

MACHINE LEARNING PRACTICAL FILE

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INDEX

S.NO	PRACTICALS	SIGN
1	Perform elementary mathematical operations in Octave/MATLAB like	
	addition, subtraction, multiplication, division and exponentiation.	
2	Perform elementary logical operations in Octave/MATLAB (like OR, AND,	
	Checking for Equality, NOT, XOR).	
3	Create, initialize and display simple variables and simple strings and use	
	simple formatting for variable.	
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.	
5	Use command to compute the size of a matrix, size/length of a particular	
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6	Perform basic operations on matrices (like addition, subtraction,	
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	room, number of balconies, number of houses of years a house has been built	
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14	Implement a classification/logistic regression problem. For example based on	
	different features of student's data, classify, whether a student is suitable for	
	a particular activity. Based on the available dataset, a student can also	
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15	Use some function for regularization of dataset based on problem 14.	
16	Use some function for neural networks, like Stochastic Gradient Descent or	
	back propagation - algorithm to predict the value of a variable based on the	
	dataset of problem 14	

Question	Code	Output
Q1) Perform	#Ques1	Enter number a :2
elementary	a = int(input("Enter number a :")	TO A MARK TO SETTLE OF A SECURITION OF THE CONTRACT OF THE CON
mathematical	<pre>b = int(input("Enter number b :")</pre>	a+b: 7 a-b: -3
	print("a+b : ",a+b)	a*b : 10
operations in	print("a-b : ",a-b)	a^b : 7
Octave/MATLAB	print("a*b : ",a*b)	
like addition,	print("a^b : ",a^b)	
subtraction,		
multiplication,		
division and		
exponentiation.		
Q2) Perform	#Ques2	Enter number a :3
elementary logical	<pre>a = int(input("Enter number a :")</pre>	
, ,	<pre>b = int(input("Enter number b :")</pre>	a AND b: 5 a OR b: 3
operations in	print("a AND b : ",a and b)	a == b : False
Octave/MATLAB	print("a OR b : ",a or b)	NOT a: -4
(like OR, AND,	print("a == b : ",a == b)	a XOR b: 6
Checking for	<pre>print("NOT a : ",~a) print("a XOR b : ",a^b)</pre>	
Equality, NOT,	p22110 (2 11011 2 1 / 2 2)	
XOR).		
Q3) Create,	#Ques3	garvit 5
initialize and	a = 5	•
display simple	a = 5 b = "garvit"	
variables and		
	print(b,a)	
simple strings and		
use simple		
formatting for		
variable.		
Q4) Create/Define	#Ques4	arrl :
single dimension /	<pre>import numpy as np arrl = np.array([10,20,30])</pre>	[10 20 30]
multi-dimension	print("arrl : \n", arrl)	arr2 : [[10]
arrays, and arrays		[20]
• •	arr2 = np.array([[10],[20],[30]]) print("arr2 : \n",arr2)	[30]]
with specific	print(arrs : (n', arrs)	arr3 : [[0. 0.]
values like array	arr3 = np.zeros((2,2))	[0. 0.]]
of all ones, all	print("arr3 : \n",arr3)	arr4 :
zeros, array with	arr4 = np.ones((1,3))	[[1. 1. 1.]]
random values	print("arr4 : \n",arr4)	arr5 : [[1. 0. 0.]
within a range, or	ACCUSED SERVICE OF REAL PROPERTY OF SERVICE SE	[0. 1. 0.]
a diagonal matrix.	arr5 = np.identity(3)	[0. 0. 1.]]
	print("arr5 : \n",arr5)	

