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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 install scikit-learn  
Requirement already satisfied: numpy>=1.19.5 in /usr/10ca1/lib/python3.11/dist-packages (1.26.1)  
Requirement already satisfied: scipy>=1.6.0 in /usr/10ca1/lib/python3.11/dist-packages (1.10.1)  
Requirement already satisfied: joblib>=1.2.0 in /usr/10ca1/lib/python3.11/dist-packages (1.3.2)  
Requirement already satisfied: scikit-learn in /usr/10ca1/lib/python3.11/dist-packages (1.3.2)  
Requirement already satisfied: scikit-learn in /usr/10ca1/lib/python3.11/dist-packages (1.3.2)  
Requirement already satisfied: scikit-learn in /usr/10ca1/lib/python3.11/dist-packages (1.3.2)
```

Requirement already	learn)	(1.5.1)	threadp001ct1>=3.1.e	in
Requirement already	/usr/10ca1/1ib/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 VI))
    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[1]
```

```
point1  =
point2
```

```
print( "Distance: '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)

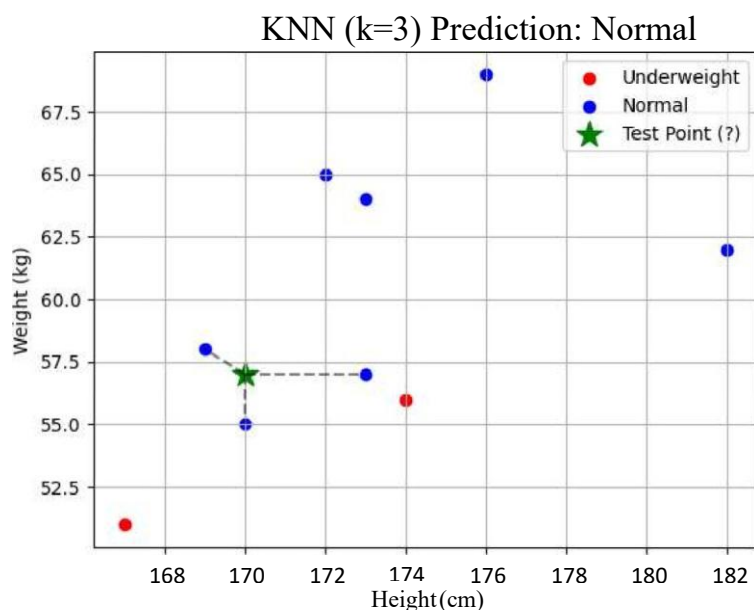
# KNN assumption def
knn_predict(train, test,
k):
    distances = [(euclidean_distance(test, row[ : row) for row
in train] distances.sort(key=lambda x: x[0]) nearest =
distances[:k] prediction = Counter([row[2] for , row in
nearest]).most_common(1) [0] [0] return prediction,
nearest

# 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}" )
Plt. legend() plt. grid (True) Plt. 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: threadpoolctl>=3.1.0 in /usr/local/lib/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.read_excel( "/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)
X = df[ '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) : \n")
display(x.head())
```

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

FileNotFoundErrorTraceback (most recent call last) [ÉmpAuython-input-1817842996.py](#) in <cell line:

```
2
3 # Load the data from the Excel file, skipping the first row
= pd.read_excel( "/content/Book1.xlsx" , 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
```

FileNotFoundError: [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
Plt .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=2ee, 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()

```

FileNotFoundError 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 ' ] ]

```

6 K = 3

3 frames

/usr/10ca1/1ib/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

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

Next steps:Explain error

```

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

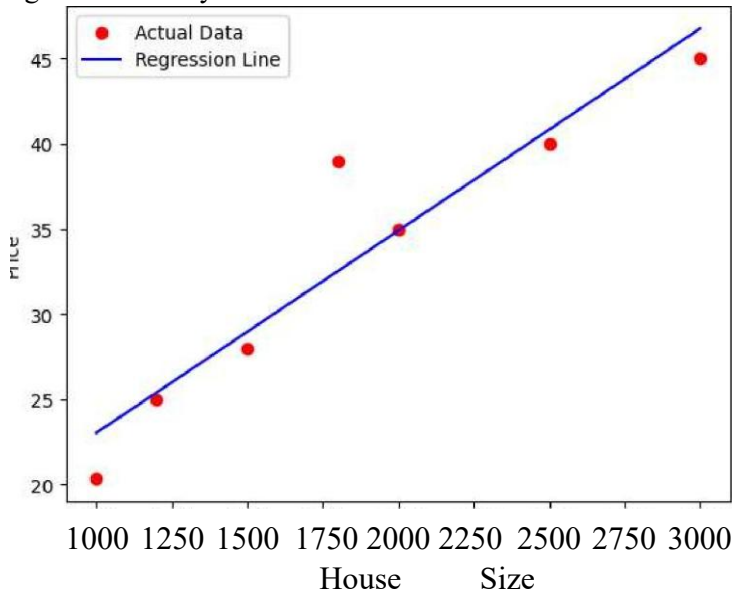
```
Y_pred
```

