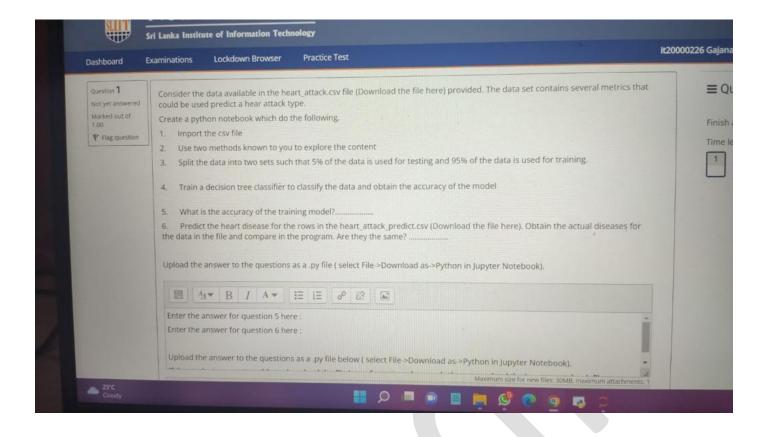
DS LAB TEST 01

Q & A





Step 01-import necessary library

- Import pandas as pd
- Import numpy as np
- from sklearn.tree import DecisionTreeClassifier
- from sklearn.model_selection import train_test_split
- Import csv file
 - a. Df=pd.read_csv("heart_attack.csv")
- 2. Explore the data
 - a. Display(df.head())
 - b. Display(df.info())
 - c. display(df.shape)
 - d. column=['<column name 1>','<column name 2>','<column name 3>']df[column].describe()
- 3. x=df.iloc[:,[<attribute column range>]]

y=df.iloc[:, <labeled column>]

X train,X test,y train,y test=train test split(x,y,test size=0.05)

Or

X train,X test,y train,y test=train test split(x,y,train size=0.95)

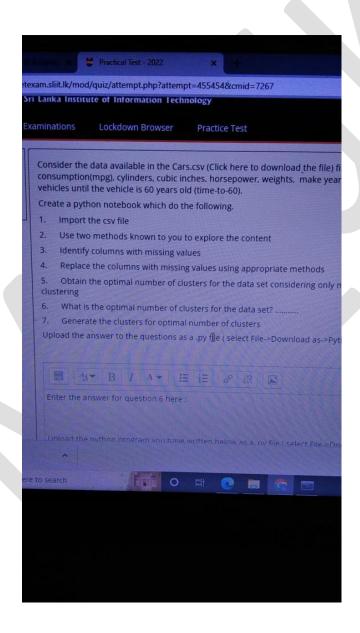
4. hart classifire=DecisionTreeClassifire(random state=0)

```
hart_classifire.fit(X_train,y_train)
accuracy=hart_classifire.score(X_test,y_test)
print("Accuracy of building classifier:",accuracy)
```

- 5. Aquracy=(out put of the above print command)
- 6. Predict=hart_classifier.predict(X_test[1:10])

Print("Predicted disease are :",predict)

Compare above out put and hart_attack_disease.csv file out put and give the answer

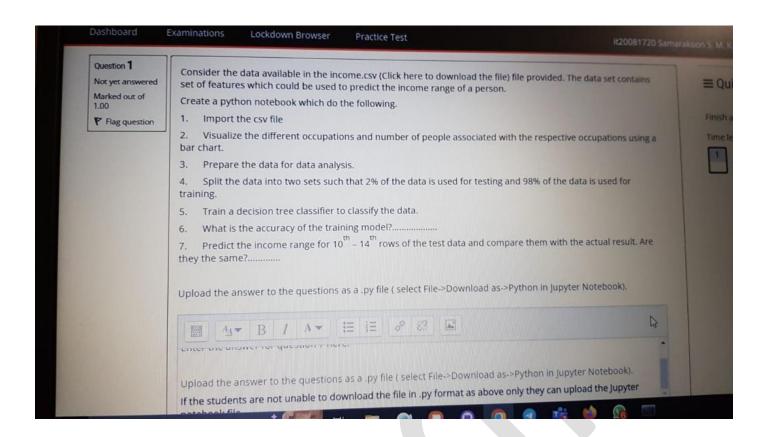


```
Import pandas as pd
   Import numpy as np
   From matplotlib import pyplot as plt
   Import seaborn as sbn
   From sklearn.preprocessing import StanderdScaler
   From sklearn.clustrt import KMeans
   Sns.set(not necessary )

    Df=pd.read_csv("Cars.csv")

2. Display(df.head())
   Display(df.info())
3. Df.isnull() or
   Df.isna()
               or
   Df.isnull().sum()
   Df,isna().sum()
   Df.info()
               or
4. Count how many missing vale are there.
           Df['<column name>'].isnull().sum()
   Fill the missing value
           Df['column name'].fillna(Unknown,inplace=True)
                           Or
           df['column name].fillna(0,inplace=True)
                           or
           df['column name].fillna(df['column name'].mean,inplace=True)
                           or
           df['column name'].fillna(df['column name'].median,inplcae=True)
                           or
           df['column name'].fillna(df['column name'].mode,inplcae=True)
```

```
5. New_df=df.iloc[:,[<column range>]]
   Ss=StanderdScaler()
   new_data = pd.DataFrame(ss.fit_transform(new_df), columns=['x axis column name','y axis column name'])
   wcss=[]
   for I in range(1:10)
           kmeans=KMeans(i)
           kmeans.fit(new_df)
           wcss_iter=kmeans.inertie_
           wcss.append(wcss_iter)
   cluster=range(1:10)
   plt.plot(cluster,wcss)
   plt.title('Elbow method')
   plt.xlable('number of clusters')
   plt.ylable('Within-cluster Sum of Squares')
   considering the elbow graph find the optimal number of cluster
7. kmeans=KMeans(<optimal no of cluster>)
   kmeans.fit(new_df)
   identify_cluster=kmeans.fit_predict(new_df)
   cluster=df.copy()
   cluster['ClusterNo']=identify_cluster
   plt.scatter(cluster['x axis column name'],cluster['y axis column
   name'],c=cluster['ClusterNo'],cmap='rainbow')
```



