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|  | **CO2160714.1 Assignment:** |
| 1. | **Analyse 5 data sets from the UCI repository. Print the following details about each data set**   1. **number of records/instances** 2. **number of incomplete records** 3. **number of attributes** |
|  | **Input:**  import fnmatch  import os  import pandas as pd  f\_n=[]  for f in os.listdir('.'):  if fnmatch.fnmatch(f, '\*.csv'):  f\_n.append(f)  print ("Total files are:",f\_n)  for i in f\_n:  print("\nFile name is ",i)  i=pd.read\_csv(i,delimiter=',')  k=i.keys()  #print(len(k))  print("Total number of records/instances:",len(i))  print("Total number of null records:",i.isnull().values.sum())  c1=0  c=i['lat'].count().sum()  for x in range(0,c):  if((i['lat'][x]< -90 or i['lat'][x]>90) and (i['long'][x]< -180 or i['long'][x]>180) ):  c1=c1+1  print("Total number of Inconsistant Data:",c1)  print("Total number of attributes:",len(k)) |
|  | **Output:**  Total files are: ['airport-codes1.csv', 'NYC\_Airbnb.csv', 'urbanGB.csv', 'worldcities.csv', 'world\_country\_and\_usa\_states\_latitude\_and\_longitude\_values.csv']  File name is airport-codes1.csv  Total number of records/instances: 99  Total number of null records: 215  Total number of Inconsistant Data: 2  Total number of attributes: 13  File name is NYC\_Airbnb.csv  Total number of records/instances: 5  Total number of null records: 2  Total number of Inconsistant Data: 0  Total number of attributes: 5  File name is urbanGB.csv  Total number of records/instances: 52  Total number of null records: 32  Total number of Inconsistant Data: 2  Total number of attributes: 3  File name is worldcities.csv  Total number of records/instances: 26569  Total number of null records: 20036  Total number of Inconsistant Data: 0  Total number of attributes: 11  File name is world\_country\_and\_usa\_states\_latitude\_and\_longitude\_values.csv  Total number of records/instances: 245  Total number of null records: 775  Total number of Inconsistant Data: 2  Total number of attributes: 8 |

