PRACTICAL 1 (SIMPLE LINEAR REGRESSION)

CODE AND OUTPUTS:

```
import numpy as np
import matplotlib.pyplot as plt
import pandas as pd
dataset = pd.read_csv("Salary_Data.csv")
X = dataset.iloc[:, :-1].values
y = dataset.iloc[:, -1].values
from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size = 1/4, random_state = 0)
from sklearn.linear model import LinearRegression
linear_regression = LinearRegression()
linear_regression.fit(X_train, y_train)
y_train_pred = linear_regression.predict(X_train)
y_test_pred = linear_regression.predict(X_test)
plt.scatter(X_train, y_train, color = "green", marker = "+", label = "Observed data")
plt.plot(X_train, y_train_pred, color = "red", label = "Predicted data")
plt.xlabel("Years of experience")
plt.ylabel("Salary")
plt.title("Years of experience v/s Salary (Training dataset)")
plt.legend()
plt.show()
               Years of experience v/s Salary (Training dataset)
                  Predicted data
```



```
plt.scatter(X_test, y_test, color = "green", marker = "+", label = "Observed data")
plt.plot(X_test, y_test_pred, color = "red", label = "Predicted data")
plt.xlabel("Years of experience")
plt.ylabel("Salary")
plt.title("Years of experience v/s Salary (Testing dataset)")
plt.legend()
```

plt.show()



PRACTICAL 2 (MULTIPLE LINEAR REGRESSION)

CODE AND OUTPUTS:

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

```
dataset = pd.read_csv('/content/50_Startups-2.csv')
x = dataset.iloc[:, :-1].values
y = dataset.iloc[:, -1].values
```

from sklearn.compose import ColumnTransformer
from sklearn.preprocessing import OneHotEncoder
ct = ColumnTransformer(transformers=[('encoder', OneHotEncoder(), [3])], remainder='passthrough')
x = np.array(ct.fit_transform(x))

from sklearn.model_selection import train_test_split
x_train, x_test, y_train, y_test = train_test_split(x, y, test_size=0.2, random_state=0)

In [10]:

from sklearn.linear_model import LinearRegression
regressor = LinearRegression()
regressor.fit(x_train, y_train)

```
y_pred = regressor.predict(x_test)
np.set_printoptions(precision=2)
print(np.concatenate((y_pred.reshape(len(y_pred),1), y_test.reshape(len(y_test),1)),1))
```

```
[[103015.2 103282.38]
[132582.28 144259.4 ]
[132447.74 146121.95]
[71976.1 77798.83]
[178537.48 191050.39]
[116161.24 105008.31]
[67851.69 81229.06]
[98791.73 97483.56]
[113969.44 110352.25]
[167921.07 166187.94]]
```

PRACTICAL 3 (SUPPORT VECTOR MACHINE)

CODE AND OUTPUTS:

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

y_pred = classifier.predict(x_test)

```
import pandas as pd
dataset = pd.read_csv('/content/Social_Network_Ads.csv')
x = dataset.iloc[:, :-1].values
y = dataset.iloc[:, -1].values
from sklearn.model_selection import train_test_split
x_train, x_test, y_train, y_test = train_test_split(x, y, test_size=0.25, random_state=0)
from sklearn.preprocessing import StandardScaler
sc = StandardScaler()
x_train = sc.fit_transform(x_train)
x_{test} = sc_{transform}(x_{test})
from sklearn.svm import SVC
classifier = SVC(kernel='linear', random_state=0)
classifier.fit(x_train, y_train)
print(classifier.predict(sc.transform([[30,200000]])))
[1]
                                                                                                          In [20]:
```

print(np.concatenate((y_pred.reshape(len(y_pred),1), y_test.reshape(len(y_test),1)),1))

```
[[0 0]
[0 0]
[0 0]
[0 0]
[0 0]
[0 0]
[0 0]
[1 1]
[0 0]
[0 0]
[0 0]
[0 0]
[0 0]
[0 0]
[0 0]
[0 0]
[0 0]
[0 0]
[1 1]
[0 0]
[0 0]
[1 1]
[0 0]
[1 1]
```

from sklearn.metrics import confusion_matrix, accuracy_score
cm = confusion_matrix(y_test, y_pred)
print(cm)
accuracy_score(y_test, y_pred)

```
[[66 2]
[824]]
0.9
```

