Practical No.1

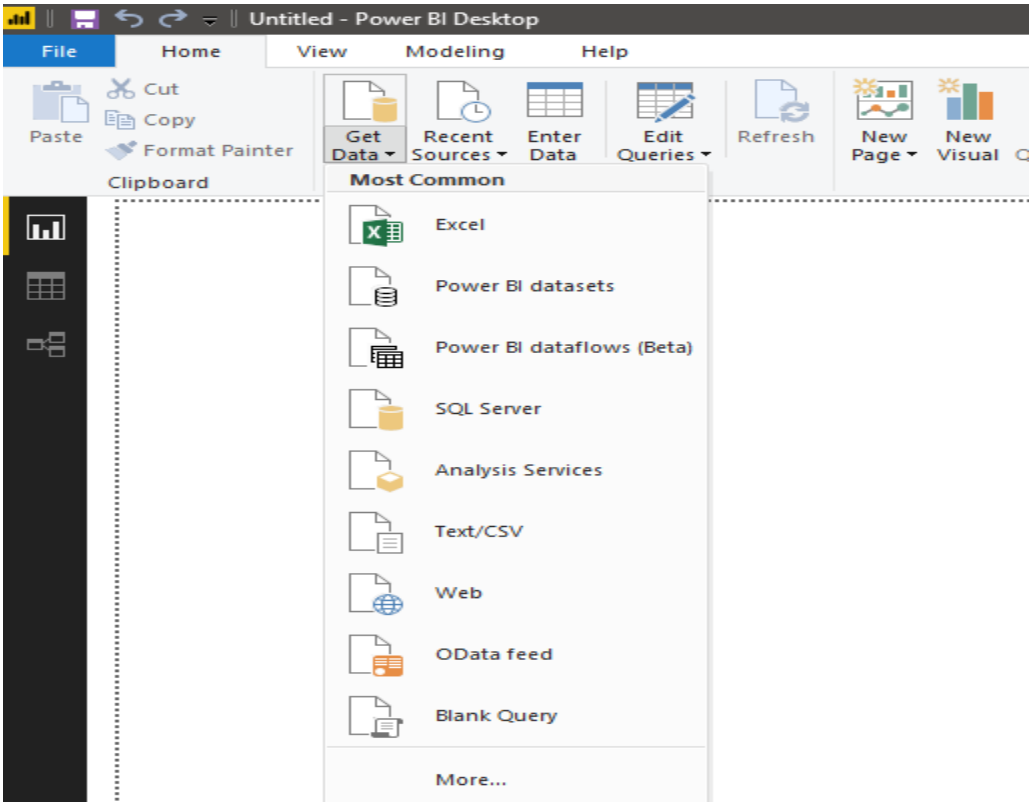
Aim: Import legacy data from Excel and load in the targeted system.

Step 1: For this first create Excel file.

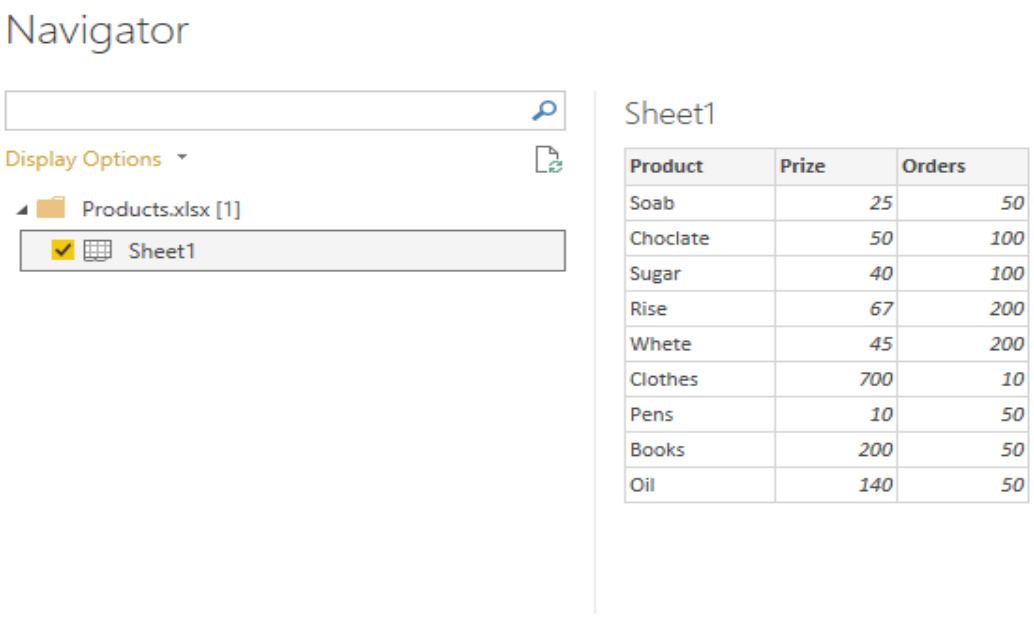
	A	B	C	D
1	Product	Prize	Orders	
2	Soab	25	50	
3	Choclate	50	100	
4	Sugar	40	100	
5	Rise	67	200	
6	Whete	45	200	
7	Clothes	700	10	
8	Pens	10	50	
9	Books	200	50	
10	Oil	140	50	
11				

Step 2: Open Power Bi Desktop.

Step 3: Go to Home -> Get data -> Excel and browse your excel file.



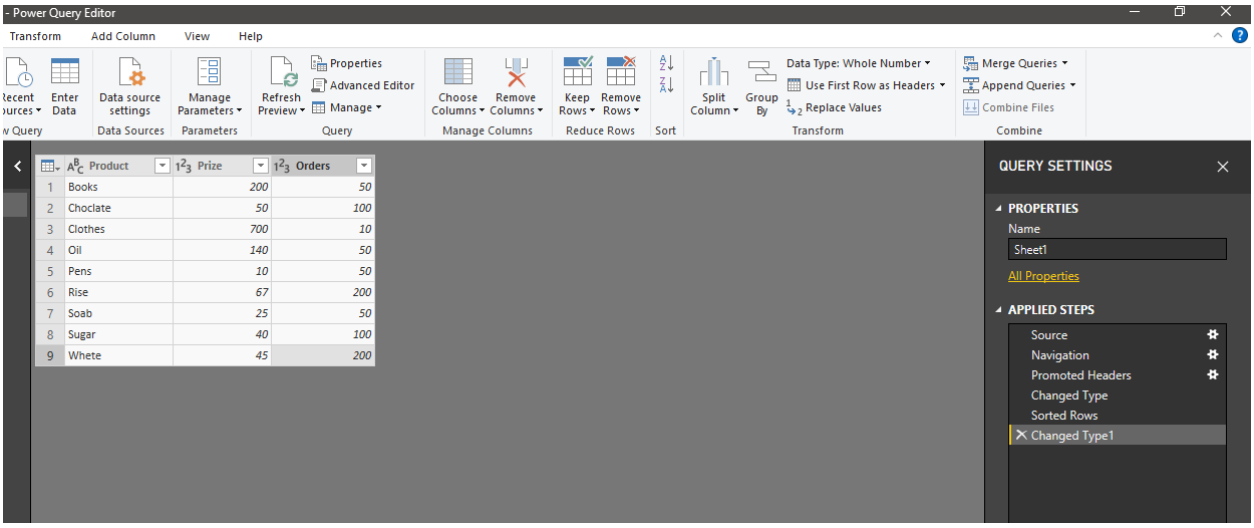
Step 4: In the navigation tab, select your (sheet1) from database (products.xlsx).



Sheet1

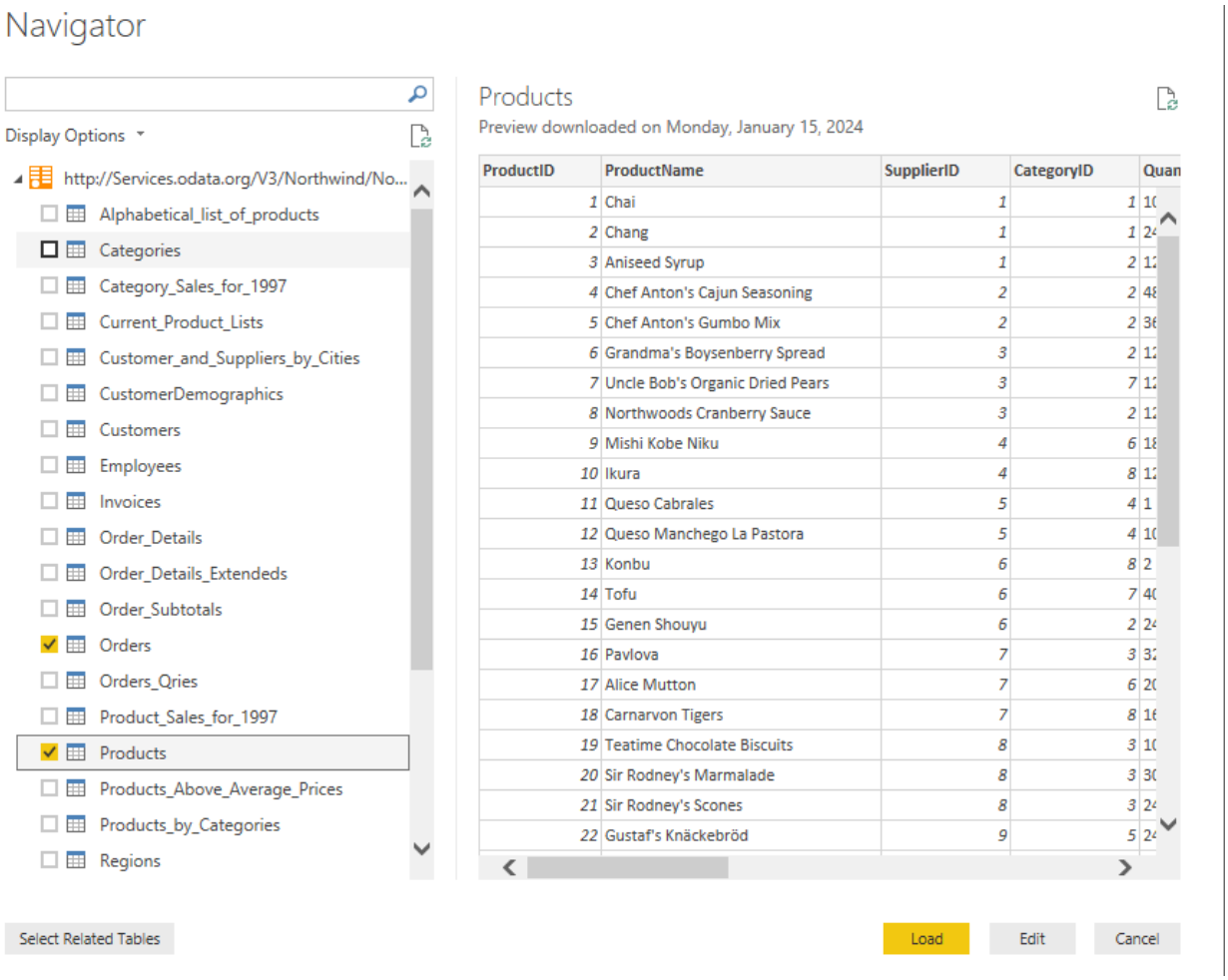
Product	Prize	Orders
Soab	25	50
Choclate	50	100
Sugar	40	100
Rise	67	200
Whete	45	200
Clothes	700	10
Pens	10	50
Books	200	50
Oil	140	50

Step 5: You will obtain this screen for queries.

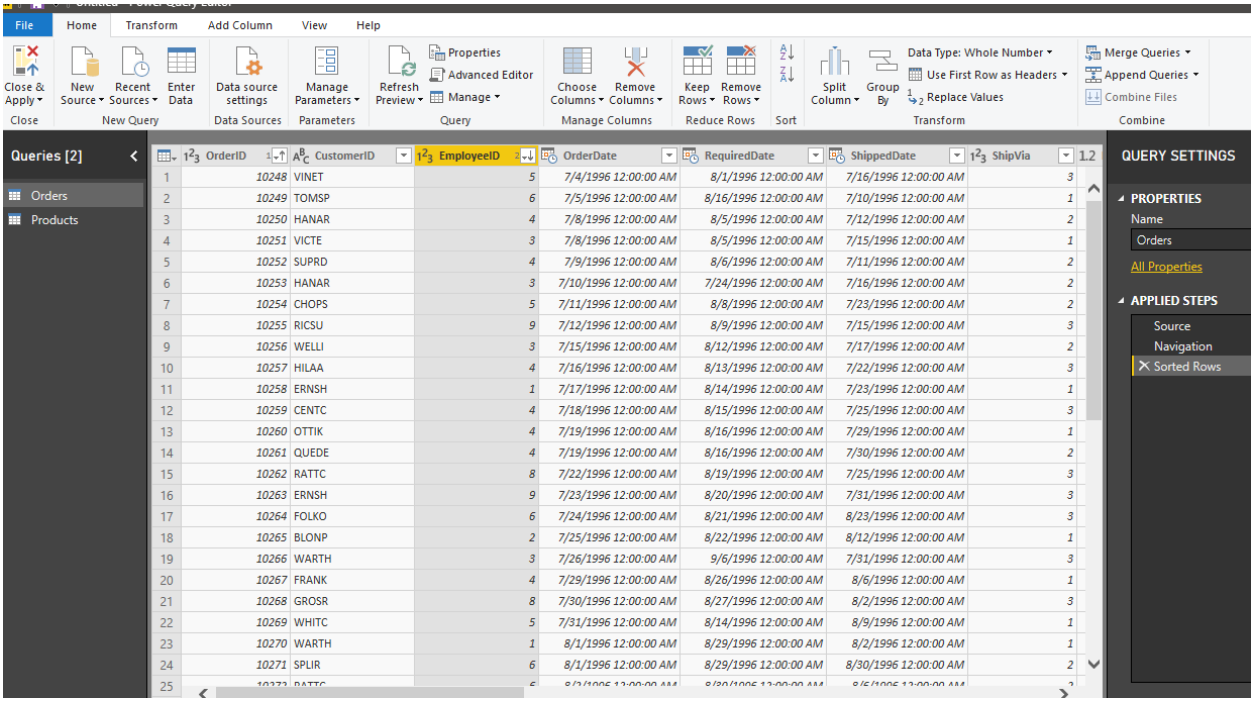


Step 6: Step 3: Go to Home -> Get data -> Odata feed and type url.

Step 7: From server Select orders and products table -> click on load



Step 8: You will obtain this screen for queries.

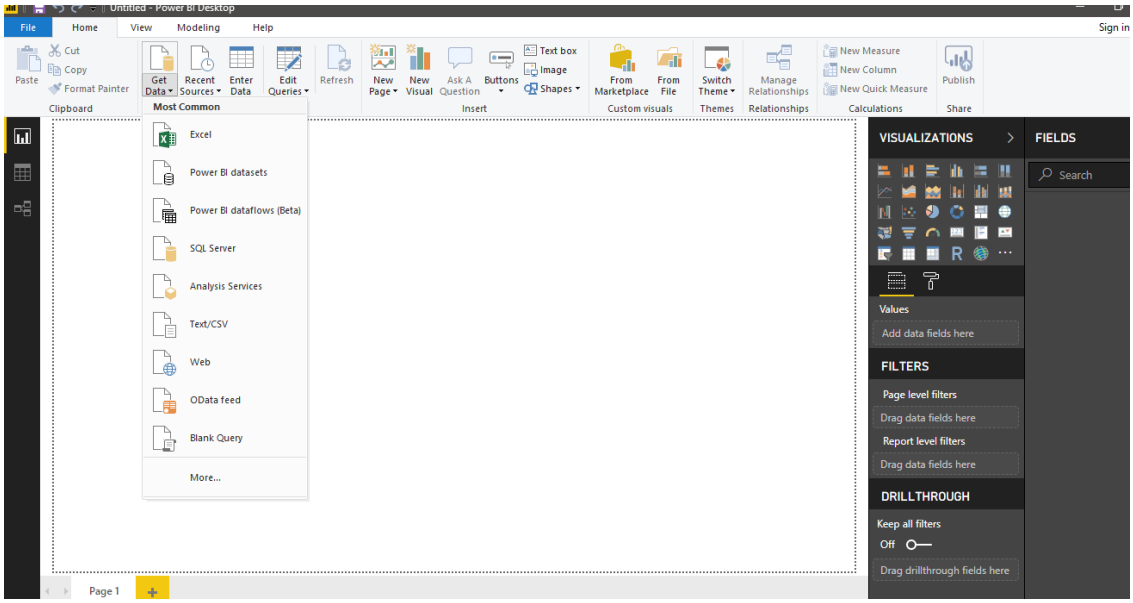


PRACTICAL 2

Aim:- Perform the extraction transformation and loading(ETL) process to construct database in the SQL server/Power BI.

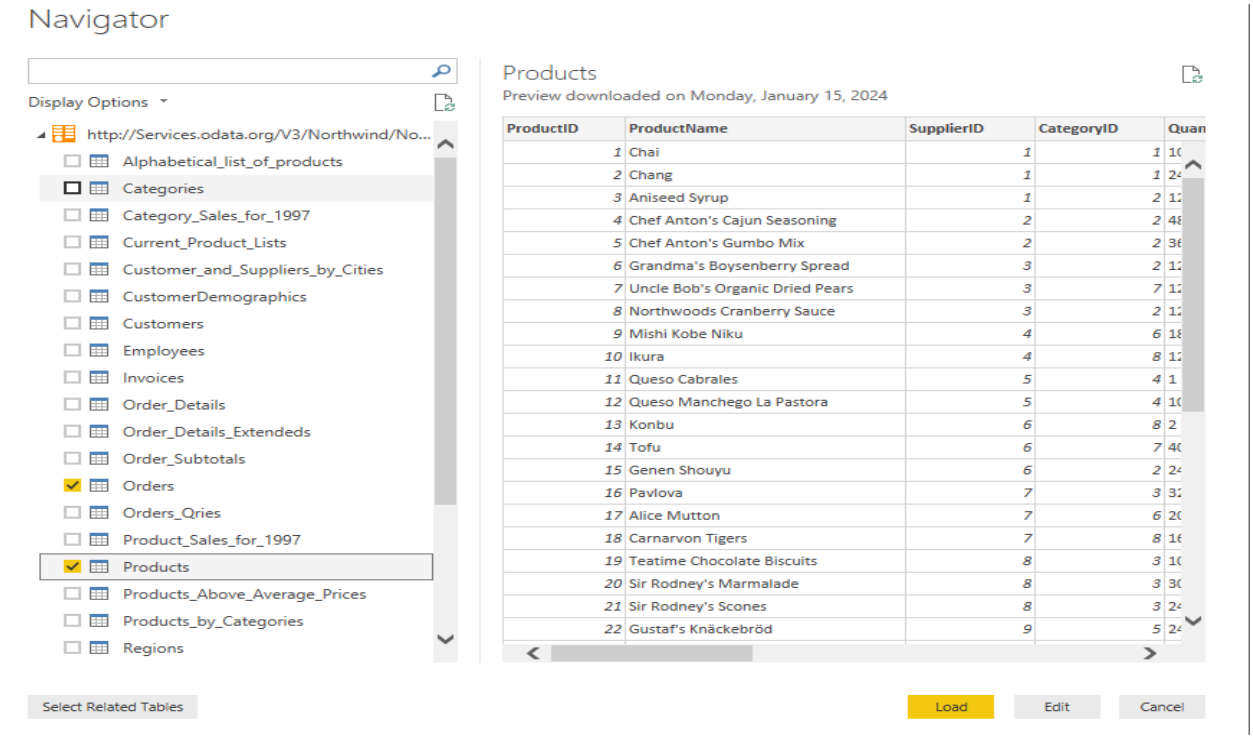
Step 1: Open Power Bi desktop.

