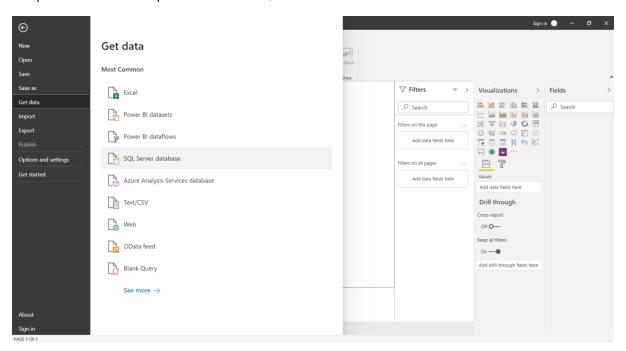
Practical No: - 1

**Aim:** - Import the legacy data from different sources such as (Excel, SqlServer, Oracle etc.) and load in the target system. (You can download sample database such as Adventureworks, Northwind, foodmart etc.)

#### Steps: -

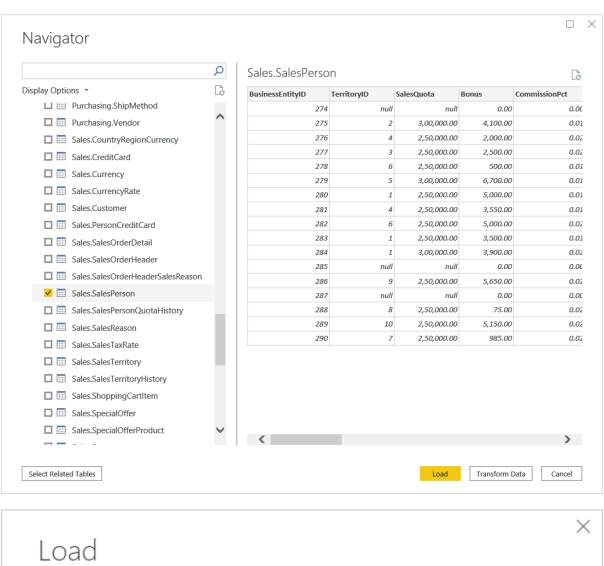
1. Open PowerBi Desktop -> Get Data -> SQL Server Database

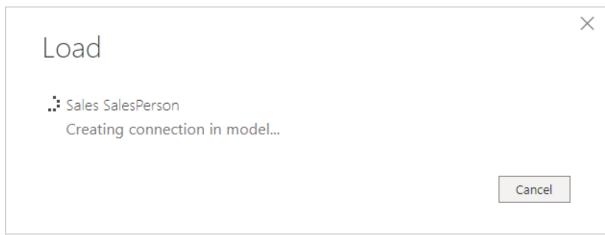


2. Type your server name -> OK

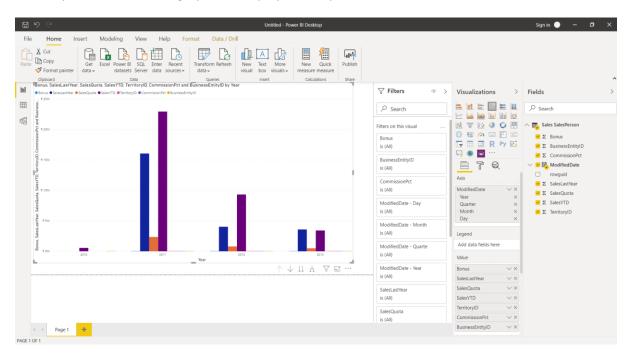


3. Navigator will show your database -> select any table -> Load





4. Finally convert data into graph and display the output.

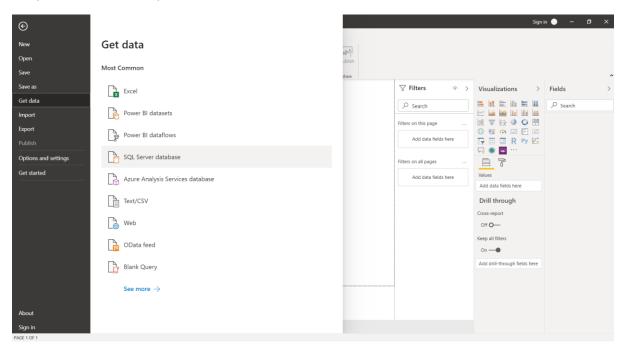


**Practical No: - 2** 

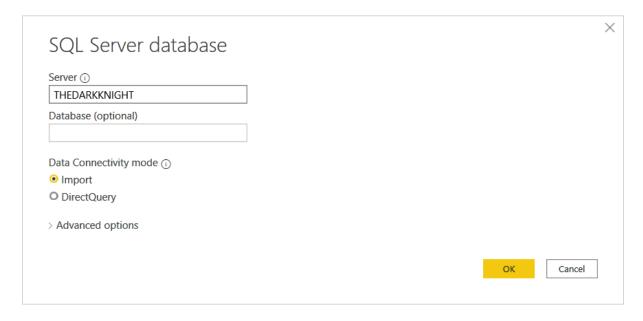
**Aim:** - Perform the Extraction Transformation and Loading (ETL) process to construct the database in the SqlServer.

#### Steps: -

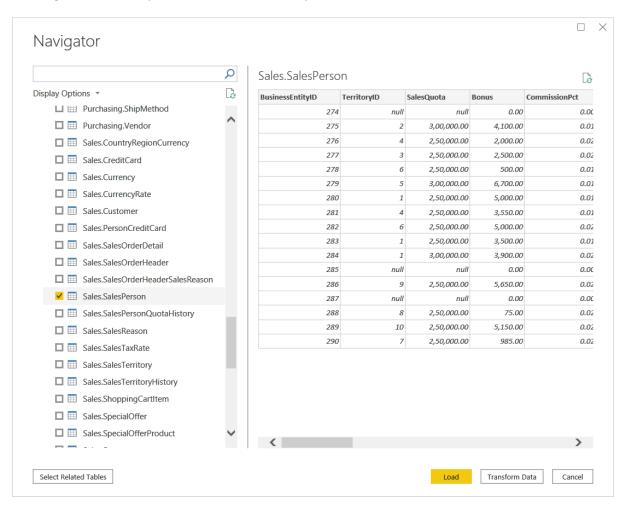
1. Open PowerBi Desktop -> Get Data -> SQL Server Database



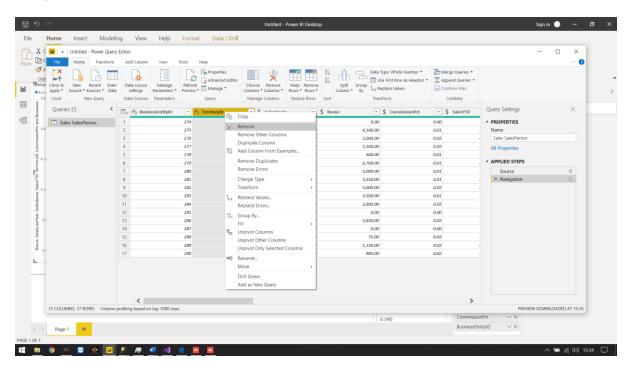
2. Type your server name -> OK



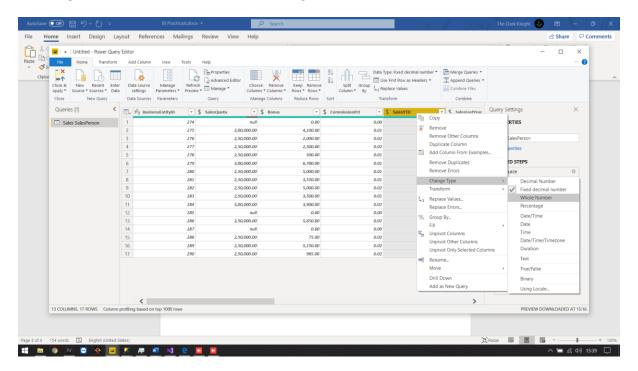
3. Navigator will show your database -> select any table -> Transform Data



