```
NAME: kalpi
R.NO: 24
SUBJECT : DATA ANALYTICS (PRACTICAL)
TERM WORK
 ****
NumPy
 ****
#Q-4 a
np.zeros(10)
 =====
output
____
array([0., 0., 0., 0., 0., 0., 0., 0., 0., 0.])
#Q-4 b
vowels = np.array([\hat{a} \in M_{\hat{a}} \in M_{\hat{
print(vowels)
 _____
output
array([\hat{a}€^{M}a\hat{a}€^{M}, \hat{a}€^{M}e\hat{a}€^{M}, \hat{a}€^{M}i\hat{a}€^{M}, \hat{a}€^{M}u\hat{a}€^{M}, \hat{a}€^{M}u\hat{a}€^{M}], dtype=\hat{a}€^{M}<U1\hat{a}€^{M})
#Q-4 c
np.ones((2,5),dtype=â€<sup>™</sup>intâ€<sup>™</sup>)
 ____
 output
____
array([[1, 1, 1, 1, 1],
[1, 1, 1, 1, 1]
#Q-4 d
list = [[2.7, -2, -19], [0, 3.4, 99.9], [10.6, 0, 13]]
 arr1 = np.array(list)
print(arr1)
 output:-
[[ 2.7 -2. -19. ]
[ 0. 3.4 99.9]
[ 10.6 0. 13. ]]
#Q-4 e
np.arange(4,5*3*4 + 1,4,dtype=\hat{a}€<sup>M</sup>float\hat{a}€<sup>M</sup>).reshape(3,5)
```

```
output
array([[ 4., 8., 12., 16., 20.],
[24., 28., 32., 36., 40.],
[44., 48., 52., 56., 60.]])
#Question 5
#(a)
zeros = np.zeros(10)
print(zeros.ndim)
print(zeros.shape)
print(zeros.size)
print(zeros.dtype)
======
output
=====
1
(10,)
10
float64
#()
vowels = np.array([\hat{a} \in Ma\hat{a} \in M,\hat{a} \in Me\hat{a} \in M,\hat{a} \in Mi\hat{a} \in M,\hat{a} \in Mo\hat{a} \in M,\hat{a} \in Mu\hat{a} \in M)
print(vowels.ndim)
print(vowels.shape)
print(vowels.size)
print(vowels.dtype)
=====
output
======
1
(5,)
5
<U1
ones = np.ones(10)
list = [[2.7, -2, -19], [0, 3.4, 99.9], [10.6, 0, 13]]
myarr1 = np.array(list)
myarr2 = np.arange(9,dtype=â€Mfloatâ€M).reshape(3,3)
\#(b)
print(ones.reshape((10,1)))
======
output
======
[[1.]]
[1.]
```

```
[1.]
[1.]
[1.]
[1.]
[1.]
[1.]
[1.]
[1.]]
#(c)
print(vowels[1:3])
=====
output
['e' 'i']
#(d)
print(myarr1[1:3])
_____
output
____
[[ 0. 3.4 99.9]
[10.6 0. 13. ]]
#(e)
print(myarr1[:,0:2])
======
output
_____
[[ 2.7 -2. ]
[ 0. 3.4]
[10.6 0. ]]
#(f)
print(myarr1[2,0:2])
=====
output
____
[10.6 0.]
#(g)
print(vowels[::-1])
_____
output
_____
```

```
['u' 'o' 'i' 'e' 'a']
#6
#(a)
print(ones / 3)
======
output
=====
#(b)
print(np.concatenate((myarr1, myarr2),axis = 1))
print("\n",np.add(myarr1, myarr2))
======
output
=====
[[ 2.7 -2. -19. 0. 1. 2. ]
[ 0. 3.4 99.9 3. 4. 5. ]
[ 10.6 0. 13. 6. 7. 8. ]]
[[ 2.7 -1. -17. ]
[ 3. 7.4 104.9]
[ 16.6 7. 21. ]]
\#(c)
print(myarr1 - myarr2)
output
======
[[ 2.7 -3. -21. ]
[ -3. -0.6 94.9]
[ 4.6 - 7. 5. ]]
\#(d)
List = []
for i in range(len(myarr1)):
List.append(myarr1[i] * myarr2[i])
print(np.array(List).reshape(3,3))
print("OR\n", myarr1 * myarr2)
_____
output
[[ 0. -2. -38. ]
[ 0. 13.6 499.5]
[ 63.6 0. 104. ]]
#(e)
```

```
myarr3 = myarr1 @ myarr2
print(myarr3)
=====
output
=====
[[-120. -138.3 -156.6]
[ 609.6 712.9 816.2]
[ 78. 101.6 125.2]]
\#(f)
print(myarr1 / myarr2)
output
======
[[ inf -2. -9.5 ]
[ 0. 0.85 19.98 ]
[ 1.76666667 0. 1.625 ]]
\#(g)
print((myarr1 ** 3) / 2)
=====
output
[[ 9.841500e+00 -4.000000e+00 -3.429500e+03]
[ 0.000000e+00 1.965200e+01 4.985015e+05]
[ 5.955080e+02 0.000000e+00 1.098500e+03]]
#(h)
print((np.round(((myarr1**(1/2)) / 2), decimals = 2)))
_____
output
=====
[[0.82 nan nan]
[0. 0.92 5.]
[1.63 0. 1.8 ]]
#7
#(a)
print(ones.transpose())
print()
print(myarr2.transpose())
======
output
[1. 1. 1. 1. 1. 1. 1. 1. 1. 1.]
[[0. 3. 6.]
[1. 4. 7.]
[2. 5. 8.]]
```

```
#(b)
print(vowels[::-1])
output
[â€Muâ€M â€Moâ€M â€Miâ€M â€Meâ€M â€Maâ€M]
#(c)
print(myarr1)
myarr1.sort(axis = 1)
print("\n",myarr1)
=====
output
[[-19. -2. 2.7]
[ 0. 3.4 13. ]
[ 0. 10.6 99.9]]
[[-19. -2. 2.7]
[ 0. 3.4 13. ]
[ 0. 10.6 99.9]]
#8 (a)
myarray2A, myarray2B, myarray2C, myarray2D, myarray2E = np.split(myarray2, 5, axis=1)
print("myarray2A:\n", myarray2A)
print("myarray2B:\n", myarray2B)
print("myarray2C:\n", myarray2C)
print("myarray2D:\n", myarray2D)
print("myarray2E:\n", myarray2E)
=====
output
=====
myarray2A:
[[4.]]
[ 8.]
[12.]]
myarray2B:
[[8.]]
[12.]
[16.]]
myarray2C:
[[12.]]
[16.]
[20.]]
myarray2D:
[[16.]]
[20.]
[24.]]
myarray2E:
```

```
[[20.]
[24.]
[28.]]
#(b)
zerosA, zerosB, zerosC, zerosD = np.split(zeros, [2, 5, 7, 8])
print("zerosA:\n", zerosA)
print("zerosB:\n", zerosB)
print("zerosC:\n", zerosC)
print("zerosD:\n", zerosD)
=====
output
======
zerosA:
[0. \ 0.]
zerosB:
[0. \ 0. \ 0.]
zerosC:
[0. \ 0.]
zerosD:
[0.]
#(c)
concatenated_array = np.concatenate((myarray2A, myarray2B, myarray2C), axis=1)
print("Concatenated Array:\n", concatenated_array)
======
output
____
Concatenated Array:
[[ 4. 8. 12.]
[ 8. 12. 16.]
[12. 16. 20.]]
#9
values = np.arange(-1, -1 + 14 * 3 * 0.25, 0.25)
myarray4 = values.reshape((14, 3))
split_arrays = np.array_split(myarray4, 3, axis=0)
for i, part in enumerate(split_arrays):
print(f"Part \{i + 1\}:")
print(part)
print()
```

```
=====
output
=====
Part 1:
[[-1. -0.75 -0.5]
[-0.25 0. 0.25]
[ 0.5 0.75 1. ]
[ 1.25 1.5 1.75]
[ 2. 2.25 2.5 ]]
Part 2:
[[2.75 3. 3.25]
[3.5 3.75 4.]
[4.25 4.5 4.75]
[5. 5.25 5.5]