Step 2: Go to Home -> Get Data -> Odata Feed

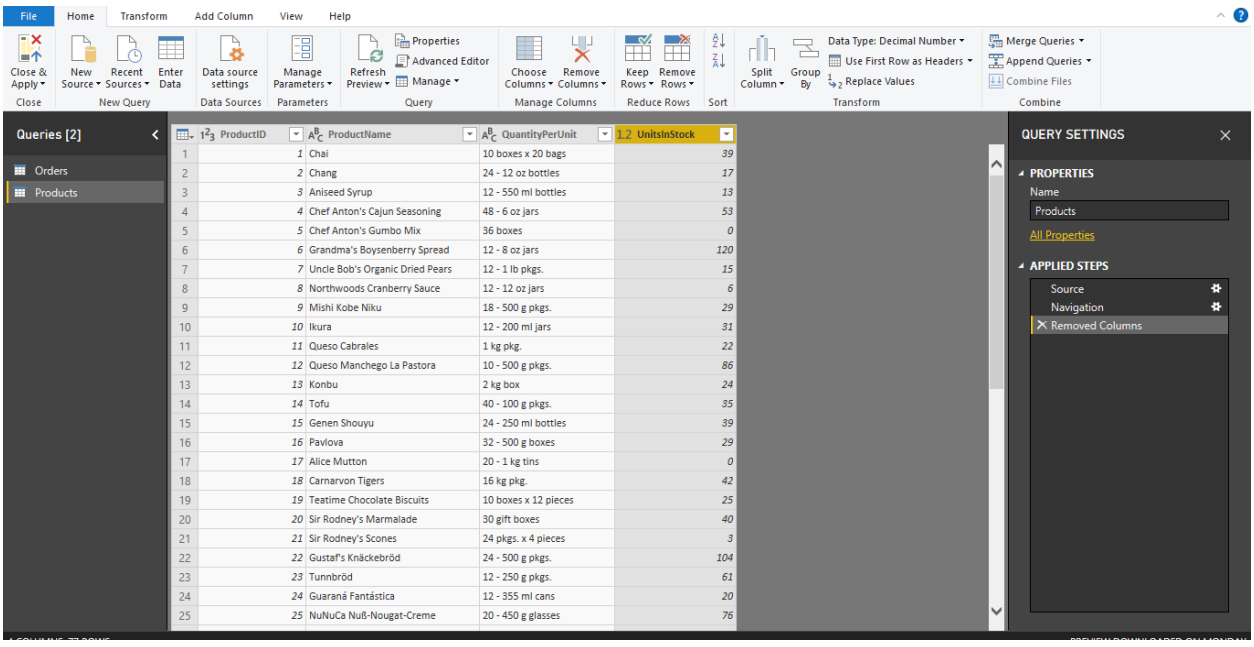


Step 3: Paste Url

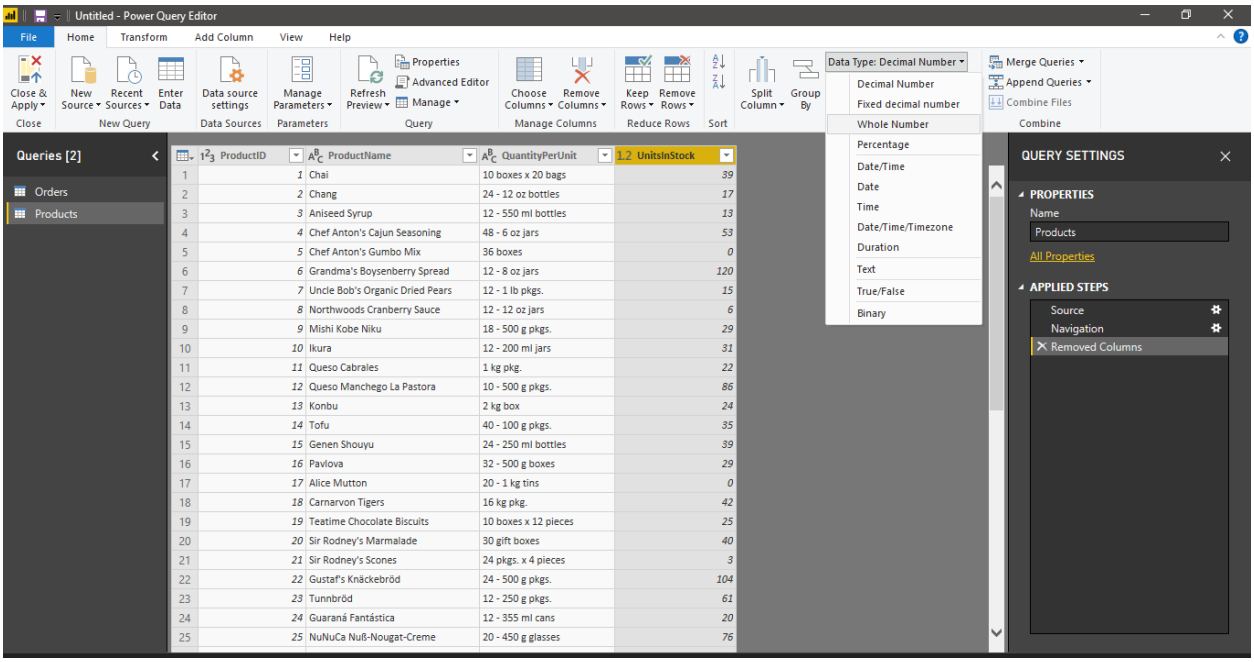
Step 4: From server Select order and products table -> click on edit .



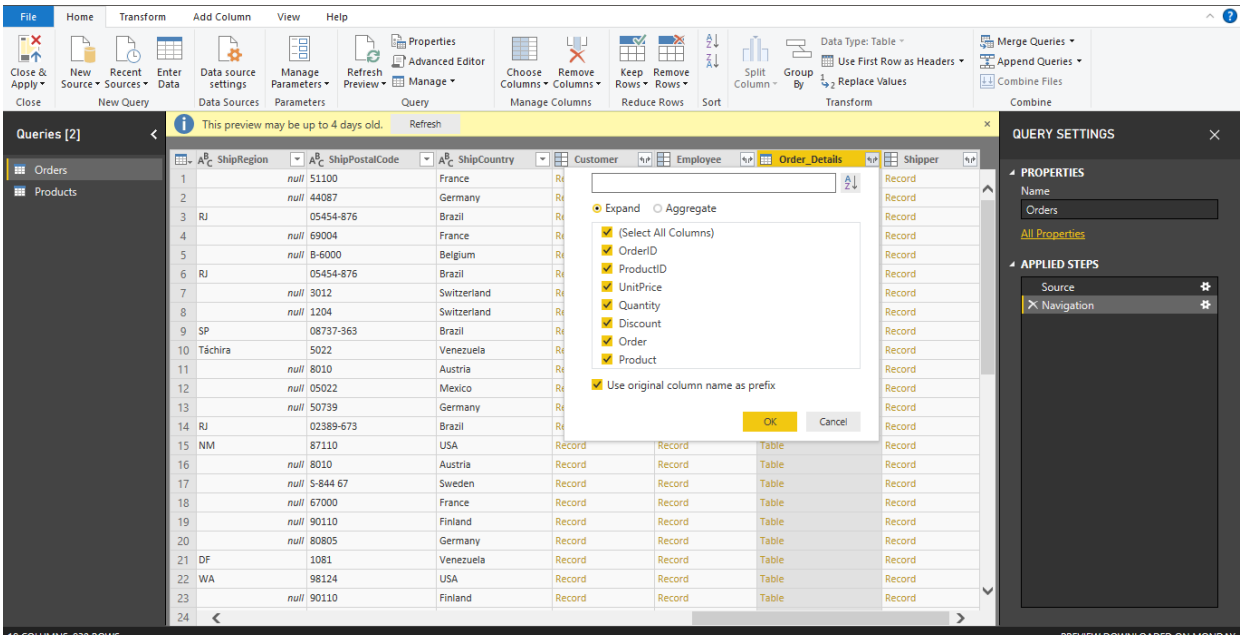
Step 5: Select Product Table -> Delete Unrequired Columns -> Keep 4 Required Column.



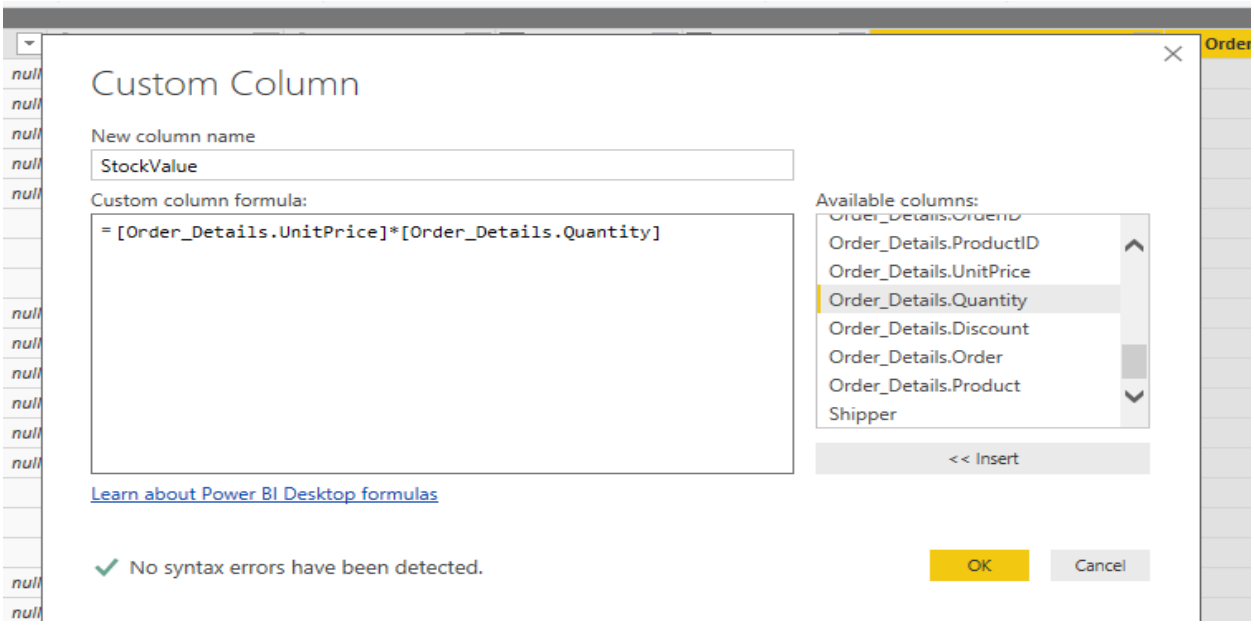
Step 6 : Change UnitsInStock Column to data type from decimal to whole number.



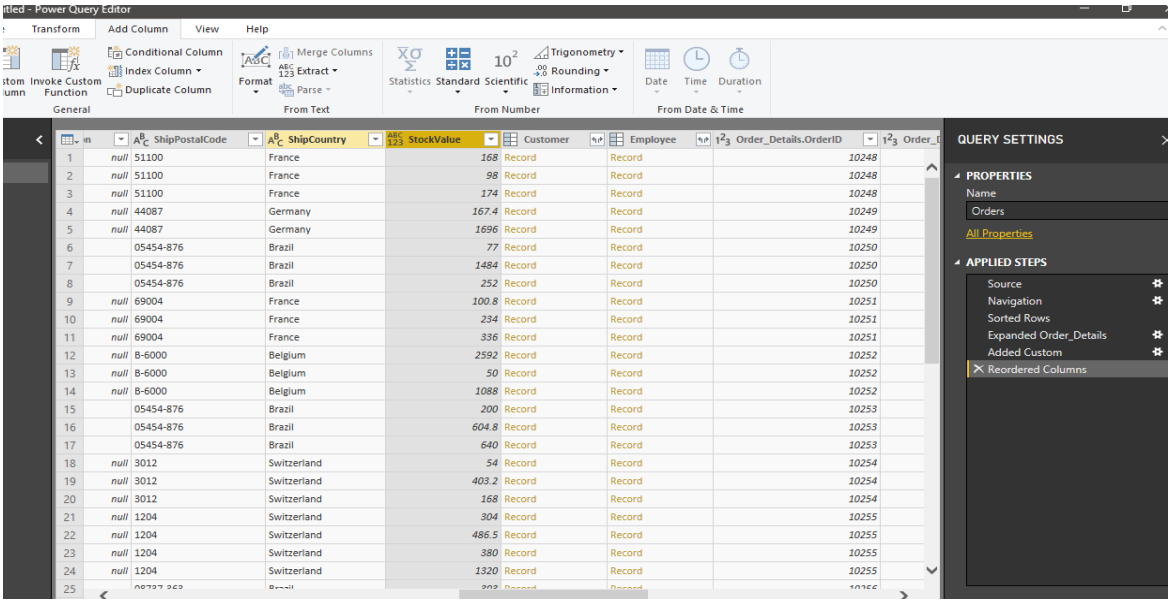
Step 7: Select Order Table -> Go to column OrderDetails -> Expand Column -> Click OK



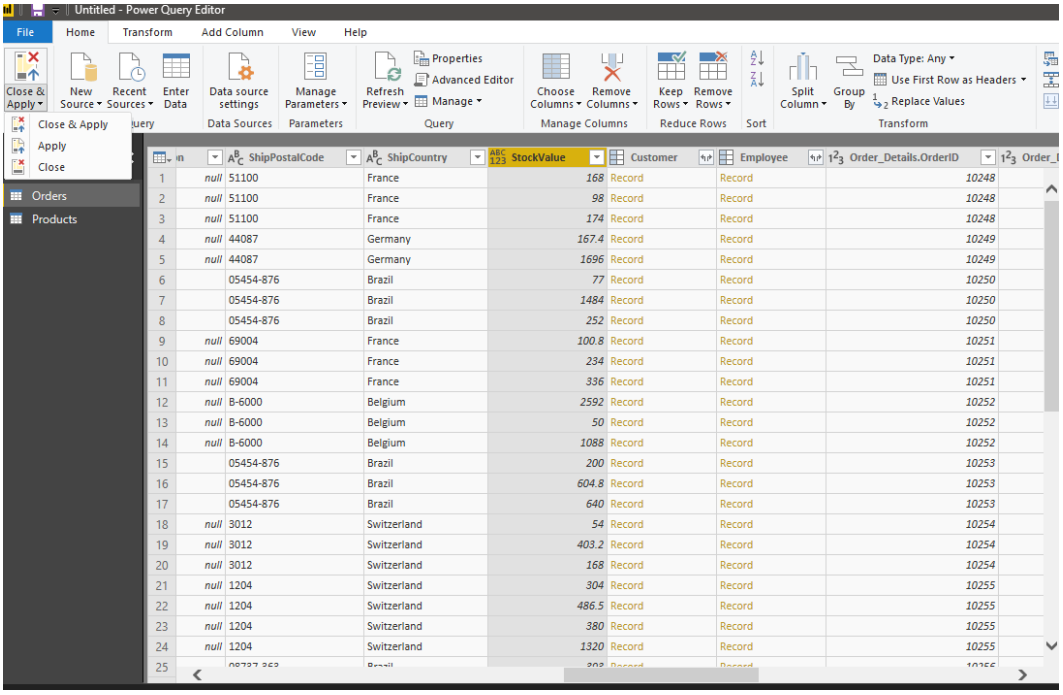
Step 8: Go to Add Column -> select Custom column -> Give name StockValue and Insert UnitPrice and Quantity -> click ok.



Step 9: Drag and drop the column StockValue.



Step 10: select Close and Apply.



Practical No 3

Aim: Create the cube with suitable dimension and fact tables based on OLAP.

1. Creating Data Warehouse Let us execute our T-SQL Script to create data warehouse with fact tables,

dimensions and

populate them with appropriate test values. Download T-SQL script attached with this article for creation of Sales Data Warehouse or

download from this article "Create First Data Warehouse" and run in your SQL Server. Downloading "Data Warehouse SQLScript.zip" from the article. https://www.codeproject.com/Articles/652108/Create-First-DataWareHouse

2. After downloading extract file in folder. Follow the given steps to run the query in SSMS (SQL Server Management Studio). Open SQL Server Management Studio 2012

3.Connect Database Engine

Click Connect. Open New Query editor

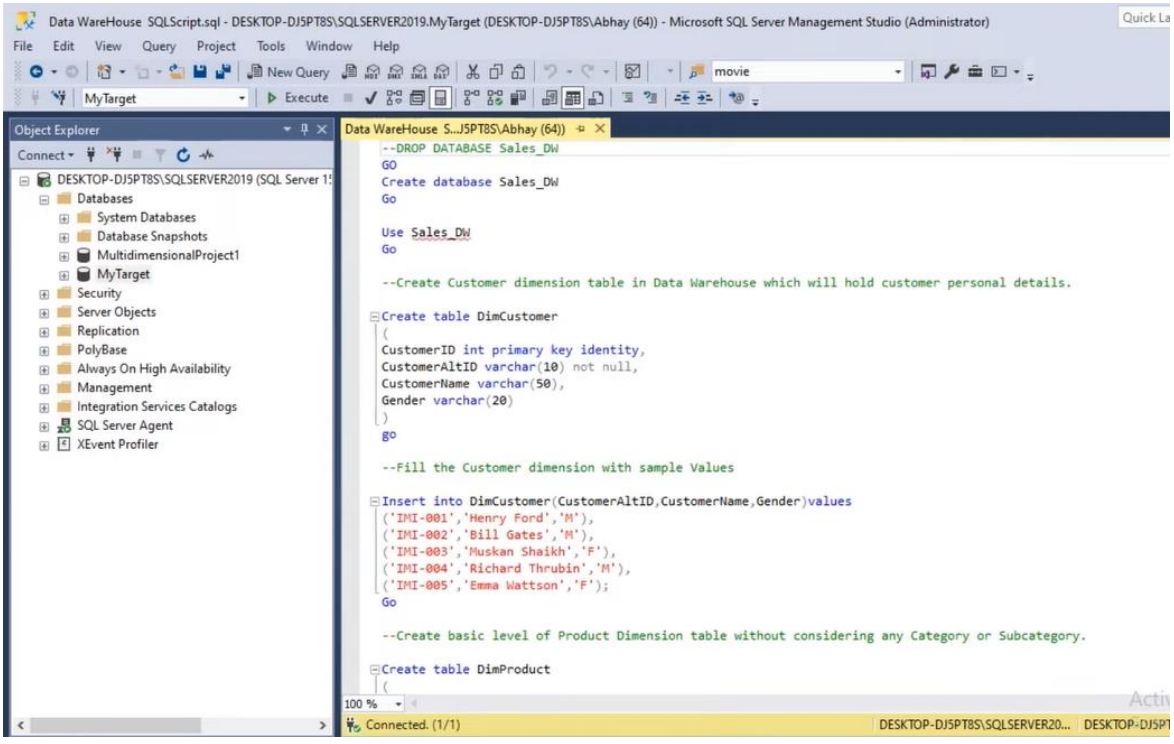
4. Copy paste Scripts given below in various steps in new query editor window one by one