PRACTICAL 4 (KNN)

CODE AND OUTPUTS:

import matplotlib.pyplot as plt
import pandas as pd
import numpy as np

In [22]:

```
dataset = pd.read_csv('/content/Social_Network_Ads.csv')
x = dataset.iloc[:, :-1].values
y = dataset.iloc[:, -1].values
```

```
from sklearn.model_selection import train_test_split
x_train, x_test, y_train, y_test = train_test_split(x, y, test_size=0.25, random_state=0)
```

```
from sklearn.preprocessing import StandardScaler
sc = StandardScaler()
x_train = sc.fit_transform(x_train)
x_test = sc.transform(x_test)
```

```
from sklearn.neighbors import KNeighborsClassifier
classifier = KNeighborsClassifier(n_neighbors=5, metric='minkowski', p=2)
classifier.fit(x_train, y_train)
```

```
KNeighborsClassifier()
```

Out[33]:

In [34]:

print(classifier.predict(sc.transform([[40, 200000]])))
[1]

```
y_pred = classifier.predict(x_test)
print(np.concatenate((y_pred.reshape(len(y_pred),1), y_test.reshape(len(y_test),1)),1))
```

[[0 0] [0 0] [0 0] [0 0] [0 0] [0 0] [0 0] [1 1] [0 0] [1 0] [0 0] [0 0] [0 0] [0 0] [0 0] [1 0] [0 0] [0 0] [1 1]

from sklearn.metrics import confusion_matrix, accuracy_score
cm = confusion_matrix(y_test, y_pred)
print(cm)
accuracy_score(y_test, y_pred)

[[64 4] [329]] 0.93

PRACTICAL 5 (HEIRARCHICAL CLUSTERING)

CODE AND OUTPUTS:

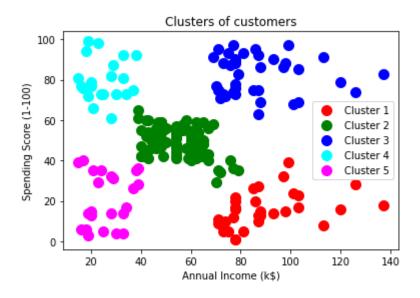
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt

In [9]:

```
dataset = pd.read_csv('/content/Mall_Customers.csv')
X = dataset.iloc[:, [3,4]].values
```

from sklearn.cluster import AgglomerativeClustering
hc = AgglomerativeClustering(n_clusters=5, affinity='euclidean', linkage='ward')
y_hc = hc.fit_predict(X)

```
 \begin{array}{l} \text{plt.scatter}(X[y\_hc==0,0],\ X[y\_hc==0,1],\ s=100,\ c='\text{red'},\ label='\text{Cluster 1'}) \\ \text{plt.scatter}(X[y\_hc==1,0],\ X[y\_hc==1,1],\ s=100,\ c='\text{green'},\ label='\text{Cluster 2'}) \\ \text{plt.scatter}(X[y\_hc==2,0],\ X[y\_hc==2,1],\ s=100,\ c='\text{blue'},\ label='\text{Cluster 3'}) \\ \text{plt.scatter}(X[y\_hc==3,0],\ X[y\_hc==3,1],\ s=100,\ c='\text{cyan'},\ label='\text{Cluster 4'}) \\ \text{plt.scatter}(X[y\_hc==4,0],\ X[y\_hc==4,1],\ s=100,\ c='\text{magenta'},\ label='\text{Cluster 5'}) \\ \text{plt.title}('\text{Clusters of customers'}) \\ \text{plt.ylabel}('\text{Annual Income (k\$)'}) \\ \text{plt.ylabel}('\text{Spending Score (1-100)'}) \\ \text{plt.show}() \end{array}
```



PRACTICAL 6 (K MEANS CLUSTERING)

CODE AND OUTPUTS:

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

In [2]:

```
dataset = pd.read_csv('Mall_Customers.csv')

X = dataset.iloc[:, [3, 4]].values
```