[5.75 6. 6.25]]
Part 3:
[[6.5 6.75 7.]
[7.25 7.5 7.75]
[8. 8.25 8.5]
[8.75 9. 9.25]]
#10
# a) Find the sum of all elements
sum_all_elements = np.sum(myarray4)
# b) Find the sum of all elements row-wise
sum row wise = np.sum(myarray4, axis=1)
# c) Find the sum of all elements column-wise
sum_column_wise = np.sum(myarray4, axis=0)
# d) Find the max of all elements
max_all_elements = np.max(myarray4)
# e) Find the min of all elements in each row
min_each_row = np.min(myarray4, axis=1)
# f) Find the mean of all elements in each row
mean_each_row = np.mean(myarray4, axis=1)
# g) Find the standard deviation column-wise
std_column_wise = np.std(myarray4, axis=0)
# Print results
print("a) Sum of all elements:", sum_all_elements)
print("b) Sum of all elements row-wise:", sum_row_wise)
print("c) Sum of all elements column-wise:", sum_column_wise)
print("d) Max of all elements:", max_all_elements)
print("e) Min of all elements in each row:", min each row)
print("f) Mean of all elements in each row:", mean_each_row)
print("g) Standard deviation column-wise:", std_column_wise)
```

```
output
 =====
a) Sum of all elements: 173.25
b) Sum of all elements row-wise: [-2.25 0. 2.25 4.5 6.75 9. 11.25 13.5 15.75 18. 20.25 22.5
24.75 27. ]
c) Sum of all elements column-wise: [54.25 57.75 61.25]
d) Max of all elements: 9.25
e) Min of all elements in each row: [-1. -0.25 0.5 1.25 2. 2.75 3.5 4.25 5. 5.75 6.5 7.25
8. 8.75]
f) Mean of all elements in each row: [-0.75 0. 0.75 1.5 2.25 3. 3.75 4.5 5.25 6. 6.75 7.5
8.25 9. ]
g) Standard deviation column-wise: [3.02334666 3.02334666 3.02334666]
Pandas - 1
#Question 5
#(a)
import string
inde = list(range(1,27))
EngAlpha = pd.Series(list(string.ascii_uppercase))
#(b)
vowels = pd.Series([0,0,0,0,0],index=[\hat{a} \in Ma\hat{a} \in M,\hat{a} \in Ma\hat{a} \in M,\hat{a} \in Ma\hat{a} \in Ma\hat{a}
print(vowels.empty)
\#(c)
friends = pd.Series({â€Mkalpiâ€M:1,â€Mvedanshiâ€M:2,â€Myashviâ€M:3,â€Mdhruviâ€M:4})
print(friends)
\#(d)
MTSeries = pd.Series()
print(MTSeries.empty)
#(e)
MonthDays = [31,29,31,30,31,30,31,30,31,30,31]
ind = list(range(1,13))
MonthDaysS = pd.Series(MonthDays, index = ind)#index = np.arange(1,13)
```

```
print(MonthDaysS)
 =====
 output
  =====
 False
kalpi 1
 vedanshi 2
 yashvi 3
 dhruvi 4
 dtype: int64
 True
  131
  2 29
 3 31
 4 30
 5 31
 6 30
 7 31
 8 31
 9 30
  1031
  11 30
  12 31
 dtype: int64
#Question 6
#(a)
# vowels[:] = 10
 vowels.loc['a':'u'] = 10
 print(vowels)
 #(b)
 vowels = vowels / 2
print(vowels)
#(C)
 vowels1 = pd.Series([2,5,6,3,8],index = [\hat{a} \in Ma\hat{a} \in M,\hat{a} \in Ma\hat{a} \in M,\hat{a} \in Ma\hat{a} \in Ma\hat
print(vowels1)
\#(d)
 vowels2 = vowels + vowels1
print(vowels2)
#(e)
print("\nSubstract \n",vowels - vowels1)
```

```
print("\nMultiply \n",vowels * vowels1)
print("\nDivide \n",vowels / vowels1)
#(f)
vowels1 = vowels1.rename({'a':'A','e':'E','i':'I','o':'Oâ
€™,'u':'U'})
print(vowels1)
print(vowels)
=====
output
=====
a 10
e 10
i 10
o 10
u 10
dtype: int64
a 5.0
e 5.0
i 5.0
o 5.0
u 5.0
dtype: float64
a 2
e 5
i 6
o 3
u 8
dtype: int64
a 7.0
e 10.0
i 11.0
o 8.0
u 13.0
dtype: float64
Substract
a 3.0
e 0.0
i -1.0
o 2.0
u -3.0
dtype: float64
Multiply
a 10.0
e 25.0
i 30.0
o 15.0
u 40.0
dtype: float64
```

```
Divide
a 2.500000
e 1.000000
i 0.833333
o 1.666667
u 0.625000
dtype: float64
A 2
E 5
I 6
O 3
U 8
dtype: int64
a 5.0
e 5.0
i 5.0
o 5.0
u 5.0
dtype: float64
#Question 7
#(a)
print("\nDimension = ",EngAlpha.ndim,"\tSize = ",EngAlpha.size,"\tValues = ",EngAlpha.values)
print("\nDimension = ",vowels.ndim,"\tSize = ",vowels.size,"\tValues = ",vowels.values)
print("\nDimension = ",friends.ndim,"\tSize = ",friends.size,"\tValues = ",friends.values)
print("\nDimension = ",MonthDaysS.ndim,"\tSize = ",MonthDaysS.size,"\tValues = ",MonthDaysS.values)
print("\nDimension = ",MTSeries.ndim,"\tSize = ",MTSeries.size,"\tValues = ",MTSeries.values)
#(b)
MTSeries.name = "SeriesEmpty"
print(MTSeries.name)
#(c)
MonthDaysS.index.name = 'monthno'
friends.index.name = 'fname'
print(MonthDaysS.index.name)
print(friends)
\#(d)
print(friends[[3,2]])
#(e)
print(EngAlpha[4:16])
#(f)
```

```
print(EngAlpha.head(10))
\#(g)
print(EngAlpha.tail(10))
#(h)
print(MTSeries)
output
Dimension = 1 Size = 26 Values = ['A' 'B' 'C' 'D' 'E' 'F' 'Gâ
 \in ^{\text{M}} \hat{a} \in ^{\text{M}} \text{H} \hat{a} \in ^{\text{M}} \hat{a
 €™Q' 'R'
â€MSâ€M â€MTâ€M â€MUâ€M â€MVâ€M â€MWâ€M â€MXâ€M â€MYâ€M â€MZâ€M]
Dimension = 1 Size = 5 Values = [5. 5. 5. 5. 5.]
Dimension = 1 Size = 5 Values = [ 8 14 27 35 36]
Dimension = 1 Size = 12 Values = [31 29 31 30 31 30 31 30 31 30 31]
Dimension = 1 Size = 0 Values = []
SeriesEmpty
monthno
fname
kalpi 1
vedandhi 2
yashvi 3
dhruvi 4
dtype: int64
fname
abc 5
def 7
dtype: int64
4 E
5 F
6 G
7 H
8 I
9 J
10 K
11 L
12 M
13 N
14 O
15 P
dtype: object
0 A
1 B
2 C
```

