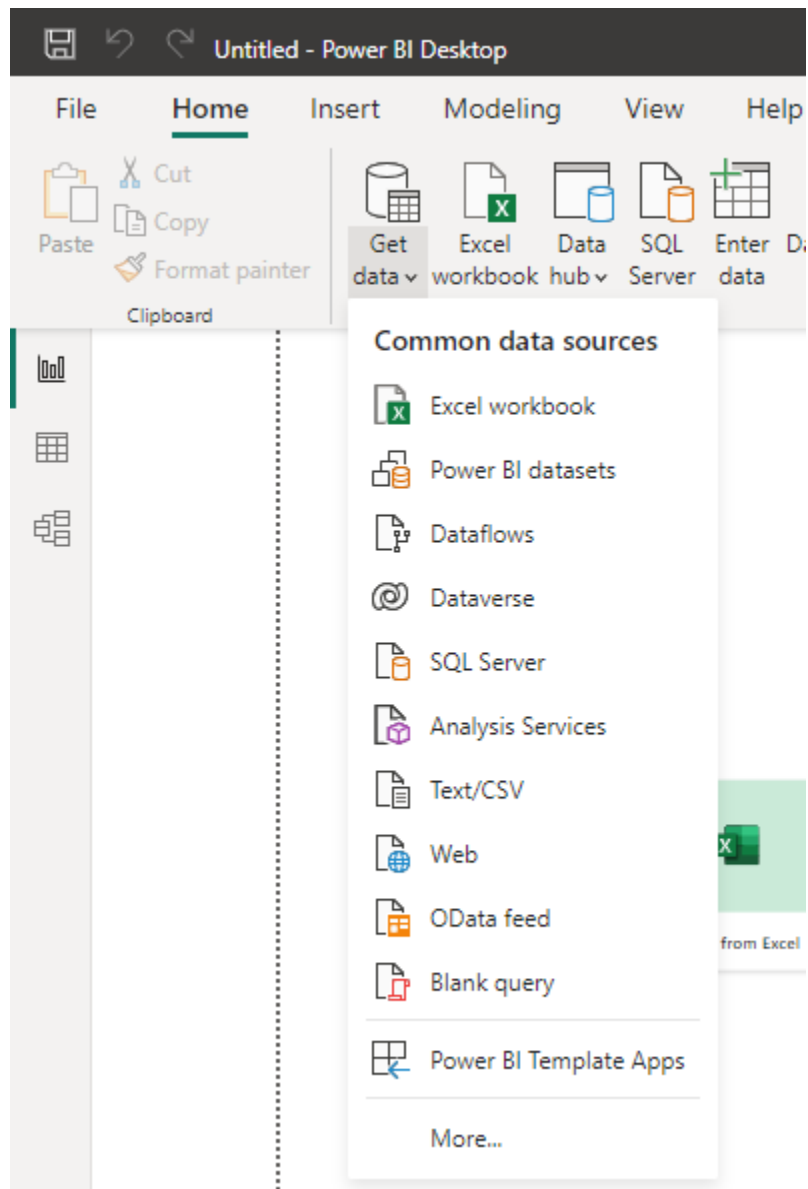


Practical 1: Import the legacy data from different sources such as (Excel, SqlServer, Oracle etc.) and load in the target system.

Importing Excel Data

1) Launch Power BI Desktop.