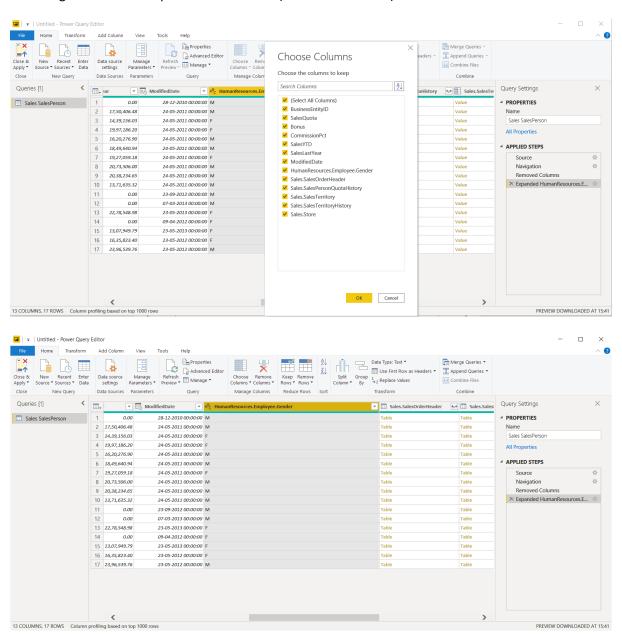
4. Manage Columns -> Remove Columns with null values and unnecessary data to display only what is required.



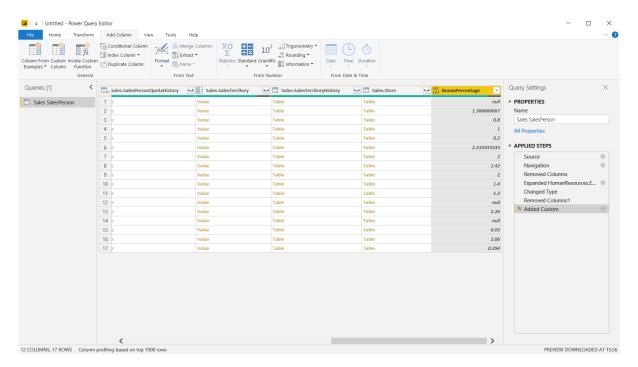
5. Manage Columns -> Convert Column -> Change data unit into Whole Number.



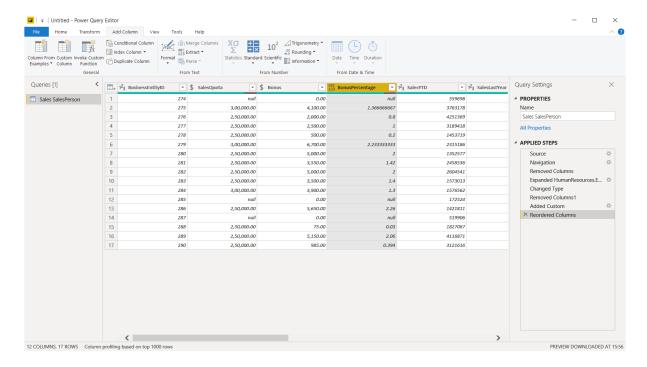
6. Manage Columns -> Expand another table (Human Resources) -> add selected columns of choice.

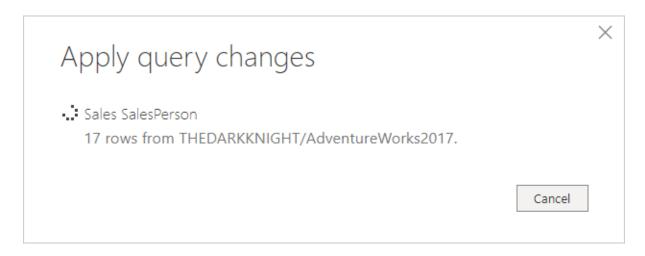


7. Add Custom Column -> Calculate a new column value (Bonus Percentage)

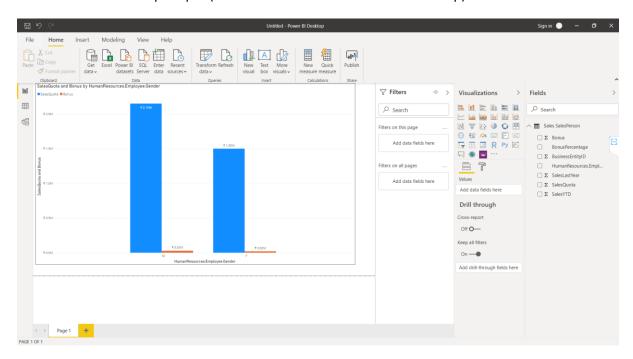


#### 8. Rename & Reorder the columns





9. Show the relationship output (Gender Based Sales & Bonus Relationship)

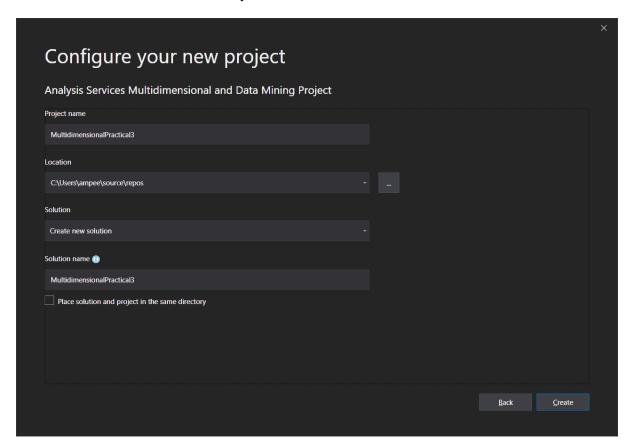


**Practical No: - 3** 

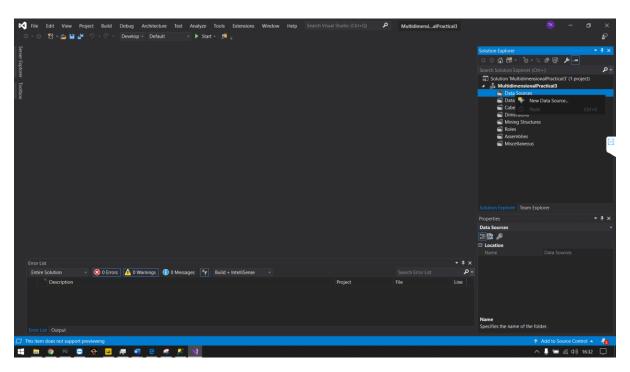
**Aim:** - Create the Data staging area for the selected database & Create the cube with suitable dimension and fact tables based on ROLAP, MOLAP and HOLAP model.