Q5) 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.

```
#Ques4
                                    [[1. 0.]
import numpy as np
                                    [0. 1.]]
arr1 = np.array([10,20,30])
                                    rows: 2
print("arrl : \n", arrl)
                                    columns: 2
arr2 = np.array([[10],[20],[30]])
print("arr2 : \n",arr2)
arr3 = np.zeros((2,2))
print("arr3 : \n",arr3)
arr4 = np.ones((1,3))
print("arr4 : \n",arr4)
arr5 = np.identity(3)
print("arr5 : \n",arr5)
```

```
size of arr : 4
 experience test score(out of 10) interview score(out of 10) salary($)
   NaN
               8.0
                                         9
     NaN
                     8.0
                                          6
                                               45000
    5.0
                     6.0
                                               60000
                     10.0
     2.0
                                          10
                                               65000
      7.0
                     9.0
                                               70000
      3.0
                      7.0
                                          10
                                               62000
                      NaN
                                               72000
     10.0
     11.0
                      7.0
                                               80000
```

Q6) Perform basic operations on matrices (like addition, subtraction, multiplication) and display specific rows or columns of the matrix.

Code:

```
#Oues6
import numpy as np
def menu():
   print ("Enter 1 foe addition")
    print ("Enter 2 foe subtraction")
    print ("Enter 3 foe multiplication")
    print ("Enter 4 to exit")
while (True):
    menu()
    c = input("Enter your choice : ")
    row = 0
    col = 0
    a = 0
    b = 0
    if(c == '4'):
    else:
       row = int(input("Enter number of rows : "))
        col = int(input("Enter number of columns : "))
        a = np.random.randint(-50,50,row*col).reshape(row,col)
        b = np.random.randint(-50,50,row*col).reshape(row,col)
        print("first matrix : ")
       print(a)
       print ("second matrix : ")
        print(b)
    if(c == '1'):
        print("Resultant matrix :\n", np.add(a,b))
    if(c == '2'):
        print("Resultant matrix :\n",np.subtract(a,b))
    if(c == '3'):
        b.reshape(col,row)
        1 = [sum([a[i][k]*b[k][j] for k in range(row)]) for i in range(row) for j in range(col)]).reshape(row,row)
        print("Resultant matrix :\n",np.array(1)
```

```
Enter 2 foe subtraction Enter 2 foe Subtraction Enter 3 for The Enter 3 for Th
                                                                                              Enter 2 foe subtraction
Enter 3 foe multiplication Enter 3 foe multiplication
Enter 4 to exit
                                                                                            Enter 4 to exit
Enter number of rows : 2
Enter number of rows: 2
Enter number of columns: 3
Enter number of columns: 3
first matrix :
                                                                                               first matrix
[[-5 -6 -2]
                                                                                               [[-16 49 -251
   [13 -9 28]]
                                                                                                  [ -6 -44 -33]
second matrix :
                                                                                                [ 3 32 -4]]
[[ 7 -45 36]
   [-38 31 -46]]
                                                                                              second matrix :
                                                                                            [[-19 -38 31]
Resultant matrix :
                                                                                             [-16 39
   [[ 2 -51 34]
                                                                                                  [ 39 -26 24]]
     [-25 22 -181]
Enter 2 foe subtraction

Enter 3 for matrix:
Enter 2 foe subtraction [[-1455 3169 -508]

Enter 3 foe multiplication [ -469 -630 -1506]
Enter 4 to exit
                                                                                              [ -725 1238 381]]
Enter your choice : 2
Enter number of rows: 3
Enter number of columns : 2
first matrix :
[[ 11 -40]
     [-26 17]
          2 17]]
    second matrix :
[[ 12 14]
     [-46 18]
    [ 22 29]]
Resultant matrix :
    [[ -1 -54]
    [ 20 -1]
    [-20 -12]]
```

Q7) Perform other matrix operations like converting matrix data to absolute values, taking the negative of matrix values, adding/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.