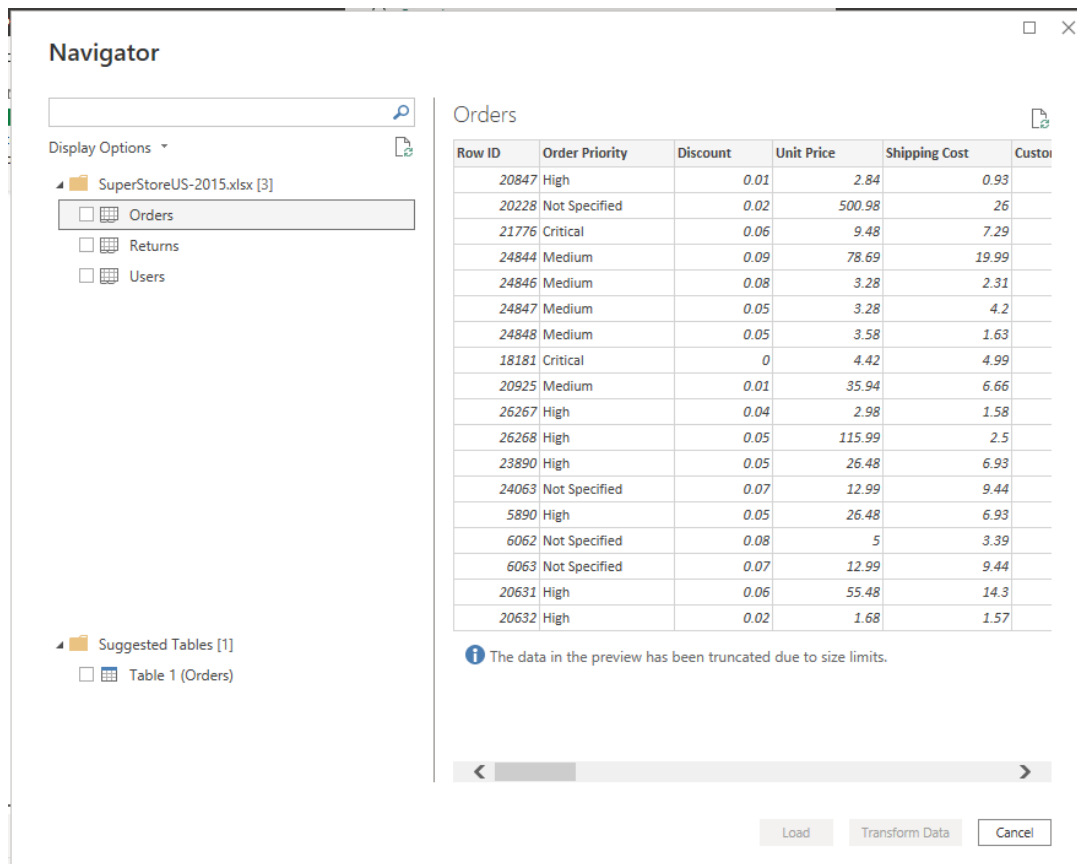
2) From the Home ribbon, select Get Data. Excel is one of the Most Common data connections, so you can select it directly from the Get Data menu



3) If you select the Get Data button directly, you can also select File > Excel and select Connect.

4) In the Open File dialog box, select the Products.xlsx file.

5) In the Navigator pane, select the Products table and then select Edit.



Importing Data from OData Feed

In this task, you'll bring in order data. This step represents connecting to a sales system. You import data into Power BI Desktop from the sample Northwind OData feed at the following URL, which you can copy (and then paste) in the steps below: <http://services.odata.org/V3/Northwind/Northwind.svc/>

Connect to an OData feed:

- 1) From the Home ribbon tab in Query Editor, select Get Data.
- 2) Browse to the OData Feed data source.
- 3) In the OData Feed dialog box, paste the URL for the Northwind OData feed.
- 4) Select OK.
- 5) In the Navigator pane, select the Orders table, and then select Edit

Navigator

Display Options ▾

SuperStoreUS-2015.xlsx [3]

☒ Orders

☒ Returns


☒ Users

Suggested Tables [1]

☒ Table 1 (Orders)

Orders

Row ID	Order Priority	Discount	Unit Price	Shipping Cost	Customer
20847	High	0.01	2.84	0.93	
20228	Not Specified	0.02	500.98	26	
21776	Critical	0.06	9.48	7.29	
24844	Medium	0.09	78.69	19.99	
24846	Medium	0.08	3.28	2.31	
24847	Medium	0.05	3.28	4.2	
24848	Medium	0.05	3.58	1.63	
18181	Critical	0	4.42	4.99	
20925	Medium	0.01	35.94	6.66	
26267	High	0.04	2.98	1.58	
26268	High	0.05	115.99	2.5	
23890	High	0.05	26.48	6.93	
24063	Not Specified	0.07	12.99	9.44	
5890	High	0.05	26.48	6.93	
6062	Not Specified	0.08	5	3.39	
6063	Not Specified	0.07	12.99	9.44	
20631	High	0.06	55.48	14.3	
20632	High	0.02	1.68	1.57	

 The data in the preview has been truncated due to size limits.