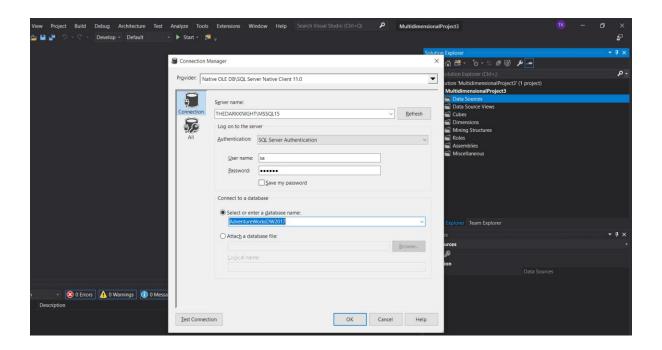
#### Steps: -

1. Create a new Multidimensional Project.

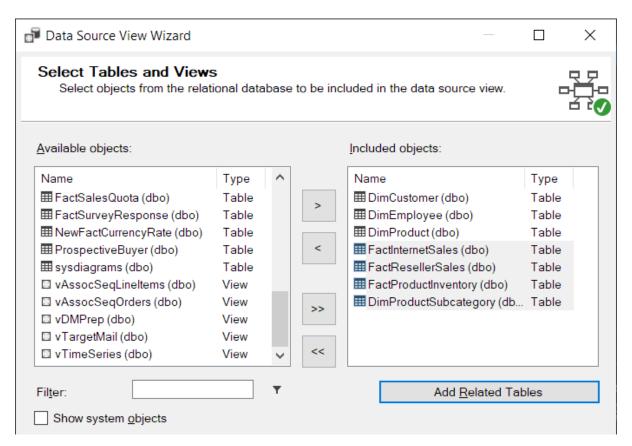


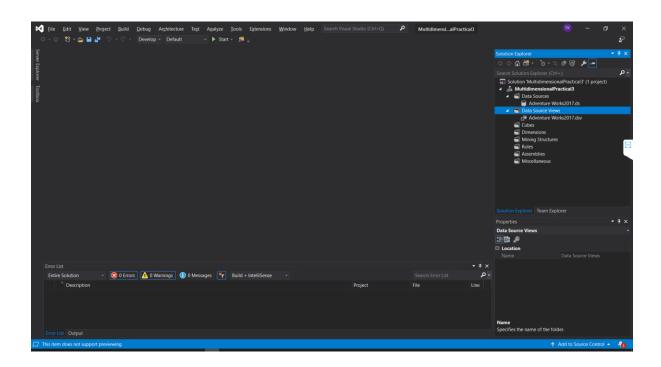
2. Go to DataSource & add a new source.



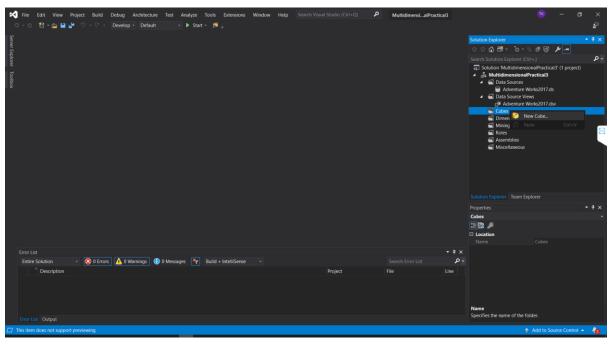


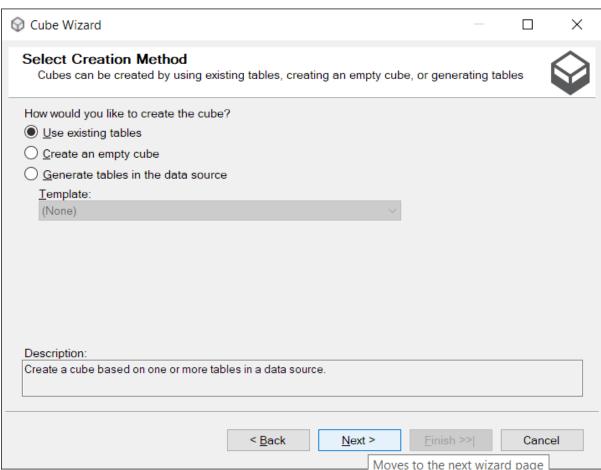
3. Create a view from the added Source.

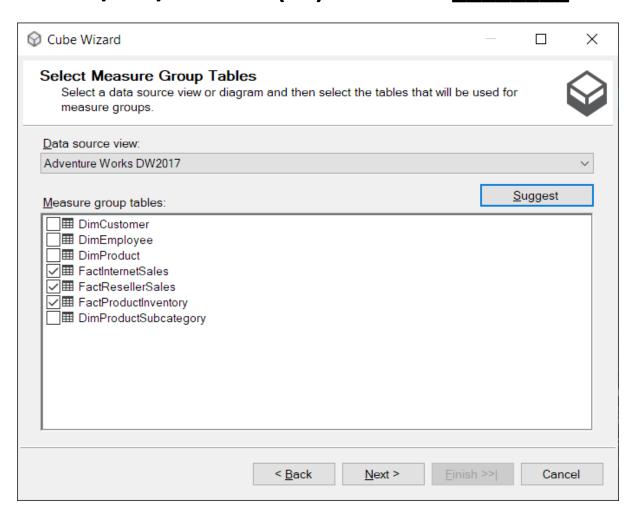




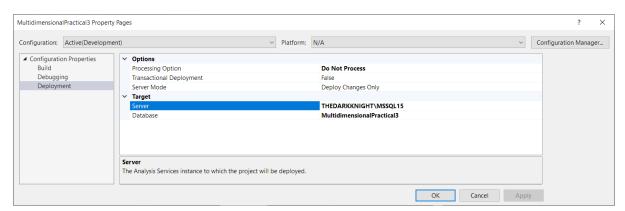
4. Create a new cube to generate data to be displayed.



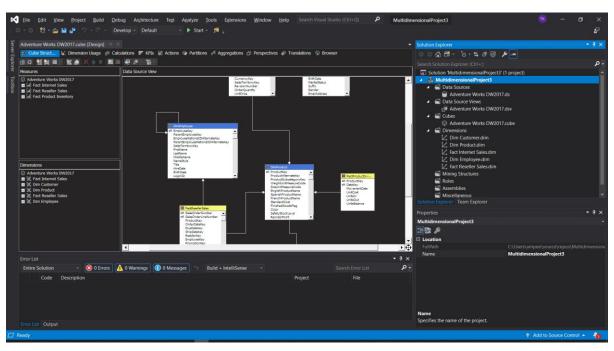


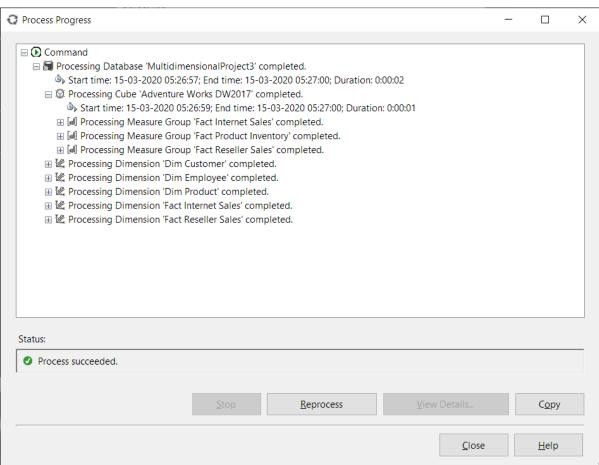


5. Change properties of Multidimensional Project.

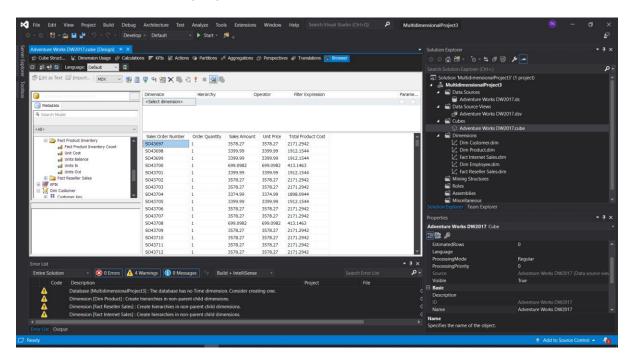


6. Show Deployment Process.





6. Click on browse and fire a query.



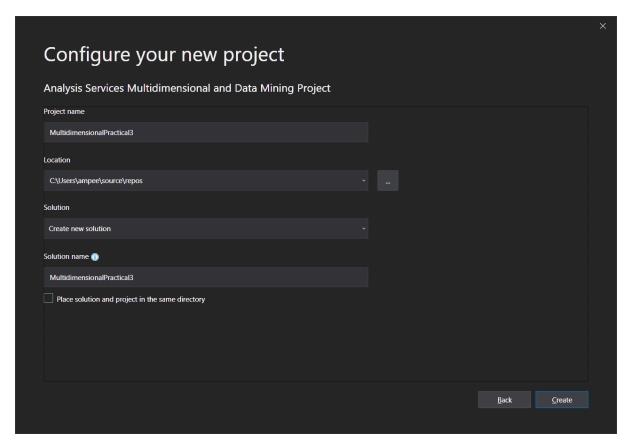
Practical No: - 4

**Aim:** - a. Create the ETL map and setup for execution. b. Execute the MDX queries to extract the data from the Datawarehouse.