In [3]:

from sklearn.cluster import KMeans

```
kmeans = KMeans(n_clusters = 5, init = 'k-means++', random_state = 42)
y_kmeans = kmeans.fit_predict(X)
print(y_kmeans)
```

```
 \begin{array}{l} plt.scatter(X[y\_kmeans == 0, 0], \, X[y\_kmeans == 0, 1], \, s = 100, \, c = 'red', \, label = 'Cluster \, 1') \\ plt.scatter(X[y\_kmeans == 1, 0], \, X[y\_kmeans == 1, 1], \, s = 100, \, c = 'blue', \, label = 'Cluster \, 2') \\ plt.scatter(X[y\_kmeans == 2, 0], \, X[y\_kmeans == 2, 1], \, s = 100, \, c = 'green', \, label = 'Cluster \, 3') \\ plt.scatter(X[y\_kmeans == 3, 0], \, X[y\_kmeans == 3, 1], \, s = 100, \, c = 'cyan', \, label = 'Cluster \, 4') \\ plt.scatter(X[y\_kmeans == 4, 0], \, X[y\_kmeans == 4, 1], \, s = 100, \, c = 'magenta', \, label = 'Cluster \, 5') \\ plt.scatter(kmeans.cluster\_centers\_[:, 0], \, kmeans.cluster\_centers\_[:, 1], \, s = 300, \, c = 'yellow', \, label = 'Centroids') \\ plt.title('Clusters of customers') \\ plt.vlabel('Annual Income (k\$)') \\ plt.ylabel('Spending Score (1-100)') \\ plt.legend() \\ plt.show() \end{aligned}
```

Clusters of customers

100

80

Cluster 1

Cluster 1

Cluster 2

Cluster 3

Cluster 4

Cluster 5

Centroids

Annual Income (k\$)

PRACTICAL 7 (ANN)

CODE AND OUTPUTS:

 $X_{\text{test}} = \text{sc.transform}(X_{\text{test}})$

ann = tf.keras.models.Sequential()

```
import numpy as np
import pandas as pd
import tensorflow as tf
                                                                                               In [3]:
dataset = pd.read_csv('Churn_Modelling.csv')
X = dataset.iloc[:, 3:-1].values
y = dataset.iloc[:, -1].values
from sklearn.preprocessing import LabelEncoder
le = LabelEncoder()
X[:, 2] = le.fit_transform(X[:, 2])
                                                                                               In [7]:
print(X)
[[619 'France' 0 ... 1 1 101348.88]
 [608 'Spain' 0 ... 0 1 112542.58]
 [502 'France' 0 ... 1 0 113931.57]
 [709 'France' 0 ... 0 1 42085.58]
 [772 'Germany' 1 ... 1 Ø 92888.52]
 [792 'France' 0 ... 1 0 38190.78]]
from sklearn.compose import ColumnTransformer
from sklearn.preprocessing import OneHotEncoder
ct = ColumnTransformer(transformers=[('encoder', OneHotEncoder(), [1])], remainder='passthrough')
X = np.array(ct.fit_transform(X))
print(X)
[[1.0 0.0 0.0 ... 1 1 101348.88]
 [0.0 0.0 1.0 ... 0 1 112542.58]
 [1.0 0.0 0.0 ... 1 0 113931.57]
  [1.0 0.0 0.0 ... 0 1 42085.58]
  [0.0 1.0 0.0 ... 1 0 92888.52]
  [1.0 0.0 0.0 ... 1 0 38190.78]]
from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size = 0.2, random_state = 0)
                                                                                              In [11]:
from sklearn.preprocessing import StandardScaler
sc = StandardScaler()
X_train = sc.fit_transform(X_train)
```