Load

Transform Data

Cancel

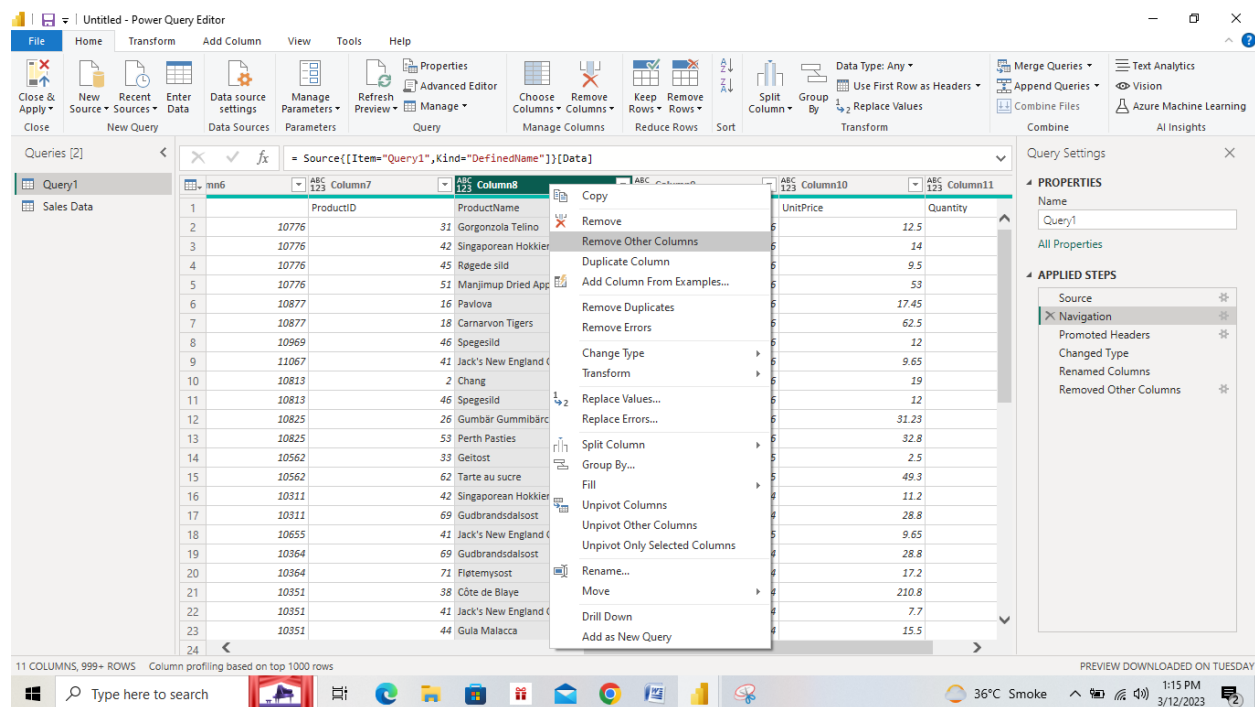
Practical 2: Perform the Extraction Transformation and Loading (ETL) Power BI.

ETL Process in Power BI

Remove other columns to only display columns of interest In this step you remove all columns except ProductID, ProductName, UnitsInStock, and QuantityPerUnit

Power BI Desktop includes Query Editor, which is where you shape and transform your data connections. Query Editor opens automatically when you select Edit from Navigator. You can also open the Query Editor by selecting Edit Queries from the Home ribbon in Power BI Desktop. The following steps are performed in Query Editor.