5. To run the given SQL Script, press F5

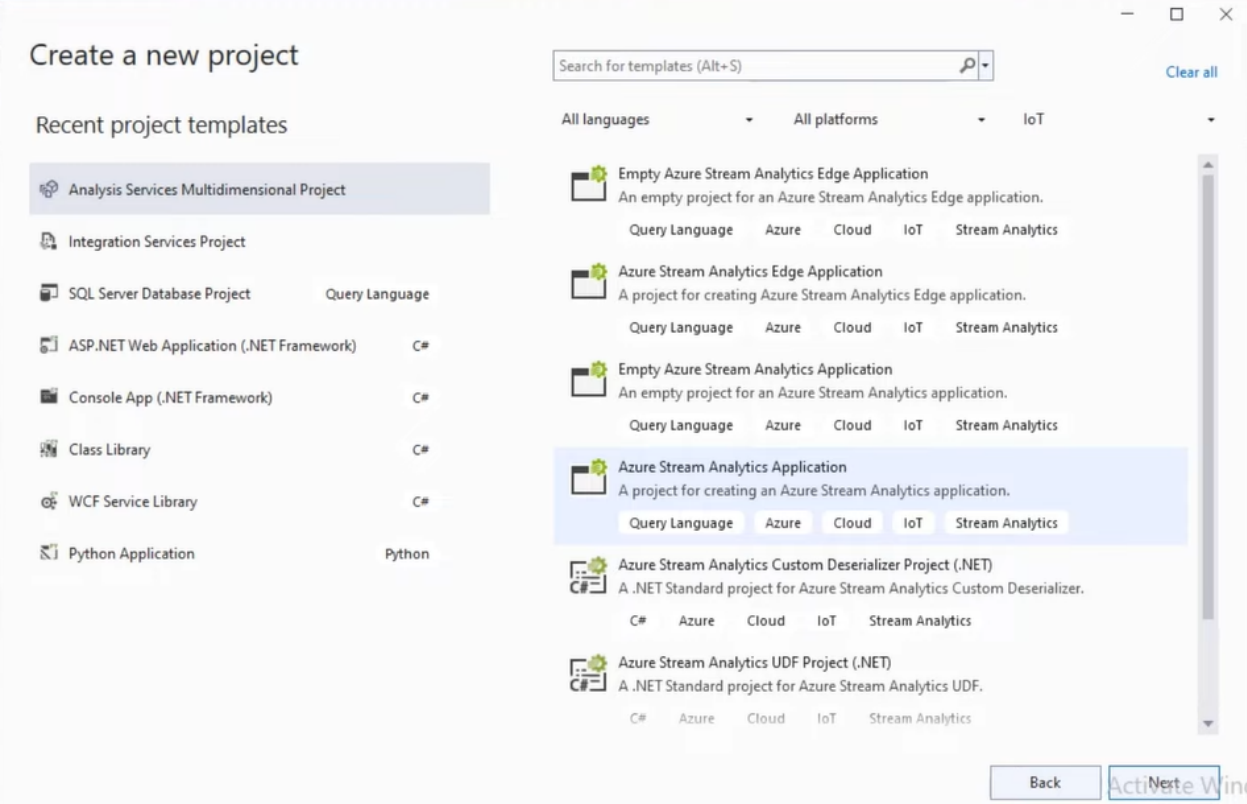
6. It will create and populate "Sales DW" database on your SQL Server

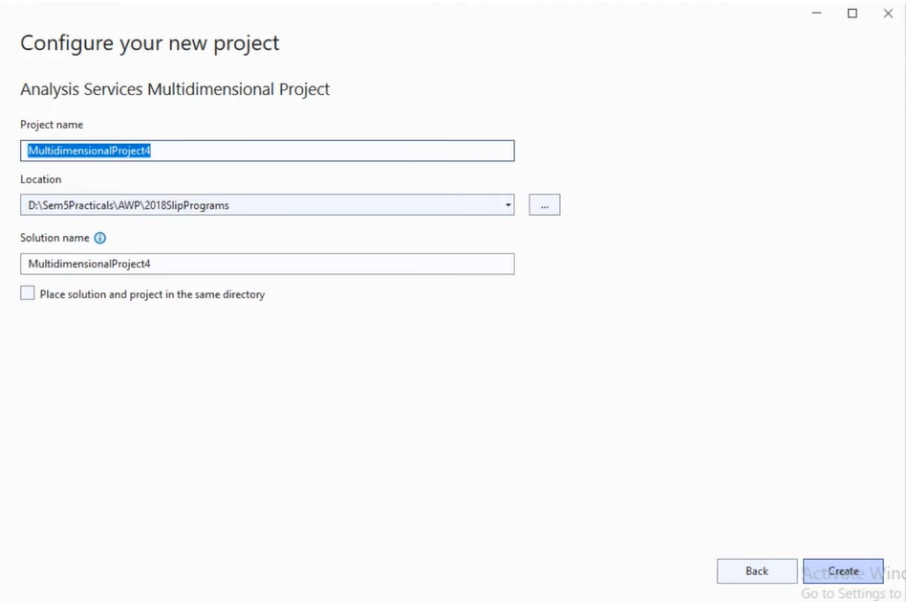
OR

1. Go to the extracted SQL file and double click on it. 2. New SQL Query Editor will be opened containing

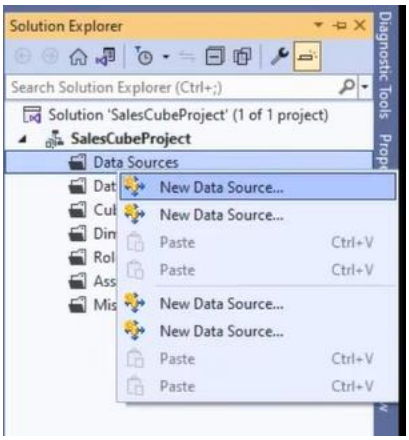


Step 2: Start SSDT environment and create New Data Source. Go to SQL Server Data Tools → Right click and run as administrator Click on File → New Project In Business Intelligence → Analysis Services Multidimensional Project → appropriate project name → click OK.

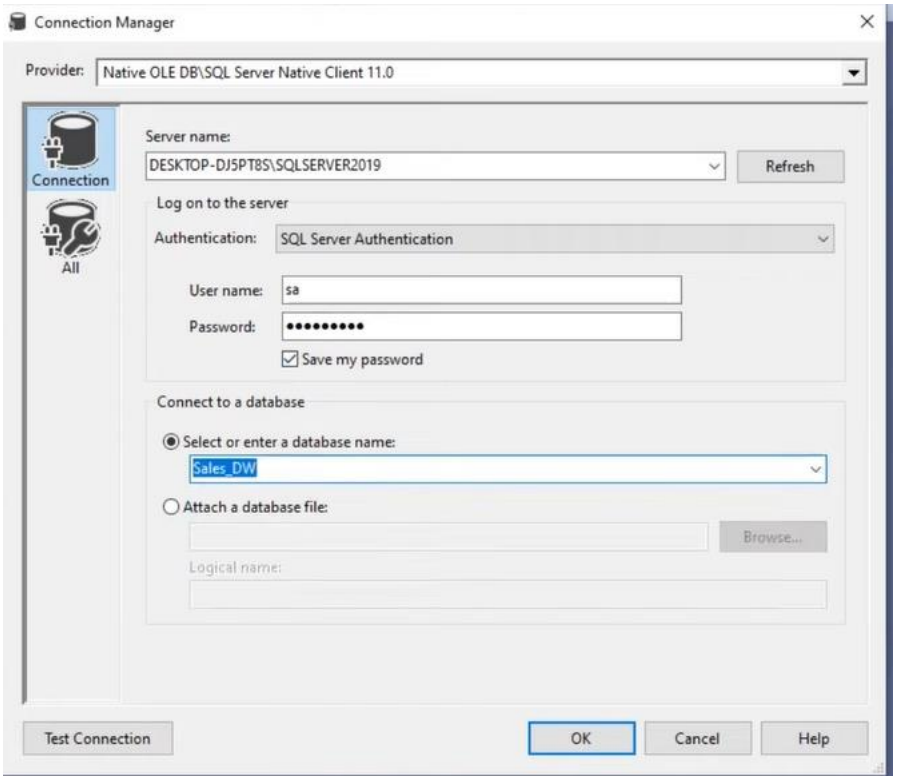




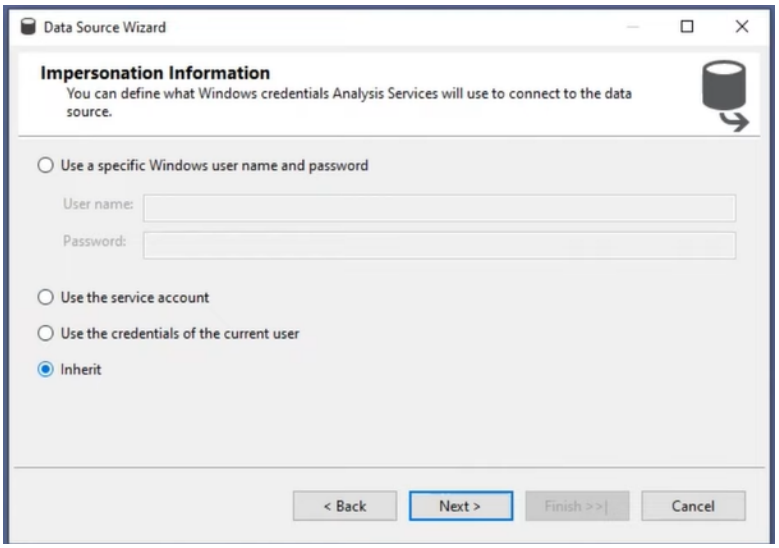
Right click on Data Sources in solution explorer → New Data Source.



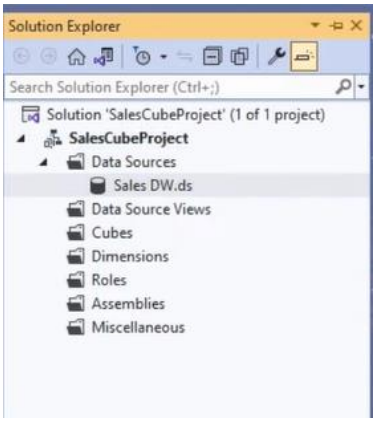
Select Server → Name select Use SQL Server Authentication → Select or enter a database name (Sales DW). Click Next



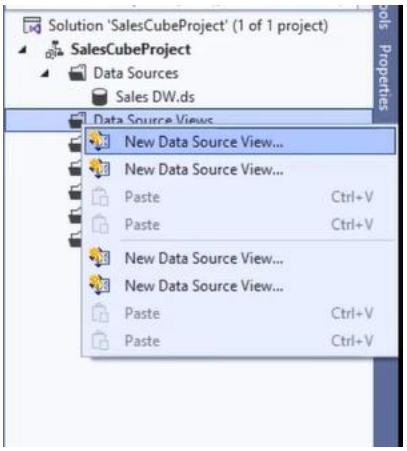
Select Inherit → Next



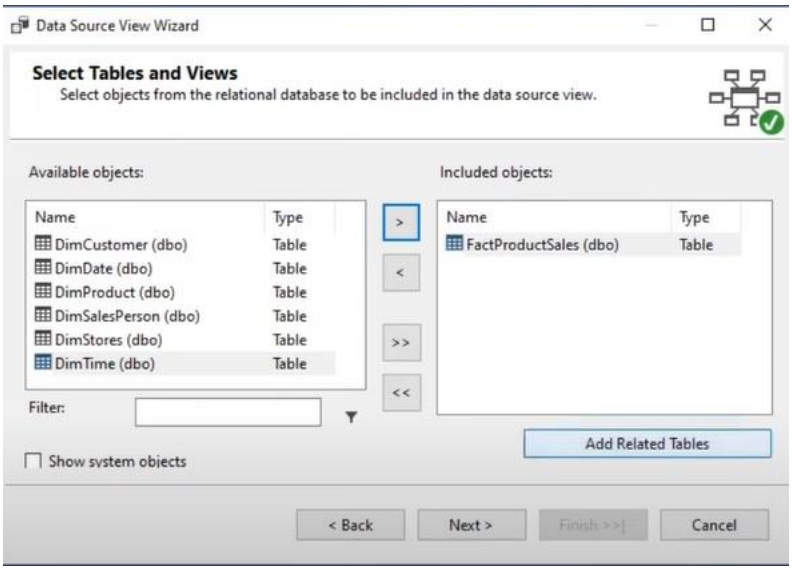
Sales_DW.ds gets created under Data Sources in Solution Explorer

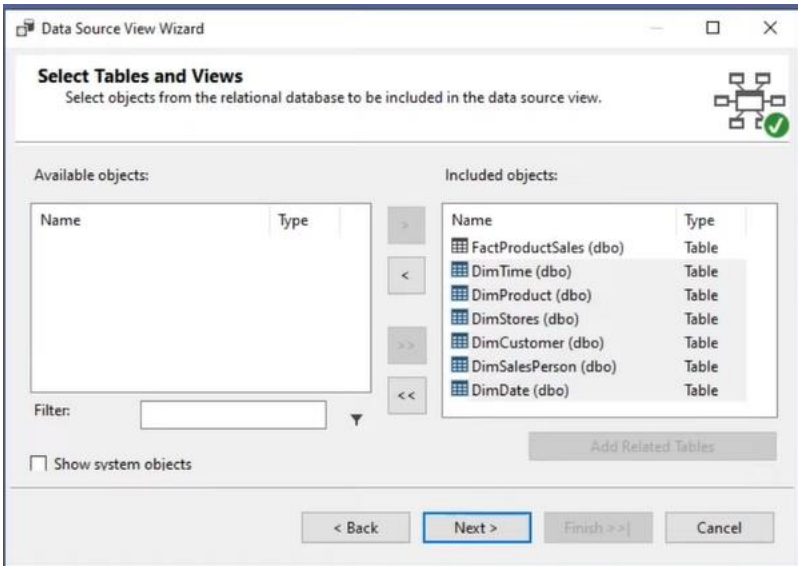


Step 3: Creating New Data Source View in Solution explorer right click on Data Source View Select New Data

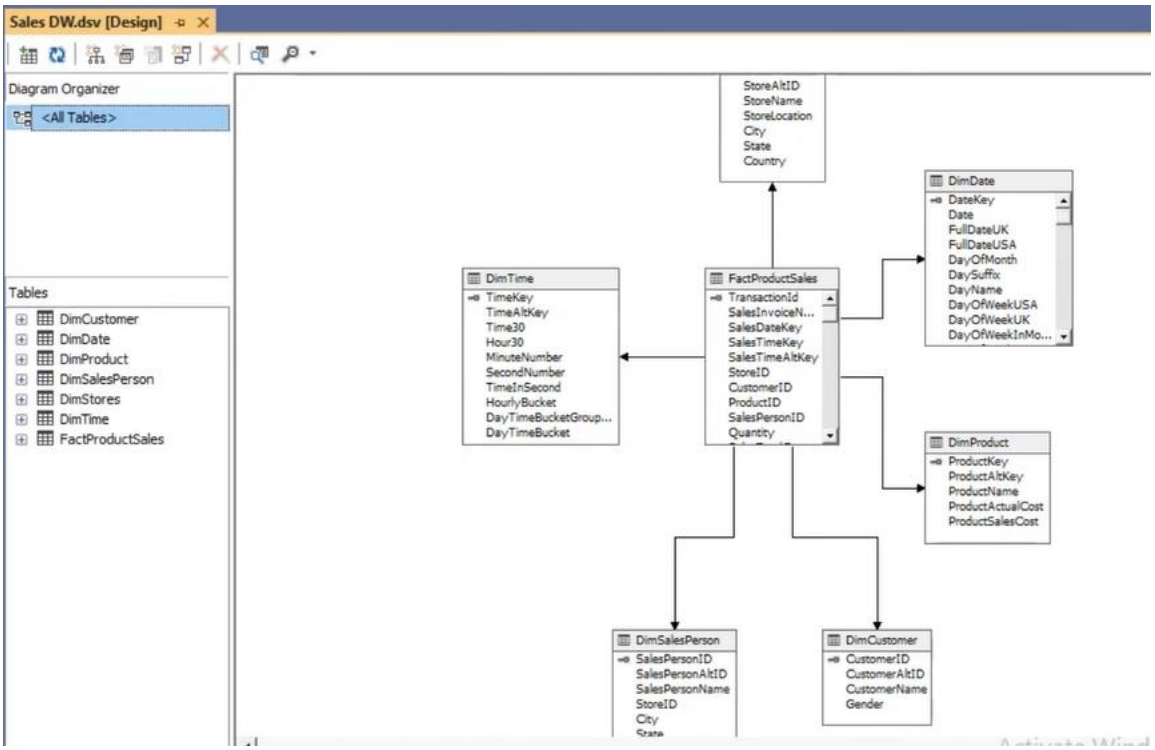


Click Next Select FactProductSales(dbo) from Available objects and put in Includes Objects by click.

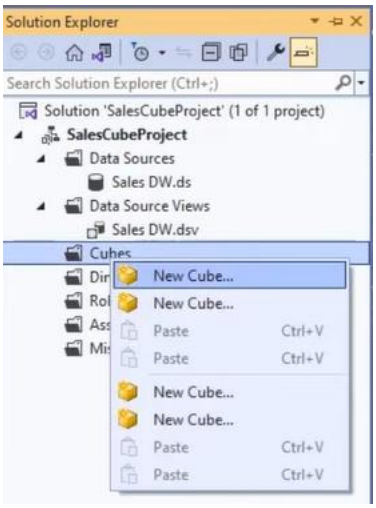




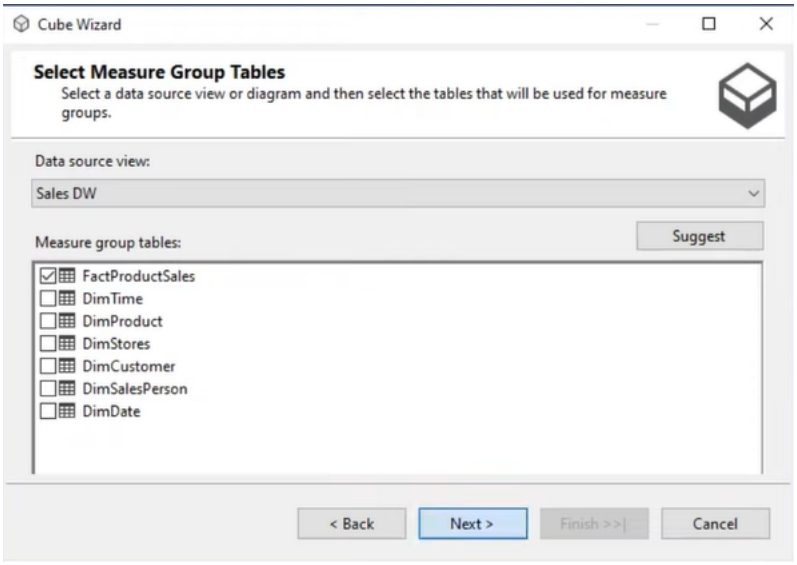
Click Finish Sales DW.dsv appears in Data Source Views in Solution Explorer



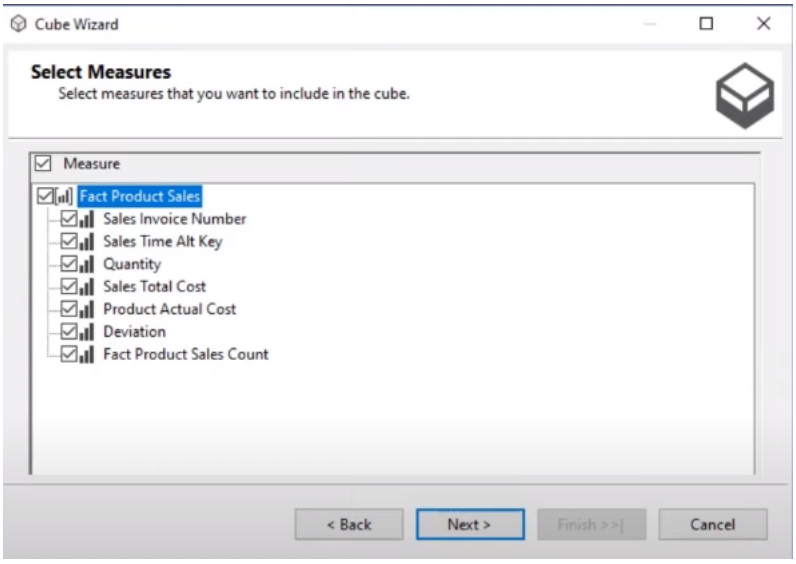
Step 4: Creating new cube Right click on cubes → New Cube