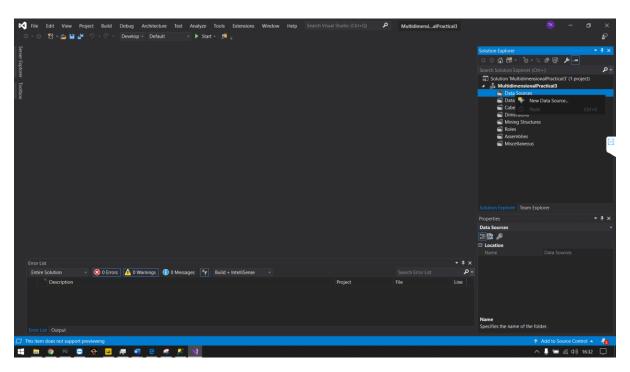
Steps: -

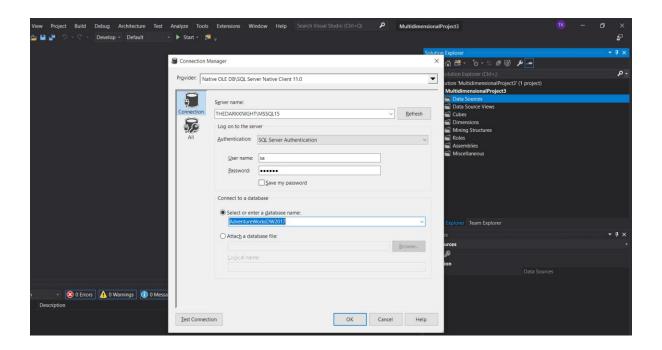
#### PART A -> Create the ETL map and setup for execution

1. Create a new Multidimensional Project.

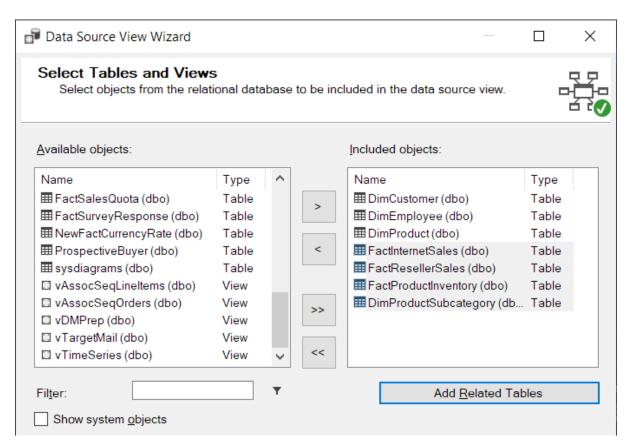


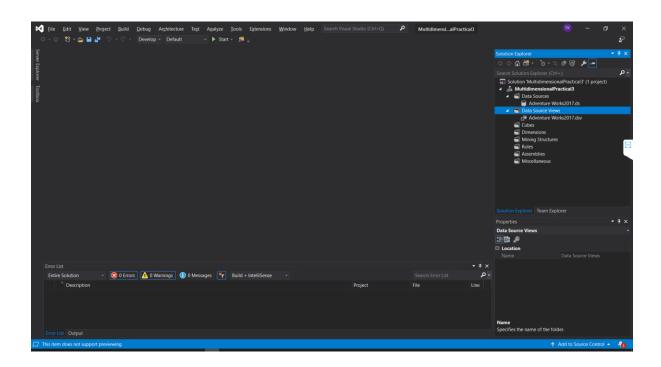
2. Go to DataSource & add a new source.



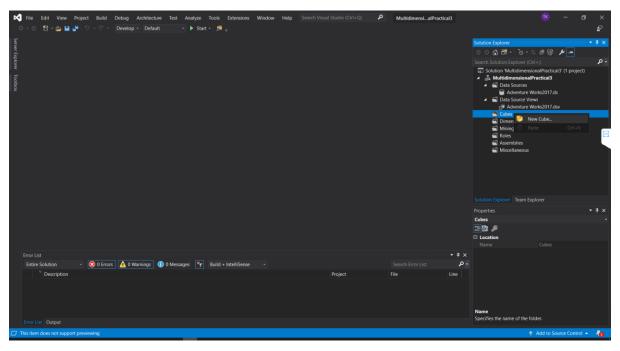


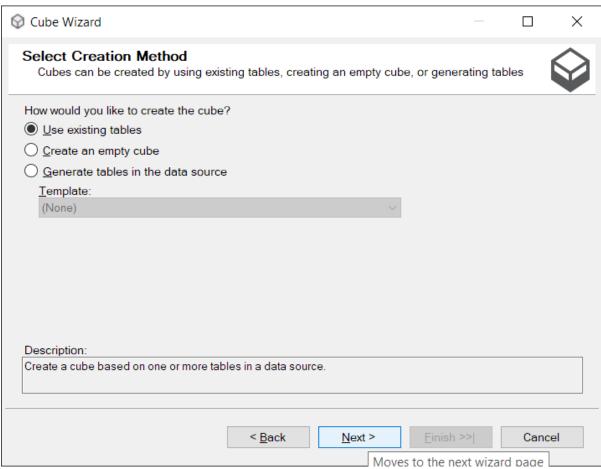
3. Create a view from the added Source.

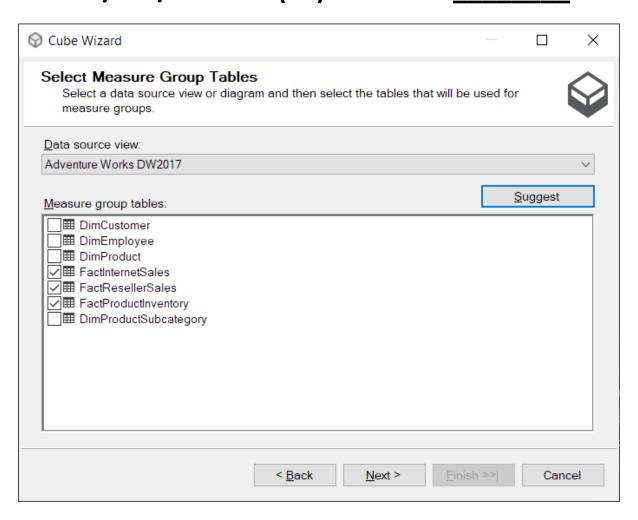




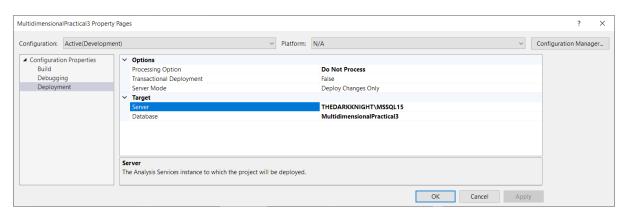
4. Create a new cube to generate data to be displayed.