Code:

```
#Ques7
import numpy as np

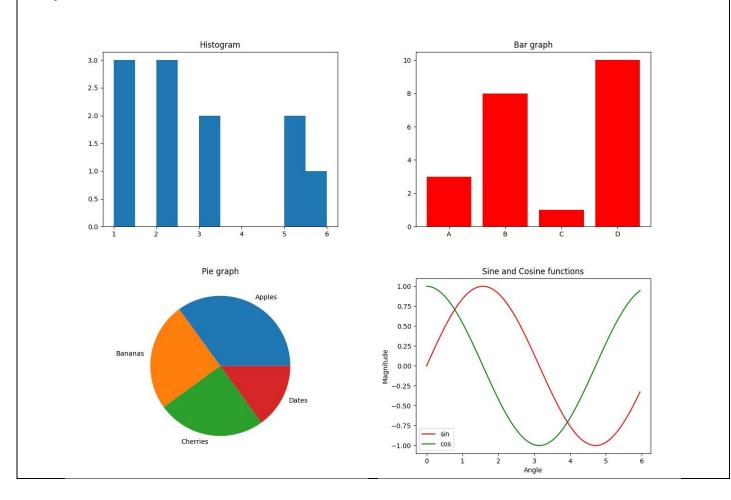
n = int(input("Enter number of rows : "))
a = np.random.randint(-50,50,n*n).reshape(n,n)
print("Matrix :\n",a)
print("Absolute : \n",np.absolute(a))
print("Negative : \n",hp.negative(a))
print("Determinant : ",np.linalg.det(a))
print("Max in rows : ",[max(a[i]) for i in range(n)])
print("Max in columns : ",[max(a.T[i]) for i in range(n)])
print("Min in rows : ",[min(a[i]) for i in range(n)])
print("Min in columns : ",[min(a.T[i]) for i in range(n)])
print("Sum of matrix : ",sum(sum(a)))
print("Sum of rows : ",[sum(a[i]) for i in range(n)])
print("Sum of columns : ",[sum(a.T[i]) for i in range(n)])
```

```
Enter number of rows: 3
Matrix :
 [[ -1 -44 -3]
 [-27 -18 -45]
 [ 36 -3 -42]]
Absolute :
[[ 1 44 3]
[27 18 45]
 [36 3 42]]
Negative :
[[ 1 44
 [ 27 18 45]
       3 4211
 [-36
Determinant: 118367.9999999997
Max in rows : [-1, -18, 36]
Max in columns : [36, -3, -3]
Min in rows : [-44, -45, -42]
Min in columns : [-27, -44, -45]
Sum of matrix: -147
Sum of rows : [-48, -90, -9]
Sum of columns : [8, -65, -90]
```

Q8) Create various types 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.

Code:

```
import matplotlib.pyplot as plt
import numpy as np
x = [1,2,3,5,1,3,5,6,2,2,1]
plt.hist(x)
plt.title('Histogram')
plt.show()
x = np.array(['A','B','C','D'])
y = np.array([3,8,1,10])
plt.bar(x,y,color = 'red')
plt.title('Bar graph')
plt.show()
y = np.array([35, 25, 25, 15])
mylabels = ['Apples', 'Bananas', 'Cherries', 'Dates']
plt.pie(y, labels = mylabels)
plt.title('Pie graph')
plt.show()
X = np.arange(0, 6, 0.05)
y = np.sin(X)
z = np.cos(X)
plt.plot(X,y,color = 'red',label = 'sin')
plt.plot(X,z,color = 'green',label = 'cos')
plt.xlabel('Angle')
plt.ylabel('Magnitude')
plt.title('Sine and Cosine functions')
plt.legend()
plt.show()
```



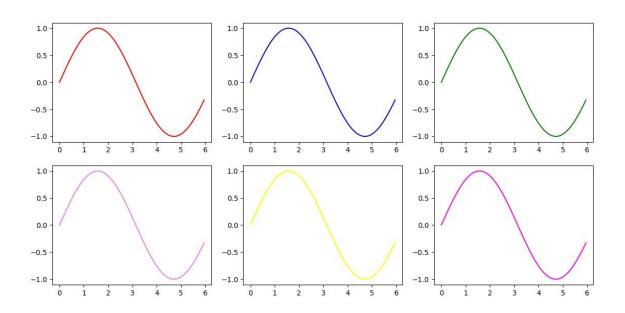
Q9) Generate different subplots from a given plot and color plot data.