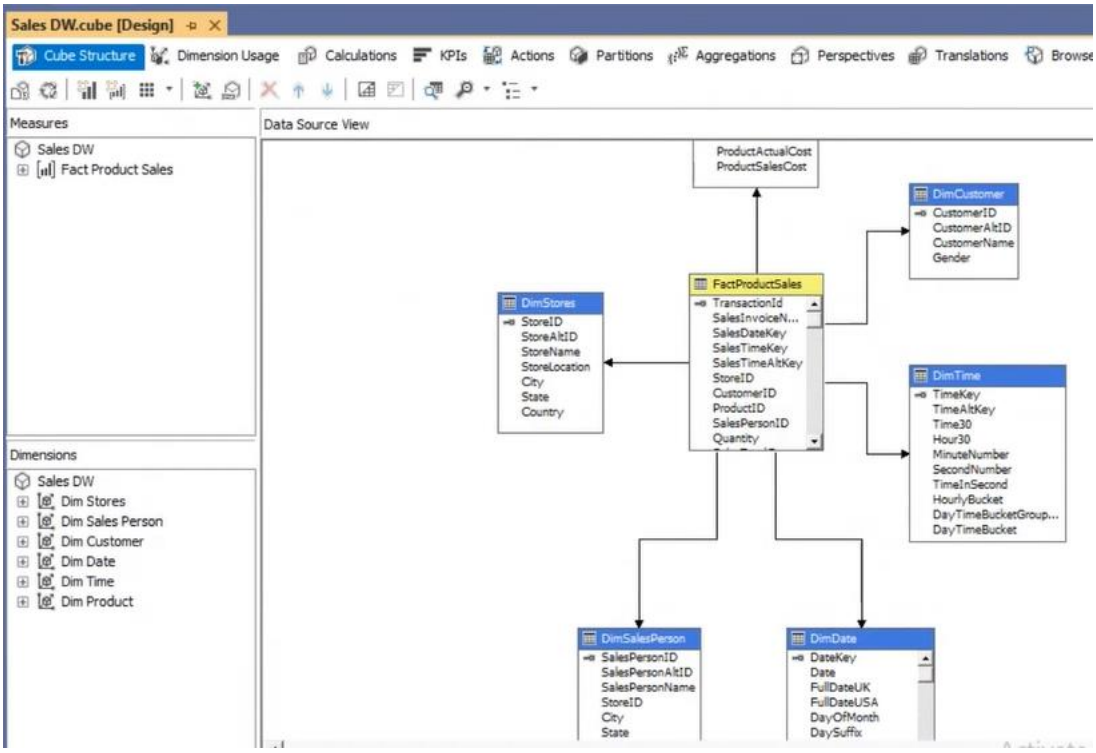
Select Use existing tables in Select Creation Method → Next In Select Measure Group Tables → Select FactProductSales → Click Next



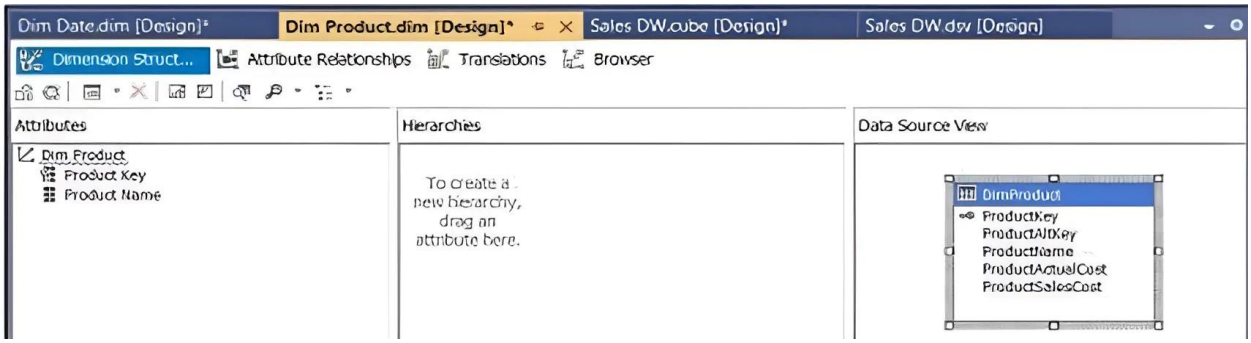
In Select Measures → check all measures → Next



In Select New Dimensions → Check all Dimensions→ Next. Click on Finish. Sales_DW.cube is created.

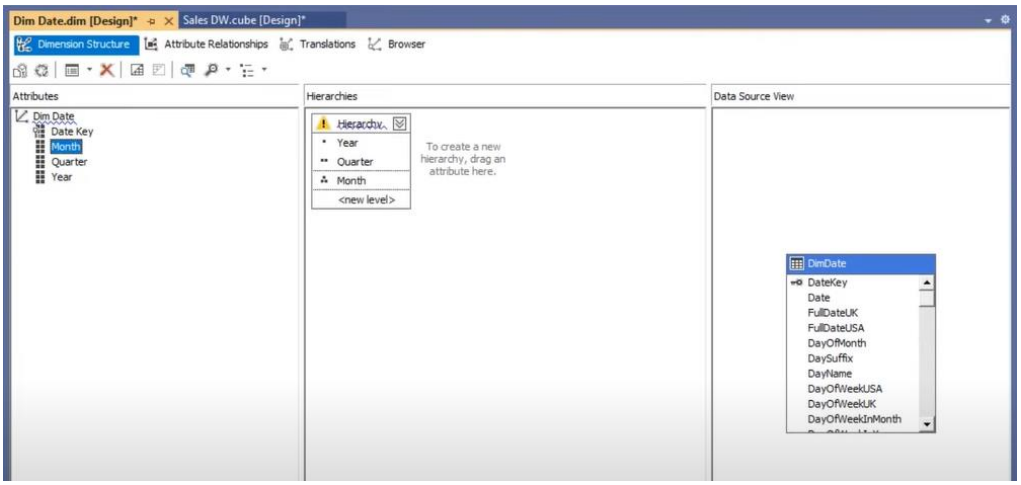


Step 5: Dimension Modification In dimension tab → Double Click Dim Product.dim Step 6: Drag and Drop Product Name from Table in Data Source View and Add in Attribute Pane at left side.

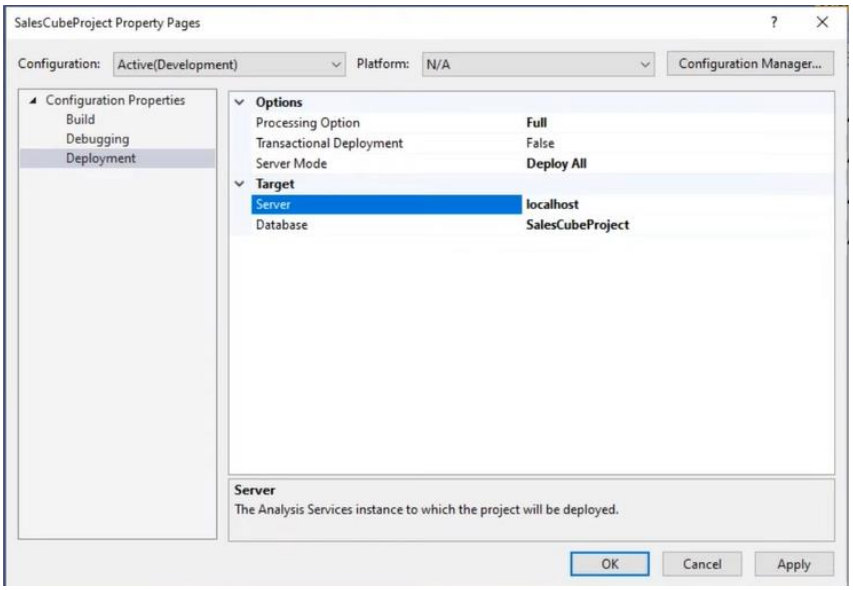


Step 7: Creating Attribute Hierarchy in Date Dimension.

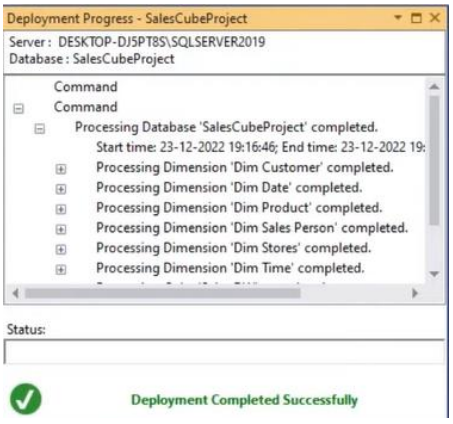
Double click on Dim Date dimension → Drag and Drop Fields from Table shown in Data Source View to Attributes → Drag and Drop attributes from leftmost pane of attributes to middle pane of Hierarchy. Drag fields in sequence from Attributes to Hierarchy window (Year, Quarter Name, Month, Name, Week of the Month, Full date UK).



Step 8: Deploy Cube Right click on Project name → Properties Do following changes and click on Apply & OK



Right click on project name → Deploy

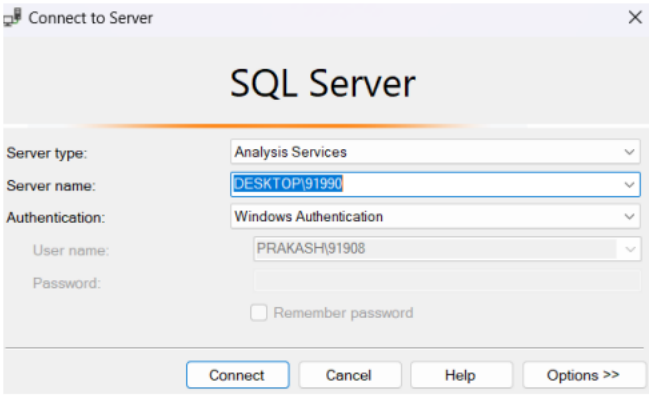


PRACTICAL NO:4

Aim: a) Create the ETL map and set up the schedule for execution.

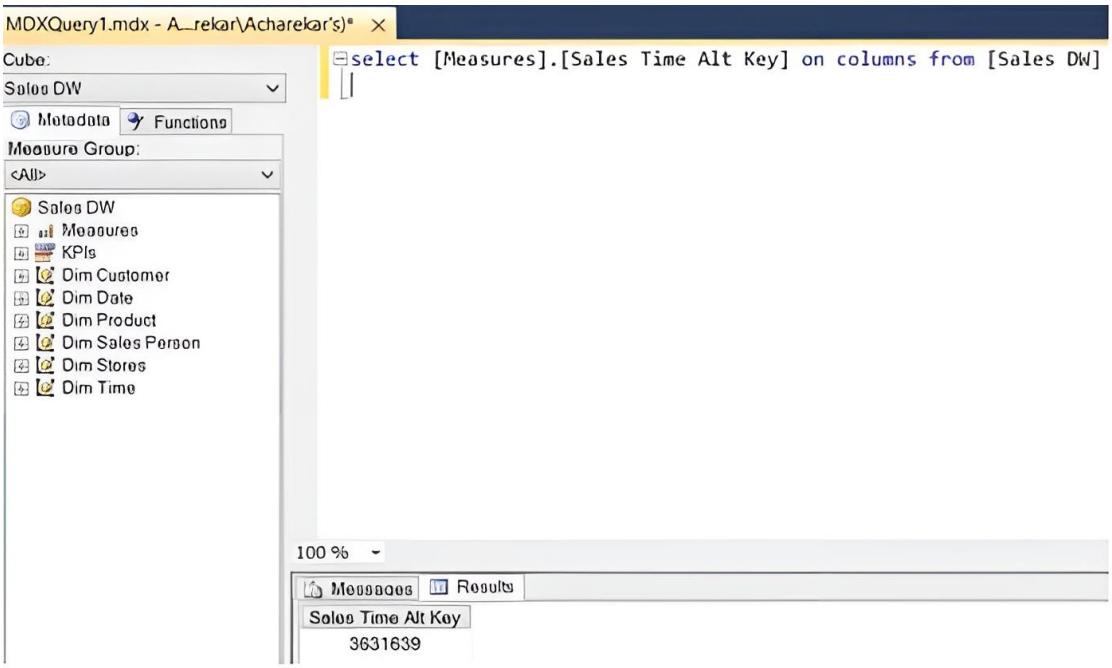
b) Execute the MDX queries to extract the data from the Datawarehouse.

Step 1: Open SQL server Management studio and Connect to Analysis Services.

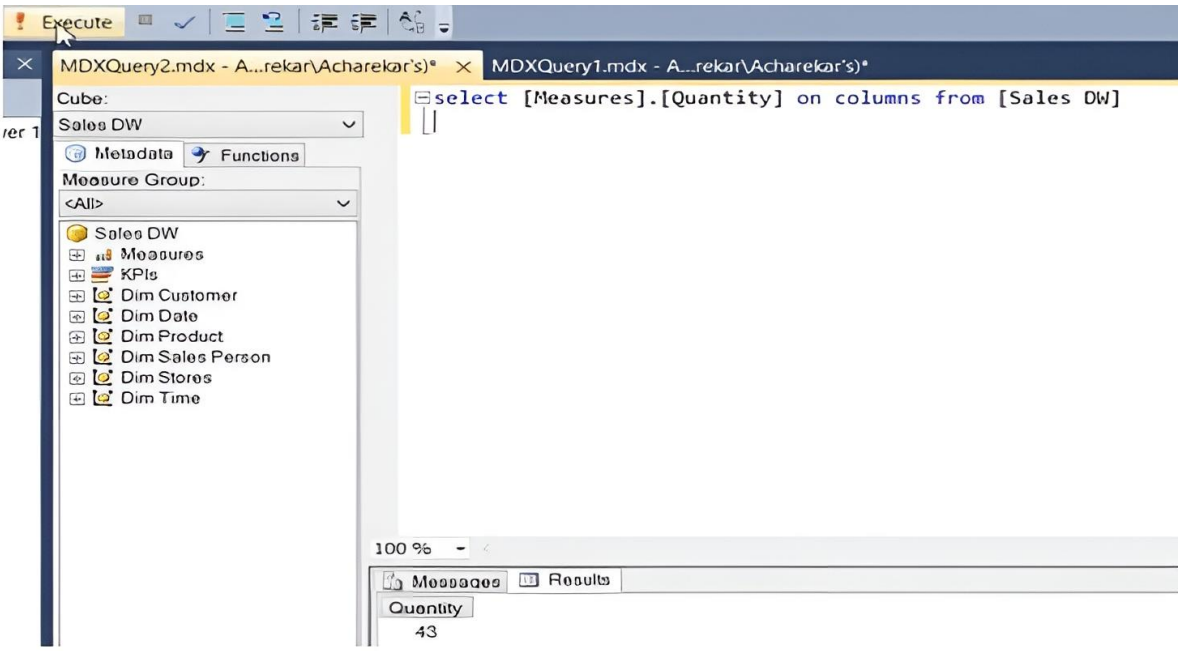


Step 2: Click on new query -> type following queries based on Sales_DW -> then click on Execute.

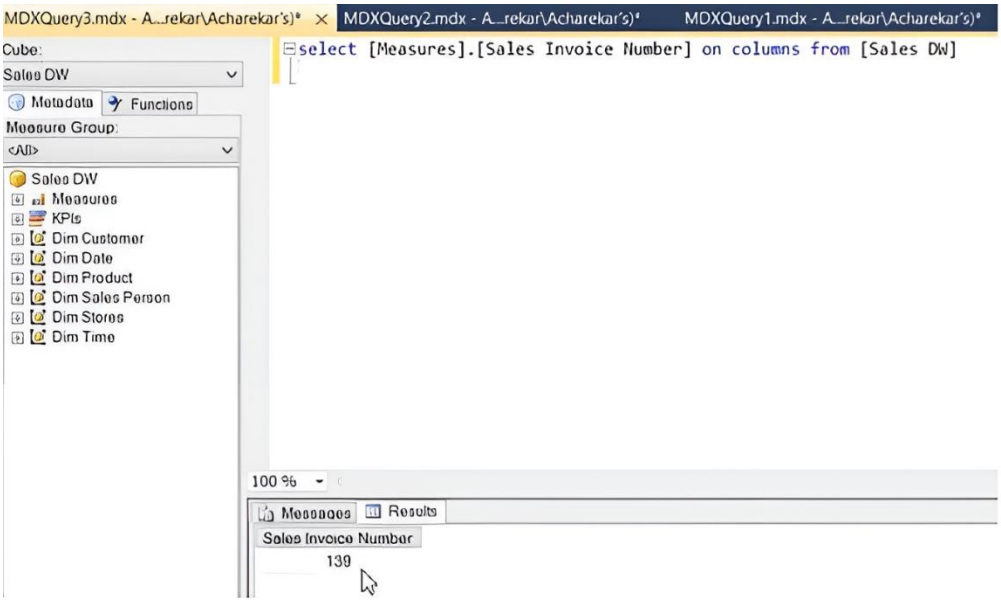
1.select[Measures].[Sales Time Alt Key] on columns from [Sales DW]



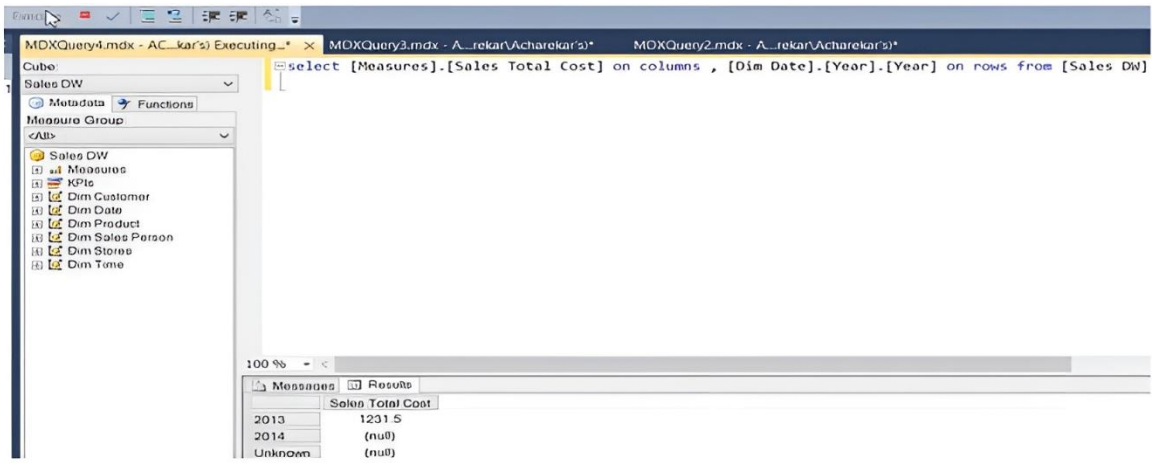
2. select [Measures].[Quantity] columns from [Sales DW]



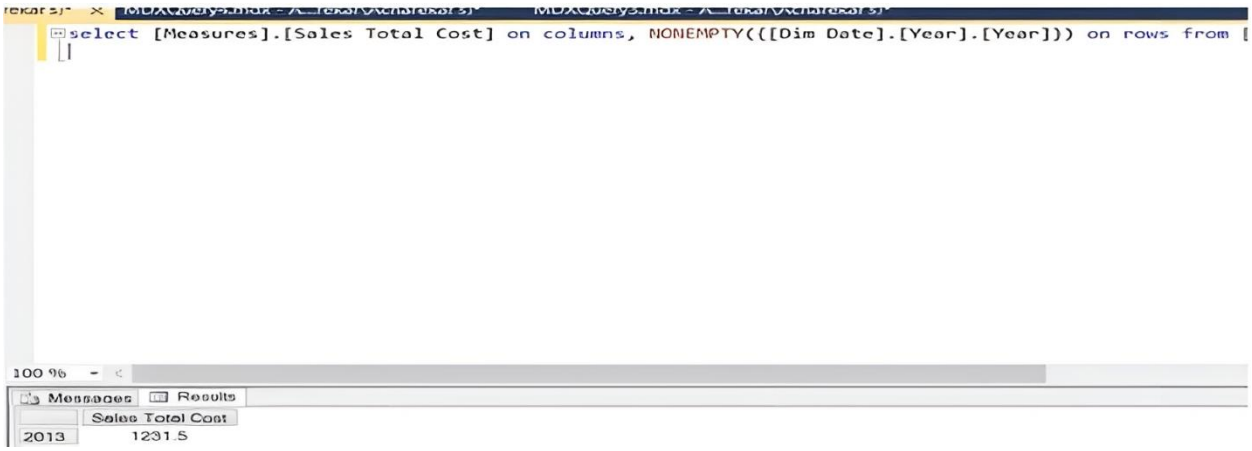
3. select [Measures].[Sales Invoice Number] columns from [Sales DW]



