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Subject: EOAIML

Batch: 12

```
import numpy as np

# Create a 1D NumPy array with 20 elements (e.g.
, 1 to 20) array = np.arange(1, 21)

print("Original Array: ")
print(array)

# Extract every second element using
# slicing every_second_element array[ :: 2]
print("Every second element:
") print
(every_second_element)
```

Original Array:  
[ 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20]  
Every second element:  
[ 1 3 5 7 9 11 13 15 17 19]

```
import pandas as pd

# Read the Excel file df = pd. /
content/ Book1.xlsx")

# Get the number of columns
and rows num columns =
df.shape[1] num rows =
df.shape[0]

print("Number of columns: "
num_columns) print("Number of
rows: " , num_rows )
```

```
# Convert each row into a
list of lists data as list of lists
= df.values.tolist()
```

```
! pip      scikit-satisficed:
install    learn   satisfied: numpy>=1.19.5 scikit-learn      in      /usr/10ca1/1ib/python3.11/dist-
Requirement already satisfied: scipy>=1.6.0 packages (1.6.1) in /usr/10ca1/1ib/python3.11/dist-
Requirement already satisfied: joblib>=1.2.0 packages (from scikit-learn) (2.0.2) in
Requirement already satisfied:/usr/10ca1/1ib/python3.11/dist-packages (from scikit-learn)
Requirement already satisfied:(1.16.1) in /usr/10ca1/1ib/python3.11/dist-packages (from scikit-
```

```
Requirementalready          learn)      (1.5.1)      threadp001ct1>=3.1.e      in
Requirementalready          /usr/10ca1/lib/python3.11/dist-packages (from scikit-learn) (3.6.0)
```

```
print("\nData as a list of
lists:           print
(data_as_list_of_lists)

Number of columns: 3
Number of rows: 10

Data as a list of lists:
[[167, 51, 'Underweight'], [182, 62, 'Normal'], [176, 69, 'Normal'], [173, 64, 'Normal'],
 [172, 65, 'Normal'], [174, 56, 'Underweight'],

import math

def cosine_similarity(v1, v2):
    dot = sum(a*b for a, b in zip(v1, v2))
    mag1 = math.sqrt(sum(a*a for a in v1))
    mag2 = math.sqrt(sum(b*b for b in v2))
    return dot / (mag1 * mag2) if mag1 and
    mag2 else 0

# Example

print("Cosine Similarity: ", cosine_similarity(VI, v2))

Cosine Similarity: 0.9746318461970762

import math

def euclidian_distance(p1, p2):
    return ((p2[0] - p1[0]) ** 2 +
            (p2[1] - p1[1]) ** 2) ** 0.5

point1 =
point2

print("Distance: " + str(euclidian_distance(point1, point2)))

Distance: 5.0

import math
from
collections import Counter

# Dataset (Height, Weight,
Class) data =
[[167, 51, "Underweight"],
 [182, 62, "Normal"],
 [176, 69, "Normal"],
 [173, 64, "Normal"],
 [172, 65, "Normal"],
 [174, 56, "Underweight"],
 [169, 58, "Normal"],
 [173, 57, "Normal"]]
```

```

[170,55,"Normal " ]
[170, 57, "?"] # Test point
train_data -- data[:-1]
test_point = data[-1][:2]

# Euclidean distance def
euclidean_distance(p1,
p2) :
    return math.sqrt((p1[0] - p2[0])**2 + (p1[1] - p2[1])**2)

# KNN assumption def
knn_predict(train, test,
k):
    distances = [(euclidean_distance(test, row[ : row
in train]) distances.sort(key=lambda x: x[0])) nearest =
distances[:k] prediction = Counter([row[2] for , row in
nearest]).most_common(1) [0] [0] return prediction,
nearest