Code:

```
#Ques9
import numpy as np
import matplotlib.pyplot as plt

x = np.arange(0,6,0.05)
y = np.sin(x)

fig,ax = plt.subplots(2,3)
ax[0,0].plot(x,y,color="red")
ax[0,1].plot(x,y,color="blue")
ax[0,2].plot(x,y,color="green")
ax[1,0].plot(x,y,color="violet")
ax[1,1].plot(x,y,color="yellow")
ax[1,2].plot(x,y,color="magenta")
plt.show()
```



Q10) Use conditional statements and different type of loops based on simple examples.

Code:

```
#Ques10
a = 5
b = 10
if(a == b):
    print('a is equal to b')

else:
    print('a is not equal to b')

print('value of i = [',end = ' ')
for i in range(0,11,2):
    print(i, end = ' ')
print(']')
print('Value from 5 to 10 = ',end = ' ')
while a<b:
    print(a, end = ' ')
    a += 1</pre>
```

```
a is not equal to b
value of i = [ 0 2 4 6 8 10 ]
Value from 5 to 10 = 5 6 7 8 9
```

Q11) Perform vectored implementation of simple matrix operation like finding the transpose of a matrix, adding, subtracting or multiplying two matrices.

Code:

```
#Questl
import numpy as np

matrix = np.array([[1,2,3],[4,5,6],[7,8,9]],ndmin = 2)
print(matrix)
print('\nTranspose:')
print(matrix.T)
print('\nMatrix 2:')
a = np.array([[9,8,1],[0,5,8],[8,7,3]])
print(a)
add = matrix + a
print('\nAdded matrix: \n', add)
sub = matrix - a
print('\nSubtracted matrix:\n', sub)
mul = matrix * a
print('\nMultiplied matrix:\n', mul)
```

```
[[1 2 3]
 [4 5 6]
 [7 8 9]]
Transpose:
[[1 4 7]
 [2 5 8]
[3 6 9]]
Matrix 2:
[[9 8 1]
 [0 5 8]
 [8 7 3]]
Added matrix:
 [[10 10 4]
 [ 4 10 14]
[15 15 12]]
Subtracted matrix:
 [[-8 -6 2]
[ 4 0 -2]
 [-1 1 6]]
Multiplied matrix:
 [[ 9 16 3]
 [ 0 25 48]
 [56 56 27]]
```

Q12) 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.

Code:

```
#Ques12
import pandas as pd
from sklearn import linear_model

df = pd.read_csv("homeprices.csv")
reg = linear_model.LinearRegression()
reg.fit(df[["area"]],df.price)
print("predicted price of house of area 2000 : ",reg.predict([[2000]]))
```

Output:

predicted price of house of area 2000 : [452191.78082192]

Q13) 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.

Code:

```
fQues13
import pandas as pd
from sklearn import linear_model

df = pd.read_csv("house.csv")
reg = linear_model.LinearRegression()
reg.fit(df[["area","bedrooms","servant_room","balconies"]],df.price)
print("predicted price of house on\narea = 8000\nbedrooms = 10\nservant_room = 3\nbalcony = 6\nprediction : ",reg.predict([[8000,10,3,6]]))
```

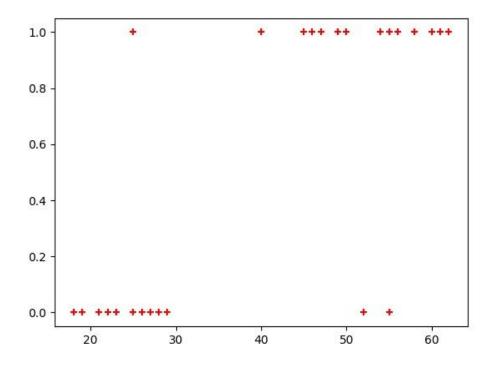
```
predicted price of house on
area = 8000
bedrooms = 10
servant_room = 3
balcony = 6
prediction : [1209999.9999991]
```

Q14) Implement a classification/ logistic regression problem. For example based on different features of student's 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.