3 D

```
5 F
6 G
7 H
8 I
9 J
dtype: object
16 Q
17 R
18 S
19 T
20 U
21 V
22 W
23 X
24 Y
25 Z
dtype: object
Series([], Name: SeriesEmpty, dtype: object)
#Question 8
#(a)
print(MonthDaysS[2:7])
#(b)
print(MonthDaysS[::-1])
data =
\{\hat{a} \in \mathbb{M}2014 \hat{a} \in \mathbb{M}: [100.5, 150.8, 200.9, 30000, 40000], \hat{a} \in \mathbb{M}2013 \hat{a} \in \mathbb{M}: [12000, 18000, 22000, 30000, 45000], \hat{a} \in \mathbb{M}2016 \hat{a} \in \mathbb{M}2014 \hat{a} \in \mathbb{M}2016 \hat{a} \in 
€™:[20000,50000,
70000,100000,125000], \hat{a} \in M2017 \hat{a} \in M:[50000,60000,70000,80000,90000]
df5 = pd.DataFrame(data)
df5.index = [â€MMadhuâ€M, â€MKusumâ€M, â€MKinshukâ€M, â€MAnkitâ€M, â€MShrutiâ€M]
print(df5)
output:-
monthno
3 31
4 30
5 31
6 30
7 31
dtype: int64
monthno
12 31
11 30
10 31
9 30
8 31
```