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| 2. | **Write a program to implement data cleaning(*incomplete, noisy, inconsistent, redundant*) on your data set. Implement each technique.**  **(a). Binning with means and/or mode** |
|  | **Input:**  import pandas as pd  import numpy as np  df1=pd.read\_csv("airport-codes1.csv")  df=[]  for i in range(0,50):  df.append(df1['elevation\_ft'][i])  print("Original Data:",df)  arr\_1D= np.sort(df)  print("Sorted Data : ",arr\_1D)  nb=input("Enter how many bins you want to create:")  size=len(df)//int(nb)  means=[]  meadians=[]  bins=[arr\_1D[i:i+size] for i in range(0, len(arr\_1D), size)]  print("Bins : ")  for x in bins:  print(x)  m=np.mean(x)  m1=np.median(x)  for y in x:  means.append(m)  meadians.append(m1)  mean\_bin=[means[i:i+size] for i in range(0, len(means), size)]  print("After Smoothing Using Mean:")  for x in mean\_bin:  print(x)  median\_bin=[meadians[i:i+size] for i in range(0, len(meadians), size)]  print("After Smoothing Using Median:")  for x in median\_bin:  print(x)  d\_b=[]  for x in bins:  lb=x[0]  rb=x[-1]  d\_b.append(lb)  for y in range(1,len(x)-1):  if((x[y]-lb)<(rb-x[y])):  d\_b.append(lb)  else:  d\_b.append(rb)  d\_b.append(rb)  bbin=[d\_b[i:i+size] for i in range(0, len(d\_b), size)]  print("Smoothing using boundaries : ")  for i in bbin:  print(i) |
|  | **Output:**  Original Data: [11.0, 3435.0, 450.0, 820.0, 237.0, 1100.0, 3810.0, 3038.0, 87.0, 3350.0, 4830.0, 53.0, 25.0, 35.0, 700.0, 957.0, 43.0, 2064.0, 3359.0, 600.0, 840.0, 634.0, 820.0, 1100.0, 1265.0, 15.0, 600.0, 12.0, 45.0, 588.0, 1365.0, 970.0, 2600.0, 105.0, 348.0, 78.0, 96.0, 1000.0, 785.0, 905.0, 960.0, 195.0, 402.0, 1301.0, 815.0, 1620.0, 150.0, 1590.0, 598.0, 600.0]  Sorted Data : [ 11. 12. 15. 25. 35. 43. 45. 53. 78. 87. 96. 105.  150. 195. 237. 348. 402. 450. 588. 598. 600. 600. 600. 634.  700. 785. 815. 820. 820. 840. 905. 957. 960. 970. 1000. 1100.  1100. 1265. 1301. 1365. 1590. 1620. 2064. 2600. 3038. 3350. 3359. 3435.  3810. 4830.]  Enter how many bins you want to create:5  Bins :  [11. 12. 15. 25. 35. 43. 45. 53. 78. 87.]  [ 96. 105. 150. 195. 237. 348. 402. 450. 588. 598.]  [600. 600. 600. 634. 700. 785. 815. 820. 820. 840.]  [ 905. 957. 960. 970. 1000. 1100. 1100. 1265. 1301. 1365.]  [1590. 1620. 2064. 2600. 3038. 3350. 3359. 3435. 3810. 4830.]  After Smoothing Using Mean:  [40.4, 40.4, 40.4, 40.4, 40.4, 40.4, 40.4, 40.4, 40.4, 40.4]  [316.9, 316.9, 316.9, 316.9, 316.9, 316.9, 316.9, 316.9, 316.9, 316.9]  [721.4, 721.4, 721.4, 721.4, 721.4, 721.4, 721.4, 721.4, 721.4, 721.4]  [1092.3, 1092.3, 1092.3, 1092.3, 1092.3, 1092.3, 1092.3, 1092.3, 1092.3, 1092.3]  [2969.6, 2969.6, 2969.6, 2969.6, 2969.6, 2969.6, 2969.6, 2969.6, 2969.6, 2969.6]  After Smoothing Using Median:  [39.0, 39.0, 39.0, 39.0, 39.0, 39.0, 39.0, 39.0, 39.0, 39.0]  [292.5, 292.5, 292.5, 292.5, 292.5, 292.5, 292.5, 292.5, 292.5, 292.5]  [742.5, 742.5, 742.5, 742.5, 742.5, 742.5, 742.5, 742.5, 742.5, 742.5]  [1050.0, 1050.0, 1050.0, 1050.0, 1050.0, 1050.0, 1050.0, 1050.0, 1050.0, 1050.0]  [3194.0, 3194.0, 3194.0, 3194.0, 3194.0, 3194.0, 3194.0, 3194.0, 3194.0, 3194.0]  Smoothing using boundaries :  [11.0, 11.0, 11.0, 11.0, 11.0, 11.0, 11.0, 87.0, 87.0, 87.0]  [96.0, 96.0, 96.0, 96.0, 96.0, 598.0, 598.0, 598.0, 598.0, 598.0]  [600.0, 600.0, 600.0, 600.0, 600.0, 840.0, 840.0, 840.0, 840.0, 840.0]  [905.0, 905.0, 905.0, 905.0, 905.0, 905.0, 905.0, 1365.0, 1365.0, 1365.0]  [1590.0, 1590.0, 1590.0, 1590.0, 1590.0, 4830.0, 4830.0, 4830.0, 4830.0, 4830.0] |