Import necessary library,

Import pandas as pd

Import numpy as np

from sklearn.tree import DecisionTreeClasifire

from sklearn.model_selection import train_test_split

df=pd.read csv("income.csv")

• Chart=df['occupations'].value_counts()

Chart.plot(kind='bar')

- Find null value and fill the null value using appropriate method and colum contain character those are convert to number using factorize method
- X=df.iloc[: , <column name>]

Y=df.iloc[:, <column name>]

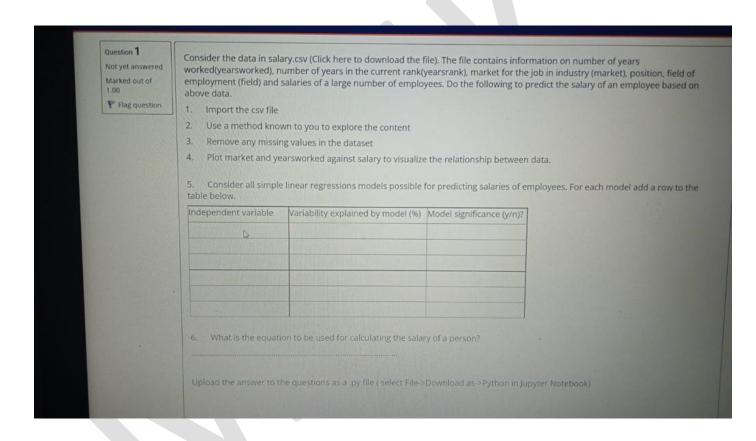
X_train,X_test,y_train,y_test=train_tset_split(X,Y,test_size=0.02)

Or

 $X_train, X_test, y_train, y_test=train_tset_split(X, Y, train_size=0.98)$

- Income_classifire=DecisionTreeClasfire(random_state=0)
 Income_classifire.fit(X_train,y_train)
- Income_classifire.score(X_test,y_test)
- Income_classifire.predict(X_test[10:14])

Y_test[10:14]



First import necessary library's

Import pandas as pd

Import numpy as np

From matplotlib import pyplot as plt

Import statsmodels.api as sm

From sklearn.linear_model import LinearRegression

- Df=pd.read_csv("salary.csv")
- 2. Df.head()

Df.info()

- 3. Df.dropna(inplace=True)
- 4. x=df.iloc[: ,<column index range>].values(plot market)

y=df.iloc[: , [column index]].values (salary)

Plt.scatter(x,y)

Model=LinearRegression()

Model.fit(x,y)

Const=sm.add_constant(x)

Model=sm.OLS(y,x).fit()

Model.summary()

5.

OLS Re	OLS Regression Results						
De	p. Variable:		у	R-s	quared:	0.957	
	Model:		OLS	Adj. R-s	quared:	0.955	
	Method:	Least Sq	uares	F-9	statistic:	622.5	
	Date:	Fri, 08 Sep	2023 I	Prob (F-s	tatistic):	1.14e-20	
	Time:	12:	09:56	Log-Lik	elihood:	-301.44	
No. Ob	servations:		30		AIC:	606.9	
Df	Residuals:		28		BIC:	609.7	
	Df Model:		1				
Covariance Type: nonrobust							
	coef	std err	t	P> t	[0.025	0.97	5]
const	2.579e+04	2273.053	11.347	0.000	2.11e+04	3.04e+0	04
x 1	9449.9623	378.755	24.950	0.000	8674.119	1.02e+0	04
C	Omnibus: 2	2.140 D u	ırbin-Wa	tson: 1	.648		

Prob(Omnibus): 0.343 Jarque-Bera (JB): 1.569

Skew: 0.363 **Prob(JB):** 0.456

Kurtosis: 2.147 **Cond. No.** 13.2

Independent variable	Variability explained by model(%)	Model Significance(Y/N)?	
	(R-squared)		
yearworked	0.957	yes	

Ptop(f-statistic) value is 1.14e-20=1.14*10⁻²⁰, this is more closer than to 0, therefore model highly significant(Yes)

Other row fill like this

Consider the data available in the examScores.csv file (Download the file here) provided. The data set include the marks of three exams given to the students to predict the marks of final exam.

Create a python notebook which do the following.

- 1. Import the csv file
- 2. Use a method known to you to explore the content
- 3. Assume that the following criteria is used to calculate grades for exams

Criteria	Grade	
Marks<45	D	
45<=Marks<55	С	
55<=Marks<75	В	
Marks>=75	A	

Write a function to calculate the final marks in each grade and show how many students have got marks in each range as a percentage

4. Develop the multiple regression model to predict the final marks of a student.

What is the most suitable equation that could be used to predict final marks of students based on the above?

3.

Def FinalMark (mark):

If mark < 45:

Return 'D'

If mark >=45 and mark < 55:

Return 'C'

If mark >=55 and mark<75:

Return 'B'

Else:

Return 'A'

Mark=df['criteria'].apply(FinalMark)

Mark.value_counts()

F=plt.figure()

Mark.value_counts().plt.pie(autopct='%1.0f%%',)

Plt.title('Student Mark')