4 E

```
7 31
6 30
5 31
4 30
3 31
2 29
1 31
dtype: int64
2014 2013 2016 2017
Madhu 100.5 12000 20000 50000
Kusum 150.8 18000 50000 60000
Kinshuk 200.9 22000 70000 70000
Ankit 30000.0 30000 100000 80000
Shruti 40000.0 45000 125000 90000
#9
dinc=\{"2014":[100.5,150.8,200.9,30000,40000],
"2015":[12000,18000,22000,30000,45000],
"2016":[20000,50000,70000,100000,125000],
"2017":[50000,60000,70000,80000,90000],
df=pd.DataFrame(dinc)
df.index=["Madhu","Kusum","Kinshuk","Ankit","Shruti"]
df.index.name = â€<sup>TM</sup>Sales Personsâ€<sup>TM</sup>
df.columns.name = 'Years'
print(df)
output:-
Years 2014 2015 2016 2017
Sales Persons
Madhu 100.5 12000 20000 50000
Kusum 150.8 18000 50000 60000
Kinshuk 200.9 22000 70000 70000
Ankit 30000.0 30000 100000 80000
Shruti 40000.0 45000 125000 90000
#10
#a
print(df.index)
print("\n",df.columns)
#c
print(df.columns.dtype)
#d
print("Dimension : ",df.ndim,"\tShape : ",df.shape,"\tSize : ",df.size , "\tValues : ",df.values)
#e
print(df.tail(2))
```

```
#f
print(df.loc[:,"2014":"2015"])
#g
dict1={"2018":pd.Series([160000,110000,500000,340000,900000],index=["Madhu","Kusum","Kinshuk","A
nkit", "Shruti"])}
df2=pd.DataFrame(dict1)
print(df2)
#h
print(df2.empty)
_____
output
_____
Index([â€MMadhuâ€M, â€MKusumâ€M, â€MKinshukâ€M, â€MAnkitâ€M, â€MShrutiâ€M], dtype=â€Mobjectâ
€™, name='Sales Persons')
Index([\hat{a} \in \mathbb{Z}^{M} 2014 \hat{a} \in \mathbb{Z}^{M}, \hat{a} \in \mathbb{Z}^{M} 2015 \hat{a} \in \mathbb{Z}^{M}, \hat{a} \in \mathbb{Z}^{M} 2016 \hat{a} \in \mathbb{Z}^{M}, \hat{a} \in \mathbb{Z}^{M} 2017 \hat{a} \in \mathbb{Z}^{M}], dtype=\hat{a} \in \mathbb{Z}^{M} object\hat{a} \in \mathbb{Z}^{M}, name=\hat{a} \in \mathbb{Z}^{M} vears\hat{a} \in \mathbb{Z}^{M}
object
Dimension: 2 Shape: (5, 4) Size: 20 Values: [[1.005e+02 1.200e+04 2.000e+04 5.000e+04]]
[1.508e+02 1.800e+04 5.000e+04 6.000e+04]
[2.009e+02 2.200e+04 7.000e+04 7.000e+04]
[3.000e+04 3.000e+04 1.000e+05 8.000e+04]
[4.000e+04 4.500e+04 1.250e+05 9.000e+04]]
Years 2014 2015 2016 2017
Sales Persons
Ankit 30000.0 30000 100000 80000
Shruti 40000.0 45000 125000 90000
Years 2014 2015
Sales Persons
Madhu 100.5 12000
Kusum 150.8 18000
Kinshuk 200.9 22000
Ankit 30000.0 30000
Shruti 40000.0 45000
2018
Madhu 160000
Kusum 110000
Kinshuk 500000
Ankit 340000
Shruti 900000
False
#11
# a
df3=pd.concat([df,df2],axis=1)
print(df3)
print("\n",df.transpose())
```

```
# c
print(df["2017"])
# d
print(df3.loc[["Madhu","Ankit"],"2017":"2018"])
# e
print(df3.loc[["Shruti"],"2016"])
#f
df3.loc[\hat{a} \in MSumit \hat{a} \in MSumit \hat{a} = [196.2,37800,52000,78438,38852]
print(df3)
#g
df3=df3.drop(columns="2014")
print(df3)
#h
df3=df3.drop("Kinshuk")
df3=df3.rename({"Ankit":"Vivaan","Madhu":"Shailesh"})
print(df3)
df3.loc["Shailesh","2018"]=100000
print(df3)
# k
df3.to_csv("data.csv",index=False,columns=None)
#1
df4 = pd.read csv("data.csv")
df4.index = ["Shailesh", "Kusum", "Kinshuk", "Vivaan", "Shruti"]
print(df4)
_____
output
____
2014 2015 2016 2017 2018
Madhu 100.5 12000 20000 50000 160000
Kusum 150.8 18000 50000 60000 110000
Kinshuk 200.9 22000 70000 70000 500000
Ankit 30000.0 30000 100000 80000 340000
Shruti 40000.0 45000 125000 90000 900000
Sales Persons Madhu Kusum Kinshuk Ankit Shruti
Years
2014 100.5 150.8 200.9 30000.0 40000.0
```