Code:

```
#Ques14
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
from sklearn import linear model
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LogisticRegression
df = pd.read csv('insurance data.csv')
plt.scatter(df.age, df.bought insurance, marker="+", color="red")
plt.show()
x_train,x_test,y_train,y_test = train_test_split(df[['age']],df.bought_insurance,test_size=0.2)
model = LogisticRegression()
model.fit(x_train, y_train)
print("\nPrediction:", end=" ")
print(model.predict(x_test))
print ("Accuracy:", end=" ")
print(model.score(x_test, y_test) * 100)
```

Output:

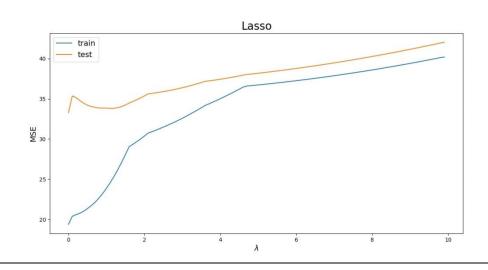


Prediction: [1 0 0 1 1 0] Accuracy: 83.33333333333334

Q15) Use some function for regularization of dataset based on problem 14.

Code:

```
#Ques15
import numpy as np
from sklearn.datasets import load boston
from sklearn.linear_model import LinearRegression
from sklearn.linear model import ElasticNet, Lasso, Ridge
from sklearn.metrics import mean squared error
import matplotlib.pyplot as plt
from sklearn.model_selection import train_test_split
X, y= load boston(return X y=True)
lr = LinearRegression()
lr.fit(X, y)
predictions = lr.predict(X)
X train, X test, y train, y test = train test split(X, y, test size=0.2)
def plot errors(lambdas, train errors, test errors, title):
    plt.figure(figsize=(16, 9))
    plt.plot(lambdas, train_errors, label="train")
    plt.plot(lambdas, test_errors, label="test")
    plt.xlabel("$\\lambda$", fontsize=14)
   plt.ylabel("MSE", fontsize=14)
   plt.title(title, fontsize=20)
   plt.legend(fontsize=14)
    plt.show()
def evaluate model (Model, lambdas):
    training errors = []
    testing errors = []
    for 1 in lambdas:
        model = Model (alpha=1, max iter=1000)
        model.fit(X train, y train)
        training_predictions = model.predict(X_train)
        training mse = mean squared error(y train, training predictions)
        training errors.append(training mse)
        testing predictions = model.predict(X test)
        testing mse = mean squared error(y test, testing predictions)
        testing errors.append(testing mse)
    return training errors, testing errors
lambdas = np.arange (0, 10, step=0.1)
lasso train, lasso test = evaluate model(Lasso, lambdas)
plot_errors (lambdas, lasso_train, lasso_test, "Lasso")
```



Q16) Use some function for neural networks, like Stochastic Gradient Descent or back propagation - algorithm to predict the value of a variable based on the dataset of problem 14

Code:

```
#Ques16
from numpy import *
import pandas as pd
from sklearn.neural network import MLPClassifier
from sklearn.model selection import train test split
import io
df = pd.read csv('diabetes.csv')
target column = ['Outcome']
predictors = list(set(list(df.columns))-set(target column))
df[predictors] = df[predictors]/df[predictors].max()
X = df[predictors].values
y = df[target_column].values
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.30, random_state=40)
mlp = MLPClassifier (hidden_layer_sizes=(8,8), activation='relu', solver='sgd', max_iter=500)
mlp.fit(X train, y train)
predict train = mlp.predict(X train)
predict test = mlp.predict(X test)
from sklearn.metrics import classification report, confusion matrix
print ('Train Data')
print(confusion matrix(y train, predict train))
print(classification_report(y_train, predict_train))
print('Test Data')
print(confusion_matrix(y_test, predict_test))
print(classification_report(y_test, predict_test))
```

	precision	recall	fl-score	support
O	0.61	1.00	0.76	142
1	0.00	0.00	0.00	89
accuracy			0.61	231
macro avg	0.31	0.50	0.38	231
weighted avg	0.38	0.61	0.47	231