```

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.ylabel("Price") plt.legend()
plt.show()

```

```
Regression Line: y = 0.0119x + 11.1535
```

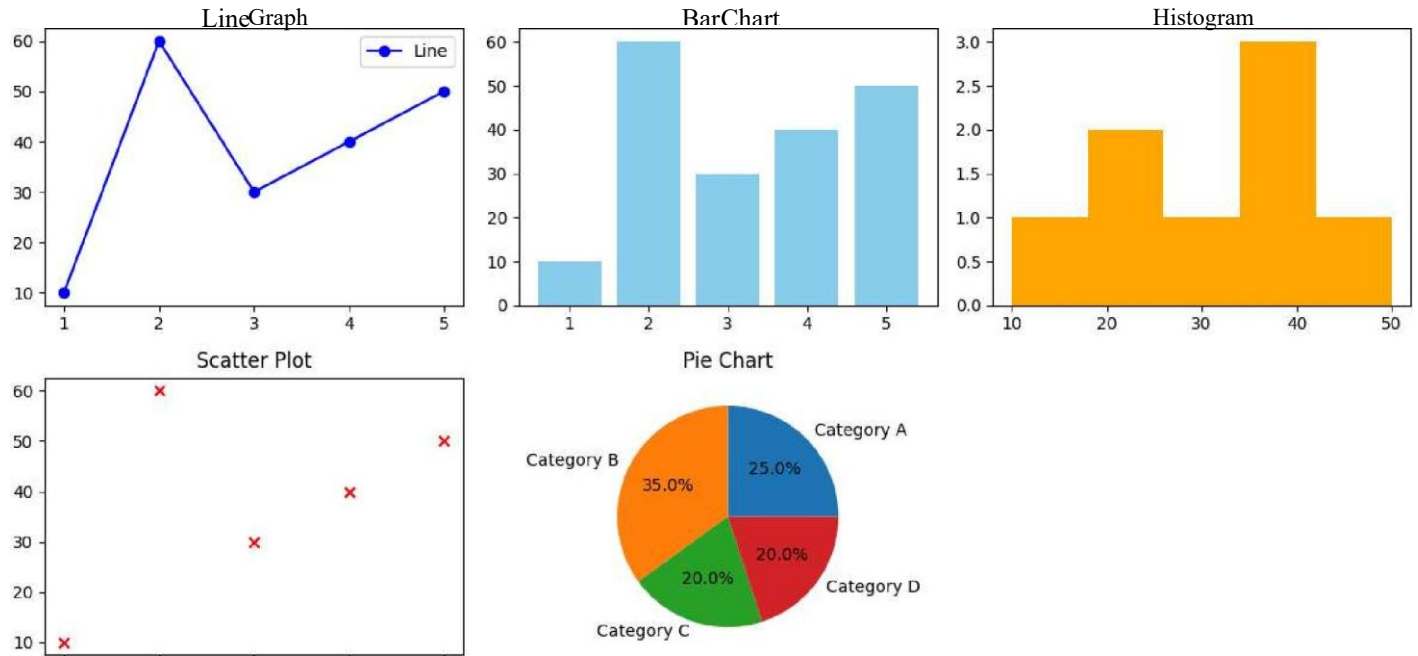


```
(sqft) import matplotlib.pyplot as plt
```

```

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[0,1] . 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='x')
ax[1,0] . set_title("Scatter Plot")
ax[1,1] .

```

```
ax[1,1] . pie(sizes, labels=labels, autopct= '
0/01 . 1 f%%,)
```

```
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
```

```
1 2 26.0 71 Data Science True
```

```
False
```

```
2 3 23.0 82 AI/ML Expert False True
```

```
3 4 21.0 45 Full Stack True
```

```
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
numpy      'Easy' , ' Moderate ]
[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[0, 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.c0lumn_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]
```