```

# Predict with k3

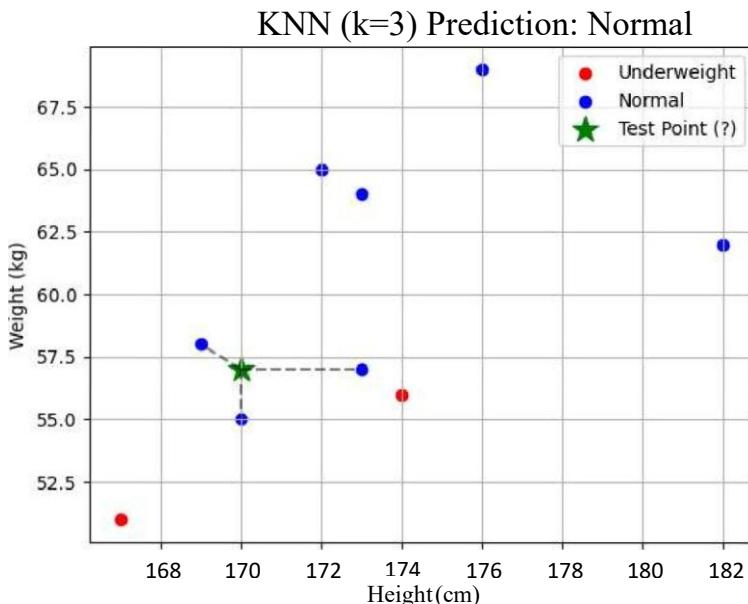
```

predicted_class, nearest_neighbors = knn_predict(train_data, test_point, k)

# Plot points colors -- {"Normal": "blue" ,
"Underweight . "red", "?" green "}
for height, weight, cls in train_data:
    plt.scatter(height, weight, color=colors[cls], label=cls if cls not in
                plt.gca().get_legend_handles_labels()[1] else "")

# Plot the test point
PI .scatter(test_point[0], test_point [1], color="green", marker="*", s=200, label="Test Point
(?)") # Connect nearest neighbors for neighbor in nearest_neighbors:
plt.plot( test_point[0], neighbor[0]], [test_point [1], neighbor[1]], 'k', alpha=e.5)
# Labels and title plt.xlabel("Height (cm)")
plt.ylabel( "Weight (kg)" ) plt.title(f'KNN
(k={k}) Prediction: {predicted_class}' )
PIt. legend() pit. grid (True) PI. show()

```



```
!pip install scikit-learn
```

```

Requirement already satisfied: scikit-learn in /usr/local/lib/python3.11/dist-packages (1.6.1)
Requirement already satisfied: numpy>=1.19.5 in /usr/local/lib/python3.11/dist-packages (from scikit-learn) (2.0.2)
Requirement already satisfied: scipy>=1.6.0 in /usr/local/lib/python3.11/dist-packages (from scikit-learn) (1.16.1)
Requirement already satisfied: joblib>=1.2.0 in /usr/local/lib/python3.11/dist-packages (from scikit-learn) (1.5.1)

```

Requirement already satisfied: threadp001ct1>=3.1.e in /usr/10ca1/1ib/python3.11/dist-packages (from scikit-learn) (3.6.0)

```
open( "/content/ Book1.xlsx" ) import pandas as pd

df=pd . read_excel( "/content/ Book1 . xlsx' )

import pandas as pd

# Load the data from the Excel file, skipping the
first row df = pd. /content/ Book1.xlsx" ,
skiprows=1)

# Manually assign the correct column
names df.columns = [ 'Height(cm)' ,
'Weight(kg)' 'Cass' ]

# Separate the features (x) and the target (y)
['Weight(kg)', 'Height(cm)' ]]

Y = df[ 'Cass' ]

# Map 'Normal' to 1 and 'Underweight' to 0 in the target
variable (y) 'Underweight : 0})
```

```
print("Features (X) : 'I ")
display(x.head())
```

```
print( "\nTarget (Y) : "
) display(y . head())
```

FileNotFoundError Traceback (most recent call last) EmpAuython-input-1817842996.py. in <cell line:

2

3 # Load the data from the Excel file, skipping the first row
= pd. , skiprows=1)

5

6 # Manually assign the correct column names

3 frames in get\_handle(path\_or\_buf, mode, encoding, compression,
memory\_map, is\_text, errors, storage\_options) 880 else:

881 # Binary mode

882 handle = open(handle, ioargs . mode)

883 handles . append (handle)