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| 2. | **Write a program to implement data cleaning(*incomplete, noisy, inconsistent, redundant*) on your data set. Implement each technique.**  **(b.) Find covariance(cov) and correlation(r), Sx and Sy are standard deviation, x̅ and ȳ are means.**  **Plot the correlation, to show whether two variables are positively correlated, negatively correlated or no relation between them.** |
|  | **Input:**  from numpy import array  import pandas as pd  import numpy as np  defcovari(x,y):  xmean = np.mean(x)  ymean = np.mean(y)  n = len(data)  ans = 0  for i in range(0,n):  ans = ans + (x[i]-xmean) \* (y[i]-ymean)  result = ans / (n-1)  return round(result,2)  defCorelation(x,y,cov):  sx = np.std(x)  sy = np.std(y)  rxy = cov / (sx \* sy)  return rxy  usecols=['S1', 'S2','S3','S4']  data = pd.read\_csv("soccer\_goals.csv",delimiter=',', names=usecols)  S1=array(data['S1'])  S2=array(data['S2'])  S3=array(data['S3'])  S4=array(data['S4'])  co=covari(S1,S2)  rel=Corelation(S1,S2,co)  print("Covarienceis:",co)  print("Corelation :",rel)  co=covari(S1,S3)  rel=Corelation(S1,S3,co)  print("Covarienceis:",co)  print("Corelation :",rel)  co=covari(S1,S4)  rel=Corelation(S1,S4,co)  print("Covarienceis:",co)  print("Corelation :",rel)  import matplotlib  import matplotlib.pyplot as plt  matplotlib.style.use('ggplot')  plt.scatter(S1, S2)  plt.xlabel("S1")  plt.ylabel("S2")  plt.title("Negative Correlation")  plt.show()  plt.scatter(S1, S3)  plt.xlabel("S1")  plt.ylabel("S3")  plt.title("Positive Correlation")  plt.show()  plt.scatter(S1, S4)  plt.xlabel("S1")  plt.ylabel("S4")  plt.title("No Correlation")  plt.show() |
|  | **Output:**  Covarience is: -841.67  Corelation : -1.01010501050105  Covarience is: 2525.0  Corelation : 1.0101010101010102  Covarience is: 2.85  Corelation : 0.027428578051746585  C:\Users\Lenovo\Desktop\pr2_2_1.PNG  C:\Users\Lenovo\Desktop\pr2_2_2.PNG |

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| 3. | **Implement chi-square test to detect whether two variables are dependent or independent variables for your dataset.** |
|  | **Input:**  import pandas as pd  import numpy as np  import scipy.stats as stats  from scipy.stats import chi2\_contingency  class ChiSquare:  def \_\_init\_\_(self, dataframe):  self.df = dataframe  self.p = None #P-Value  self.chi2 = None  self.dof = None  self.dfObserved = None  self.dfExpected = None  def \_print\_chisquare\_result(self, colX, alpha):  result = ""  if self.p<alpha:  result="{0} is IMPORTANT for Prediction".format(colX)  else:  result="{0} is NOT an important predictor. (Discard {0} from model)".format(colX)  print(result)  defTestIndependence(self,colX,colY, alpha=0.09):  X = self.df[colX].astype(str)  Y = self.df[colY].astype(str)  self.dfObserved = pd.crosstab(Y,X)  chi2, p, dof, expected = stats.chi2\_contingency(self.dfObserved.values)  self.p = p  self.chi2 = chi2  self.dof = dof  self.dfExpected = pd.DataFrame(expected, columns=self.dfObserved.columns, index = self.dfObserved.index)  self.\_print\_chisquare\_result(colX, alpha)  n=['ident','type','name','elevation\_ft','continent','iso\_country','iso\_region','municipality','gps\_code','iata\_code','local\_code','lat','long']  df = pd.pandas.read\_csv("airport-codes1.csv",delimiter=",",names=n)  #Initialize ChiSquare Class  cT = ChiSquare(df)  #Feature Selection  testColumns = ['type','local\_code','lat','long']  for var in testColumns:  cT.TestIndependence(colX=var,colY="ident" ) |
|  | **Output:**  type is NOT an important predictor. (Discard type from model)  local\_code is NOT an important predictor. (Discard local\_code from model)  lat is NOT an important predictor. (Discard lat from model)  long is NOT an important predictor. (Discard long from model) |
| 4. | **Write a program to implement normalization techniques (a)min max (b) z-score (c) decimal scaling on your data set.** |
|  | **Input:**  import statistics as st  import pandas as pd  usecols=['ident','type','name','elevation\_ft','continent','iso\_country','iso\_region','municipality','gps\_code','iata\_code','local\_code','lat','long']  data1 = pd.read\_csv("airport-codes1.csv",delimiter=',',names=usecols)  data=data1['lat']  newmin = 0  newmax = 1  v = 10  minmax = newmin+((v-min(data))\*(newmax-newmin)/(max(data)-min(data)))  print("minmax normalization : ",round(minmax,2))  zscore = (v-st.mean(data))/st.stdev(data)  print("z-score : ", zscore)  decimalscale = v/100  print("decimal scaling : ",decimalscale) |
|  | **Output:**  minmax normalization : 2.11  z-score : 5.31864647094237  decimal scaling : 0.1 |

