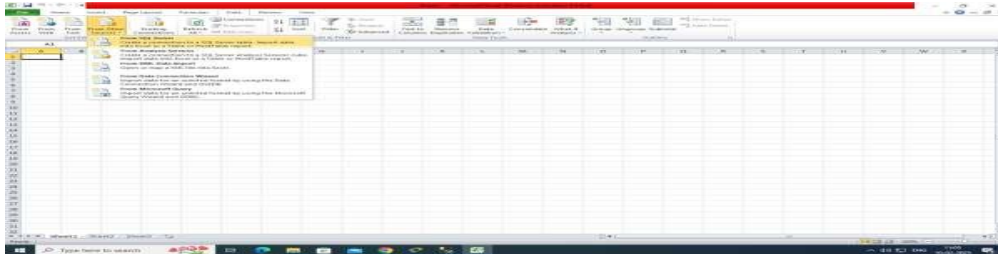


PRACTICAL NO. 01

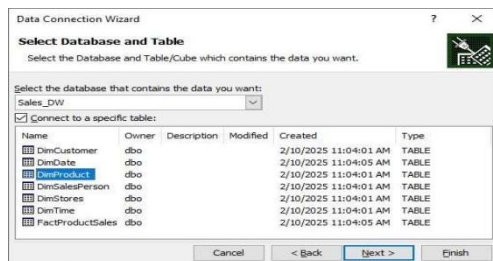
Aim: 1a) Import the data warehouse data in Microsoft Excel and create the Pivot Table and Pivot Chart.

Step 1: In excel, click on Data Menu -> from other sources->from SQL Server



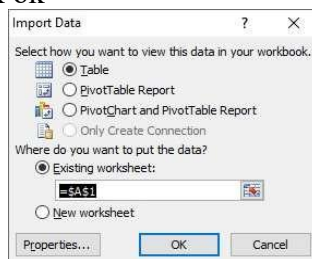
Step 2: Enter your SQL Server name and log on to window authentication and click next.

Step 3: In Data Connection Wizard, select database as Sales_DW and also select table.

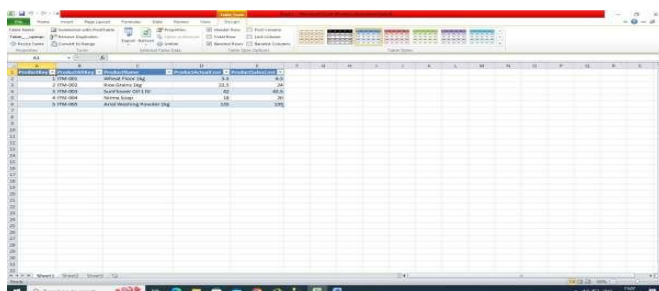


Step 4: Save file with default name and click finish.

Step 5: In this step, default Table Option and existing worksheet are already selected -> click ok

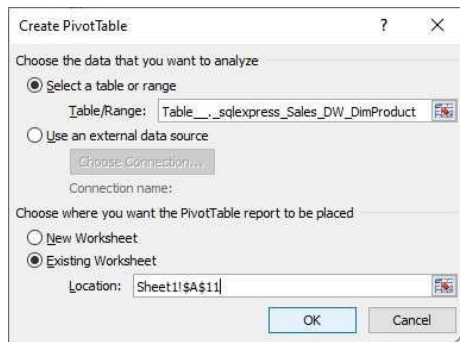


Step 6: It will open table in existing worksheet.

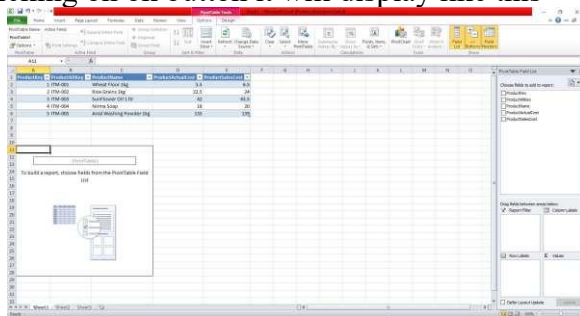


Step 7: Now to create pivot table click on Insert Menu -> Pivot Table

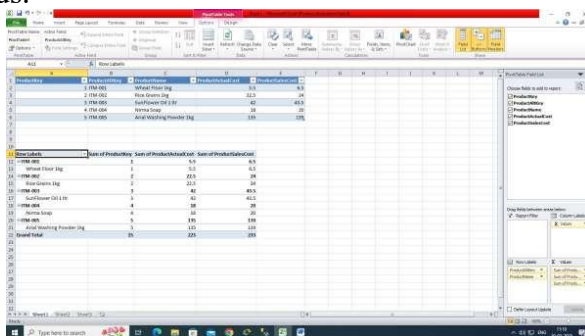
Step 8: Now in create pivot table window select existing worksheet option and also select proper location where you want to display pivot table and click ok.



After clicking on ok button it will display like this

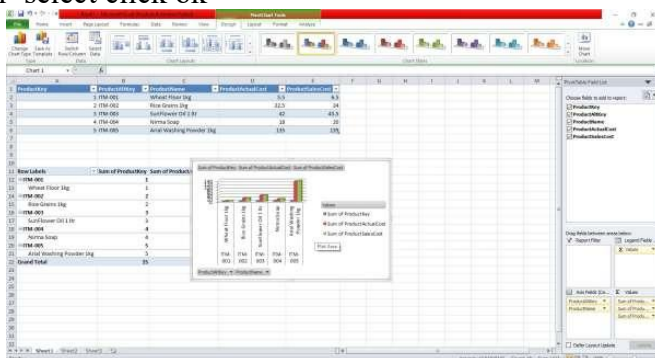


Step 9: In select pivot table fields we can choose fields to add to report -> here we are selecting all the fields.



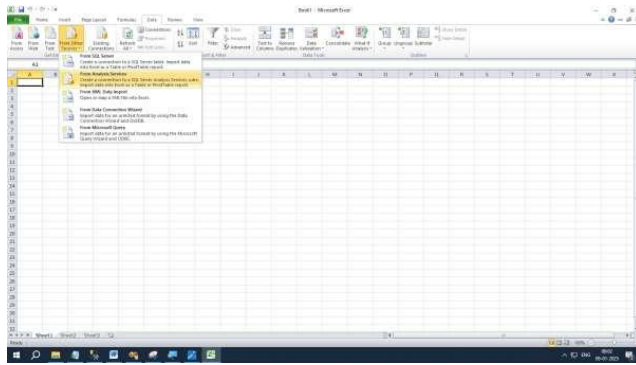
Step 10: To create Pivot Chart again go to Insert Menu-> Pivot Chart

Step 11: Here select chart type of your interest-> we are selecting default clustered column>select click ok



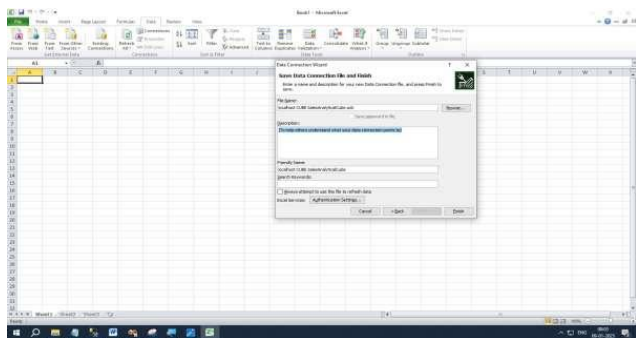
Aim: 1b) Import the cube in Microsoft Excel and create the Pivot Table and Pivot Chart to perform data analysis.