884

FileNotFoundException: [Errno 2] No such file or directory: '/content/Book1. xlsx'

Next steps: Explain error

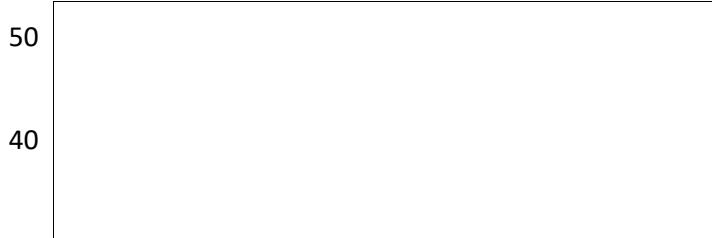
```
import numpy as np
import matplotlib.pyplot
as plt from sklearn.cluster
import KMeans data =
np.array([
[0, 1], [1, 3], [6, 22], [9, 11], [10, 10],
[0, 40], [41], [1, 42], [3, 45], [4, 46], [5, 47], [6, 48], [7, 50], [8, 44], [9, 49],
[35, 35], [36, 36], [37, 40], [39, 42], [40, 41], [45, 50], [50, 50], [50, 45]
```

```

3 kmeans = KMeans(n_clusters=K, .fit(data) labels =
kmeans.labels centroids = kmeans.cluster centers plt.scatter(data[ :,0],
data[ :,1], c=labels, cmap= 'rainbow', s=50) centroids[ :,1]
c='black', marker= 's-200, label= 'Centroids
PIt xlabel ( "dim1" ) plt . y
label ( "dim2" ) plt.title(f'K-
Means Clustering (K={K})')
plt . legend () plt . show()

```

K-Means Clustering (K=3)



```

import pandas as pd import
matplotlib.pyplot as plt from
sklearn.cluster import KMeans df
= pd.      .xlsx")
'dim2' ]]

kmeans = KMeans(n_clusters=K, random_state=e, n _ 1 • nit• = 'auto' )
kmeans . fit (X) cluster_assignments kmeans . labels cluster centers =
kmeans. cluster centers plt.scatter(X[ 'dim1' ], X[ 'dim2' ],
c=cluster_assignments, s=50, cmap= 'viridis' ) plt . : , 0] , cluster_centers [
: , 1], c='red', s=200, marker= 'x' , label= 'Centroids plt.title(f"K-Means
Clustering (k={K})") plt.xlabel("Dimension 1") plt.ylabel("Dimension
2") plt . legend() plt.grid(True) plt . show()

```

FileNotFoundException Traceback (most recent call last)

Limp.Lipython-input-2207411179.p\_y. in <cell line:

```

2 import matplotlib.pyplot as plt
3 from sklearn.cluster import KMeans
----> 4 df = pd . read _excel ( "cluster _data . xlsx" )
      'dim2' ]

```

6K = 3

3 frames

/usr/10cal/1ib/python3.12/dist-packagesLpandas/io/common.py\_ in get\_handle(path\_or\_buf, mode,
encoding, compression, memory\_map, is\_text, errors, storage\_options)

```

880
881      # Binary mode
882      handle    open (handle, ioargs . mode)
883      handles . append (handle)
884

```

FileNotFoundException: [Errno 2] No such file or directory: 'cluster data.xlsx'

Next steps: Explain error

```

import numpy as np
import matplotlib.pyplot
as PIt
X = np. 1200, 1500, 1800, 2000, 2500, 3000] ) Y =
np. 25, 28, 39, 35, 40, 45]

