear_model
In [14]:
           # Reading csv file to dataframe
           df = pd.read_csv('houseprices2.csv')
Out[14]:
             area bedrooms age
                                   price
          0 2600
                         3.0
                              20 550000
          1 3000
                              15 565000
                         4.0
          2 3200
                              18 610000
                        NaN
          3 3600
                         3.0
                              30 595000
          4 4000
                         5.0
                               8 760000
          5 4100
                         6.0
                               8 810000
```

Data Preprocessing: Fill NA values with median value of a column

```
df.bedrooms.median()
In [15]:
Out[15]: 4.0
In [16]:
           df.bedrooms = df.bedrooms.fillna(df.bedrooms.median())
Out[16]:
              area
                   bedrooms
                             age
                                    price
           0 2600
                              20 550000
                         3.0
             3000
                         4.0
                              15 565000
           2 3200
                         4.0
                              18 610000
           3 3600
                          3.0
                               30 595000
           4 4000
                         5.0
                                8 760000
           5 4100
                          6.0
                                8 810000
```

Applying Linear Regression

```
In [17]:    reg = linear_model.LinearRegression()
    reg.fit(df.drop('price',axis='columns'), df.price)
Out[17]: LinearRegression()
```

```
In [18]:
          m1, m2, m3 = reg.coef_
          c = reg.intercept_
          print('Coefficients, \
          \n = {}, \
          \n = {}, \
          n\tan = {}'.format(m1, m2, m3))
          print('Intercept, c = ', c)
          Coefficients,
                 m1 = 112.06244194213456,
                 m2 = 23388.880077939153,
                 m3 = -3231.717908632967
          Intercept, c = 221323.00186540443
         Predictions
In [19]:
          ans1 = reg.predict([[3000, 3, 40]])
          print('(1) Price of home with 3000 sqr ft area, 3 bedrooms, 40 year old: ', ar
          (1) Price of home with 3000 sqr ft area, 3 bedrooms, 40 year old: [498408.251
          58031]
In [20]:
          y1 = m1*3000 + m2*3 + m3*40 + c
           print('\ty1 = m1*x1 + m2*x2 + m3*x3 + c = \n\t', y1)
                  y1 = m1*x1 + m2*x2 + m3*x3 + c =
                   498408.2515803069
In [21]:
          reg.predict([[2500, 5, 10]])
          print('(2) Price of home with 2500 sqr ft area, 5 bedrooms, 10 year old: ', ar
          (2) Price of home with 2500 sqr ft area, 5 bedrooms, 10 year old: [934886.383
          34779]
          y1 = m1*2500 + m2*5 + m3*10 + c
In [22]:
          print('\ty1 = m1*x1 + m2*x2 + m3*x3 + c = \n\t', y1)
                  y1 = m1*x1 + m2*x2 + m3*x3 + c =
```

14.

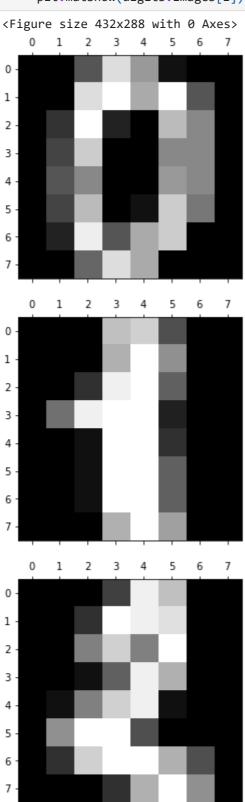
586106.3280241069

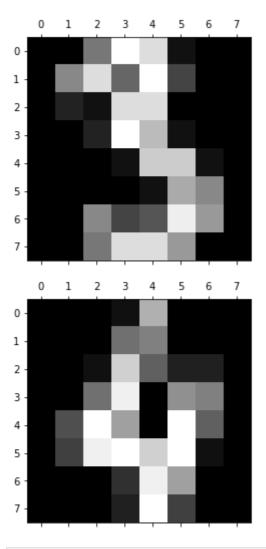
Implement a classification/ logistic regression problem. For example based on different features of students data, classify, whether a student is suitable for a particular activity. Based on the available dataset, a student can also implement another classification problem like checking whether an email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

```
In [23]: # Import and Load digits dataset
    from sklearn.datasets import load_digits
    digits = load_digits()
# Import matplotlib
```

In [24]:

```
# Plot 2D matrix data of digits
plt.gray()
for i in range(5):
    plt.matshow(digits.images[i])
```





```
In [25]: # Get the attributes/columns of digits dataset
    dir(digits)
```

Out[25]: ['DESCR', 'data', 'feature_names', 'frame', 'images', 'target', 'target_name s']

Creating and training the logistic regression model

```
In [26]: # Import the model
    from sklearn.linear_model import LogisticRegression
    model = LogisticRegression(max_iter=3000)

In [27]: # Import train_test_split
    from sklearn.model_selection import train_test_split
    # Split the dataset into training and testing datasets
    X_train, X_test, y_train, y_test = train_test_split(digits.data,digits.target,)