Step 1: First in excel click on data menu -> select from other sources -> from analysis services



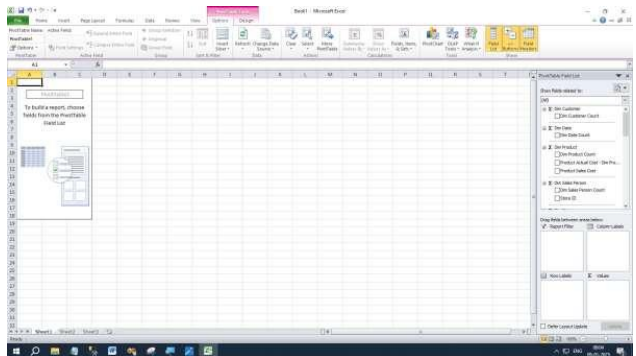
Step 2: Enter your SQL Server name and log on to windows authentication and click next

Step 3: In Data Connection Wizard, select database as cubepract4 and also select your cube name as SalesAnalyticalCube and click Next.



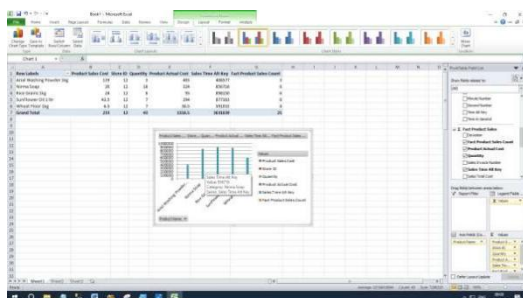
Step 4: Save file with default name and click finish

Step 5: Now click on ok



Step 6: In select pivot table fields we can choose fields to add to report -> here we are selecting all fields

Step 7: To create Pivot Chart again go to Insert Menu -> Pivot Chart and select pie chart -> click ok.



PRACTICAL NO. 02

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

A book store and have 100 books in storage. You sell a certain % for the highest price of \$50 and a certain % for the lower price of \$20.

=B4*(1-C4)				
	A	B	C	D
1	Book Store			
2				
3		total number of books	% sold for the highest price	
4		100	60%	
5				
6			number of books	unit profit
7		highest price	60	\$50
8		lower price	40	\$20
9				
10			total profit	\$3,800
11				

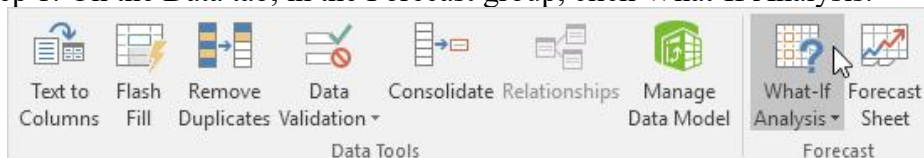
If you sell 60% for the highest price, cell D10 calculates a total profit of $60 * \$50 + 40 * \$20 = \$3800$.

Create Different Scenarios

But what if you sell 70% for the highest price? And what if you sell 80% for the highest price? Or 90%, or even 100%? Each different percentage is a different scenario. You can use the Scenario Manager to create these scenarios.

Note: You can simply type in a different percentage into cell C4 to see the corresponding result of a scenario in cell D10. However, what-if analysis enables you to easily compare the results of different scenarios. Read on.

Step 1: On the Data tab, in the Forecast group, click What-If Analysis.

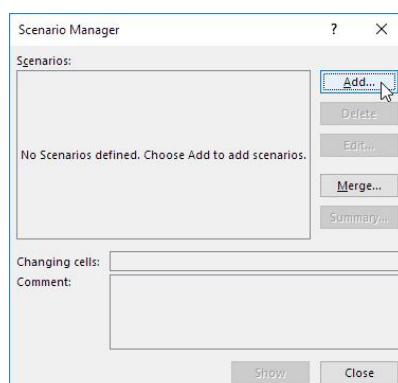


Step 2: Click Scenario Manager.

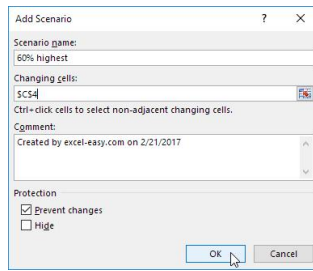


The Scenario Manager dialog box appears.

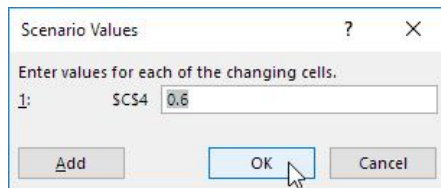
Step 3: Add a scenario by clicking on Add.



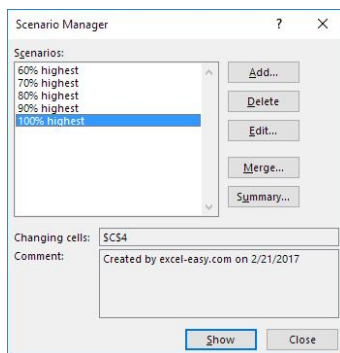
Step 4: Type a name (60% highest), select cell C4 (% sold for the highest price) for the Changing cells and click on OK.



Step 5: Enter the corresponding value 0.6 and click on OK again.



Step 6: Next, add 4 other scenarios (70%, 80%, 90% and 100%). Finally, your Scenario Manager should be consistent with the picture below:



PRACTICAL NO. 03

Aim: Perform the data classification using classification algorithm using R/Python.

Code:

Consider the annual rainfall details at a place starting from January 2012. We create an R time series object for a period of 12 months and plot it.

```
# Get the data points in form of a R vector.
rainfall <-
c(799,1174.8,865.1,1334.6,635.4,918.5,685.5,998.6,784.2,985,882.8,1071)

# Convert it to a time series object.
rainfall.timeseries <- ts(rainfall,start = c(2012,1),frequency = 12)

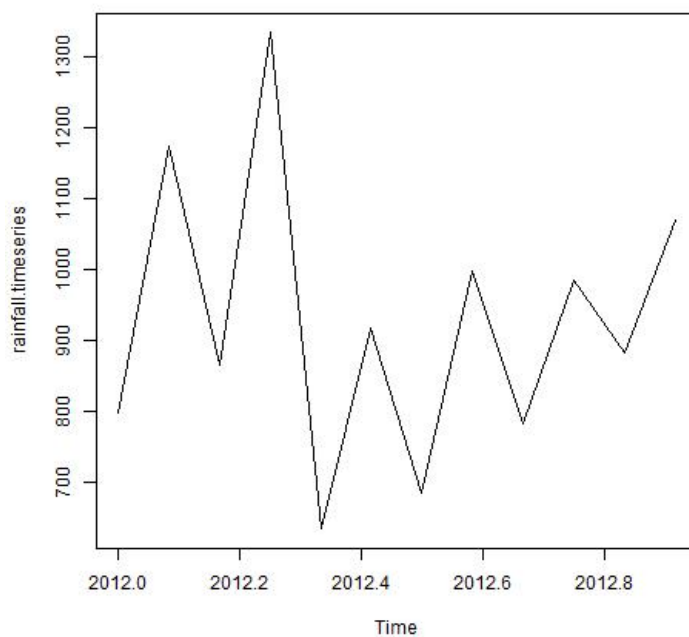
# Print the timeseries data.
print(rainfall.timeseries)

# Give the chart file a name.
png(file = "rainfall.png")

# Plot a graph of the time series.
plot(rainfall.timeseries)

# Save the file.
dev.off()
```

Output:



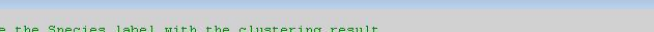
PRACTICAL NO. 04

Aim: Perform the data clustering using clustering algorithm using R/Python.