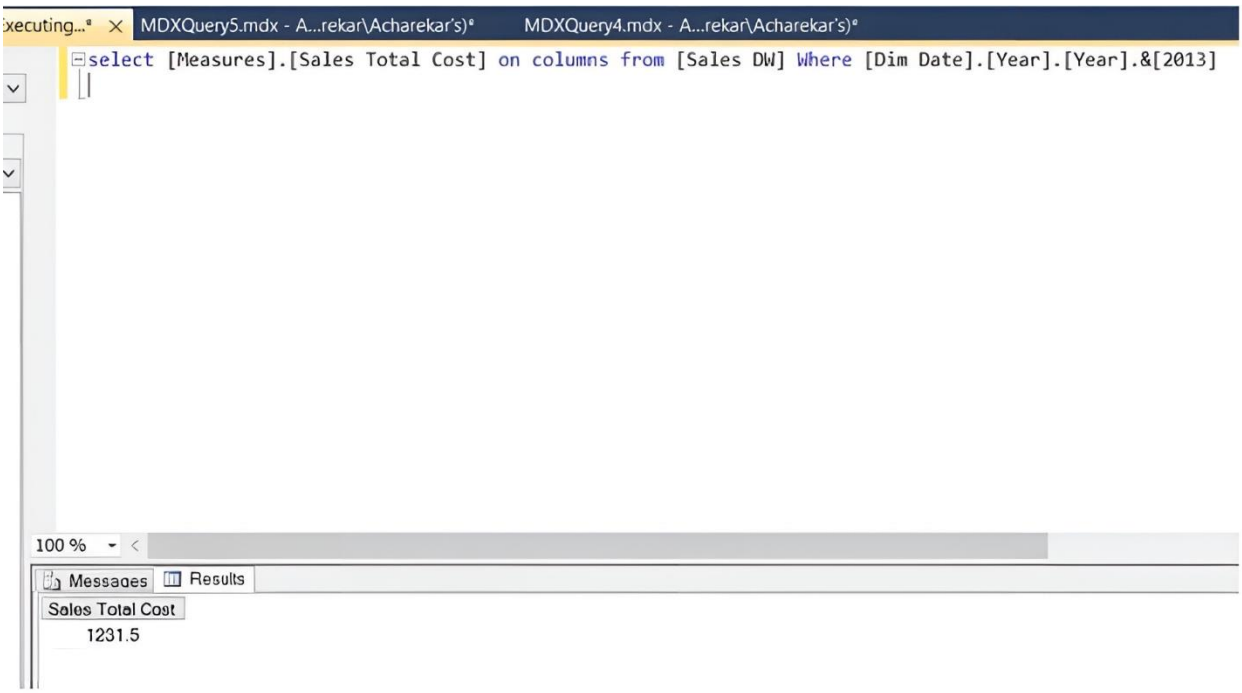
4. Select [Measures].[Sales Total Cost] on columns, [Dim Date].[Year].[Year] on rows from [Sales_DW]



5. Select [Measures].[Sales Total Cost] on columns, NONEMPTY({[Dim Date].[Year].[Year]}) on rows from [Sales_DW]



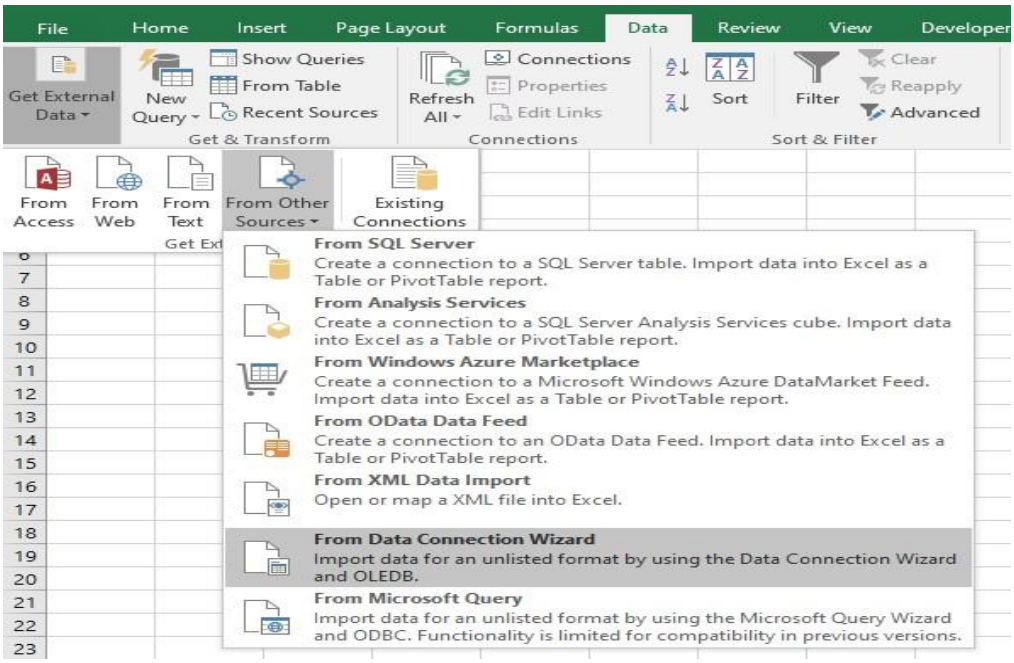
6.Select [Measures].[Sales Total Cost] on columns from [Sales_DW] where [Dim Date].[Year].[Year].&[2013]



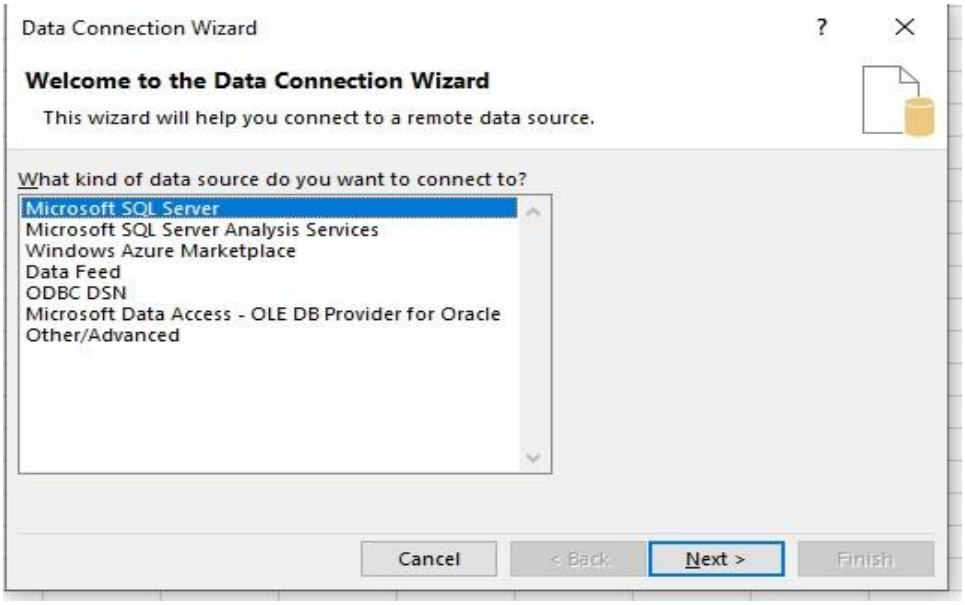
PRACTICAL NO: 5(A)

Aim: Import the Data Warehouse data in Microsoft Excel and create the Pivot table and Pivot Chart.

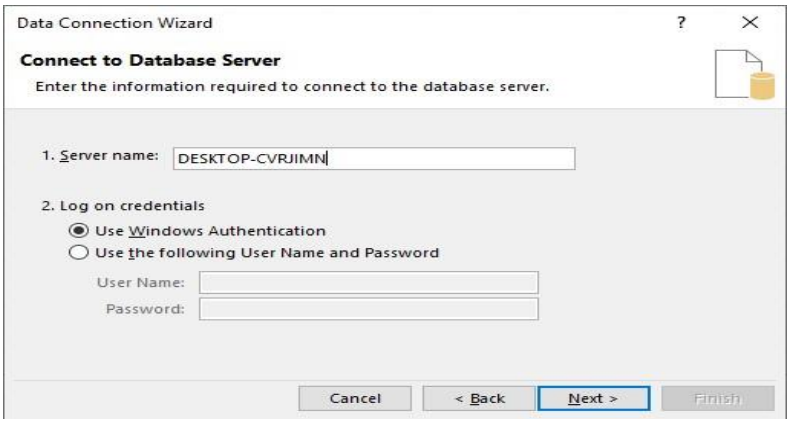
Step 1: Open Excel Go to Data tab → Get External Data → From Other Sources → From Data Connection Wizard



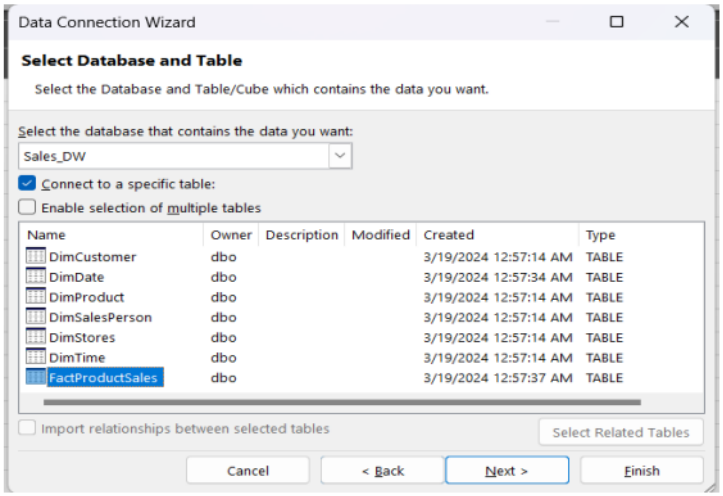
Step 2: In Data Connection Wizard → Select Microsoft SQL Server → Click on Next



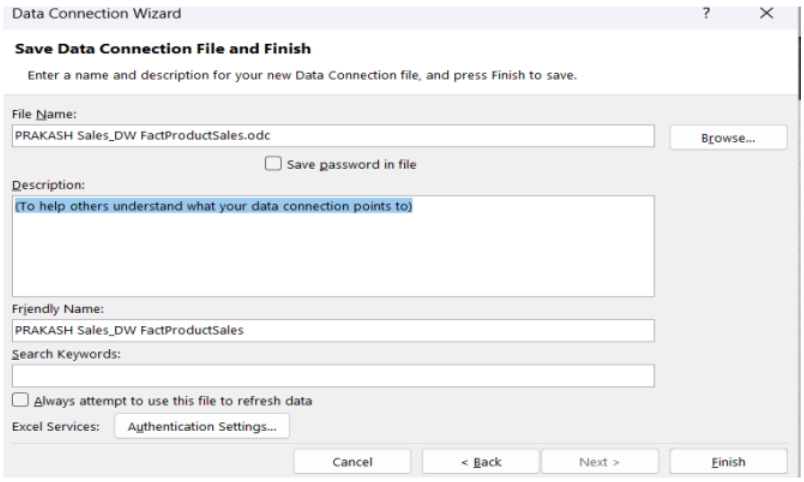
Step 3: In connect to Database Server provide Server name-> Click on Next



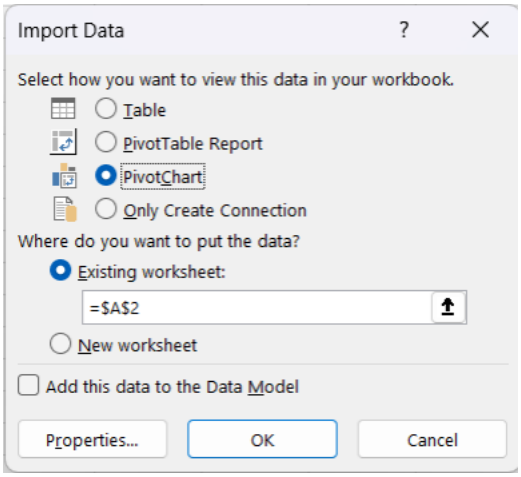
Step 4: In Select Database and Table→ Select Adventure works (already created in SQL) → check all dimensions and import relationships between selected tables -> Click on next.



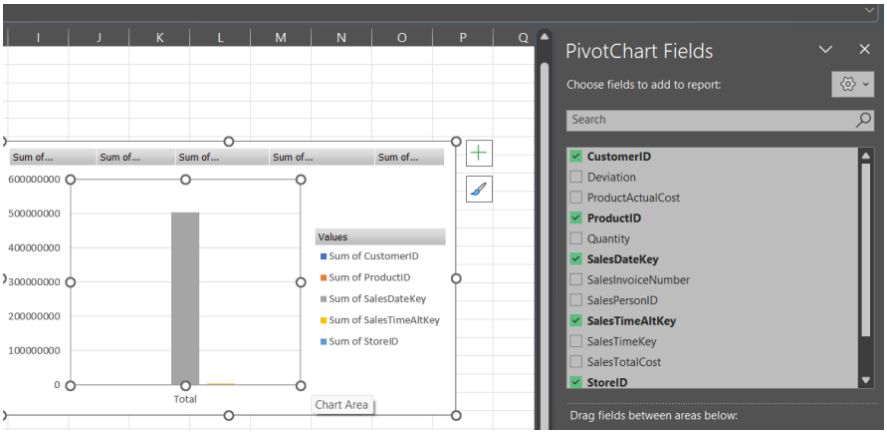
Step 5: In save data connection files browse path and click on Finish



Step 6: In import data select Pivot Chart and click on OK

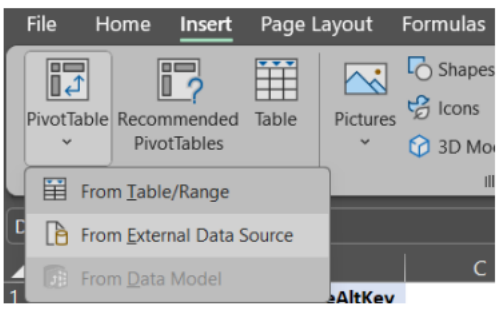


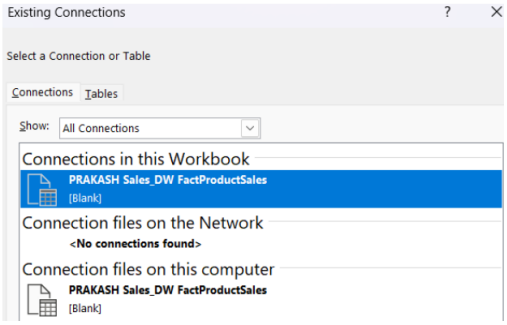
Step 7: Select field to create Pivot Chart



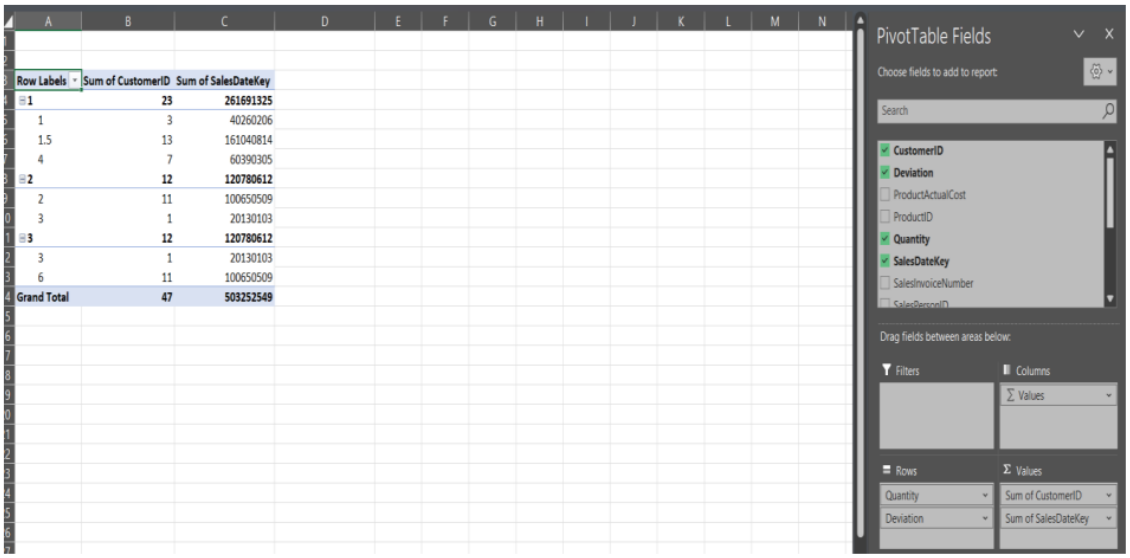
Step 8: In Insert Tab -> Go to Pivot Table

Step 9: Click on choose Connection to select existing connection with Sales_DW and click on open.





Step 10: Select Field to create Pivot Table. You will get this.



Row Labels	Sum of CustomerID	Sum of SalesDateKey
1	**23**	**261691325**
1	3	40260206
1.5	13	161040814
4	7	60390305
2	**12**	**120780612**
2	11	100650509
3	1	20130103
3	**12**	**120780612**
3	1	20130103
6	11	100650509
Grand Total	**47**	**503252549**

PRACTICAL NO: 6

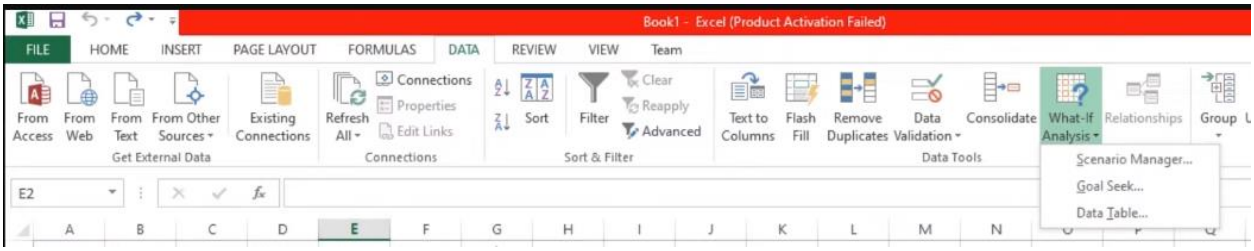
Aim : Apply the what – if Analysis for data visualization. Design and generate necessary reports based on the data warehouse data.

1) Goal Seek Method :

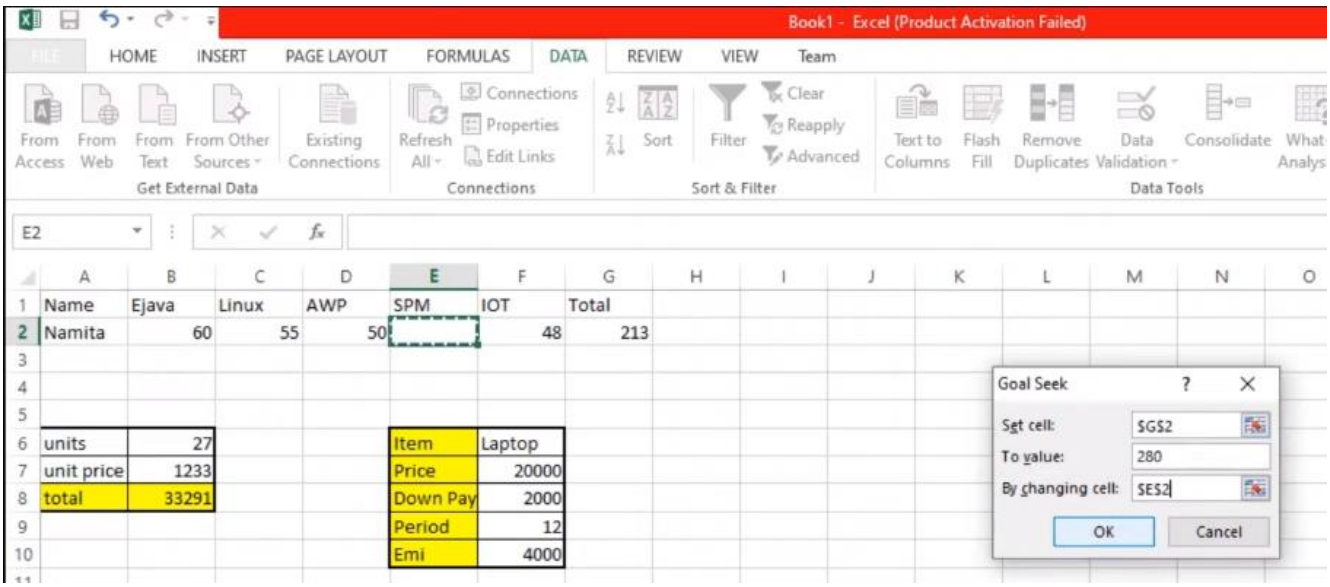
Step 1 : Go to Data Tab -> Select the What- if Analysis -> click on goal seek

	A	B	C	D	E	F	G
1	Name	Ejava	Linux	AWP	SPM	IOT	Total
2	Namita	60	55	50		48	213

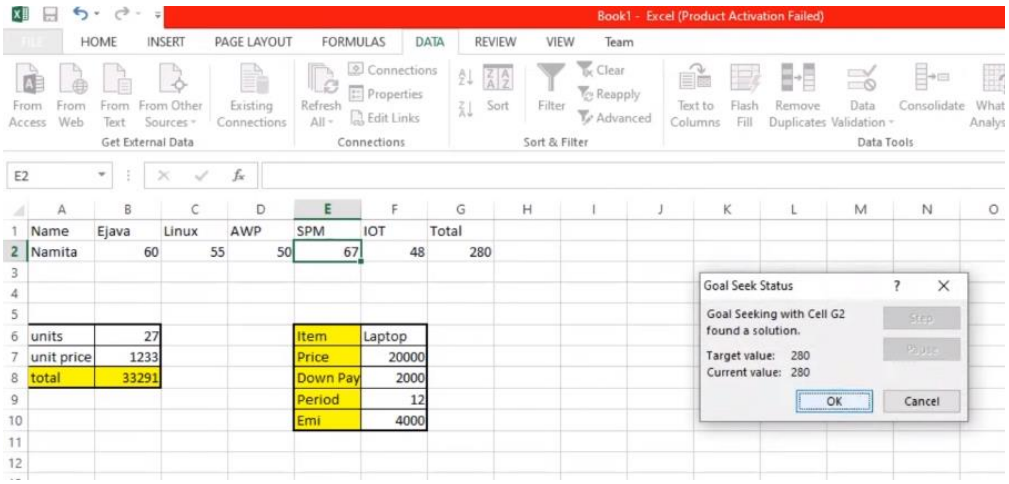
Step 2: Go to Data tab -> What-if Analysis->Click on Goal Seek



Step 3: In “set cells” select the total column and in “By changing sell” select the missing value and in “to value” give any random number.