In [28]: print(len(X_train), len(X_test), len(y_train), len(y_test))
    1437 360 1437 360

In [29]: # Training the model
    model.fit(X_train, y_train)
```

plt.ylabel('Truth')

Measuring accuracy of our model

```
In [30]:
         model.score(X_test, y_test)
Out[30]: 0.9611111111111111
         Predictions
          model.predict(digits.data[0:5])
In [31]:
Out[31]: array([0, 1, 2, 3, 4])
In [32]: y_predicted = model.predict(X_test)
           y_predicted
Out[32]: array([0, 1, 1, 4, 3, 4, 6, 8, 6, 2, 1, 6, 8, 6, 7, 2, 7, 2, 1, 3, 5, 4,
                 3, 9, 4, 9, 6, 7, 7, 7, 1, 6, 8, 5, 6, 4, 4, 1, 4, 4, 4, 2, 9, 3,
                 6, 9, 1, 7, 8, 3, 6, 9, 8, 0, 9, 1, 2, 7, 8, 9, 8, 8, 9, 1, 3, 8,
                 8, 5, 7, 6, 3, 6, 7, 4, 6, 7, 6, 6, 8, 8, 0, 5, 4, 9, 7, 9, 0, 1,
                   7, 2, 1, 6, 3, 8, 0, 2, 1, 1, 5, 0, 0, 7, 3, 5, 1, 5, 6, 0, 5,
                       2, 0, 0, 7, 3, 0, 7, 4, 5, 9, 0, 6, 5, 9, 1, 7, 8, 9, 3, 8,
                   3, 9, 0, 0, 0, 9, 0, 4, 0, 7, 5, 3, 0, 7, 1, 1, 9, 3, 0, 5, 5,
                    7, 6, 8, 7, 9, 8, 7, 6, 5, 9, 4, 8, 2, 6, 2, 9, 3, 0, 4, 6, 9,
                       2, 0, 2, 0, 1, 4, 0, 4, 1, 6, 1, 3, 5, 1, 9, 0, 3, 3, 9, 2,
                       2, 5, 1, 4, 9, 9, 2, 7, 2, 6, 0, 9, 0, 4, 4, 3, 7, 4, 5, 0,
                      7, 0, 9, 3, 1, 4, 3, 6, 5, 7, 2, 3, 5, 7, 2, 9, 7, 4, 2, 1,
                   6, 4, 3, 8, 6, 1, 8, 2, 5, 8, 7, 5, 5, 0, 5, 8, 9, 7, 3, 6, 0,
                   1, 3, 5, 7, 0, 8, 7, 9, 1, 3, 9, 9, 5, 3, 9, 2, 1, 7, 9, 5, 6,
                   3, 0, 7, 9, 5, 8, 4, 5, 8, 1, 8, 1, 9, 4, 9, 3, 7, 7, 3, 7, 9,
                         7, 0, 9, 2, 3, 9, 0, 0, 1, 9, 5, 9, 3, 3, 5, 4, 2, 1, 2,
                      2,
                 2, 4, 1, 3, 8, 4, 7, 8, 7, 9, 3, 0, 4, 4, 4, 5, 9, 1, 3, 2, 6, 6,
                 9, 0, 0, 6, 2, 7, 3, 5])
         Confusion matrix
          from sklearn.metrics import confusion matrix
In [33]:
           cm = confusion_matrix(y_test, y_predicted)
           cm
                               0,
                                   0,
                                       0,
                                           0,
                                               0,
                                                   0,
                                                       0],
Out[33]: array([[39,
                      0,
                          0,
                                   0,
                                          1,
                                                       1],
                 [ 0, 33, 0,
                              0,
                                       0,
                                               0,
                                                  1,
                              0,
                                   0,
                                       0,
                                              0,
                                                       0],
                      1, 29,
                 [ 0,
                                          0,
                                                  0,
                      0,
                                  0,
                                      1,
                                          0,
                                              0,
                                                       0],
                          0, 38,
                 [ 0,
                                                  0,
                      1,
                                 36,
                                      0,
                                          0,
                                              1,
                                                  0,
                                                       0],
                 [ 0,
                          0, 0,
                      0,
                             0,
                                          0,
                                              0,
                                                  0,
                                                       0],
                                   0, 32,
                 [ 0,
                          0,
                                              0,
                      1,
                                   0,
                                                       0],
                 [ 0,
                          0, 0,
                                      0, 30,
                                                   0,
                                      0,
                      0,
                                   0,
                                          0, 38,
                                                  0,
                                                       2],
                 [ 0,
                          0, 1,
                       1,
                          0,
                              0,
                                   0,
                                          0,
                                                      0],
                 [ 0,
                                      0,
                                              0, 29,
                                   0,
                                      1,
                 [ 0,
                               0,
                                          0,
                                              1, 0, 42]], dtype=int64)
           import seaborn as sn
In [34]:
           plt.figure(figsize = (10,7))
           sn.heatmap(cm, annot=True)
           plt.xlabel('Predicted')
```

Out[34]: Text(69.0, 0.5, 'Truth')