Code:

[illegible]

Compare the Species label with the clustering result




The screenshot shows an R Console window with the following text:

```
> #Compare the Species label with the clustering result
> table (iris$Species,kc$cluster)

      1  2  3
setosa  0  0 50
versicolor 48  2  0
virginica 14 36  0
```

Plot the clusters and their centre

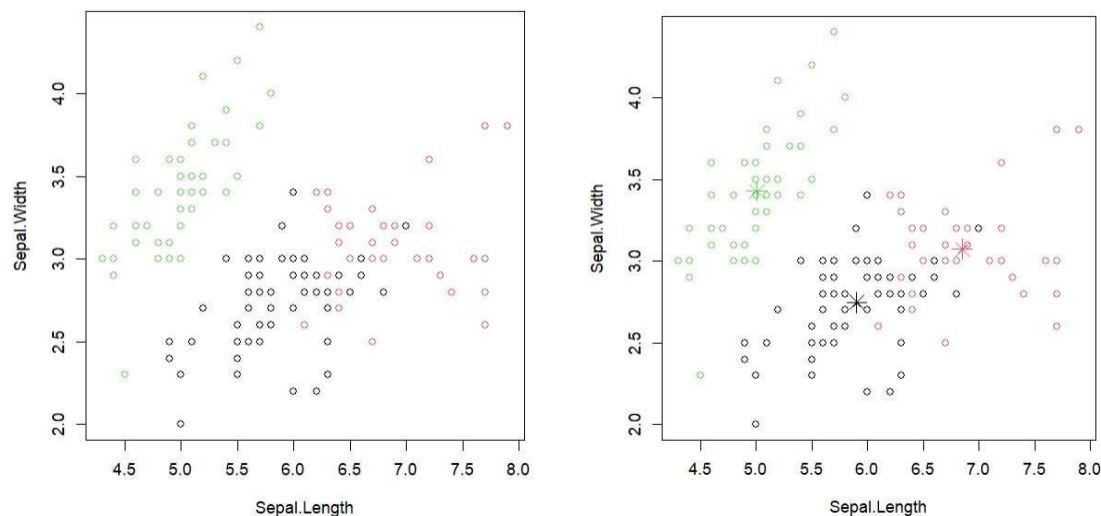


```

> # Plot the clusters and their centers
> plot(newiris[c("Sepal.Length", "Sepal.Width")], col=kc$cluster)
> points(kc$centers[,c("Sepal.Length", "Sepal.Width")], col=1:3, pch=8, cex=2)
>

```

Output:



PRACTICAL NO. 05

Aim: Perform Linear Regression on the given data warehouse data using R/Python.

Input Data

Below is the sample data representing the observations –

```
# Values of height
151, 174, 138, 186, 128, 136, 179, 163, 152, 131
# Values of weight.
63, 81, 56, 91, 47, 57, 76, 72, 62, 48
```

lm() Function

This function creates the relationship model between the predictor and the response variable.

Syntax

The basic syntax for lm() function in linear regression is –

```
lm(formula,data)
```

Following is the description of the parameters used –

- formula is a symbol presenting the relation between x and y.
- data is the vector on which the formula will be applied.

Create Relationship Model & get the Coefficients

```
x <- c(151, 174, 138, 186, 128, 136, 179, 163, 152, 131)
y <- c(63, 81, 56, 91, 47, 57, 76, 72, 62, 48)
```

```
# Apply the lm() function.
```

```
relation <- lm(y~x)
```

```
print(relation)
```

When we execute the above code, it produces the following result –

Call:

```
lm(formula = y ~ x)
```

Coefficients:

```
(Intercept)      x
-38.4551      0.6746
```

Get the Summary of the Relationship

```
x <- c(151, 174, 138, 186, 128, 136, 179, 163, 152, 131)
```

```
y <- c(63, 81, 56, 91, 47, 57, 76, 72, 62, 48)
```

```
# Apply the lm() function.
```

```
relation <- lm(y~x)
```

```
print(summary(relation))
```

When we execute the above code, it produces the following result –

call:

```
lm(formula = y ~ x)
```

Residuals:

Min	1Q	Median	3Q	Max
-6.3002	-1.6629	0.0412	1.8944	3.9775

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	-38.45509	8.04901	-4.778	0.00139 **
x	0.67461	0.05191	12.997	1.16e-06 ***

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 3.253 on 8 degrees of freedom
Multiple R-squared: 0.9548, Adjusted R-squared: 0.9491
F-statistic: 168.9 on 1 and 8 DF, p-value: 1.164e-06
predict() Function

Syntax

The basic syntax for predict() in linear regression is –
predict(object, newdata)

Following is the description of the parameters used –

- object is the formula which is already created using the lm() function.
- newdata is the vector containing the new value for predictor variable.

Predict the weight of new persons

```
# The predictor vector.  
x <- c(151, 174, 138, 186, 128, 136, 179, 163, 152, 131)  
# The response vector.  
y <- c(63, 81, 56, 91, 47, 57, 76, 72, 62, 48)  
  
# Apply the lm() function.  
relation <- lm(y~x)  
# Find weight of a person with height 170.  
a <- data.frame(x = 170)  
result <- predict(relation,a)  
print(result)
```

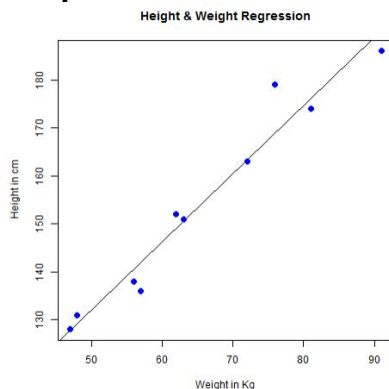
Result:

1
76.22869

Visualize the Regression Graphically

```
# Create the predictor and response variable.  
x <- c(151, 174, 138, 186, 128, 136, 179, 163, 152, 131)  
y <- c(63, 81, 56, 91, 47, 57, 76, 72, 62, 48)  
relation <- lm(y~x)  
# Give the chart file a name.  
png(file = "linearregression.png")  
# Plot the chart.  
plot(y,x,col = "blue",main = "Height & Weight Regression",  
abline(lm(x~y)),cex = 1.3,pch = 16,xlab = "Weight in Kg",ylab = "Height in  
cm")  
# Save the file.  
dev.off()
```

Output:



PRACTICAL NO. 06

Aim: Perform the Logistic Regression on the given data warehouse data using R/Python.

Create Regression Model

We use the glm() function to create the regression model and get its summary for analysis

```
> am.data=glm(formula = am~cyl+hp+wt,data = input,family = binomial())  
> print(summary(am.data))
```

Output:

Call: glm(formula = am ~ cyl + hp + wt, family = binomial(), data = input)

Deviance Residuals:

Min	1Q	Median	3Q	Max
-2.17272	-0.14907	-0.01464	0.14116	1.27641

Coefficients:

	Estimate	Std. Error	z value	Pr(> z)
(Intercept)	19.70288	8.11637	2.428	0.0152 *
cyl	0.48760	1.07162	0.455	0.6491 hp
	0.03259	0.01886	1.728	0.0840 . wt -
	9.14947	4.15332	-2.203	0.0276 *

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

(Dispersion parameter for binomial family taken to be 1)

Null deviance: 43.2297 on 31 degrees of freedom

Residual deviance: 9.8415 on 28 degrees of freedom

AIC: 17.841

Number of Fisher Scoring iterations: 8

PRACTICAL NO. 07

Aim: Write a Python Program to read data from CSV file, perform sample data analysis and generate basic insights. (Use pandas as python library).