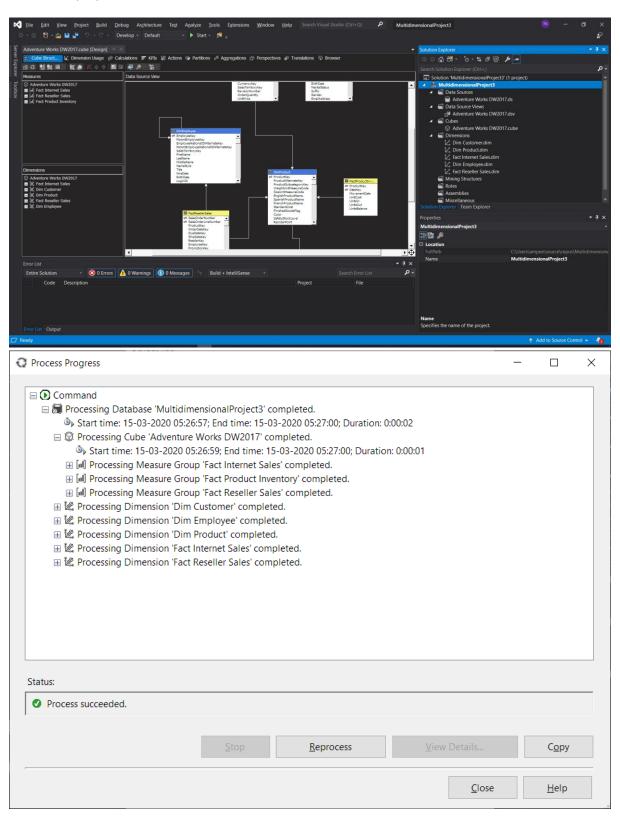




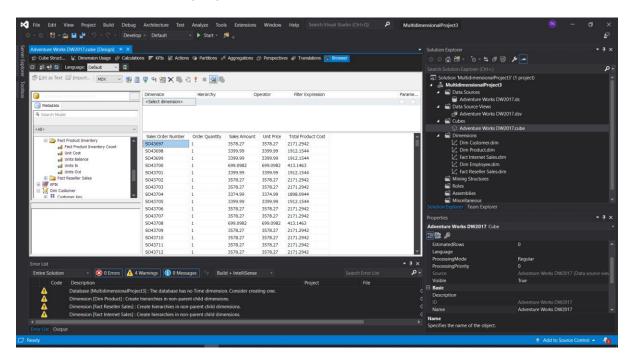
5. Change properties of Multidimensional Project.



6. Show Deployment Process.

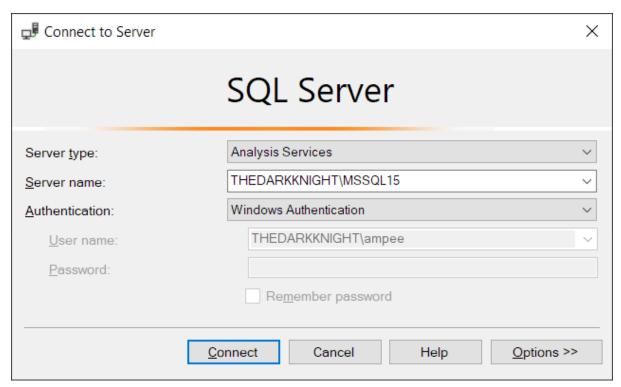


6. Click on browse and fire a query.

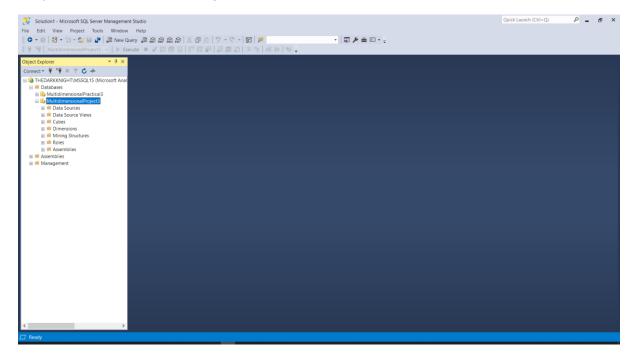


PART B -> Execute MDX Query to extract data from Warehouse.

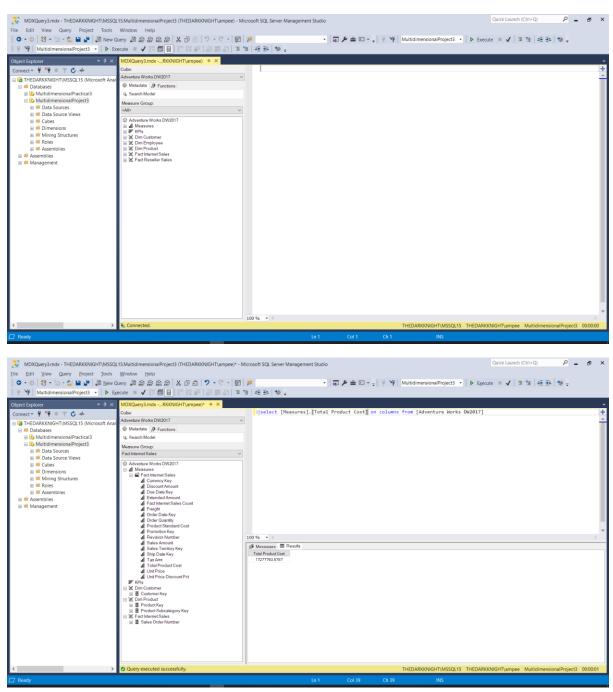
1. Open SQL Server Management Studio -> Select Analysis Server Engine -> Connect.



2. Open the Multi Dimension Project from earlier and load it.



3. Execute MDX Query to fetch result from Data Warehouse. (Select total production cost).



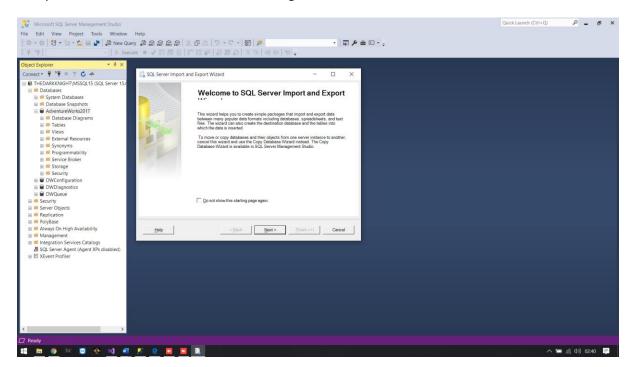
Practical No: - 5

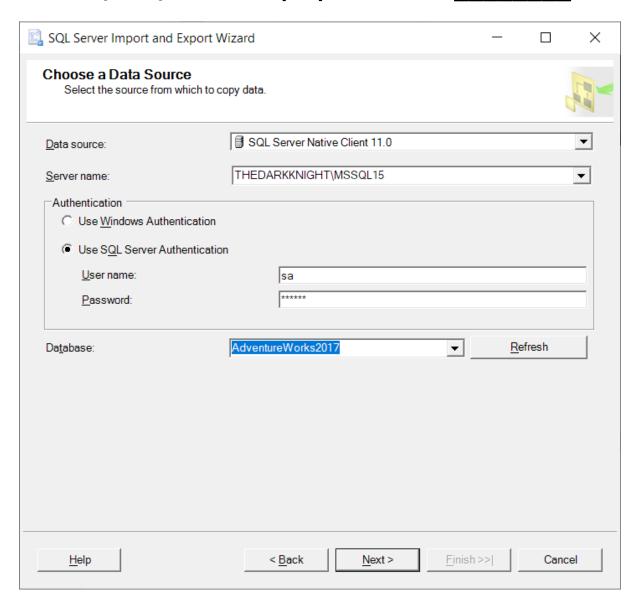
**Aim:** - a. Import the Datawarehouse data in Microsoft Excel and create the Pivot table and Pivot Chart. b. Import the cube in Microsoft Excel and create the Pivot table and Pivot Chart to perform data analysis.