2015 12000.0 18000.0 22000.0 30000.0 45000.0 2016 20000.0 50000.0 70000.0 100000.0 125000.0 2017 50000.0 60000.0 70000.0 80000.0 90000.0 Sales Persons Madhu 50000 Kusum 60000 Kinshuk 70000 Ankit 80000 Shruti 90000 Name: 2017, dtype: int64 2017 2018 Madhu 50000 160000 Ankit 80000 340000 Shruti 125000 Name: 2016, dtype: int64 2014 2015 2016 2017 2018 Madhu 100.5 12000.0 20000.0 50000.0 160000.0 Kusum 150.8 18000.0 50000.0 60000.0 110000.0 Kinshuk 200.9 22000.0 70000.0 70000.0 500000.0 Ankit 30000.0 30000.0 100000.0 80000.0 340000.0 Shruti 40000.0 45000.0 125000.0 90000.0 900000.0 Sumit 196.2 37800.0 52000.0 78438.0 38852.0 2015 2016 2017 2018 Madhu 12000.0 20000.0 50000.0 160000.0 Kusum 18000.0 50000.0 60000.0 110000.0 Kinshuk 22000.0 70000.0 70000.0 500000.0 Ankit 30000.0 100000.0 80000.0 340000.0 Shruti 45000.0 125000.0 90000.0 900000.0 Sumit 37800.0 52000.0 78438.0 38852.0 2015 2016 2017 2018 Shailesh 12000.0 20000.0 50000.0 160000.0 Kusum 18000.0 50000.0 60000.0 110000.0 Vivaan 30000.0 100000.0 80000.0 340000.0 Shruti 45000.0 125000.0 90000.0 900000.0 Sumit 37800.0 52000.0 78438.0 38852.0 2015 2016 2017 2018 Shailesh 12000.0 20000.0 50000.0 100000.0 Kusum 18000.0 50000.0 60000.0 110000.0 Vivaan 30000.0 100000.0 80000.0 340000.0 Shruti 45000.0 125000.0 90000.0 900000.0 Sumit 37800.0 52000.0 78438.0 38852.0 2015 2016 2017 2018