1) Male and Female ratio on the Titanic

Code:

```
import pandas as pd

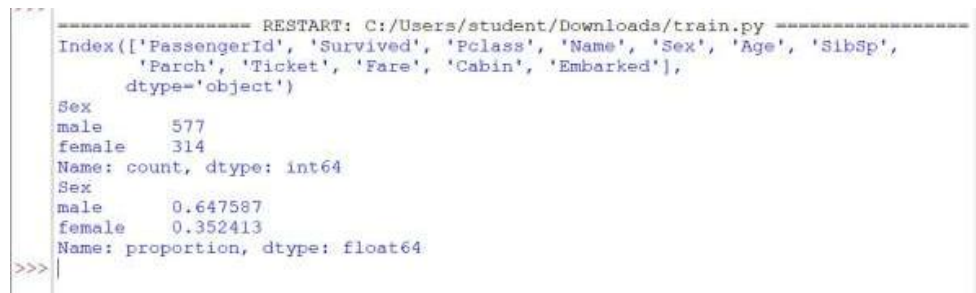
#load the csv file
df = pd.read_csv('train.csv')

#Columns Names
print(df.columns)

#Count unique values in Sex column
print(df['Sex'].value_counts())

#Percentage of male and female passengers
print(df['Sex'].value_counts(normalize=True))
```

Output:



```
----- RESTART: C:/Users/student/Downloads/train.py -----
Index(['PassengerId', 'Survived', 'Pclass', 'Name', 'Sex', 'Age', 'SibSp',
      'Parch', 'Ticket', 'Fare', 'Cabin', 'Embarked'],
      dtype='object')
Sex
male      577
female    314
Name: count, dtype: int64
Sex
male      0.647587
female    0.352413
Name: proportion, dtype: float64
>>>
```

Insights:

The above analysis shows that 65% of people on Titanic were Male and 35% were female.

2) Surviving male to female ratio on the titanic

Code:

```
import pandas as pd

#load the csv file
df = pd.read_csv('train.csv')

#Column Names
print(df.columns)

#Count unique values in Sex column
print(df[df["Survived"] == 1]['Sex'].value_counts())

#Percentage of surviving male and female
passengers print(df[df["Survived"] == 1]['Sex'].value_counts(normalize=True))
```

Output:

```
===== RESTART: C:/Users/student/Downloads/train.py =====
Index(['PassengerId', 'Survived', 'Pclass', 'Name', 'Sex', 'Age', 'SibSp',
      'Parch', 'Ticket', 'Fare', 'Cabin', 'Embarked'],
      dtype=object)
Sex
female    233
male      109
Name: count, dtype: int64
Sex
female    0.681287
male      0.318713
Name: proportion, dtype: float64
```

Insights:

The above analysis shows that 68% of surviving people on the Titanic were female.

3) Median age of each sex

Code:

```
import pandas as pd
#load the csv file
df = pd.read_csv('train.csv')

#median age of each sex
median_age_men=df[df['Sex']=='male']['Age'].median()
median_age_women=df[df['Sex']=='female']['Age'].median()

print(f"The median age of men is {median_age_men}")
print(f"The median age of women is {median_age_women}")
```

Output:

```
>>>
===== RESTART: C:/Users/student/Downloads/train.py =====
Index(['PassengerId', 'Survived', 'Pclass', 'Name', 'Sex', 'Age', 'SibSp',
      'Parch', 'Ticket', 'Fare', 'Cabin', 'Embarked'],
      dtype=object)
The median age of men is 29.0
The median age of men is 27.0
>>>
```

Insights:

The above analysis shows that median age of male was 29 whereas median age of female was 27.

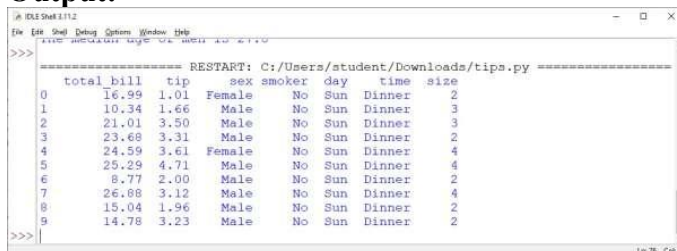
PRACTICAL NO. 08

Aim: 8a) Perform data visualization using Python on any sales data.

To install pandas library type, below command in the terminal.
pip3 install pandas

```
import pandas as pd
#reading the database
data = pd.read_csv("tips.csv")
#printing the top 10 rows.
print(data.head(10))
```

Output:

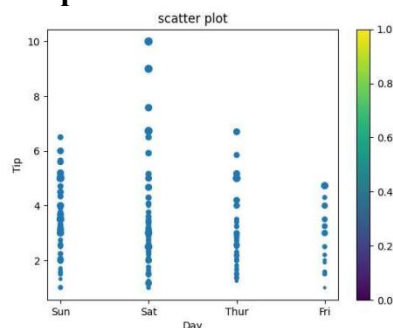


```
RESTART: C:/Users/student/Downloads/tips.py -----
total_bill  tip  sex  smoker  day  time  size
0    16.99   1.01 Female    No  Sun  Dinner    2
1    10.34   1.66  Male    No  Sun  Dinner    3
2    21.01   3.50  Male    No  Sun  Dinner    3
3    23.68   3.31  Male    No  Sun  Dinner    2
4    24.59   3.61 Female    No  Sun  Dinner    4
5    25.29   4.71  Male    No  Sun  Dinner    4
6     8.77   2.00  Male    No  Sun  Dinner    2
7    26.68   3.12  Male    No  Sun  Dinner    4
8    15.04   1.96  Male    No  Sun  Dinner    2
9    14.78   3.23  Male    No  Sun  Dinner    2
```

Scatter Plot

```
import pandas as pd
import matplotlib.pyplot as plt
#reading the database
data = pd.read_csv("tips.csv")
#Scatter plot with day against tip plt.scatter(data['day'], data['tip'],
c=data['size'], s=data['total_bill'])
#Adding Title to the plot
plt.title("Scatter plot")
#Setting the X and Y
labels plt.xlabel('Day')
plt.ylabel('Tip')
plt.colorbar() plt.show()
```

Output:

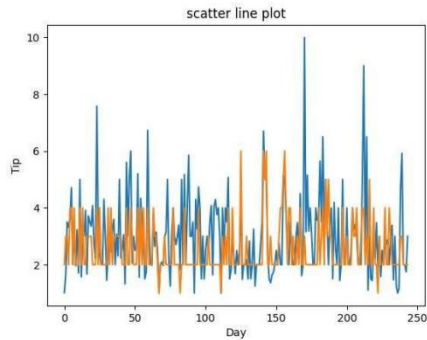


1) Line Chart

```
import pandas as pd
import matplotlib.pyplot as plt
#reading the database
data = pd.read_csv("tips.csv")
#Scatter plot with day against
tip plt.plot(data['tip'])
plt.plot(data['size'])
```

```
#Adding Title to the plot
plt.title("Scatter Plot")
#Setting the X and Y labels
plt.xlabel('Day')
plt.ylabel('Tip')
plt.show()
```

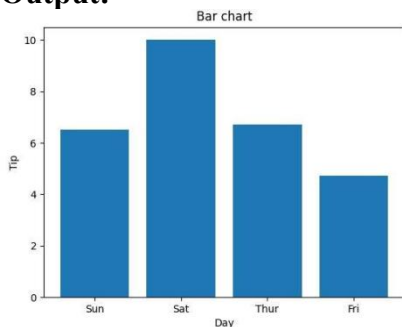
Output:



2) Bar Chart

```
import pandas as pd
import matplotlib.pyplot as plt
#reading the database
data = pd.read_csv("tips.csv")
#Bar chart with day against tip
plt.bar(data['day'], data['tip'])
plt.title("Bar Chart")
#Setting the X and Y labels
plt.xlabel('Day')
plt.ylabel('Tip')
#Adding the legends
plt.show()
```

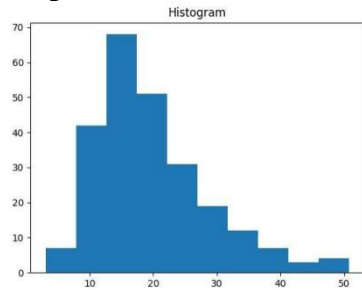
Output:



3) Histogram

```
import pandas as pd
import matplotlib.pyplot as plt
#reading the database
data = pd.read_csv("tips.csv")
#histogram of total_bills
plt.hist(data['total_bill'])
plt.title("Histogram")
#Adding the legends
plt.show()
```

Output:



4) Pie Chart

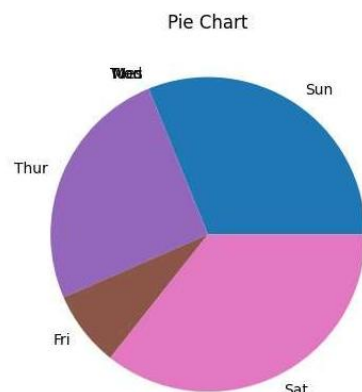
5)

```
import pandas as pd
import matplotlib.pyplot as plt
import array
df = pd.read_csv("tips.csv")

count_Sun = df['day'].value_counts().get('Sun',0)
print("Occurrences of 'Sun':",count_Sun) count_Mon = df['day'].value_counts().get('Mon',0)

print("Occurrences of 'Mon':",count_Mon) count_Tues
= df['day'].value_counts().get('Tues',0)
print("Occurrences of 'Tues':",count_Tues) count_Wed
= df['day'].value_counts().get('Wed',0)
print("Occurrences of 'Wed':",count_Wed) count_Thu =
df['day'].value_counts().get('Thu',0)
print("Occurrences of 'Thu':",count_Thu) count_Fri =
df['day'].value_counts().get('Fri',0)
print("Occurrences of 'Fri':",count_Fri) count_Sat =
df['day'].value_counts().get('Sat',0)
print("Occurrences of 'Sat':",count_Sat)
mylabel=["Sun","Mon","Tue","Wed","Thu","Fri","Sat"]
e=(0.2, 0, 0, 0, 0, 0, 0)
arr=array.array('i',[count_Sun,count_Mon,count_Tue,count_Wed,count_Thu,count_Fri,count_Sat])
plt.pie(arr,explode=e,labels=mylabel)
#Title to the plot
plt.title("Pie Chart")
plt.show()
```

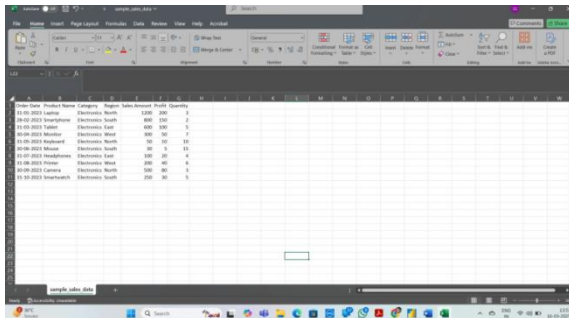
Output:



Aim: 8b) Perform data visualization using Power BI on any sales data

Step 1:- Download & Install Power BI

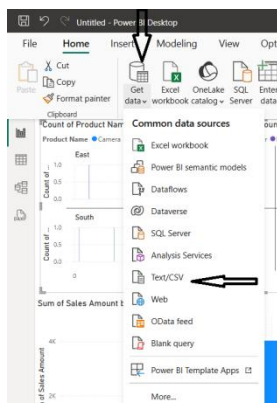
Step 2:- create a sample data



Order Date	Product Name	Category	Region	Sub-Region	Product ID	Quantity
11-01-2013	Laptop	Electronics	North	USA	1001	2
11-01-2013	Smartphone	Electronics	South	USA	1002	3
11-01-2013	Tablet	Electronics	East	USA	1003	5
11-01-2013	Monitor	Electronics	West	USA	1004	7
11-01-2013	Keyboard	Electronics	North	USA	1005	10
11-01-2013	Mouse	Electronics	South	USA	1006	15
11-01-2013	Printer	Electronics	West	USA	1007	8
11-01-2013	Camera	Electronics	North	USA	1008	4
11-01-2013	Smartwatch	Electronics	South	USA	1009	6

Step 3 :- Import the Sales Dataset

- Click "Home" > "Get Data".
- Choose your data source:
 - o Excel (XLSX/CSV)
 - o SQL Server
 - o Online Services (Google Sheets, SharePoint, etc.)
- Browse and select the dataset.
- Click "Load" to import.



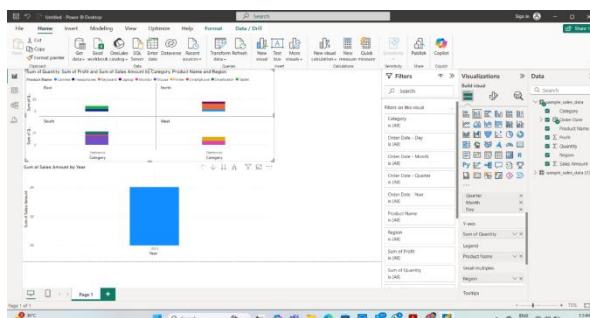
Step 4 :- Clean & Transform Data (Power Query)

Step 5 :- Create Data Visualizations

- Click on "Line Chart" in the "Visualizations" pane.
- Drag Order Date to the X-axis.
- Drag Sales Amount to the Y-axis.
- Customize the chart (format labels, add title, etc.).

Step 6:- Add Filters & Interactivity

Output:



PRACTICAL NO. 09

Aim: Create the Data staging area for the selected database using SQL.

Step 1:- Create a Staging Database

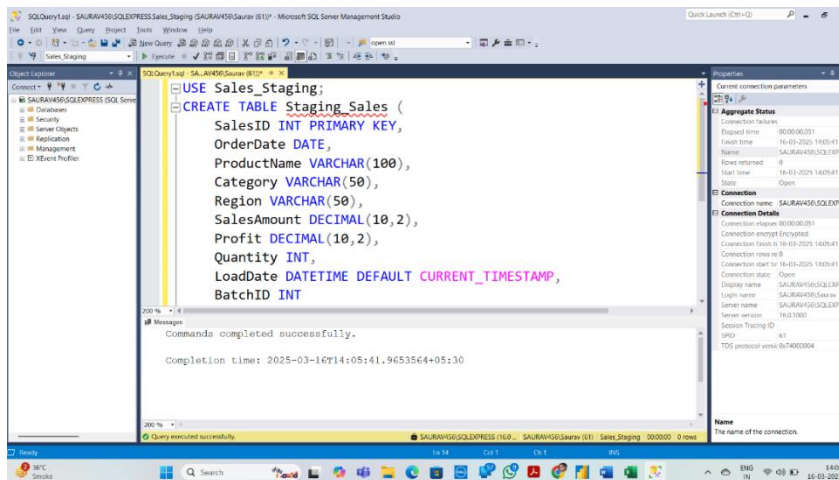
First, create a staging database to store raw sales data.

```
CREATE DATABASE Sales_Staging;  
USE Sales_Staging;
```