```

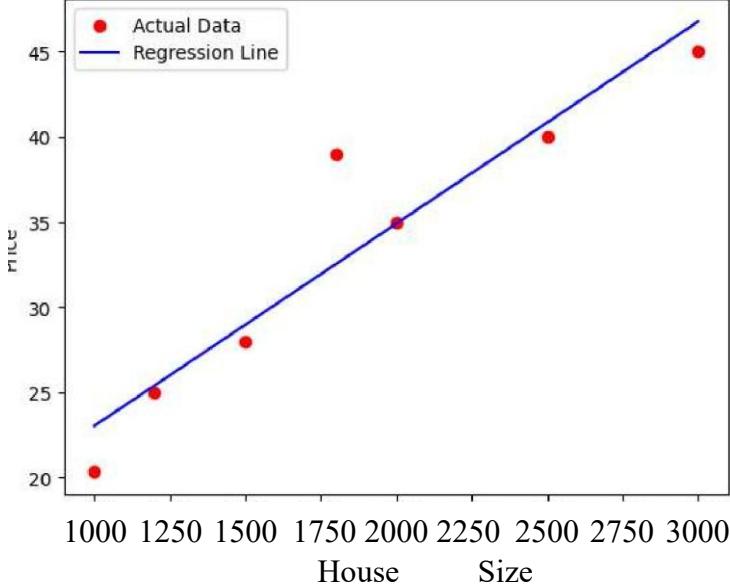
```

x_mean = np.mean(X) Y_mean =
np.mean(Y) num = x_mean) * (Y
y_mean) ) den = np.sum((X x_mean) m
= num / den c = y_mean m x mean

print(f'Regression Line: Y
Y_pred

plt.scatter(X, Y, color='red' label='Actual Data
') plt.plot(X, Y_pred, color= 'blue' ,
label='Regression Line' ) plt.xlabel("House
Size (sqft)") plt . y label( "Price") plt . legend()
plt . show()
Regression Line: y = 0.0119x + 11.1535

```

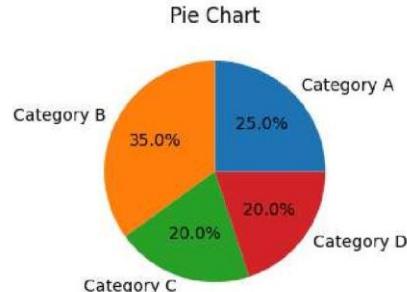
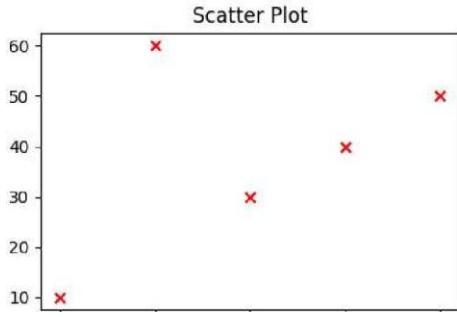
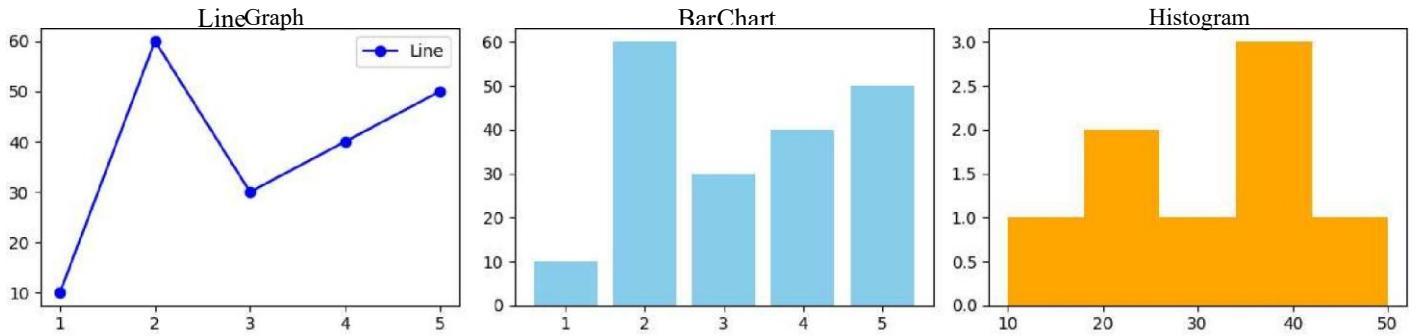


```
(sqft) import matplotlib.pyplot as plt
```

```

x = [1, 2, 3, 4, 5]
y =[10, 60, 30, 40, 50] data = [10, 20, 20, 30, 40, 40,
40, 50] sizes = [25, 35, 20, 20] labels = [ 'Category
A', 'Category B', 'Category C', 'Category D' ]
fig,ax = plt.subplots(2, 3, figsize=(12, 6))
ax[0,0] . plot (x, y, marker='o' label=
color='blue' ,Line' )
ax[0,0] set_title("Line Graph")
ax[0,e ] . legend()
ax[0,1] . bar(x, y, color='skyblue'
)
ax[0,1] set_title("Bar Chart")
ax[0,2] . hist (data, bins=5,
color= 'orange' )
ax[0,2] . set_title( "Histogram" )
ax[1,0] . scatter (x, Y, color='red' ,
marker=
)
ax[e ] .set_title("Scatter Plot")
[1,