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| 5. | **Write a program to implement data reduction techniques for your data.** |
|  | **Input:**  import pandas as pd  usecols=['ident','type','name','elevation\_ft','continent','iso\_country','iso\_region','municipality','gps\_code','iata\_code','local\_code','lat','long']  df = pd.read\_csv('airport-codes1.csv',names= usecols)  print("BEFORE REDUCTION")  print("Rows:",df.shape[0]," Columns: ",df.shape[1])  print("SIZE:",df.size)  columns = ['ident','type','name','iso\_country','iso\_region']  df.drop(columns,axis=1, inplace = True)  print("\n AFTER REDUCTION")  print("Rows:",df.shape[0]," Columns: ",df.shape[1])  print("SIZE:",df.size) |
|  | **Output:**  STRING/DATA REDUCTION  BEFORE REDUCTION  Rows: 14 Columns: 13  SIZE: 182  AFTER REDUCTION  Rows: 14 Columns: 8  SIZE: 112​ |

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| 6. | **Write a program to implement any method of data discretization.** |
|  | **Input:**  #using label  from numpy import array  import pandas as pd  usecols=['ident','type','name','elevation\_ft','continent','iso\_country','iso\_region','municipality','gps\_code','iata\_code','local\_code','lat','long']  data = pd.read\_csv("airport-codes1.csv",delimiter=',', names=usecols)  c=[]  card1=data['type']  for x in range(0,13):  c.append(card1[x])  print("Your Data")  print(c)  heliport=[]  closed=[]  small\_airport=[]  for x in c:  if(x=='heliport'):  heliport.append(x)  elif(x=='closed'):  closed.append(x)  else:  small\_airport.append(x)  print("\nheliport : ",heliport)  print("\nsmall\_airport : ",small\_airport)  print("\nclosed : ",closed) |
|  | **Output:**  Your Data  ['heliport', 'small\_airport', 'small\_airport', 'small\_airport', 'closed', 'small\_airport', 'small\_airport', 'small\_airport', 'small\_airport', 'heliport', 'closed', 'small\_airport', 'heliport']  heliport : ['heliport', 'heliport', 'heliport']  small\_airport : ['small\_airport', 'small\_airport', 'small\_airport', 'small\_airport', 'small\_airport', 'small\_airport', 'small\_airport', 'small\_airport']  closed : ['closed', 'closed'] |