Kusum 18000.0 50000.0 60000.0 110000.0 Kinshuk 30000.0 100000.0 80000.0 340000.0 Vivaan 45000.0 125000.0 90000.0 900000.0 Shruti 37800.0 52000.0 78438.0 38852.0

Shailesh 12000.0 20000.0 50000.0 100000.0

Shruti 37800.0 52000.0 78438.0 38852.0

Pandas 2

```
#13
#(a)
Product = pd.DataFrame({â€MItemâ€M:[â€MTVâ€M,â€MTVâ€M,â€MTVâ€M,â€MACâ€M], â€MCompanyâ
€M:[â€MLGâ€M,â€MVIDEOCONâ€M,â€MLGâ€M,â€MSONYâ€M],
\hat{a}€™Rupees\hat{a}€™:[12000, 10000, 15000, 14000], \hat{a}€™USD\hat{a}€™:[700, 650, 800, 750]})
print(Product)
\#(b)
Product.loc[5] = [â€MACâ€M,â€MSONYâ€M, 32000, 2200]
print(Product)
\#(c)
print(Product[Product['Company'] == 'LG'].max())
\#(d)
print(Product.groupby("Item").sum())
#(e)
ProductTemp = Product[Product[â€MCompanyâ€M] == â€MSONYâ€M]
print(ProductTemp[â€MUSDâ€M].median())
#(f)
Product_sorted_by_ruppee = Product.sort_values(â€MRupeesâ€M, ascending = True, axis = 0)
print(Product_sorted_by_ruppee)
\#(g)
engine=sqlalchemy.create engine(â€Mmysql+pymysql://root:@localhost:3306/practiceâ€M)
Product.to_sql(â€Mproductâ€M, if_exists = â€Mappendâ€M, con = engine, index = False)
output
======
Item Company Rupees USD
0 TV LG 12000 700
1 TV VIDEOCON 10000 650
2 TV LG 15000 800
3 AC SONY 14000 750
Item Company Rupees USD
0 TV LG 12000 700
1 TV VIDEOCON 10000 650
2 TV LG 15000 800
3 AC SONY 14000 750
5 AC SONY 32000 2200
Company LG
Rupees 15000
USD 800
dtype: object
Company Rupees USD
Item
AC SONYSONY 46000 2950
```

TV LGVIDEOCONLG 37000 2150

1475.0

```
Item Company Rupees USD
1 TV VIDEOCON 10000 650
0 TV LG 12000 700
3 AC SONY 14000 750
2 TV LG 15000 800
5 AC SONY 32000 2200
import pandas as pd
import numpy as np
import pymysql as py
import sqlalchemy
#14
#(a)
Data = pd.DataFrame({â€MNameâ€M:[â€MAparnaâ€M,â€MPankajâ€M,â€MRamâ€M,â€MRameshâ€M, â
€<sup>™</sup>Naveenâ€<sup>™</sup>, â€<sup>™</sup>Krrishanavâ€<sup>™</sup>, â€<sup>™</sup>Bhawnaâ€<sup>™</sup>],
'Degree':['MBA','BCA','M.Tech','MBA',None,'BCA','MBAâ
€M].
\hat{a}€ Score \hat{a}€ (90, None, 80, 98, 97, 78, 89)
print(Data)
#(b)
print(Data.groupby(â€MDegreeâ€M).max())
Data['Score'].fillna(76, inplace = True)
print(Data)
\#(d)
Data.set index = â€<sup>TM</sup>Nameâ€<sup>TM</sup>
print(Data)
#(e)
print(Data.loc[:,[\hat{a} \in MDegree \hat{a} \in M,\hat{a} \in MScore \hat{a} \in M]].groupby(by = \hat{a} \in MDegree \hat{a} \in M).mean())
print(Data.groupby([â€MNameâ€M,â€MDegreeâ€M])[â€MScoreâ€M].mean())
#(f)
print("Total MBA Students :",Data[Data[â€MDegreeâ€M] == â€MMBAâ€M].shape[0])
\#(g)
d = (Data[Data[â€MDegreeâ€M] == â€MBCAâ€M][â€MScoreâ€M])
print("Mode : ",d.mode()[0])
output
_____
Name Degree Score
0 Aparna MBA 90.0
1 Pankaj BCA NaN
2 Ram M.Tech 80.0
3 Ramesh MBA 98.0
4 Naveen None 97.0
```