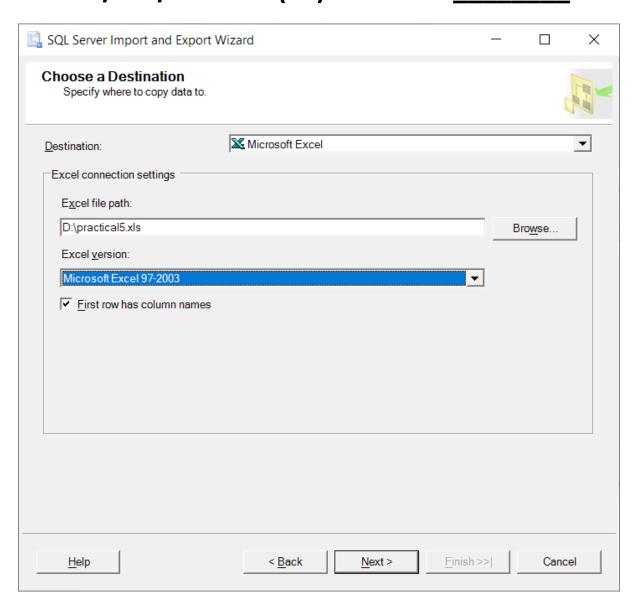
PART A -> Import the Datawarehouse data in Microsoft Excel and create the Pivot table and Pivot Chart

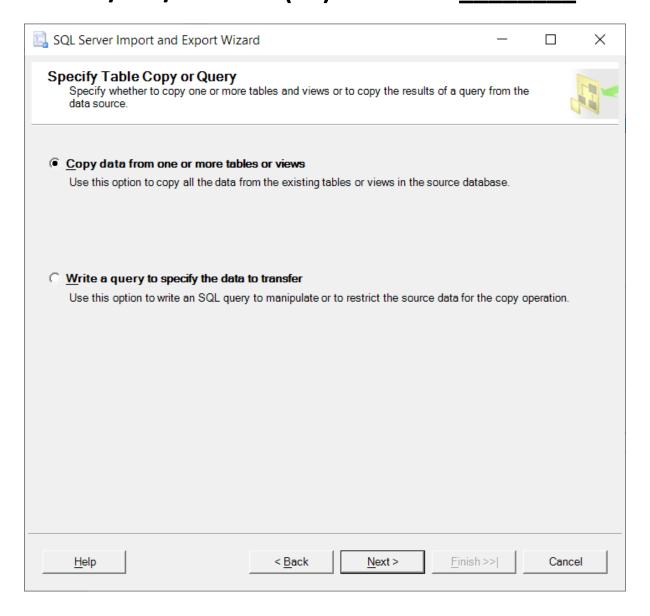
#### Steps: -

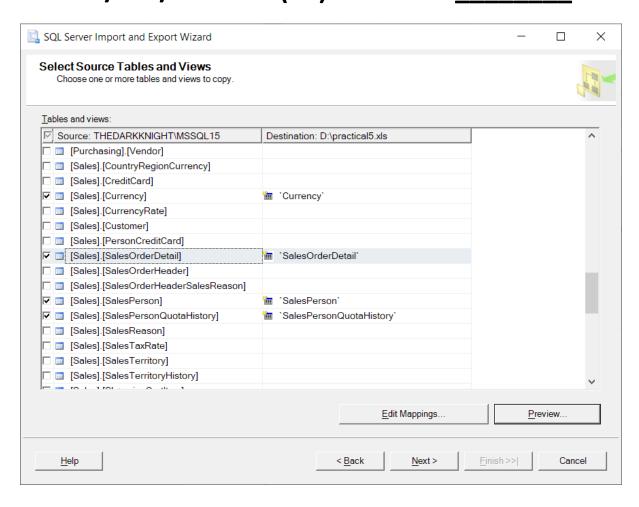
1. Export dB from Microsoft SQL Server Management Studio.

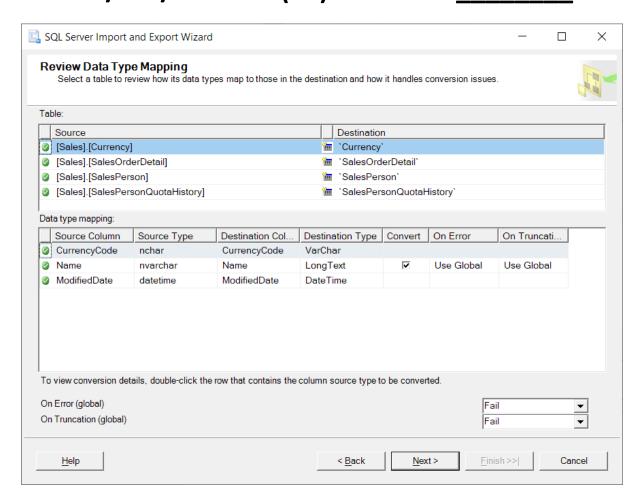


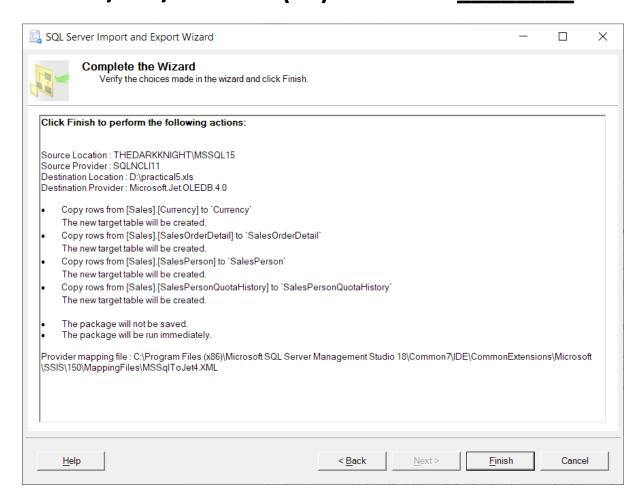




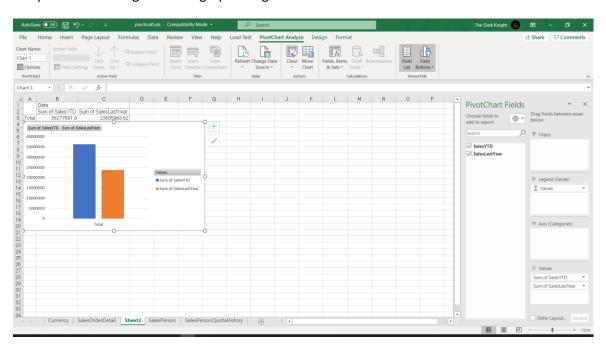








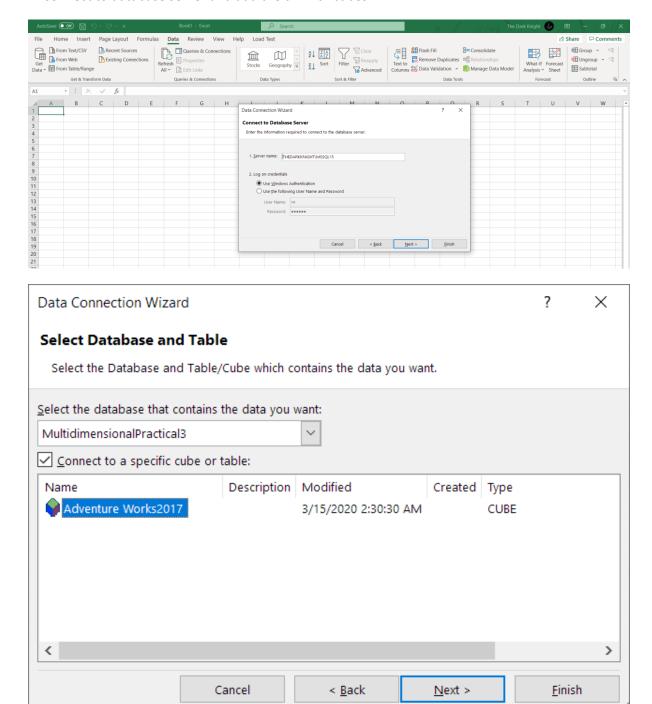
2. Open excel file & generate a graph using the data available.