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|  | **CO2160714.2 Assignment:** |
| 7. | **Implement apriori algorithm and show the output as candidate sets in each iteration, as well as show association rules generated.** |
|  | **Input:**  # apriory  import itertools  support = int(input("Please enter support value in %: "))  confidence = int(input("Please enter confidence value in %: "))  C1 = {}  transactions = 0  D = []  T = []  with open("foods\_type.csv", "r") as f:  for line in f:  T = []  transactions += 1  for word in line.split(','):  T.append(word)  if word not in C1.keys():  C1[word] = 1  else:  count = C1[word]  C1[word] = count + 1  D.append(T)  print ("Dataset:",D)  #Computing frequent dataitems-1  L1 = []  for key in C1:  if (100 \* C1[key]/transactions) >= support:  list = []  list.append(key)  L1.append(list)  print ("\nFrequentItemSet: 1")  print (L1)  defapriori\_gen(Lk\_1, k):  length = k  Ck = []  for list1 in Lk\_1:  for list2 in Lk\_1:  count = 0  c = []  if list1 != list2:  while count < length-1:  if list1[count] != list2[count]:  break  else:  count += 1  else:  if list1[length-1] < list2[length-1]:  for item in list1:  c.append(item)  c.append(list2[length-1])  if not has\_infrequent\_subset(c, Lk\_1, k):  Ck.append(c)  c = []  return Ck  deffindsubsets(S,m):  return set(itertools.combinations(S, m))  defhas\_infrequent\_subset(c, Lk\_1, k):  list = []  list = findsubsets(c,k)  for item in list:  s = []  for l in item:  s.append(l)  s.sort()  if s not in Lk\_1:  return True  return False  deffrequent\_itemsets():  k = 2  Lk\_1 = []  Lk = []  L = []  count = 0  transactions = 0  for item in L1:  Lk\_1.append(item)  while Lk\_1 != []:  Ck = []  Lk = []  Ck = apriori\_gen(Lk\_1, k-1)  for c in Ck:  count = 0  transactions = 0  s = set(c)  for T in D:  transactions += 1  t = set(T)  if s.issubset(t) == True:  count += 1  if (100 \* count/transactions) >= support:  c.sort()  Lk.append(c)  Lk\_1 = []  print ("\nFrequentItemset: %d" % k)  print (Lk)  for l in Lk:  Lk\_1.append(l)  k += 1  if Lk != []:  L.append(Lk)  return L  defgenerate\_association\_rules():  s = []  r = []  length = 0  count = 1  inc1 = 0  inc2 = 0  num = 1  m = []  L= frequent\_itemsets()  print ("\nAssociation Rules:\n\n")  for list in L:  for l in list:  length = len(l)  count = 1  while count < length:  s = []  r = findsubsets(l,count)  count += 1  for item in r:  inc1 = 0  inc2 = 0  s = []  m = []  for i in item:  s.append(i)  for T in D:  if set(s).issubset(set(T)) == True:  inc1 += 1  if set(l).issubset(set(T)) == True:  inc2 += 1  if 100\*inc2/inc1 >= confidence:  for index in l:  if index not in s:  m.append(index)  print ("%d. %s ==> %s\nSupport: %d\nConfidence: %d\n" %(num,s, m, 100\*inc2/len(D), 100\*inc2/inc1))  num += 1  generate\_association\_rules() |
|  | **Output:**  Please enter support value in %: 50  Please enter confidence value in %: 50  Dataset: [['Cheese', 'Milk', 'Cookies\t\n'], ['Butter', 'Milk', 'Bread\t\n'], ['Cheese', 'Butter', 'Milk', 'Bread\n'], ['Butter', 'Bread\t\t']]  Frequent ItemSet: 1  [['Cheese'], ['Milk'], ['Butter']]  Frequent Itemset: 2  [['Cheese', 'Milk'], ['Butter', 'Milk']]  Frequent Itemset: 3  []  Association Rules:  1. ['Milk'] ==> ['Cheese']  Support: 50  Confidence: 66  2. ['Cheese'] ==> ['Milk']  Support: 50  Confidence: 100  3. ['Milk'] ==> ['Butter']  Support: 50  Confidence: 66  4. ['Butter'] ==> ['Milk']  Support: 50  Confidence: 66 |

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|  | **CO2160714.3 Assignment:** |
| 8 | Write programs to implement the following Classification methods.  (a)distance based  (b)statistics based  (c)tree based  (d)neural Network based |
|  | (a)  Input:  import pandas as pd  from sklearn.model\_selection import train\_test\_split  from sklearn.preprocessing import LabelEncoder  from sklearn.preprocessing import StandardScaler  from sklearn.neighbors import KNeighborsClassifier  from sklearn.metrics import classification\_report, confusion\_matrix  df=read\_csv(‘airport.csv’)  l=LabelEncoder()  df["elevation\_ft"]=l.fit\_transform(df["elevation\_ft"])  df["elevation\_ft"].unique()  df["ident"]=l.fit\_transform(df["ident"])  df["ident"].unique()  df["coordinates"]=l.fit\_transform(df["coordinates"])  df["coordinates"].unique()  x=df[["ident","coordinates"]]  y=df[["elevation\_ft"]]  scaler = StandardScaler()  scaler.fit(x\_train)  x\_train, x\_test, y\_train, y\_test = train\_test\_split(x, y, test\_size=0.20)  x\_train = scaler.transform(x\_train)  x\_test = scaler.transform(x\_test)  classifier = KNeighborsClassifier(n\_neighbors=5)  classifier.fit(x\_train, y\_train)  y\_pred = classifier.predict(x\_test)  print(confusion\_matrix(y\_test, y\_pred))  print(classification\_report(y\_test, y\_pred)) |
|  | Output:  prac8a.PNG |