Step 2:- Create Staging Tables

Create tables that match the structure of raw sales data but include additional fields like load date and batch ID.

```
CREATE TABLE Staging_Sales (  
    SalesID INT PRIMARY KEY,  
    OrderDate DATE,  
    ProductName VARCHAR(100),  
    Category VARCHAR(50),  
    Region VARCHAR(50),  
    SalesAmount DECIMAL(10,2),  
    Profit DECIMAL(10,2),  
    Quantity INT,  
    LoadDate DATETIME DEFAULT CURRENT_TIMESTAMP,  
    BatchID INT );
```

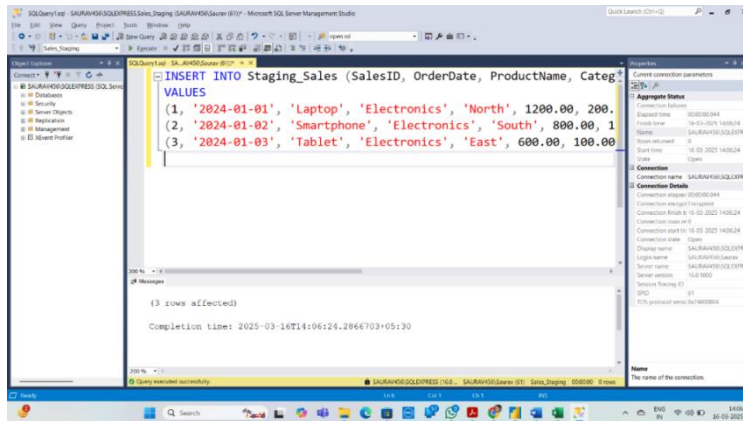


Step 3:- Load Raw Data into the Staging Table

Simulating data load from a CSV file, API, or external source:

```
INSERT INTO Staging_Sales (SalesID, OrderDate, ProductName, Category, Region, SalesAmount,  
Profit, Quantity, BatchID) VALUES
```

```
(1, '2024-01-01', 'Laptop', 'Electronics', 'North', 1200.00, 200.00, 3, 101),  
(2, '2024-01-02', 'Smartphone', 'Electronics', 'South', 800.00, 150.00, 2, 101),  
(3, '2024-01-03', 'Tablet', 'Electronics', 'East', 600.00, 100.00, 5, 101);
```



Step 4:- Perform Data Cleansing & Transformation

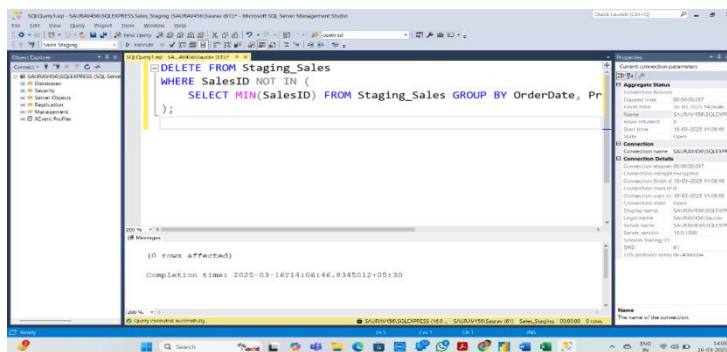
• Remove Duplicates

DELETE FROM Staging_Sales

WHERE SalesID NOT IN (

SELECT MIN(SalesID) FROM Staging_Sales GROUP BY OrderDate, ProductName, Region

);

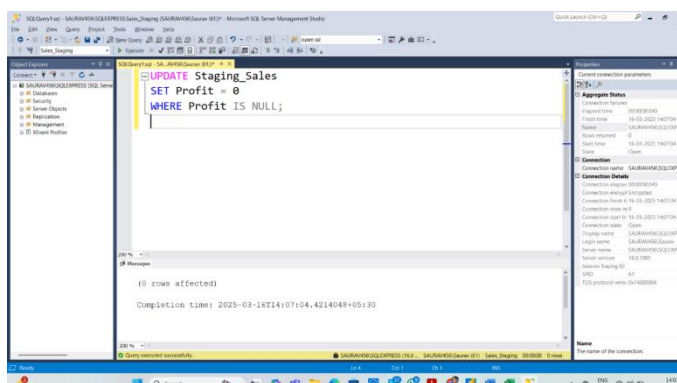


• Handle Null Values

UPDATE Staging_Sales

SET Profit = 0

WHERE Profit IS NULL;



Step 5:- Transfer Clean Data to the Final Sales Table Move the cleaned data into the Data Warehouse (DWH).

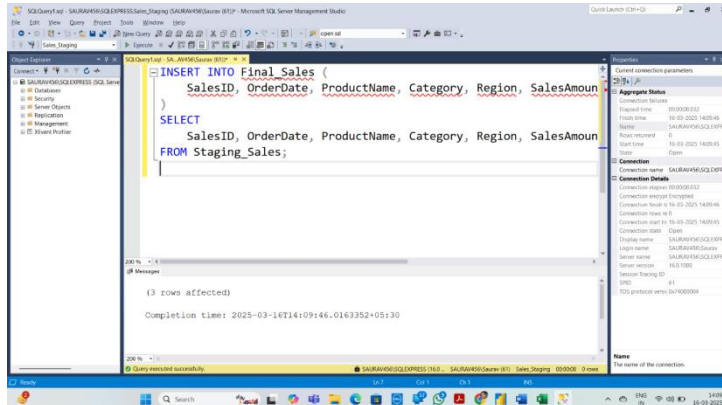
INSERT INTO Final_Sales (

SalesID, OrderDate, ProductName, Category, Region, SalesAmount, Profit, Quantity

)

SELECT

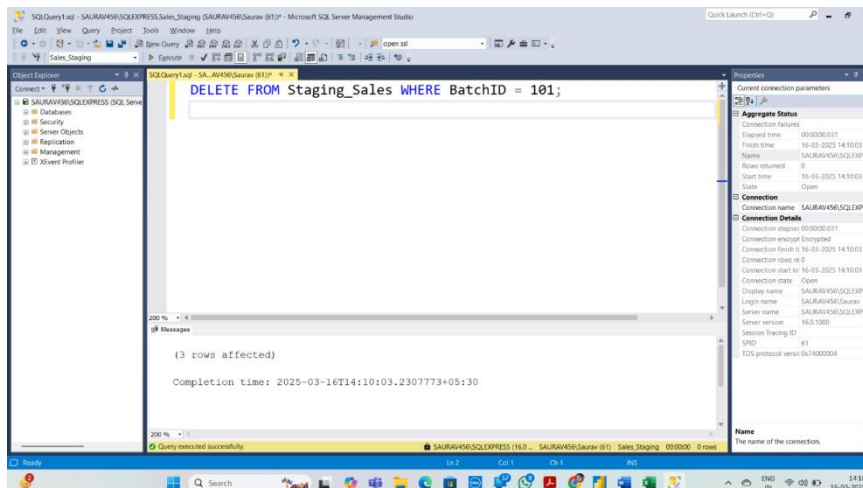
SalesID, OrderDate, ProductName, Category, Region, SalesAmount, Profit, Quantity
FROM Staging_Sales;



Step 6:- Archive or Delete Processed Data

Once data is loaded, either archive it or delete it from the staging area.