Step 4: The empty value is been filled according to the given condition.



	A	B	C	D	E	F	G
1	Name	Ejava	Linux	AWP	SPM	IOT	Total
2	Namita	60	55	50	67	48	280

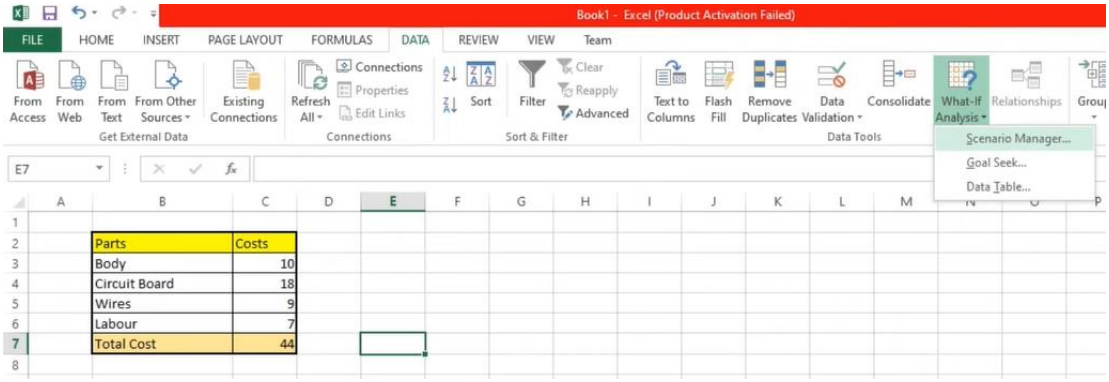
6	units	27
7	unit price	1233
8	total	33291

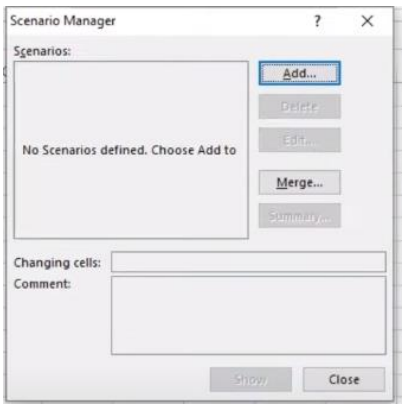
	Item	Laptop
	Price	20000
	Down Pay	2000
	Period	12
	Emi	4000

Step 5: Also create one more table.

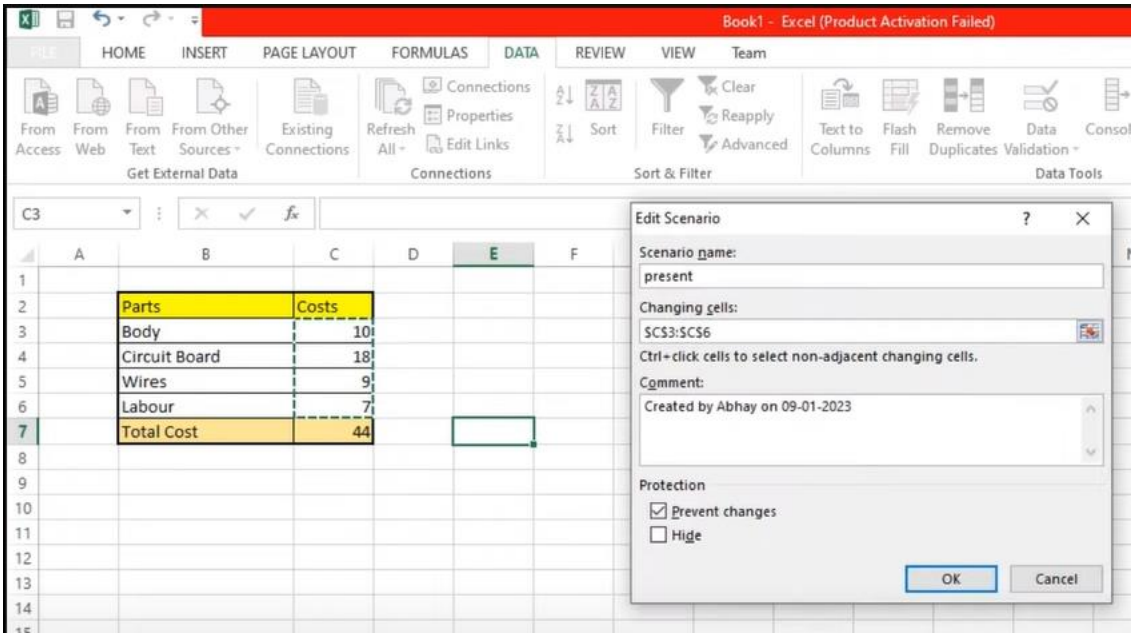
Parts	Costs
Body	10
Circuit Board	18
Wires	9
Labour	7
Total Cost	44

Step 6: Go to Data tab -> What-if Analysis -> Click on Scenario Manager -> Then Click on Add.

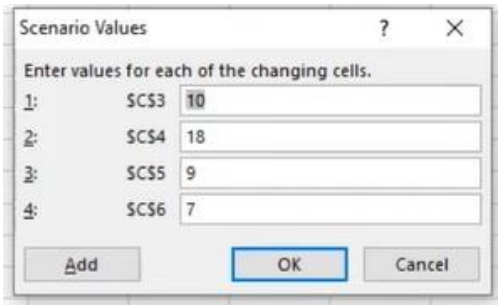


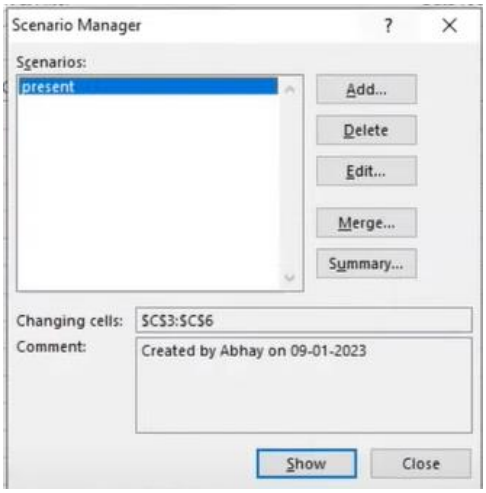


Step 7: Give the scenario name Present -> select the changing cells-> Click on ok

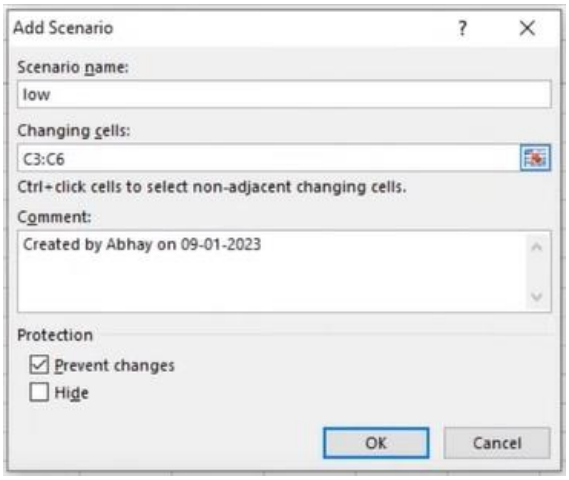


Step 8: Type Scenario present values -> Click on ok

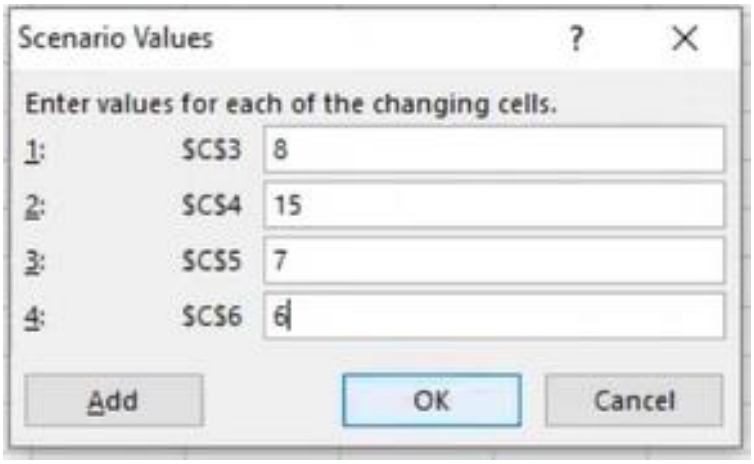


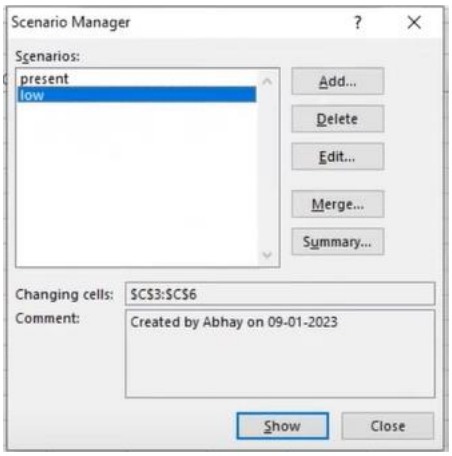


Step 9: Click on Add -> Give the scenario name Low -> select the changing cells -> Click on ok.

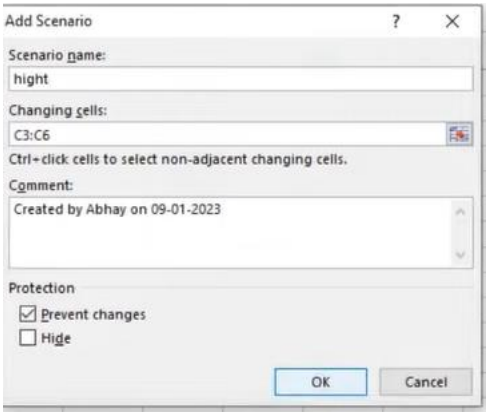


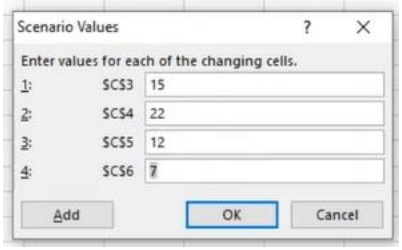
Step 10: Type scenario Low values -> click ok

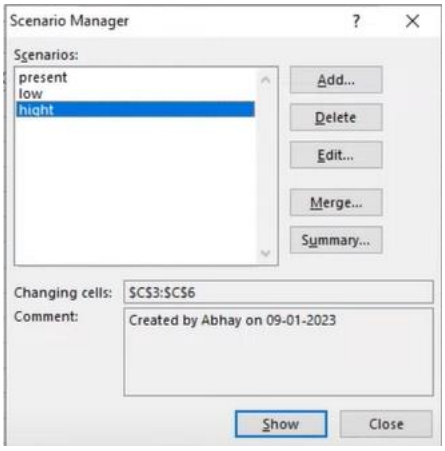




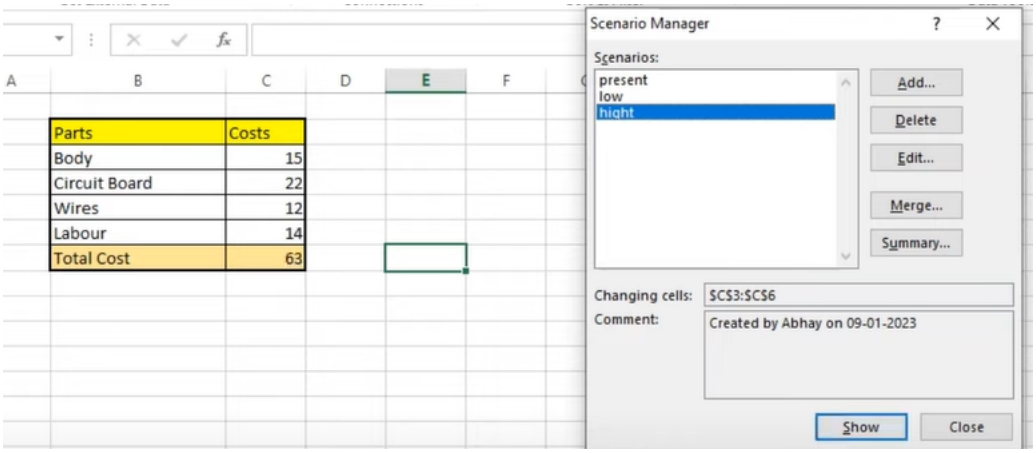
Step 11: Click on Add -> Give the scenario name High ->Select the changing cells -> Click on ok.



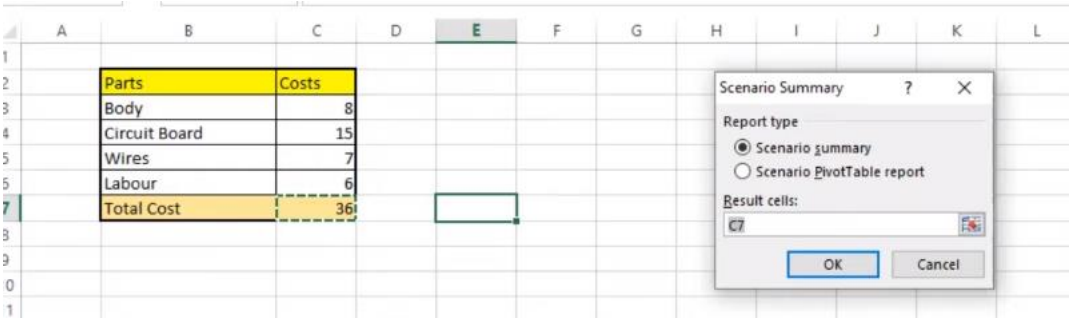




Step 13: Click on Show



Step 14: Click on Summary -> select total cost cell in result cells -> Click on ok.



Step 15: You will get Scenario Summary

Scenario Summary				
	Current Values:	present	low	hight
Changing Cells:				
C3	8	10	8	15
C4	15	18	15	22
C5	7	9	7	12
C6	6	7	6	14
Result Cells:				
C7	36	44	36	63

Notes: Current Values column represents values of changing cells at time Scenario Summary Report was created. Changing cells for each scenario are highlighted in gray.

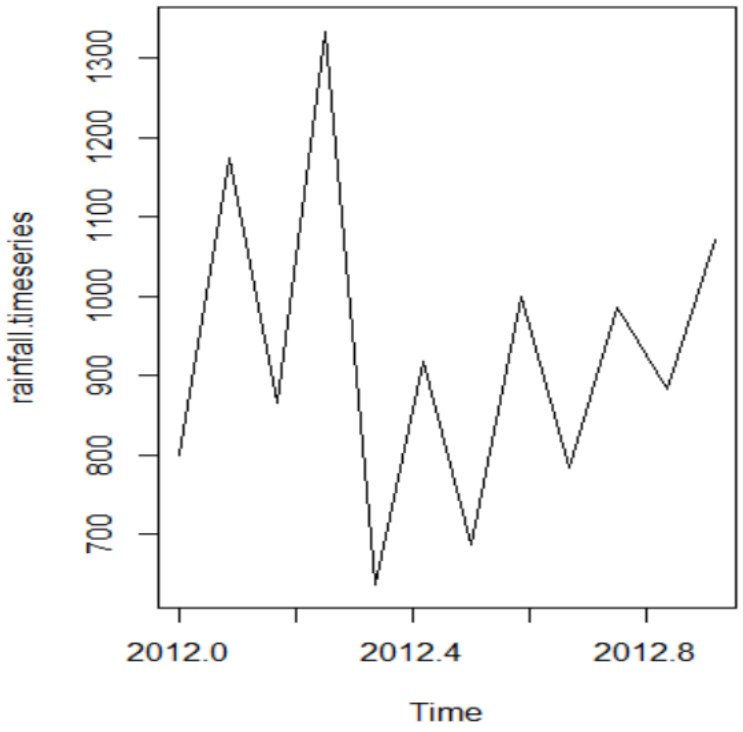
PRACTICAL NO: 7

AIM : Perform the data classification using classification algorithm.

Step 1: write this code in Rstudio ide

```
> rainfall <- c(799,1174.8,865.1,1334.6,635.4,918.5,685.5,998.6,784.2,985,882.8,1071)
> rainfall.timeseries<-ts(rainfall,start=c(2012,1),frequency = 12)
> print(rainfall.timeseries)
        Jan    Feb    Mar    Apr    May    Jun    Jul    Aug    Sep    Oct    Nov    Dec
2012  799.0 1174.8  865.1 1334.6  635.4  918.5  685.5  998.6  784.2  985.0  882.8 1071.0
> png(file="rainfall.png")
> plot(rainfall.timeseries)
> dev.off()
null device
          1
> plot(rainfall.timeseries)
> dev.off()
null device
          1
> plot(rainfall.timeseries)
>
```

Final Output



PRACTICAL No: 8

Aim: Perform the data clustering using a clustering algorithm.

Step 1: Create Excel file with Age and Expenditure column -> save as CSV.

Age	expenditure
19	4000
20	5500
23	5400
21	6500
22	4300
23	3400
25	5400
43	5500
34	7000
45	5400
45	3500
32	6500
34	4300
23	3400
27	2300
37	5000
32	4500
54	7500
32	5400
30	3500

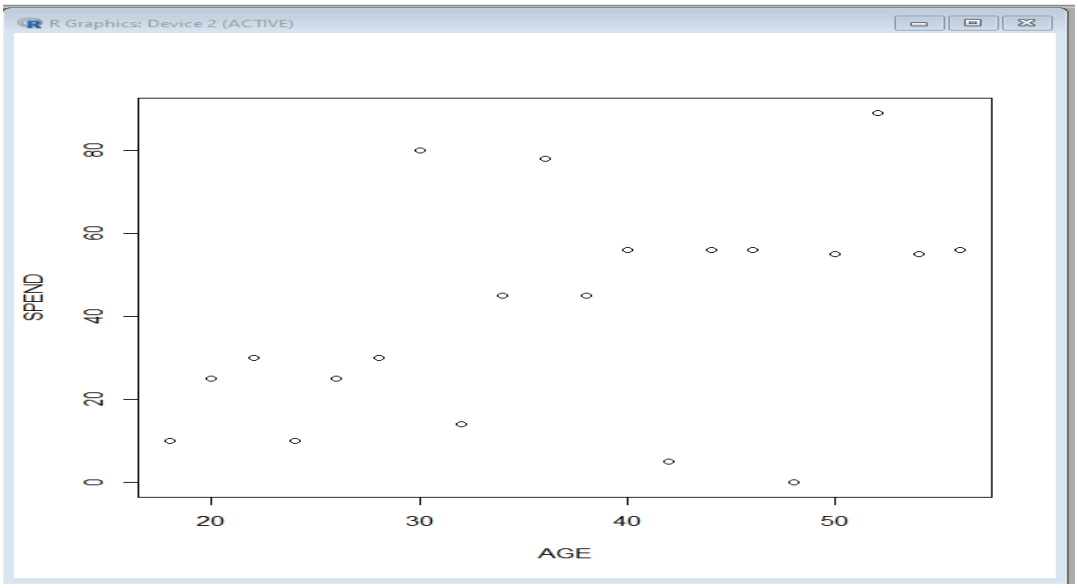
Step 2: Open R Studio -> Go to R console and type code.