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|  | (b) statistics based  Input:  import pandas as pd  from sklearn.model\_selection import train\_test\_split  from sklearn.preprocessing import LabelEncoder  from sklearn.preprocessing import StandardScaler  from sklearn.naive\_bayes import GaussianNB  df=read\_csv(‘airport.csv’)  l=LabelEncoder()  df["elevation\_ft"]=l.fit\_transform(df["elevation\_ft"])  df["elevation\_ft"].unique()  df["ident"]=l.fit\_transform(df["ident"])  df["ident"].unique()  df["coordinates"]=l.fit\_transform(df["coordinates"])  df["coordinates"].unique()  x=df[["ident","coordinates"]]  y=df[["elevation\_ft"]]  scaler = StandardScaler()  scaler.fit(x\_train)  x\_train, x\_test, y\_train, y\_test = train\_test\_split(x, y, test\_size=0.20)  x\_train = scaler.transform(x\_train)  x\_test = scaler.transform(x\_test)  M=GaussianNB()  M.fit(x\_train,y\_train)  y\_pred=M.predict(x\_test)  accuracy=accuracy\_score(y\_test,y\_pred)\*100  print(confusion\_matrix(y\_test, y\_pred))  print(accuracy\_score(y\_test, y\_pred)\*100) |
|  | Output:  prac9b.PNG |

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|  | (b)tree base  Input:  import pandas as pd  from sklearn.model\_selection import train\_test\_split  from sklearn.preprocessing import LabelEncoder  from sklearn.preprocessing import StandardScaler  from sklearn.tree import DecisionTreeClassifier  df=read\_csv(‘airport.csv’)  l=LabelEncoder()  df["elevation\_ft"]=l.fit\_transform(df["elevation\_ft"])  df["elevation\_ft"].unique()  df["ident"]=l.fit\_transform(df["ident"])  df["ident"].unique()  df["coordinates"]=l.fit\_transform(df["coordinates"])  df["coordinates"].unique()  x=df[["ident","coordinates"]]  y=df[["elevation\_ft"]]  scaler = StandardScaler()  scaler.fit(x\_train)  x\_train, x\_test, y\_train, y\_test = train\_test\_split(x, y, test\_size=0.20)  x\_train = scaler.transform(x\_train)  x\_test = scaler.transform(x\_test)  clf\_gini = DecisionTreeClassifier(criterion = "gini",random\_state = 100,max\_depth=3, min\_samples\_leaf=5)  clf\_gini.fit(x\_train, y\_train)  clf\_entropy = DecisionTreeClassifier(criterion = "entropy", random\_state = 100,max\_depth = 3, min\_samples\_leaf = 5)  clf\_entropy.fit(x\_train, y\_train)  y\_pred = clf\_object.predict(X\_test)  print("Predicted values:")  print(y\_pred)  print("Confusion Matrix: ",  confusion\_matrix(y\_test, y\_pred))    print ("Accuracy : ",  accuracy\_score(y\_test,y\_pred)\*100)    print("Report : ",  classificatiprint("Confusion Matrix: ",  confusion\_matrix(y\_test, y\_pred))  x, y, x\_train, x\_test, y\_train, y\_test = splitdataset(data)  clf\_gini = train\_using\_gini(x\_train, x\_test, y\_train)  clf\_entropy = tarin\_using\_entropy(x\_train, x\_test, y\_train)  y\_pred\_gini = prediction(X\_test, clf\_gini)  cal\_accuracy(y\_test, y\_pred\_gini) |
|  | Output:  prac9ca.PNG  prac9cb.PNG |