DELETE FROM Staging_Sales WHERE BatchID = 101;

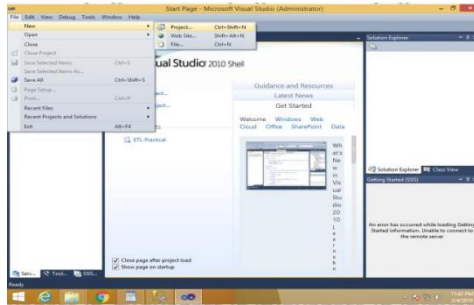


PRACTICAL NO. 10

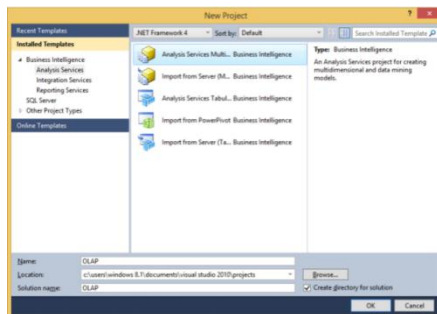
Aim: Create the cube with suitable dimension and fact tables based on ROLAP, MOLAP and HOLAP model.

Step 1: Creating Data Warehouse
CREATED IN PRACTICAL 9

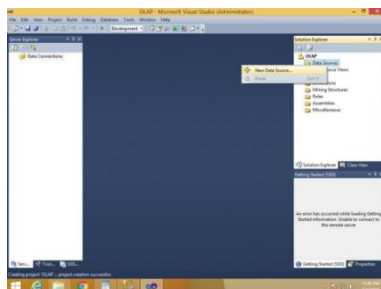
Step 2: Start SSDT environment and create New Data Source
Go to Sql Server Data Tools --> Right click and run as administrator
Click on File → New → Project



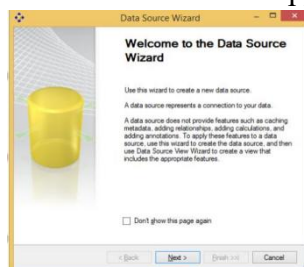
In Business Intelligence → Analysis Services Multidimensional and Data Mining models → appropriate project name → click OK



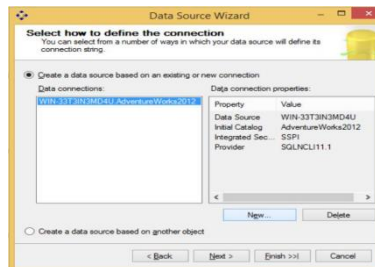
Right click on Data Sources in solution explorer → New Data Source



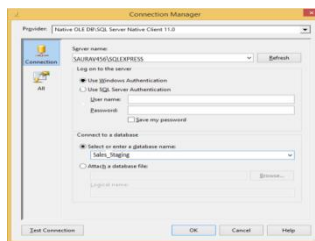
Data Source Wizard appears



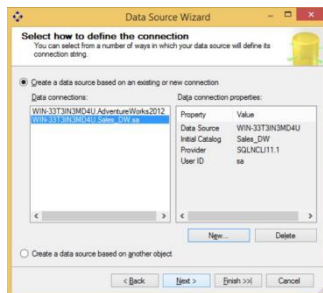
Click on New



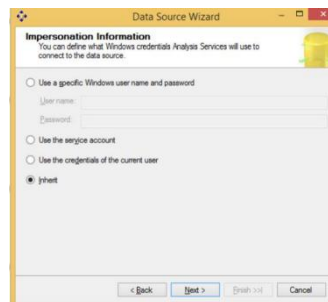
Select Server Name → select Use SQL Server Authentication → Select or enter a database name (Sales_Staging)



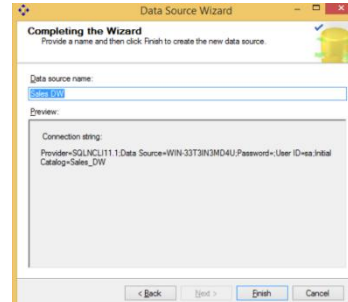
Click ok



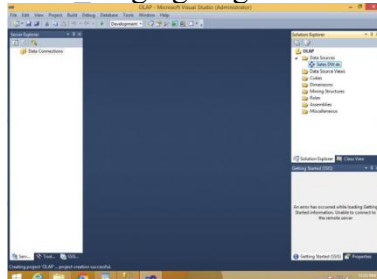
Select Inherit → Next



Click Finish

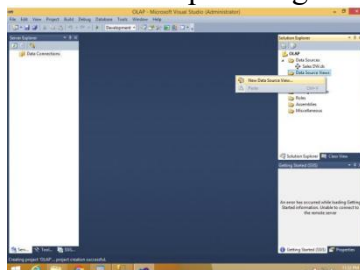


Sales_Staging.ds gets created under Data Sources in Solution Explorer



Step 3: Creating New Data Source View

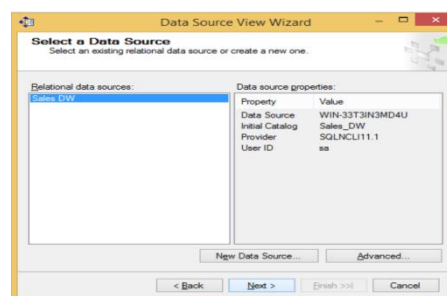
In Solution explorer right click on Data Source View → Select New Data Source View



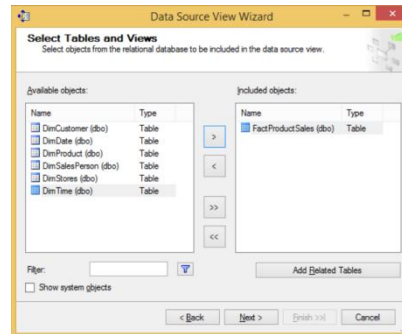
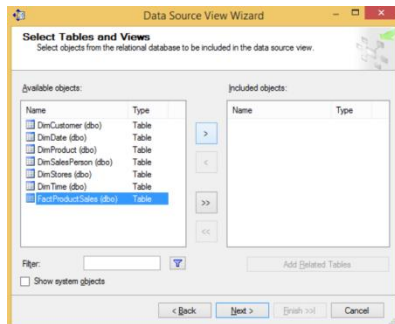
Click Next



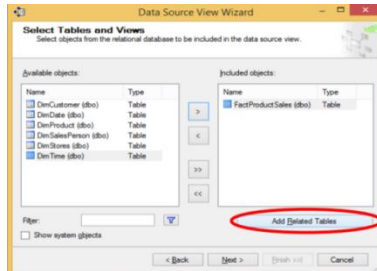
Click Next



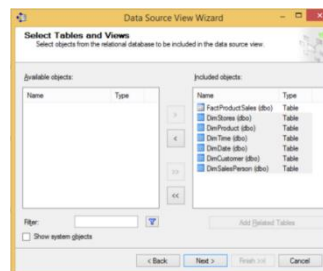
Select FactProductSales(dbo) from Available objects and put in Includes Objects by clicking on



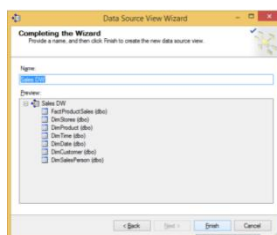
Click on Add Related Tables



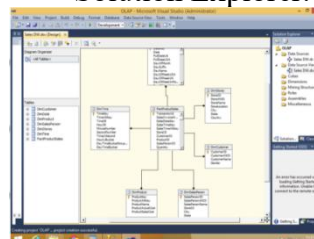
Click Next



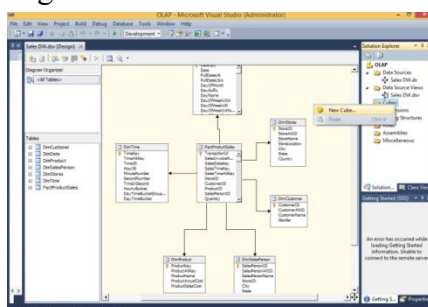
Click Finish



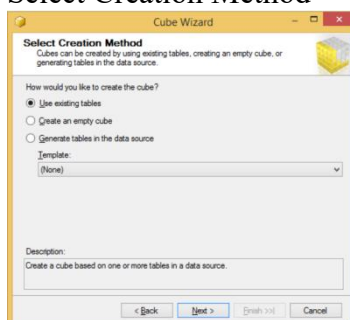
Sales_Staging.dsv appears in Data Source Views in Solution Explorer.