1.df=read.csv("C:/Users/admin/Documents/AGE.csv")

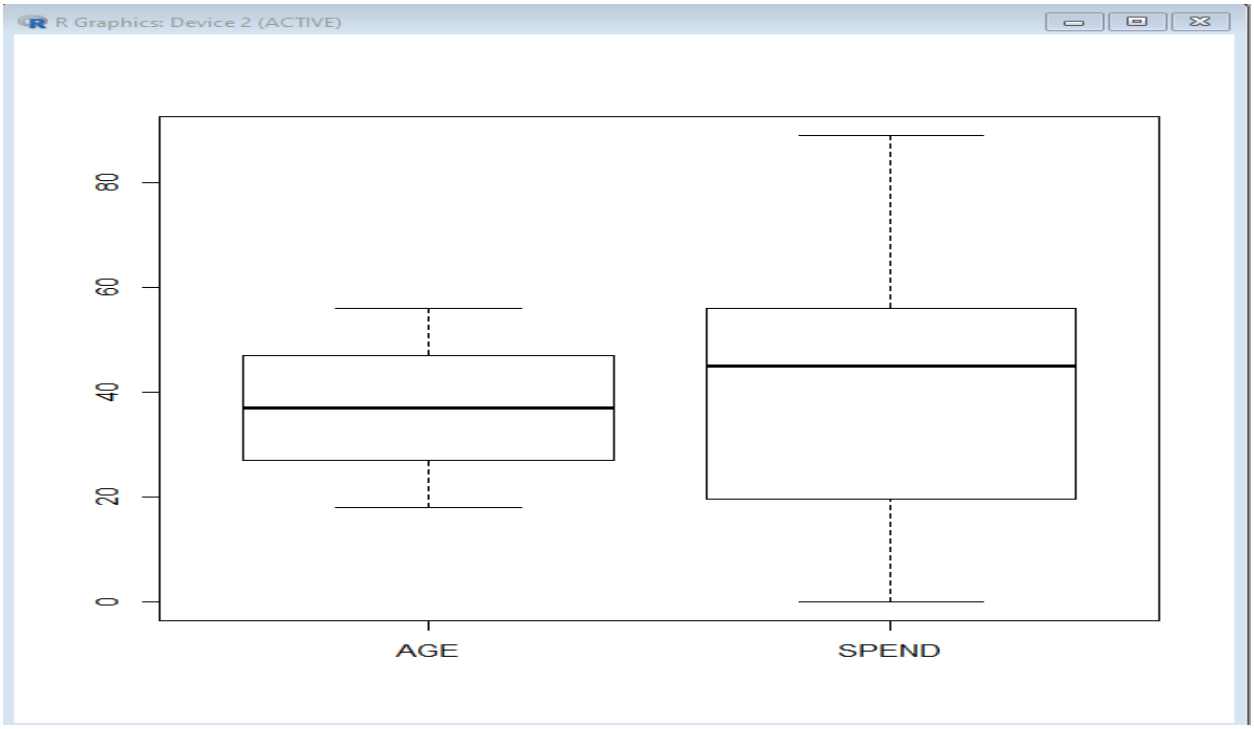
df

```
> df=read.csv("C:/Users/admin/Documents/AGE.csv")
> df
   AGE SPEND
1   18    10
2   20    25
3   22    30
4   24    10
5   26    25
6   28    30
7   30    80
8   32    14
9   34    45
10  36    78
11  38    45
12  40    56
13  42     5
14  44    56
15  46    56
16  48     0
17  50    55
18  52    89
19  54    55
20  56    56
```

4.plot(df)



5.boxplot(df)



6.set.seed(20)

> c1=kmeans(df[,1:2],3)

> c1

```
> set.seed(20)
> cl=kmeans(df[,1:2],3)
> cl
K-means clustering with 3 clusters of sizes 3, 8, 9

Cluster means:
       AGE    SPEND
1 39.33333 82.33333
2 45.25000 53.00000
3 28.88889 16.55556

Clustering vector:
 [1] 3 3 3 3 3 3 1 3 2 1 2 2 3 2 2 3 2 1 2 2

Within cluster sum of squares by cluster:
[1]  327.3333  595.5000 1829.1111
 (between_SS / total_SS =  82.3 %)

Available components:

[1] "cluster"      "centers"      "totss"        "withinss"     "tot.withinss"
[6] "betweenss"    "size"         "iter"         "ifault"
```

7.iris

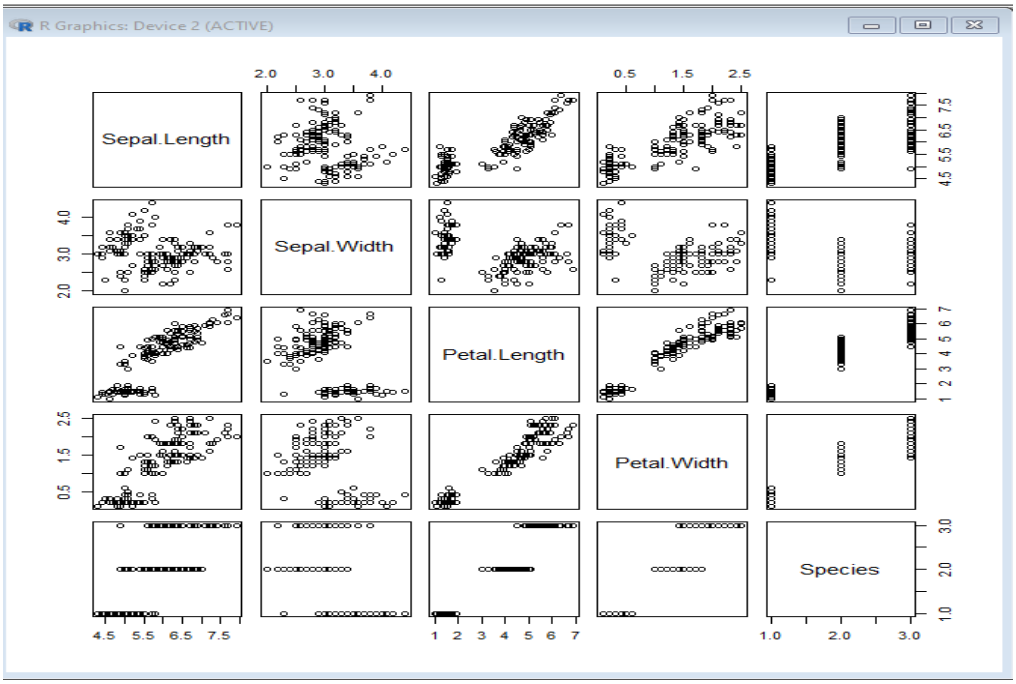
```
> iris
    Sepal.Length Sepal.Width Petal.Length Petal.Width    Species
1            5.1         3.5          1.4         0.2     setosa
2            4.9         3.0          1.4         0.2     setosa
3            4.7         3.2          1.3         0.2     setosa
4            4.6         3.1          1.5         0.2     setosa
5            5.0         3.6          1.4         0.2     setosa
6            5.4         3.9          1.7         0.4     setosa
7            4.6         3.4          1.4         0.3     setosa
8            5.0         3.4          1.5         0.2     setosa
9            4.4         2.9          1.4         0.2     setosa
10           4.9         3.1          1.5         0.1     setosa
11           5.4         3.7          1.5         0.2     setosa
12           4.8         3.4          1.6         0.2     setosa
13           4.8         3.0          1.4         0.1     setosa
```

8.head(iris)

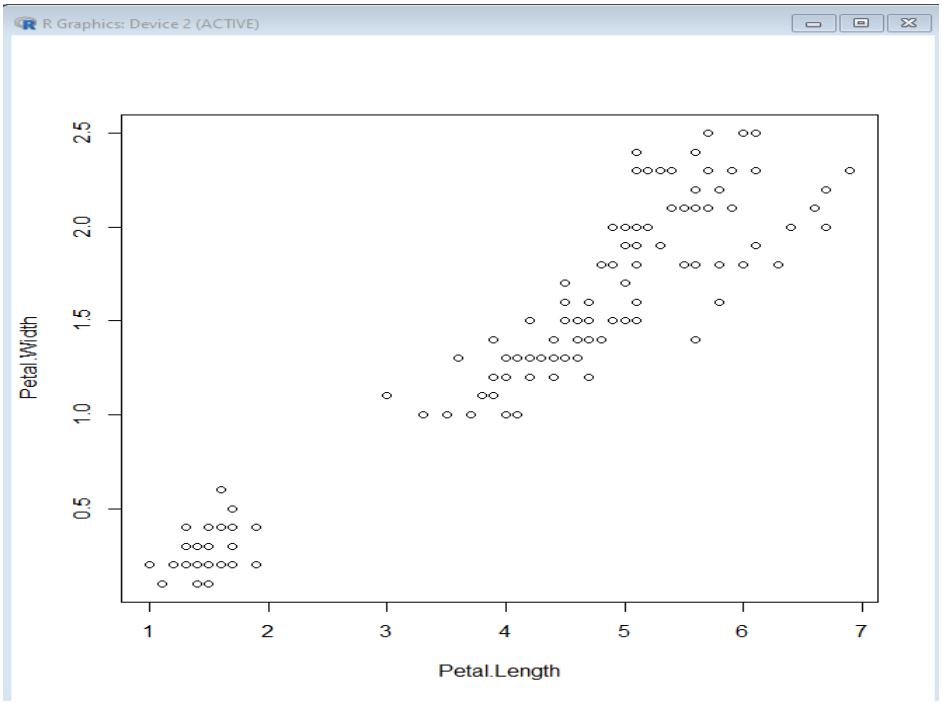
9.summary(iris)

```
> View(iris)
> head(iris)
  Sepal.Length Sepal.Width Petal.Length Petal.Width Species
1          5.1         3.5          1.4         0.2  setosa
2          4.9         3.0          1.4         0.2  setosa
3          4.7         3.2          1.3         0.2  setosa
4          4.6         3.1          1.5         0.2  setosa
5          5.0         3.6          1.4         0.2  setosa
6          5.4         3.9          1.7         0.4  setosa
> summary(iris)
  Sepal.Length    Sepal.Width     Petal.Length    Petal.Width
 Min.   :4.300   Min.   :2.000   Min.   :1.000   Min.   :0.100
 1st Qu.:5.100   1st Qu.:2.800   1st Qu.:1.600   1st Qu.:0.300
 Median :5.800   Median :3.000   Median :4.350   Median :1.300
 Mean   :5.843   Mean   :3.057   Mean   :3.758   Mean   :1.199
 3rd Qu.:6.400   3rd Qu.:3.300   3rd Qu.:5.100   3rd Qu.:1.800
 Max.   :7.900   Max.   :4.400   Max.   :6.900   Max.   :2.500
       Species
 setosa     :50
 versicolor:50
 virginica :50
```

10.plot(iris)



11.plot(iris[,3:4])



12.kmeansc1=kmeans(iris[,3:4],3)

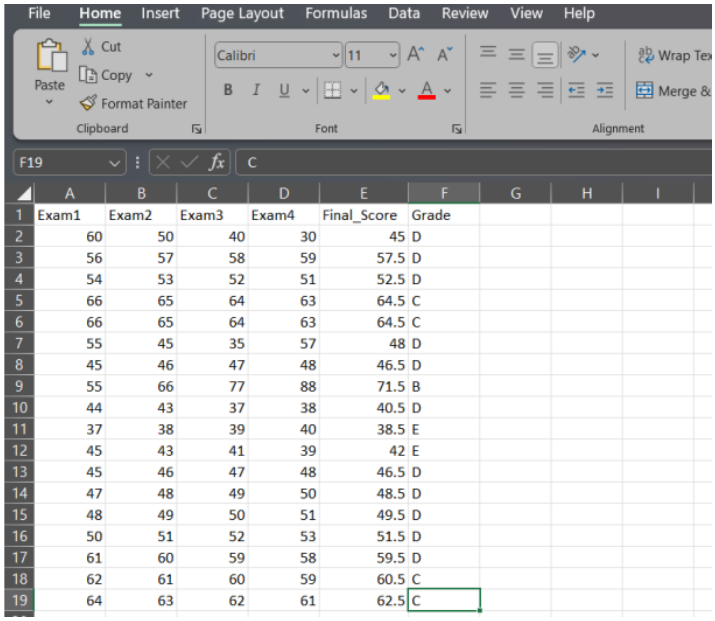
kmeansc1

```
> kmeansc1=kmeans(iris[,3:4],3)
> kmeansc1
K-means clustering with 3 clusters of sizes 50, 46, 54

Cluster means:
  Petal.Length Petal.Width
1     1.462000    0.246000
2     5.626087    2.047826
3     4.292593    1.359259

Clustering vector:
  [1] 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1
 [38] 1 1 1 1 1 1 1 1 1 1 1 1 1 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3
 [75] 3 3 3 2 3 3 3 3 3 2 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 2 2 2 2 2 2 3 2 2 2 2 2
[112] 2 2 2 2 2 2 2 2 3 2 2 2 3 2 2 3 3 2 2 2 2 2 2 2 2 2 3 2 2 2 2 2 2 2 2 2 2
[149] 2 2

Within cluster sum of squares by cluster:
[1]  2.02200 15.16348 14.22741
 (between_SS / total_SS =  94.3 %)

Available components:

[1] "cluster"      "centers"      "totss"        "withinss"     "tot.withinss"
[6] "betweenss"    "size"         "iter"         "ifault"
> |
```

PRACTICAL NO: 9

Aim: Perform the Linear regression on the given data warehouse data.

Step 1: Open Excel-> create 4 columns with names Exam1, Exam2, Exam3, Exam4, Final Score, Grade.

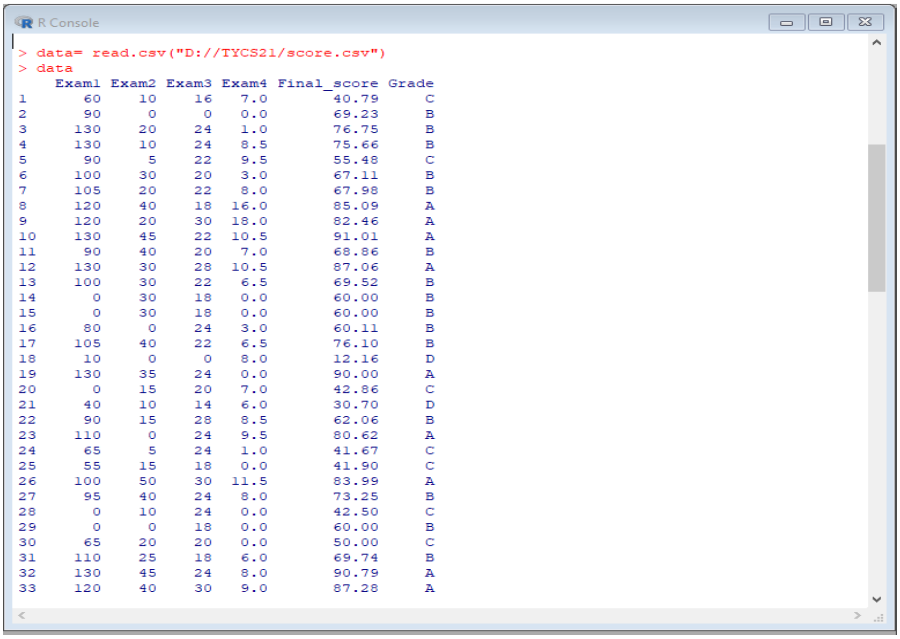


	A	B	C	D	E	F
1	Exam1	Exam2	Exam3	Exam4	Final_Score	Grade
2	60	50	40	30	45	D
3	56	57	58	59	57.5	D
4	54	53	52	51	52.5	D
5	66	65	64	63	64.5	C
6	66	65	64	63	64.5	C
7	55	45	35	57	48	D
8	45	46	47	48	46.5	D
9	55	66	77	88	71.5	B
10	44	43	37	38	40.5	D
11	37	38	39	40	38.5	E
12	45	43	41	39	42	E
13	45	46	47	48	46.5	D
14	47	48	49	50	48.5	D
15	48	49	50	51	49.5	D
16	50	51	52	53	51.5	D
17	61	60	59	58	59.5	D
18	62	61	60	59	60.5	C
19	64	63	62	61	62.5	C

Step 2: Open R Studio -> Go on R console and type code

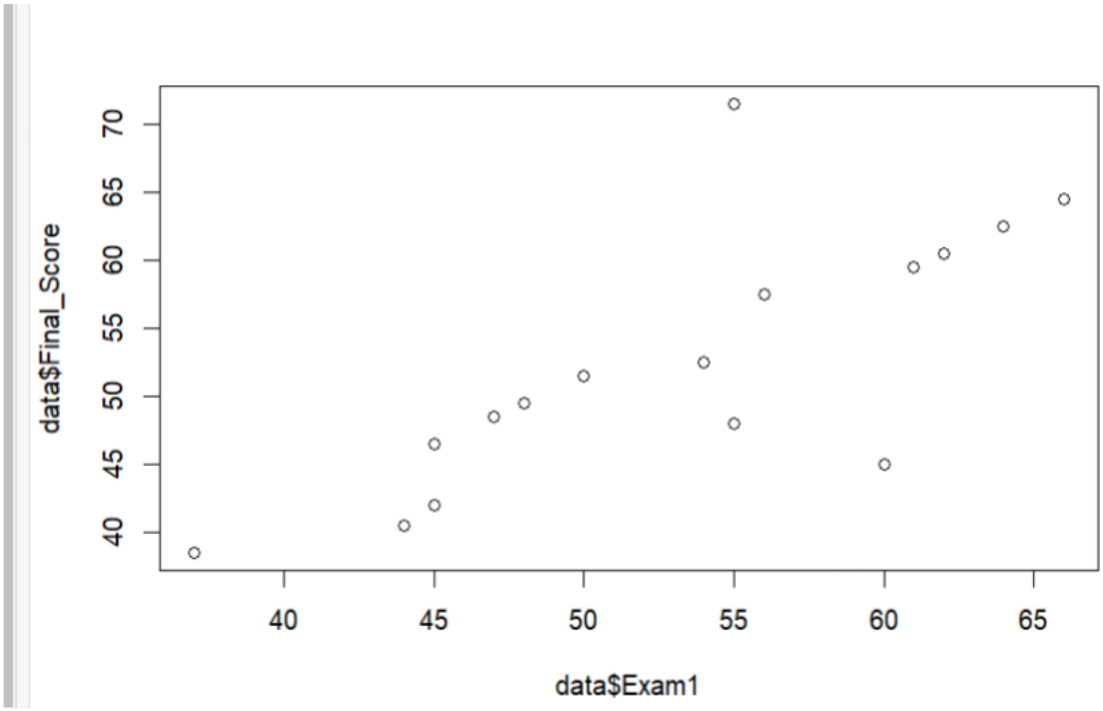
1.Data = read.csv("D://TYCS21/score.csv")