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|  | (C)  Input:  import pandas as pd  import numpy as np  import matplotlib.pyplot as plt  import sklearn  from sklearn.neural\_network import MLPClassifier  from sklearn.neural\_network import MLPRegressor  from sklearn.model\_selection import train\_test\_split  from sklearn.metrics import mean\_squared\_error  from math import sqrt  from sklearn.metrics import r2\_score  from sklearn.preprocessing import LabelEncoder  from sklearn.neural\_network import MLPClassifier  from sklearn.metrics import classification\_report,confusion\_matrix  df = pd.read\_csv('covid.csv')  encoder=LabelEncoder()  df=df.apply(encoder.fit\_transform).astype(float)  target\_column = ['Diabetes']  predictors = list(set(list(df.columns))-set(target\_column))  df[predictors] = df[predictors]/df[predictors].max()  df.describe().transpose()  X = df[predictors].values  y = df[target\_column].values  X\_train, X\_test, y\_train, y\_test = train\_test\_split(X, y, test\_size=0.30, random\_state=40)  print(X\_train.shape); print(X\_test.shape)  mlp = MLPClassifier(hidden\_layer\_sizes=(8,8,8), activation='relu', solver='adam', max\_iter=500)  mlp.fit(X\_train,y\_train)  predict\_train = mlp.predict(X\_train)  predict\_test = mlp.predict(X\_test)  print(confusion\_matrix(y\_train,predict\_train))  print(classification\_report(y\_train,predict\_train)) |
|  | Output:  Capture.PNG |

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| 9 | Write a program to implement following Prediction methods: (a)Linear (b)logistic regression |
|  | (a)Linear regression  Input:  import pandas as pd  from sklearn.preprocessing import LabelEncoder  from sklearn.model\_selection import train\_test\_split  from sklearn.linear\_model import LinearRegression  from sklearn.metrics import r2\_score  df=pd.read\_csv("airport.csv")  encoder=LabelEncoder()  df['ident']=encoder.fit\_transform(df['ident'])  df['ident'].unique()  df['elevation\_ft']=encoder.fit\_transform(df['elevation\_ft'])  df['elevation\_ft'].unique()  x=df[["ident"]]  y=df[["elevation\_ft"]]  x\_train,x\_test,y\_train,y\_test=train\_test\_split(y,y,test\_size=0.5)  R=LinearRegression()  R.fit(x\_train,y\_train)  pred=R.predict(x\_test)  r2\_score(pred,y\_test)  print('Coefficients: ', R.coef\_)   |  | | --- | |  | |
|  | Output:  prac10.PNG |

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|  | (b)logistic Regression  Input:  import pandas as pd  from sklearn.preprocessing import LabelEncoder  from sklearn.model\_selection import train\_test\_split  from sklearn.linear\_model import LogisticRegression  from sklearn.metrics import classification\_report  df=pd.read\_csv("airport.csv")  encoder=LabelEncoder()  df['ident']=encoder.fit\_transform(df['ident'])  df['ident'].unique()  df['elevation\_ft']=encoder.fit\_transform(df['elevation\_ft'])  df['elevation\_ft'].unique()  x=df[["ident"]]  y=df[["elevation\_ft"]]  x\_train,x\_test,y\_train,y\_test=train\_test\_split(y,y,test\_size=0.5)  log=LogisticRegression()  log.fit(x\_train,y\_train)  pred=log.predict(X\_test)  classification\_report(pred,Y\_test) |
|  | Outpu:  prac10a.PNG |

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|  | **CO2160714.4 Assignment:** |
| 10 | Using weka tool for your data set, make the following table for classification and clustering algorithms. |
|  | Input:  Classification algorithm:   |  |  |  |  |  |  | | --- | --- | --- | --- | --- | --- | | no | Algorithm | Precision | Recall | F-measure | Execution Time | | 1 | Randomforest | 0.993 | o.979 | 0.986 | 0.25 | | 2 | Decision tree | 0.910 | 0.979 | 0.982 | 0.6 | | 3 | Naivebayes | 0.929 | 0.930 | 0.926 | 0.4 |   Randomforest:  randomforest.PNG  Decision tree:  desicion tree.PNG  Naivebayes:  naivebayes.PNG |
|  | clustering algorithms:  K-mean:  cluster1kmean.PNG  Makedenstinybase:  cluster2makedensitybased.PNG  Filtercluster:  Clusterfiltered.PNG |

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|  | **CO2160714.5 Assignment:** |
| 11 | Implement any 2 **unsupervised clustering** algorithms. |
|  | (1)K-mean:  kmean1.PNG |
|  | kmeans2.PNG  (2) hierarchical clustering:  hierarchical clusterer.PNG |