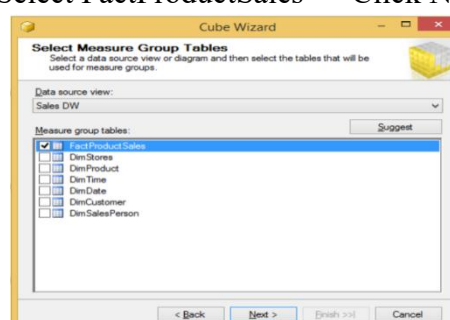
Step 4: Creating new cube
Right click on Cubes → New Cube



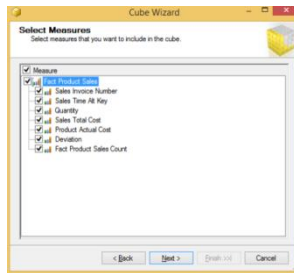
Select Use existing tables in
Select Creation Method → Next



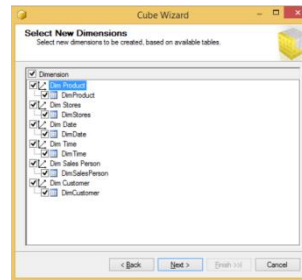
In Select Measure Group Tables →
Select FactProductSales → Click Next



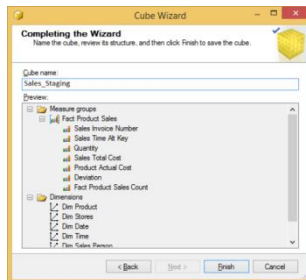
In Select Measures →
check all measures → Next



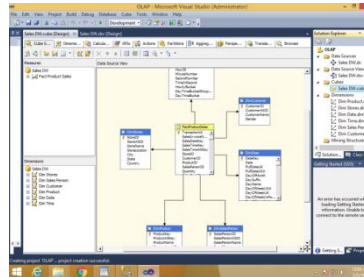
In Select New Dimensions → Check all
Dimensions → Next



Click on Finish

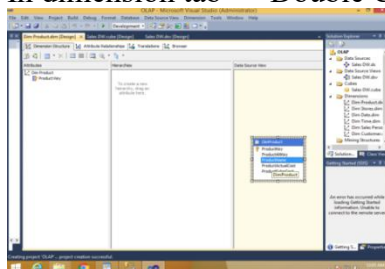


Sales_Staging.cube is created

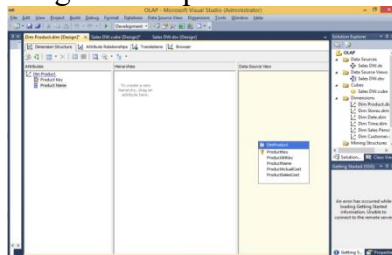


Step 5: Dimension Modification

In dimension tab → Double Click Dim Product.dim



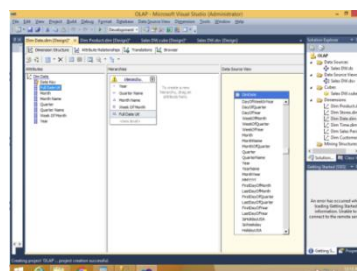
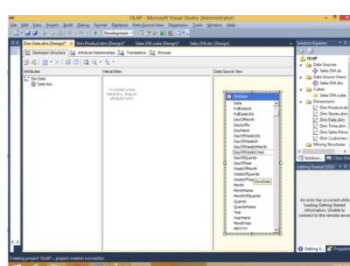
Drag and Drop Product Name from Table in Data Source View and Add in Attribute Pane at left side



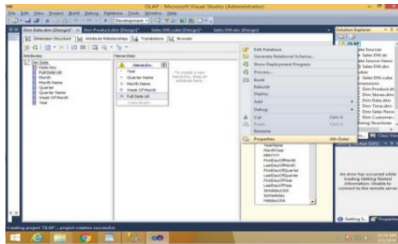
Step 6: Creating Attribute Hierarchy in Date Dimension

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.

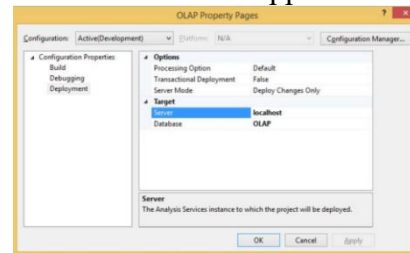
Drag fields in sequence from Attributes to Hierarchy window (Year, Quarter Name, Month Name, Week of the Month, Full Date UK)



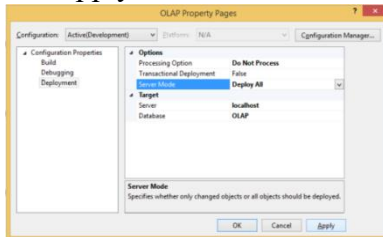
Step 7: Deploy Cube .
Right click on Project name → Properties



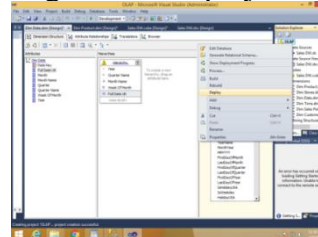
This window appears



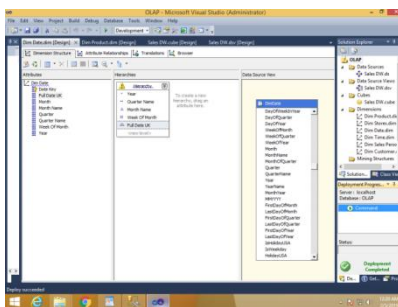
Do following changes and click on Apply & ok



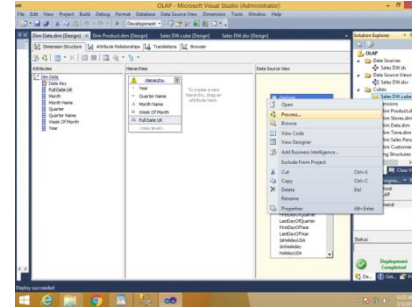
Right click on project name → Deploy



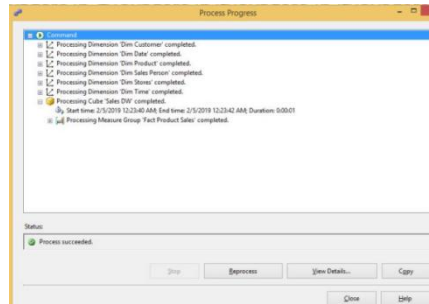
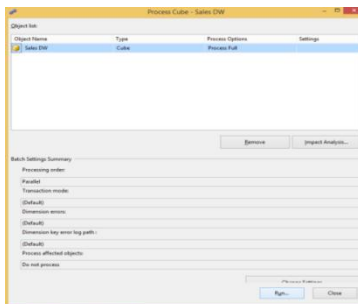
Deployment successful



To process cube right click on Sales_DW.cube → Process



Click run



Browse the cube for analysis in solution explorer