5 Krrishanav BCA 78.0 6 Bhawna MBA 89.0 Name Score

Degree

BCA Pankaj 78.0

M.Tech Ram 80.0

MBA Ramesh 98.0

Name Degree Score

O Anorma MD A OO (

0 Aparna MBA 90.0

1 Pankaj BCA 76.0

2 Ram M.Tech 80.0

3 Ramesh MBA 98.0

4 Naveen None 97.0

5 Krrishanav BCA 78.0

6 Bhawna MBA 89.0

Name Degree Score

0 Aparna MBA 90.0

1 Pankaj BCA 76.0

2 Ram M.Tech 80.0

3 Ramesh MBA 98.0

4 Naveen None 97.0

5 Krrishanav BCA 78.0

6 Bhawna MBA 89.0

Score

Degree

BCA 77.000000

M.Tech 80.000000

MBA 92.333333

Name Degree

Aparna MBA 90.0

Bhawna MBA 89.0

Krrishanav BCA 78.0

Pankaj BCA 76.0

Ram M.Tech 80.0

Ramesh MBA 98.0

Name: Score, dtype: float64 Total MBA Students : 3

Mode: 76.0

Matplotlib

import matplotlib.pyplot as plt import numpy as np

categories = $[\hat{a} \in MA\hat{a} \in MA\hat{a}$

```
y = np.random.randint(1, 10, 10)
#Bar Chart
plt.figure(figsize = (10, 8))
plt.subplot(3, 3, 1)
plt.bar(categories, values)
plt.title('Bar Chart')
#Scatter Plot
plt.subplot(3, 3, 2)
plt.scatter(x, y)
plt.title("Scatter Chart")
#Line Chart
plt.subplot(3, 3, 3)
plt.plot(x, y)
plt.title("Line Chart")
#Area Chart
plt.subplot(3, 3, 4)
plt.fill_between(x, y, color = "skyblue", alpha=0.5)
plt.title("Area Chart")
#Pie Chart
plt.subplot(3, 3, 5)
plt.pie(values, labels=categories, autopct='%1.1f%%')
plt.title("Pie Chart")
#Histogram
plt.subplot(3, 3, 6)
plt.hist(y, bins=5)
plt.title('Histogram')
#Cumulative Distribution (Ogive)
plt.subplot(3, 3, 7)
plt.hist(y, bins=5, cumulative=True, histtype=â€Mstepâ€M, color=â€Mredâ€M)
plt.title('Ogive')
#Dot Plot
plt.subplot(3, 3, 8)
plt.plot(x, y, â€<sup>TM</sup>gpâ€<sup>TM</sup>, markersize=10)
plt.title('Dot Plot')
# Stem-and-Leaf Display
plt.subplot(3, 3, 9)
plt.stem(x, y,)
plt.title('Stem-and-Leaf')
plt.tight_layout()
plt.show()
```