Data

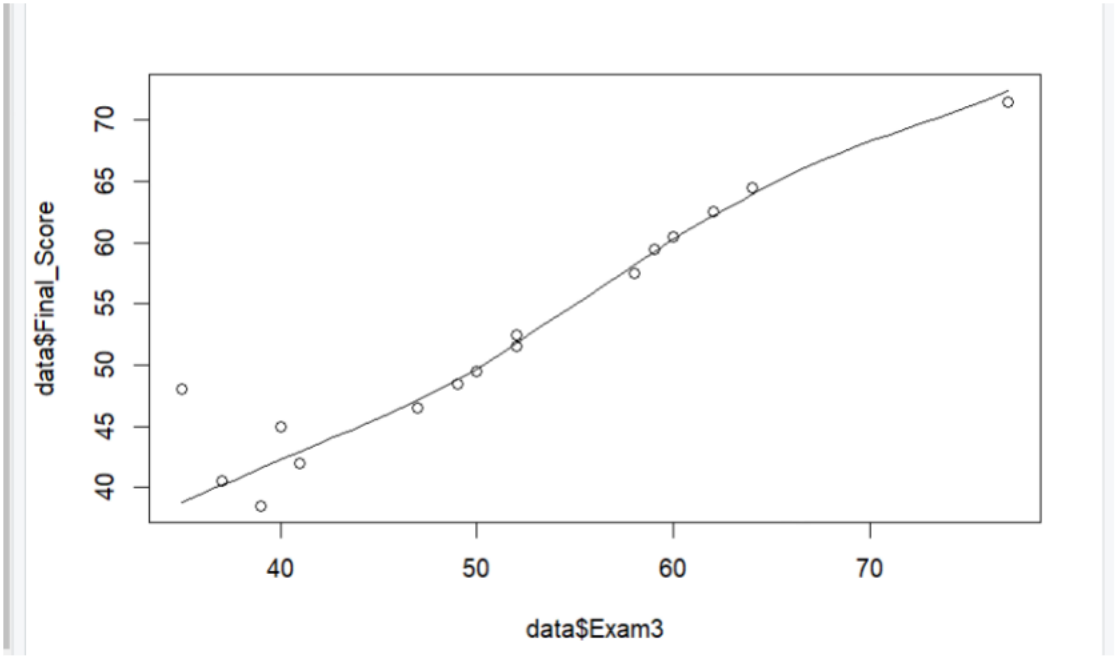


```
> data= read.csv("D://TYCS21/score.csv")
> data
   Exam1 Exam2 Exam3 Exam4 Final_score Grade
1     60    10    16   7.0       40.79     C
2     90     0     0   0.0       69.23     B
3    130    20    24   1.0       76.75     B
4    130    10    24   8.5       75.66     B
5     90     5    22   9.5       55.48     C
6    100    30    20   3.0       67.11     B
7    105    20    22   8.0       67.98     B
8    120    40    18  16.0       85.09     A
9    120    20    30  18.0       82.46     A
10   130    45    22  10.5       91.01     A
11    90    40    20   7.0       68.86     B
12   130    30    28  10.5       87.06     A
13   100    30    22   6.5       69.52     B
14     0    30    18   0.0       60.00     B
15     0    30    18   0.0       60.00     B
16    80     0    24   3.0       60.11     B
17   105    40    22   6.5       76.10     B
18    10     0     0   8.0       12.16     D
19   130    35    24   0.0       90.00     A
20     0    15    20   7.0       42.86     C
21    40    10    14   6.0       30.70     D
22    90    15    28   8.5       62.06     B
23   110     0    24   9.5       80.62     A
24    65     5    24   1.0       41.67     C
25    55    15    18   0.0       41.90     C
26   100    50    30  11.5       83.99     A
27    95    40    24   8.0       73.25     B
28     0    10    24   0.0       42.50     C
29     0     0    18   0.0       60.00     B
30    65    20    20   0.0       50.00     C
31   110    25    18   6.0       69.74     B
32   130    45    24   8.0       90.79     A
33   120    40    30   9.0       87.28     A
```

2.plot(x=data$Exam1,y=data$Final_score)



3.scatter.smooth(x=data$Exam3,y=data$Final_score)

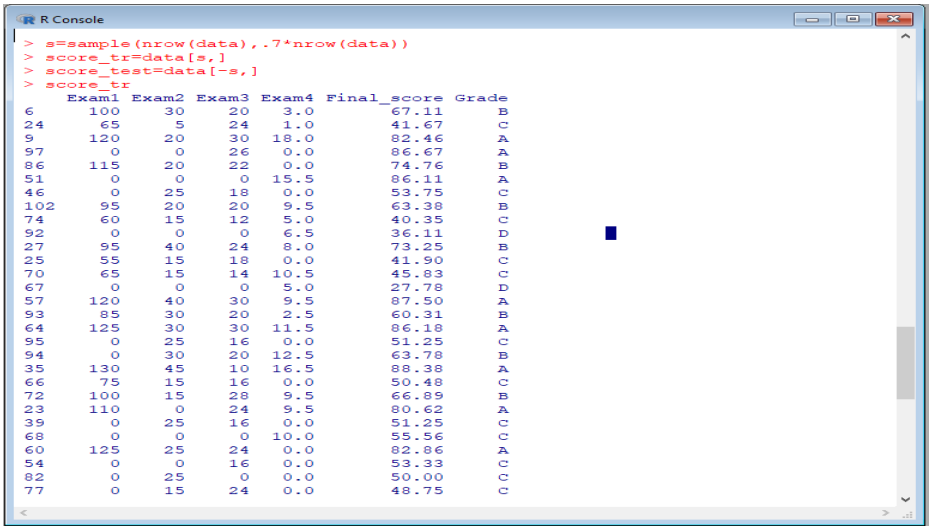


4.s=sample(nrow(data),.7*nrow(data))

>score_tr=data[s,]

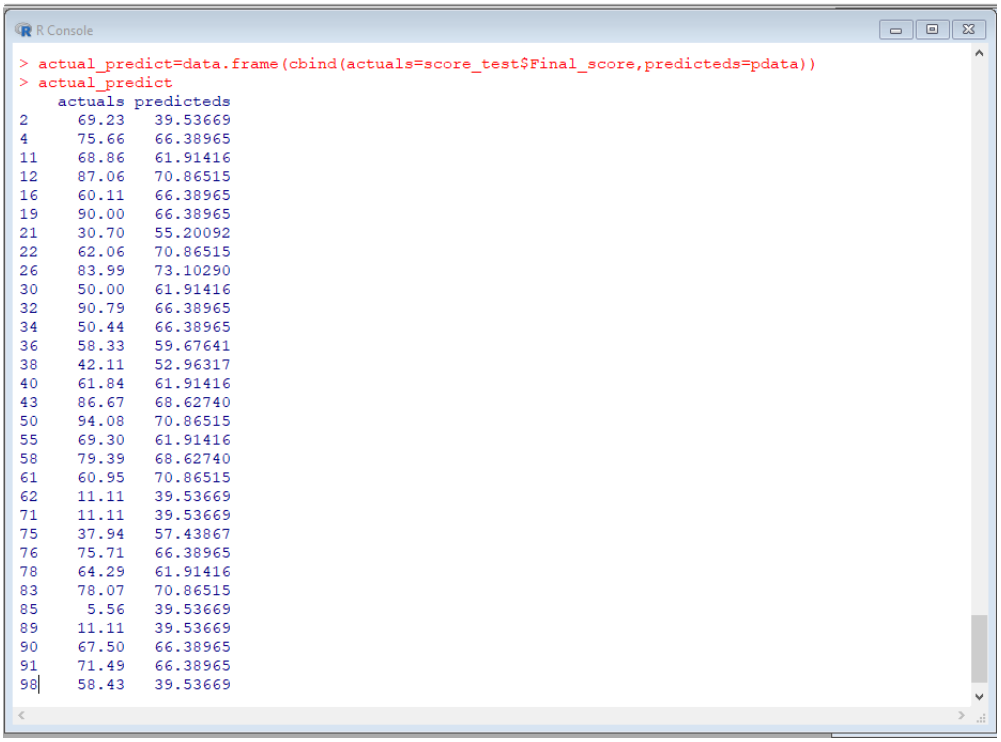
>score_test=[-s,]

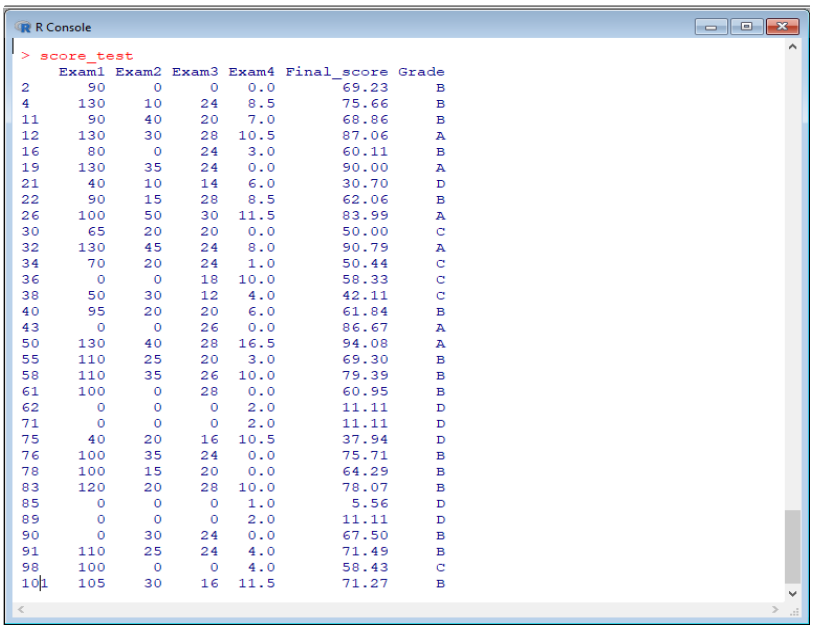
Score_tr



5. linmod =lm(Final_score-Exam3,data=score_tr)

print(linmod)






```
> cor(actual_predict$actual,actual_predict$predict)
[1] 0.7674963
>
```

```
> mape= mean(abs((actual_predict$predicteds - actual_predict$actual))/ actual_predict$actual)*100
> mape
[1] 60.6191
> mape= mean(abs((actual_predict$predicteds - actual_predict$actual))/ actual_predict$actual)
> mape
[1] 0.606191
>
```

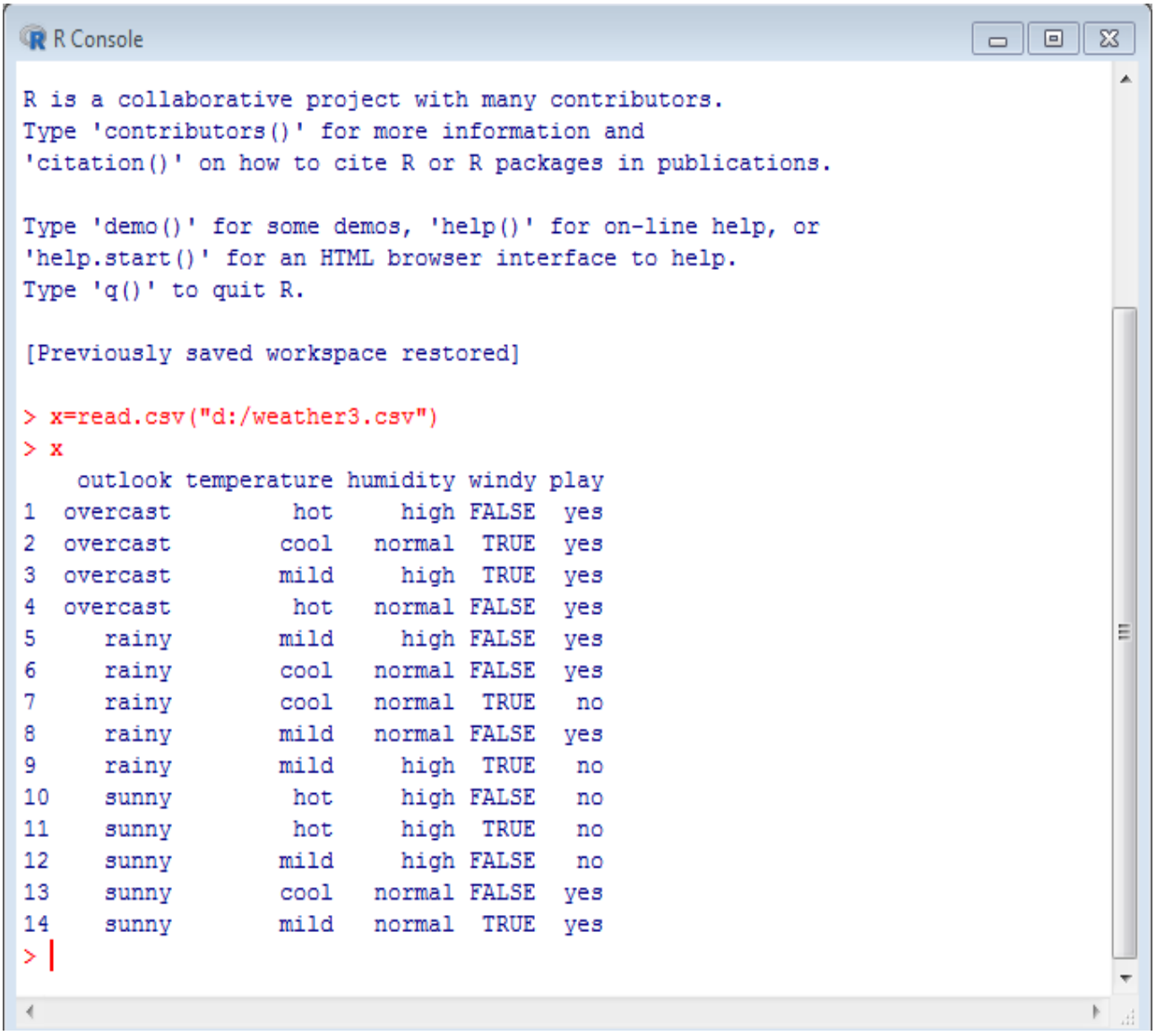
Practical N0: 10

Aim: Perform the logistic regression on the given data warehouse data

Step1: Open R studio -> Go on R console and type code.

1.X<-read.csv("C:/Users/Admin/Documents/SampleStudentData.csv")

> X



2.>x$humidity=ifelse(test=x$humidity=="high",yes=1,no=0)

>x

```
> x$humidity=ifelse(test=x$humidity=="high",yes=1,no=0)
> x
    outlook temperature humidity windy play
1  overcast         hot        1 FALSE  yes
2  overcast        cool        0  TRUE  yes
3  overcast        mild        1  TRUE  yes
4  overcast         hot        0 FALSE  yes
5     rainy        mild        1 FALSE  yes
6     rainy        cool        0 FALSE  yes
7     rainy        cool        0  TRUE   no
8     rainy        mild        0 FALSE  yes
9     rainy        mild        1  TRUE   no
10    sunny         hot        1 FALSE   no
11    sunny         hot        1  TRUE   no
12    sunny        mild        1 FALSE   no
13    sunny        cool        0 FALSE  yes
14    sunny        mild        0  TRUE  yes
```

3.x$play=ifelse(test=x$play=="yes",yes=1,no=0)

>x

```
> x$play=ifelse(test=x$play=="yes",yes=1,no=0)
> x
    outlook temperature humidity windy play
1  overcast         hot        1 FALSE    1
2  overcast        cool        0  TRUE    1
3  overcast        mild        1  TRUE    1
4  overcast         hot        0 FALSE    1
5     rainy        mild        1 FALSE    1
6     rainy        cool        0 FALSE    1
7     rainy        cool        0  TRUE    0
8     rainy        mild        0 FALSE    1
9     rainy        mild        1  TRUE    0
10    sunny         hot        1 FALSE    0
11    sunny         hot        1  TRUE    0
12    sunny        mild        1 FALSE    0
13    sunny        cool        0 FALSE    1
14    sunny        mild        0  TRUE    1
```

4.x$windy=ifelse(test=x$windy=="FALSE",yes=0,no=1)

>x

```
> x$windy=ifelse(test=x$windy=="FALSE",yes=0,no=1)
> x
    outlook temperature humidity windy play
1  overcast         hot        1     0    1
2  overcast        cool        0     1    1
3  overcast        mild        1     1    1
4  overcast         hot        0     0    1
5     rainy        mild        1     0    1
6     rainy        cool        0     0    1
7     rainy        cool        0     1    0
8     rainy        mild        0     0    1
9     rainy        mild        1     1    0
10    sunny         hot        1     0    0
11    sunny         hot        1     1    0
12    sunny        mild        1     0    0
13    sunny        cool        0     0    1
14    sunny        mild        0     1    1
> 
```

5. s=sample(nrow(x),.7*nrow(x))

>x_tr=x[s,]

>x_test=x[-s,]

>nrow(x)

>nrow(x_tr)

>nrow(x_test)

```
> s=sample(nrow(x),.7*nrow(x))
> x_tr=x[s,]
> x_test=x[-s,]
> nrow(x)
[1] 14
> nrow(x_tr)
[1] 9
> nrow(x_test)
[1] 5
> 
```

6.lmod=glm(play~windy,data=x_tr,family=binomial,control=list(maxit=100))

>lmod

```
> lmod=glm(play~windy,data=x_tr,family=binomial,control=list(maxit=100))
> lmod

Call:  glm(formula = play ~ windy, family = binomial, data = x_tr, control = list(maxit = 100))

Coefficients:
(Intercept)        windy
      20.57       -19.87

Degrees of Freedom: 8 Total (i.e. Null);  7 Residual
Null Deviance:      6.279
Residual Deviance: 3.819        AIC: 7.819
>
```

```
> summary(lmod)

Call:
glm(formula = play ~ windy, family = binomial, data = x_tr, control = list(maxit = 100))

Deviance Residuals:
     Min        1Q    Median        3Q       Max
-1.48230   0.00005   0.00005   0.00005   0.90052

Coefficients:
            Estimate Std. Error z value Pr(>|z|)
(Intercept)    20.57    7238.39   0.003    0.998
windy         -19.87    7238.39  -0.003    0.998

(Dispersion parameter for binomial family taken to be 1)

    Null deviance: 6.2790  on 8  degrees of freedom
Residual deviance: 3.8191  on 7  degrees of freedom
AIC: 7.8191

Number of Fisher Scoring iterations: 19

>
```

>summary(lmod)

```
> lmod=glm(play~humidity,data=x_tr,family=binomial,control=list(maxit=100))
> summary(lmod)

Call:
glm(formula = play ~ humidity, family = binomial, data = x_tr,
    control = list(maxit = 100))

Deviance Residuals:
     Min        1Q    Median        3Q       Max
-1.97277   0.00008   0.55525   0.55525   0.55525

Coefficients:
            Estimate Std. Error z value Pr(>|z|)
(Intercept)    1.792      1.080   1.659   0.0971 .
humidity      17.774   7604.236   0.002   0.9981
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

(Dispersion parameter for binomial family taken to be 1)

    Null deviance: 6.2790  on 8  degrees of freedom
Residual deviance: 5.7416  on 7  degrees of freedom
AIC: 9.7416

Number of Fisher Scoring iterations: 18

>
```

>lmod=glm(play~temperature,data=x_tr,family=binomial,control=list(maxit=100))

>summary(lmod)

```
> lmod=glm(play~temperature,data=x_tr,family=binomial,control=list(maxit=100))
> summary(lmod)

Call:
glm(formula = play ~ temperature, family = binomial, data = x_tr,
    control = list(maxit = 100))

Deviance Residuals:
     Min        1Q    Median        3Q       Max
-1.66511   0.00005   0.00005   0.75853   0.75853

Coefficients:
                Estimate Std. Error z value Pr(>|z|)
(Intercept)        1.099      1.155   0.951    0.341
temperaturehot    19.467  12537.265   0.002    0.999
temperaturemild   19.467  10236.634   0.002    0.998

(Dispersion parameter for binomial family taken to be 1)

    Null deviance: 6.2790  on 8  degrees of freedom
Residual deviance: 4.4987  on 6  degrees of freedom
AIC: 10.499

Number of Fisher Scoring iterations: 19

>
```

8.p=predict(lmod,x_test,type="response")

>p

```
> p=predict(lmod,x_test,type="response")
> p
           3            9           10           11           12
1.000000e+00 5.800756e-11 1.000000e+00 1.000000e+00 1.000000e+00
>
```