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Linear Regression
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1. Simple Linear Regression
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a) Importing the Libraries:-
import pandas as pd
import numpy as mp
import matplotlib.pyplot as plt
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b) Importing Dataset:-

dataset = pd. read_csv ('Data.csv')

X = dataset.iloc[:,:-1].values

y = dataset.iloc[:,-1].values
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(integer based)

iloc = locate indexes of rows and
columns both

: = range (when specified without
lower/upper bound, taking
everything in range)
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- c) Splitting the dataset into Training set and Testing set: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)
- d) Training the simple linear Regression model on the Training set:from sklearn.linear\_model import LinearRegression
  regressor = LinearRegression()
  regressor.fit (X\_train, y\_train)
- e) Predicting the Test set:regressor.predict(x\_test)
- f) Visualizing the Training set results:

  plt.scatter (X-train, y-train, color = 'red')

  plt.plot (X-train, regressor.predict (X-train), color = 'blue')

  plt.title ('Training set Results')

  plt.xlabel ('Experience')

  plt.ylabel ('Salary')

  plt.show()

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9) Visualizing the Test set results:-

plt·scatter(x_test, y_test, color = 'red')

plt·plot(x_train, regressor. predict(x_train), color = 'blue')

plt·title('Test set Result')

plt·xlabel('Experience')

plt·ylabel('Salary')

plt·show()
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2) Multiple Linear Regression:
 a) Import Libraries
 b) Import Dataset:-
        csv = 50_ startups. csv
 c) Encoding Categorical Data:
      from sklearn compose import ColumnTransformer
      from sklearn preprocessing import one Hot Encoder
       ct = ColumnTransformer (transformers = [('encoder', One HotEncoder(), [3])],
                                            remainder = (pass-through)
       X = np.array (ct.fit_transform (x))
     Encoding the Dependent Variable :-
        from sklearn. preprocessing import label Encoder
          le = labelEncoder ()
          y = le.fit_transform(y)
  d) Splitting the dataset into Training and Testing set.
  e) Training MIR model on Training dataset:
         from skleam. linear_model import Linear Regression ()
         regressor = Linear Regression ()
         regressor.fit (X-train, y-train)
  f) Predicting the Test set results :-
          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))
  3) Polynomial Regression:
   a) Import libraries
   b) Import Dataset
   c) Training PR model on whole dataset:-
         from skleam. preprocessing import Polynomial Features
          poly-reg = Polynomial Features (degree = 2)
          poly- reg. fit_transform (x)
          lin-leg = Linear Regression ()
          Lin_seq.fit (x_polynomial, y)
    d) Visualizing the results:
          (i) Linear Regression
                                                       (ii) Polynomial Regression
                                                           pit.scatter (x,y, color = 'red')
         pit.scatter (x, y, color = 'red')
                                                         pit. show (x, lin_leg.predict (poly-
        pit.plot (x, lin_seg1.predict(x), color='blue')
                                                               leg. fit_transform(x)), color
         plt. show ()
                                                                = 'blue')
                                                          plt.show()
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4) Decision Tree :-
  a) Regression Model:-
  (i) Import libraries
 (ii) Import Dataset
        dataset = pd read_csv ('PositionSalaries csv')
            x = dataset · iloc [:, 1:-1] · values
            y = dataset · iloc [:,-1] · values
  (iii) Training DTR on whole dataset:
         from sklearn tree import DecisionTree Regressor
         regressor = DecisionTree Regressor ()
         regressor.fit (x,y)
  (iv) Predicting a new result :-
          regressor.predict ([[7.5]])
   b) Classification Model:-
    (i) Import Libraries
    (ii) Import Dataset
    (iii) Split Lataset into Training and Testing set
    (iv) Training DIC on training dataset:
            from sklearn. tree import Decision Tree Classifier
            classifier = DecisionTreeClassifier (criterion = 'entropy', random_state = 0)
            classifier. fit (x-train, y-train)
    (V) Predict Results :-
          · Test set results
            y-pred = classifier.predict (x-test)
            print(np.concatenate((y-pred.reshape(len(y-pred),1), y-test.reshape(len(y-test),1))
                                                                                         1))
          · Predicting new result
            print(classifier.predict(sc.transform([[30,87000]])))
    (vi) Making Confusion Matrix :-
           from skleam metrics import confusion matrix, accuracy score
             cm = confusion_matrix (y-test, y-pred)
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accuracy-score (y-test, y-pred)