PART B -> Import the cube in Microsoft Excel and create the Pivot table and Pivot Chart to perform data analysis

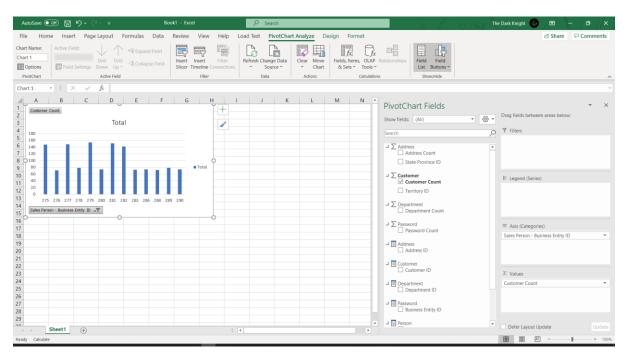
#### Steps: -

1. Connect to database server and add the dB with cubes.



| Data Connection Wizard  | ?            | ×    |
|---|--------------|------|
| Save Data Connection File and Finish  |              |      |
| Enter a name and description for your new Data Connection file, and press Finish to save. |              |      |
| File Name:  |              |      |
| THEDARKKNIGHT_MSSQL15 MultidimensionalPractical3 Adventure Works2017.odc                  | B <u>r</u> o | wse  |
| Save password in file   |              |      |
| <u>D</u> escription:  |              |      |
| Friendly Name:  |              |      |
| THEDARKKNIGHT_MSSQL15 MultidimensionalPractical3 Adventure Works2017                      | 1            |      |
| Search Keywords:  Always attempt to use this file to refresh data                         |              |      |
| Excel Services: A <u>u</u> thentication Settings  |              |      |
| Cancel < <u>B</u> ack Next >  | <u>E</u> i   | nish |

#### 2. Render the graph

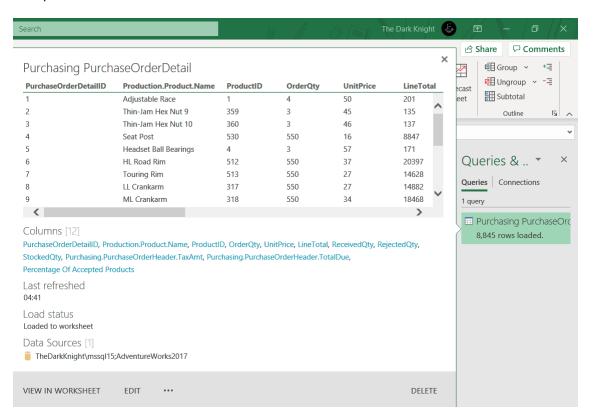


#### **Practical No: - 6**

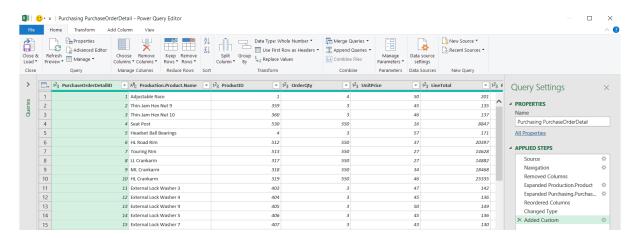
**Aim:** - Apply the what – if Analysis for data visualization. Design and generate necessary reports based on the data warehouse data.

#### Steps: -

1. Import data from dB to excel.

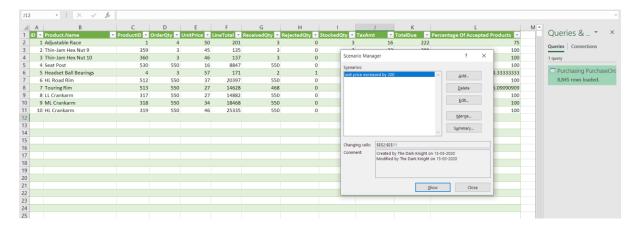


#### 2. ETL the data.



3. Select Data -> What If Analysis -> Scenario Manager -> Add New Scenario.

Scenario: all unit prices increased by 200



4. Click on Summary to check the result.

| В         | С        | D               | Е                           |
|-----------|----------|-----------------|-----------------------------|
| Scenario  | Summ     | ary             |                             |
|           |          | Current Values: | unit price increased by 200 |
| Changing  | g Cells: |                 |                             |
|           | \$E\$2   | 50              | 250                         |
|           | \$E\$3   | 45              | 245                         |
|           | \$E\$4   | 46              | 246                         |
|           | \$E\$5   | 16              | 216                         |
|           | \$E\$6   | 57              | 257                         |
|           | \$E\$7   | 37              | 237                         |
|           | \$E\$8   | 27              | 227                         |
|           | \$E\$9   | 27              | 227                         |
|           | \$E\$10  | 34              | 234                         |
|           | \$E\$11  | 46              | 246                         |
| Result Ce | lls:     |                 |                             |
|           | \$F\$2   | 200             | 1000                        |
|           | \$F\$3   | 135             | 735                         |
|           | \$F\$4   | 138             | 738                         |
|           | \$F\$5   | 8800            | 118800                      |
|           | \$F\$6   | 171             | 771                         |
|           | \$F\$7   | 20350           | 130350                      |
|           | \$F\$8   | 14850           | 124850                      |
|           | \$F\$9   | 14850           | 124850                      |
|           | \$F\$10  | 18700           | 128700                      |
|           | \$F\$11  | 25300           | 135300                      |

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.

PART A -> Perform logistic regression in the given data warehouse.

#### Code: -

```
input <- mtcars [, c("am","cyl","hp","wt")]
print(head(input))
am.data=glm (formula=am~cyl+hp+wt, data=input, family=binomial)
print(summary(am.data))</pre>
```

```
R Console
                                                                   - © X
Call:
glm(formula = am ~ cyl + hp + wt, family = binomial, data = input)
Deviance Residuals:
    Min 1Q
                    Median
                                 3Q
-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 *
           0.48760
                               0.455
                      1.07162
                                       0.6491
cyl
           0.03259 0.01886 1.728
                                       0.0840 .
hp
           -9.14947 4.15332 -2.203 0.0276 *
wt
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
>
                                                                 - - X

    C:\Users\ampee\Documents\practical9.R - R Editor

input <- mtcars[,c("am","cyl","hp","wt")]
print(head(input))
am.data=glm(formula=am~cyl+hp+wt,data=input,family=binomial)
print(summary(am.data))
```

#### PART B -> Perform the data classification using classification algorithm

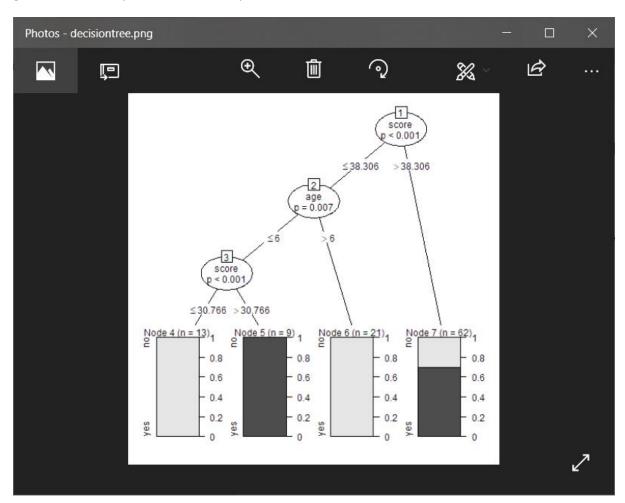
#### Code: -

dev.off()

```
library(party)
print(head(readingSkills))
input<-readingSkills [c (1:105),]
print(head(input))
png(file="decisiontree.png")
output. tree<-ctree (nativeSpeaker~age+shoeSize+score, data=input)
plot (output. tree)
dev off ()
```

```
dev.off ()
                                                                       - - X
 😱 R Console
 6: package 'sandwich' was built under R version 3.5.3
 > print(head(readingSkills))
  nativeSpeaker age shoeSize
                                score
           yes
                  5 24.83189 32.29385
 2
            yes
                  6 25.95238 36.63105
 3
             no 11 30.42170 49.60593
                  7 28.66450 40.28456
            ves
            yes 11 31.88207 55.46085
            yes 10 30.07843 52.83124
 > input<-readingSkills[c(1:105),]
 > print(head(input))
  nativeSpeaker age shoeSize
                               score
            yes 5 24.83189 32.29385
                 6 25.95238 36.63105
 2
            yes
             no 11 30.42170 49.60593
                  7 28.66450 40.28456
            yes
            yes 11 31.88207 55.46085
             yes 10 30.07843 52.83124
 > png(file="decisiontree.png")
 > output.tree<-ctree(nativeSpeaker~age+shoeSize+score,data=input)
 > plot(output.tree)
 > dev.off()
 null device
 >
                                                                      C:\Users\ampee\Documents\practical9.R - R Editor
 print(head(readingSkills))
 input<-readingSkills[c(1:105),]
 print(head(input))
 png(file="decisiontree.png")
 output.tree<-ctree(nativeSpeaker~age+shoeSize+score,data=input)
 plot(output.tree)
```