```



```
ax[1,1].pie(sizes, labels=labels, autopct='%.1f%%')
plt.show()
```

```
ax[1,1].set_title("Pie Chart")
```

```
ax[1,2].axis('off')
```

```
plt.
```

```
tight_layout()
```

```
) plt.show()
```

```
import pandas as pd
```

```
# Create the
```

```
DataFrame data
```

```
'SAP ID':
```

```
'AGE': [25, 26, 23, 21, 24.5, 21.5, 20, 19], 'MARKS': [70, 71, 82, 45, 39, 78, 91, 90], 'FUTURE DREAM':
```

```
'Cyber Security', 'Data Science' 'AI/ML Expert'
```

```
,
```

```
' Full Stack', 'Dev Ops', 'Cyber Security',
```

```
'Data Science', 'Cloud Expert'
```

```
'LABEL': [ 'Hard', 'Easy', 'Hard', 'Easy', 'Easy', 'Hard',
```

```
df = pd.DataFrame(data) encoded df = pd.get_dummies(df, columns=[ 'LABEL' ])
print(encoded_df)
```

```
SAP ID AGE MARKS FUTURE DREAM
```

```
LABEL_Easy LABEL_Hard 1 25.0 70 Cyber Security False True
```

	SAP ID	AGE	MARKS	FUTURE	DREAM	LABEL_Easy	LABEL_Hard
1	2	26.0	71	Data Science	True	False	True
2	3	23.0	82	AI/ML Expert	False	True	False
3	4	21.0	45	Full Stack	True	False	False

```

4      5 24.5   39      Dev Ops      True
      False
5      6 21.5   78 Cyber Security    False  True
6      7 20.0   91      Data Science  True
      False
7      8       19.0   90      Cloud      Expert
      False  False
import  as                                'Easy' , 'Moderate ]
numpy
[40,  50,48, 55,  80]
45,      70,
2, 5, 8, 11]                               LABEL Moderate
                                                False
                                                False
                                                False
                                                False
                                                False
                                                False
                                                False
                                                True

```

```

#      Covariance
matrix cov matrix =
np.cov(X,      Y)
cov_xy      =
cov_matrix[Ø, 1]
# Correlation coefficient
corr_xy = np.corrcoef()(,
Y) [0, 1] print(
"Covariance , cov_xy")
print("Correlation -
corr_xy)

```

Covariance = 52.88095238095238  
Correlation = e. 9936740547587747

```

import numpy as np X
= np.array([2, 3, 4, 5])
Y = np.array([3, 4, 5,
6])      data      =
np.column_stack((X,
Y))

mean     = np.mean(data,
axis=e) centered data = data
mean

cov matrix = np. cov (centered_data, rowvar=False)

```

```
eigvals, eigvecs = np.linalg.eig(cov_matrix)

order = np.argsort(eigvals)[::-1]
principal_axes = eigvecs[:, order]

pc1 = principal_axes[:, 0]
projections = centered_data.dot(pc1)

print("Principal axes (each column):")
print(principal_axes)
print("Projections along first principal axis:")
print(projections)
```

```
Principal axes (each column):
[[ 0.70710678 -0.70710678]
 [ 0.70710678  0.70710678]]
Projections along first principal axis:
[-2.12132034 -0.70710678  0.70710678  2.12132034]
```

```
import numpy as np
import pandas as pd
from sklearn.decomposition import PCA

X = [2, 3, 4, 5]
```