## (5) Naive Baye's Classifier:a) Import libraries b) Import Dataset c) Split into training and testing set d) Training Naive Bayes Model on training set from sklearn naive bayes import GaussianNB classifier = GaussianNB() classifier.fit (x\_train, y\_train) e) Predicting the Test Result: print(classifier.predict (sc.transform ([[30,87000]]))) f) Predicting the Test Results:y-pred = classifier.predict (x-test) print (np. contatenate ((y-pred. reshape (len (y-pred), 1), y-test. reshape (len(y-test),1)),1)) 9) Making the confusion matrix :from sklearn. metrics import confusion matrix, accuracy - score cm = confusion\_matrix (y\_test, y\_pred) accuracy-score (y-test, y-pred) 6) K-Nearest Neighbour :a) Import Libraries b) Import Dataset c) Split into training and testing set d) Training K-NN Model on Training set from skleam. neighbors import KNeighbors Classifier classifier = KNeighbors Classifier (n-neighbors = 5, metric = 'minkowski', p=2) classifier. fit (x-train, y-train)

- e) Predicting new test data result
- f) Predicting the Test Results
- g) Making confusion Matrix

## 7) Support Vector Machine :-

- a) Import Libraries
- b) Import Rataset
- c) Split into training and testing set
- d) Training svM Model on Training set

  from sklearn.svm import svc

  classifier = svc (kernel = 'linear', landom\_state = 0)

  classifier.fit (X\_train, y\_train)
- e) Predicting new test data result
- f) Predicting the test data result
- 9) Make Confusion Matrix

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8) Artificial Neural Network :-
  a) Import Libraries :-
      import pandas as pd
      import numpy as np
      import tensorflow as tf
   b) Import Dataset :-
        dataset = pd. read_csv ('Churn_Modelling.csv')
           X = dataset.iloc[:,3:-1].values
           y = dataset.iloc[:,-1].values
    c) Encoding Categorical data :-
       (i) label encoding 'Gender' column
           from sklearn preprocessing import Label Encoder
            le = labelEncoder()
            x[:,2] = le.fit_transform (x[:,2])
       (ii) One Hot Encoding 'Geography' column
             from sklearn compose import ColumnTransformer
            from sklearn preprocessing import OneHotEncoder
            et = ColumnTransformers(transformers = [ ('encoder', One Hot Encoder(), [1])],
                 remainder'='passthrough')
             X = np. array (ct. fit_transform(x))
     d) Split Dataset into Training and Testing set
     e) Feature Scaling :-
          from sklearn preprocessing import Standard Scaler
            sc = Standard Scaler()
           X-train = sc.fit_transform (x-train)
           X_test = sc. transform (x_test)
   f) Initialize ANN:-
         ann = tf.keras.models.Sequential ()
    g) Adding input layer and 1st Hidden layer:-
         ann.add (tf.keras.layers.Dense(units = 6, activation = 'selu'))
   h) Adding second hidden layer: -
        ann.add(tf. keras.layers. Dense (units = 6, activation = 'selu'))
    i) Adding the output layer:-
         anniadd (tf. keras. layers. Dense (units = 1, activation = 'sigmoid'))
   j) Training the ANN:-
       (i) Compiling the ANN :-
          ann.compile (optimizer = 'adam', loss = 'binary_crossentropy', metrics = ['accuracy'])
      (ii) Training the ANN on training set :-
          ann.fit (X-train, y-train, batch_size = 32, epochs = 100)
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K) Making Predictions and Evaluating the Model:
      e.g. Geography = France, Credit Score = 600, Gender = Male, Age = 40 years,
          Tenure = 3 years, Balance = 60000, Number of Products = 2, credit Card = Yes,
           Active Member = les , Estimated Salary = 50000, Goodbye = ?
    print (ann. predict (sc. transform ([[],0,0,600,1,40,3,60000,2,1,1,50000]])))
  1) Predicting the Test Set results :-
          y-pred = ann.predict (x-test)
           y-pred = (4-pred > 0.5)
          print (np. concatenate ((y-pred. reshape (len(y-pred), +), y-test. reshape (len(
                                                                       y-test),1)),1))
  m) Make Confusion Matrix :-
       from sklearn metrics import confusion matrix, accuracy score
        cm = confusion-matrix (y-test, y-pred)
       accuracy_score (y-test, y-pred)
9) K-Means Clustering:
   a) Import Libraries
   b) Import Dataset:-
         dataset = pd. read_csv ('Mall_customers.csv')
          X = dataset.iloc[:,[3,4]].values
    c) Using the elbow method to find optimal no of clusters :-
           from sklearn cluster import KMeans
            wess = []
            for i in range (1,11):
                kmeans = KMeans (n_clusters = i , init = 'k-means++', random_state = 0)
                kmeans.fit(x)
                wcss.append(kmeans.inertia_)
            pit.plot (range (1,11), wess)
            plt. show ()
    d) Training the k-Means Model on the dataset:
         kmeans = KMeans (n_clusters = 5, init = k-means++', random_state = 0)
         y_kmeans = kmeans · fit_predict(x)
         print(y-kmeans)
```