In [12]:

```
In [13]:
ann.add(tf.keras.layers.Dense(units=6, activation='relu'))
                                                                                         In [14]:
ann.add(tf.keras.layers.Dense(units=6, activation='relu'))
                                                                                         In [15]:
ann.add(tf.keras.layers.Dense(units=1, activation='sigmoid'))
                                                                                         In [16]:
ann.compile(optimizer = 'adam', loss = 'binary_crossentropy', metrics = ['accuracy'])
                                                                                         In [17]:
ann.fit(X train, y train, batch size = 32, epochs = 100)
Epoch 1/100
250/250 [============ ] - 1s 1ms/step - loss: 0.5750 - accuracy: 0.7490
Epoch 2/100
250/250 [============ ] - Øs 1ms/step - loss: 0.4712 - accuracy: 0.7960
Epoch 3/100
250/250 [=========== ] - Øs 2ms/step - loss: 0.4428 - accuracy: 0.7986
Epoch 4/100
250/250 [============ ] - Øs 2ms/step - loss: 0.4296 - accuracy: 0.8075
Epoch 5/100
250/250 [============ ] - 0s 2ms/step - loss: 0.4212 - accuracy: 0.8149
print(ann.predict(sc.transform([[1, 0, 0, 600, 1, 40, 3, 60000, 2, 1, 1, 50000]])) >0.5)
[[False]]
y_pred = ann.predict(X_test)
y pred = (y \text{ pred} > 0.5)
print(np.concatenate((y_pred.reshape(len(y_pred),1), y_test.reshape(len(y_test),1)),1))
 [[0 0]]
  [0 1]
  [0 0]
  . . .
  [0 0]
  [0 0]
  [0 0]]
from sklearn.metrics import confusion_matrix, accuracy_score
cm = confusion_matrix(y_test, y_pred)
print(cm)
accuracy_score(y_test, y_pred)
 [[1499
           961
  [ 186 219]]
 0.859
```

PRACTICAL 8 (CNN)

CODE AND OUTPUTS:

import tensorflow as tf

from keras.preprocessing.image import ImageDataGenerator

In [46]:

train_datagen = ImageDataGenerator(rescale=1./255, shear_range=0.2, zoom_range=0.2, horizontal_flip=**True**) training_set = train_datagen.flow_from_directory('/content/drive/MyDrive/small_dataset/training_set', target_size=(64,64), batch_size=32, class_mode='binary')

Found 10 images belonging to 2 classes.

In [48]:

train_datagen = ImageDataGenerator(rescale=1./255, shear_range=0.2, zoom_range=0.2, horizontal_flip=**True**) test_set = train_datagen.flow_from_directory('/content/drive/MyDrive/small_dataset/test_set', target_size=(64,64), batch_size=32, class_mode='binary')

Found 10 images belonging to 2 classes.

In [49]:

cnn = tf.keras.models.Sequential()

In [50]:

 $cnn. add (tf. keras. layers. Conv2D (filters=32, kernel_size=3, activation='relu', input_shape=[64, 64, 3]))$

In [51]:

cnn.add(tf.keras.layers.MaxPool2D(pool_size=2, strides=2))

In [52]:

cnn.add(tf.keras.layers.Conv2D(filters=32, kernel_size=3, activation='relu')) cnn.add(tf.keras.layers.MaxPool2D(pool_size=2, strides=2))

In [53]:

cnn.add(tf.keras.layers.Flatten())

cnn. add (tf. keras. layers. Dense (units=128, activation='relu'))

In [55]:

cnn.add(tf.keras.layers.Dense(units=1, activation='sigmoid'))

In [56]:

cnn.compile(optimizer='adam', loss='binary_crossentropy', metrics=['accuracy'])

In [57]:

cnn.fit(x=training_set, validation_data=test_set, epochs=25)

| Epoch 1/25 |
|--|
| 1/1 [=================================== |
| Epoch 2/25 |
| 1/1 [=================================== |
| Epoch 3/25 |
| 1/1 [=================================== |
| Epoch 4/25 |
| 1/1 [=================================== |
| Epoch 5/25 |
| 1/1 [=================================== |
| |

import numpy as np

from keras.preprocessing import image

test_image=image.load_img('/content/drive/MyDrive/small_dataset/single_prediction/cat_or_dog_1.jpg', target_size=(64,64))

test_image=image.img_to_array(test_image)

In [63]:

```
test_image=np.expand_dims(test_image, axis=0)
result=cnn.predict(test_image)
training_set.class_indices
if result[0][0]==1:
prediction='dog'
else:
prediction='cat'

print(prediction)
```