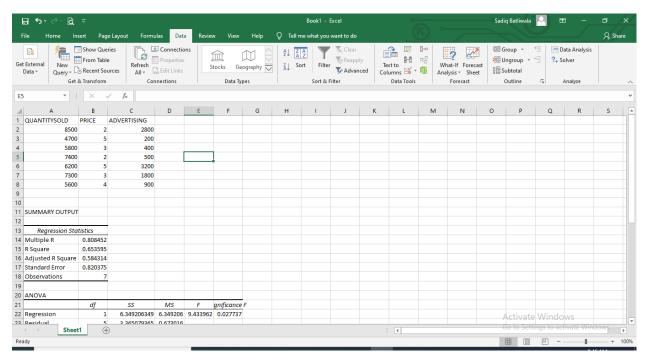
From the decision tree, we can conclude the people with reading skills less than 38.3 and also age is greater than 6 they are not a motive speaker.



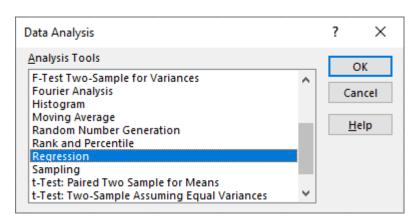
#### PART C -> Linear Regression with Excel

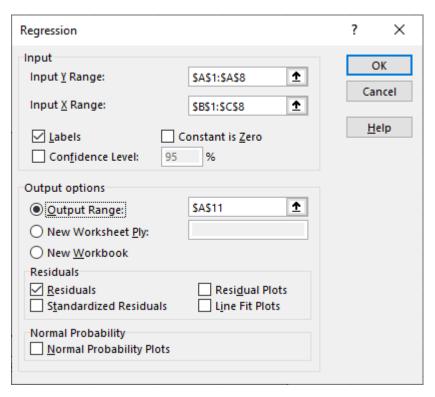
#### Steps: -

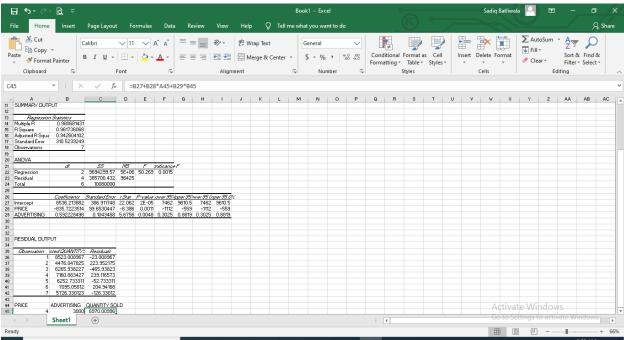
1. Import data into excel



#### 2. Apply Data Analysis -> Regression







3. Render the Output & Check the result -> APPLY Formula=B17+B18\*A36+B19\*B36

#### **SUMMARY OUTPUT**

| Regression Statistics |          |  |  |  |
|-----------------------|----------|--|--|--|
| Multiple R            | 0.980681 |  |  |  |
| R Square              | 0.961736 |  |  |  |
| Adjusted R            |          |  |  |  |
| Square                | 0.942604 |  |  |  |
| Standard              |          |  |  |  |
| Error                 | 310.5239 |  |  |  |
| Observations          | 7        |  |  |  |

#### ANOVA

|            |    |   |          |          |          | Significance |
|------------|----|---|----------|----------|----------|--------------|
|            | df |   | SS       | MS       | F        | F            |
| Regression |    | 2 | 9694300  | 4847150  | 50.26854 | 0.001464     |
| Residual   |    | 4 | 385700.4 | 96425.11 |          |              |
| Total      |    | 6 | 10080000 |          |          |              |

|             |              | Standard |          |          |           | Upper    | Lower    | Upper    |
|-------------|--------------|----------|----------|----------|-----------|----------|----------|----------|
|             | Coefficients | Error    | t Stat   | P-value  | Lower 95% | 95%      | 95.0%    | 95.0%    |
| Intercept   | 8536.214     | 386.9117 | 22.06243 | 2.5E-05  | 7461.975  | 9610.453 | 7461.975 | 9610.453 |
| Price       | -835.722     | 99.65304 | -8.38632 | 0.001106 | -1112.4   | -559.041 | -1112.4  | -559.041 |
| Advertising | 0.592228     | 0.104347 | 5.675579 | 0.004755 | 0.302515  | 0.881942 | 0.302515 | 0.881942 |

#### RESIDUAL OUTPUT 🎡

| Predicted   |     |             |           |  |  |
|-------------|-----|-------------|-----------|--|--|
| Observation | Qua | ıntity Sold | Residuals |  |  |
| 1           |     | 8523.009    | -23.009   |  |  |
| 2           |     | 4476.048    | 223.9522  |  |  |
| 3           |     | 6265.938    | -465.938  |  |  |
| 4           |     | 7160.883    | 239.1166  |  |  |
| !           | 5   | 6252.733    | -52.7333  |  |  |
|             | 6   | 7095.058    | 204.9419  |  |  |
|             | 7   | 5726.33     | -126.33   |  |  |

|       |               | quantity |
|-------|---------------|----------|
| Price | advertisement | sold     |
| 4     | 4 3000        | 6970.01  |

#### Practical No: - 9

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

#### Code: -

```
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^xx)

print(relation)

print(summary(relation))

a <-data. frame(x=170)

result <- predict (relation, a)

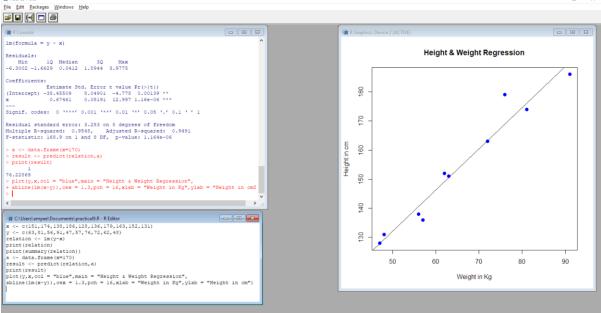
print(result)

plot (y, x, col = "blue", main = "Height & Weight Regression",

abline(lm(x^yy)), cex = 1.3, pch = 16, xlab = "Weight in Kg", ylab = "Height in cm")

Red (64-b0)

Red (64-b0)
```



Practical No: - 10

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

#### Code: -

bodysize=rnorm (20,30,2)
bodysize=sort(bodysize)
survive=c (0,0,0,0,1,0,1,0,0,1,1,0,1,1,1,0)
dat=as.data. frame (cbind (bodysize, survive))
dat
plot (bodysize, survive, xlab="Body size", ylab="Probability of survival")
g=glm (survive~bodysize, family=binomial, dat)
curve (predict (g, data. frame(bodysize=x), type="resp"), add=TRUE)
points (bodysize, fitted(g), pch=20)