Linear Regression import numpy as np import pandas as pd import matplotlib.pyplot as plt import seaborn as sns sns.set() import time def linear(X, b0, b1): return [b0+b1*x for x in X] #b0 - Intercept def intercept(X, Y, b1): $x_{-} = np.mean(X)$ $y_{-} = np.mean(Y)$ return y_ - b1 * x_ #b1 - slope def slope(X, Y): $x_{-} = np.mean(X)$ $y_{-} = np.mean(Y)$ rise = $sum([(x-x_) * (y-y_) for x, y in zip(X,Y)])$ $run = sum([(x-x_)**2 \text{ for } x \text{ in } X])$ return rise / run data = pd.read_csv(â€TMAutomobile_data.csvâ€TM) data.head() data = data.loc[data[â€Mpriceâ€M] != â€M?â€M] data[[â€Mpriceâ€M]] = data[[â€Mpriceâ€M]].astype(int) for i in data.columns: print(i) data = data.loc[data[i] != â€TM?â€TM] final_data = data[[â€Mengine-sizeâ€M, â€Mpriceâ€M]] predictor = data['engine-size'] target = data[â€[™]priceâ€[™]] plt.figure(figsize=(8,5)) plt.title("Price vs engine-size") plt.scatter(predictor, target, color = "#247ba0") plt.xlabel('engine-size') plt.ylabel(â€TMpriceâ€TM) b1 = slope(predictor, target)

```
b0 = intercept(predictor, target, b1)
predicted = linear(predictor, b0, b1)
print(predicted)
plt.figure(figsize = (8, 5))
plt.plot(predictor, predicted, color = '#f25f5c')
plt.scatter(predictor, predicted, color = '#f25f5c')
plt.title(â€MPredicted values bt Linear Regressionâ€M, fontsize = 15)
plt.xlabel(â€Mengine-sizeâ€M)
plt.ylabel('price')
plt.scatter(predictor, target, color = '#247ba0')
plt.show()
Pre-processing and Laptop-price prediction
import pandas as pd
import numpy as np
from sklearn, preprocessing import MinMaxScaler, StandardScaler, OneHotEncoder
from sklearn.preprocessing import LabelEncoder
from scipy import stats
df = pd.read_csv('laptop_price---dataset.csv')
print(df.head())
print(df.info())
print(df.isnull().sum())
print(df.drop_duplicates())
df[â€MCompany_Labelâ€M] = LabelEncoder().fit_transform(df[[â€MCompanyâ€M]])
df.head()
df[â€MCompanyâ€M] = df[â€MCompanyâ€M].str.lower()
z = \text{np.abs}(\text{stats.zscore}(\text{df}[\hat{a} \in \text{PMPrice}(\text{Euro})\hat{a} \in \text{PM}]))
print(z)
threshold_z = 2
outlier_indices = np.where(z > threshold_z)[0]
no outliers = df.drop(outlier indices)
print("Original DataFrame Shape:", df.shape)
print("DataFrame Shape after Removing Outliers:", no_outliers.shape)
import seaborn as sns
import matplotlib.pyplot as plt
```

```
corr = no_outliers.corr(numeric_only = True)
sns.heatmap(corr, annot = True)
df = no outliers
scaler = MinMaxScaler()
df[â€MPrice (Euro)_MinMaxScaledâ€M] = scaler.fit_transform(np.array(df[â€MPrice (Euro)â€M]).reshape(-1,1))
print(df[â€MPrice (Euro) MinMaxScaledâ€M].head())
scaler = StandardScaler()#Data is in range of
df[â€MPrice (Euro)_StandardScalerâ€M] = scaler.fit_transform(np.array(df[â€MPrice (Euro)â€M]).reshape(-1,1))
print(df[â€MPrice(Euro) StandardScalerâ€M].head())
df[â€MPrice (Euro) StandardScalerâ€M].min()
_____
Prediction
_____
from sklearn.model_selection import train_test_split
X = df[\hat{a} \in \mathbb{M}RAM (GB)\hat{a} \in \mathbb{M}]
Y = df['Price (Euro)']X_train, X_test, Y_train, Y_test = train_test_split(X, Y, test_size=0.33,
random state=42)
print(X train, X test, Y train, Y test)
from sklearn.linear model import LinearRegression
lr = LinearRegression()
lr.fit(np.array(X train).reshape(-1,1), Y train)
lr.intercept_
lr.coef
from sklearn.metrics import mean_squared_error, r2_score
y pred = lr.predict(np.array(X test).reshape(-1,1))
print(â€MMeanSquaredErrorâ€M, np.sqrt(mean squared error(Y test,y pred)))
print('R2 Score', r2 score(Y test, y pred))
USA Housing
import matplotlib.pyplot as plt
import pandas as pd
import numpy as np
import seaborn as sns
dfHouse = pd.read csv("./USA Housing.csv")
dfHouse.head()
```

```
dfHouse.info()
dfHouse.drop_duplicates(inplace=True)
dfHouse.isnull().sum()
import pandas as pd
import matplotlib.pyplot as plt
def PrintChart(dfHouse):
float cols = dfHouse.select dtypes(include=[â€Mfloat64â€M]).columns
fig, axes = plt.subplots(len(float_cols), 1, figsize=(10, 5 * (len(float_cols))))
# Box plots for each float64 column
for idx, col in enumerate(float cols):
axes[idx].boxplot(dfHouse[col])
axes[idx].set title(f"Box Plot of {col}")
axes[idx].set_ylabel(f"Range of {col}")
axes[idx].set xlabel(col)
# Adjust layout
plt.tight layout()
plt.show()
PrintChart(dfHouse)
# Convert it into ranges
chartDfHouse = dfHouse
for i in chartDfHouse.columns:
if chartDfHouse[i].dtype == â€<sup>M</sup>float64â€<sup>M</sup>:
q1 = np.percentile(chartDfHouse[i], 25)
q3 = np.percentile(chartDfHouse[i], 75)
iqr = q3 - q1
upperBound = q3 + 1.5 * iqr
lowerBound = q1 - 1.5 * iqr
chartDfHouse = chartDfHouse[(chartDfHouse[i] > lowerBound) & (chartDfHouse[i] < upperBound)]
PrintChart(chartDfHouse)
numericChartDfHouse = chartDfHouse.select dtypes(include='number').corr()
# Create a heatmap
plt.figure(figsize=(10, 8))
sns.heatmap(numericChartDfHouse, cmap='coolwarm', linewidths=.5)
plt.title('Sample Heatmap')
```

plt.show()