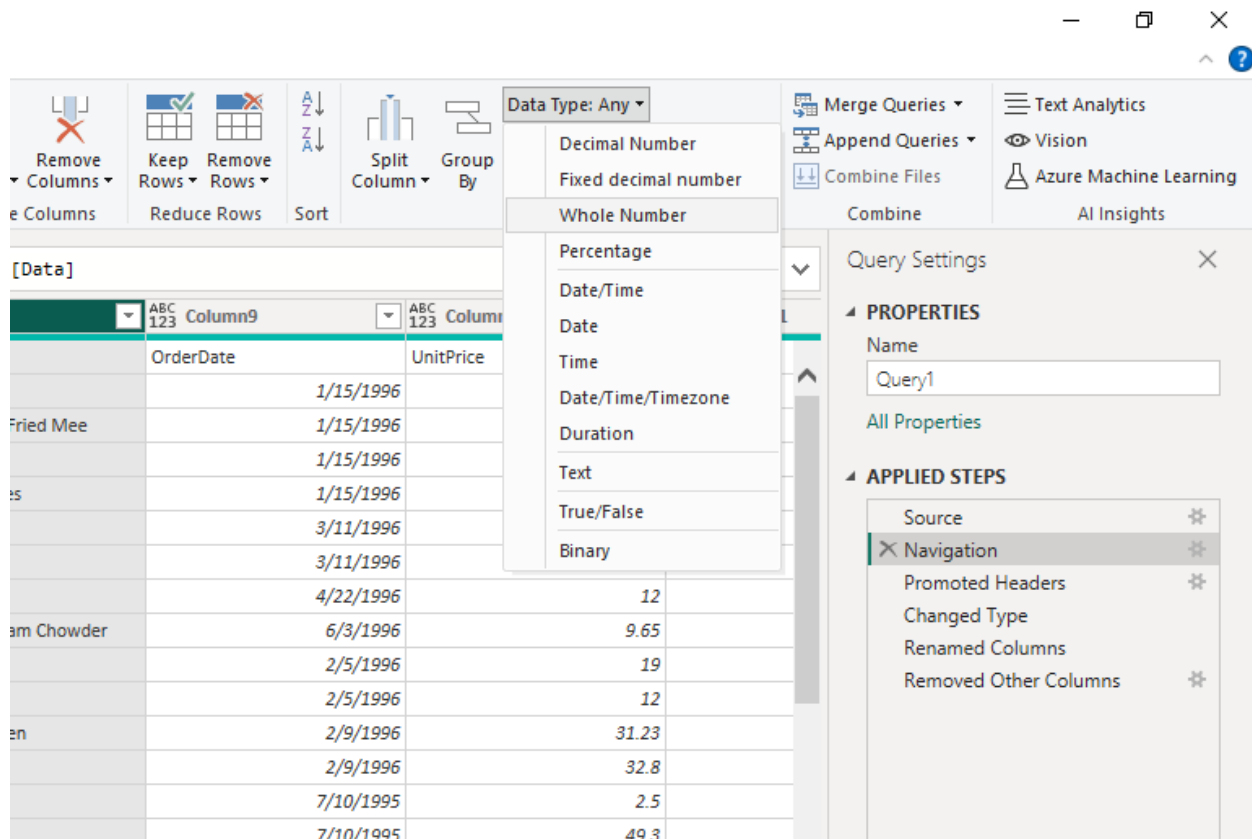
1. In Query Editor, select the ProductID, ProductName, QuantityPerUnit, and UnitsInStock columns (use Ctrl+Click to select more than one column, or Shift+Click to select columns that are beside each other).
2. Select Remove Columns > Remove Other Columns from the ribbon, or right-click on a column header and click Remove Other Columns.



3. Change the data type of the UnitsInStock column When Query Editor connects to data, it reviews each field and to determine the best data type.

For the Excel workbook, products in stock will always be a whole number, so in this step you confirm the UnitsInStock column's datatype is Whole Number.

1. Select the UnitsInStock column.
2. Select the Data Type drop-down button in the Home ribbon.
3. If not already a Whole Number, select Whole Number for data type from the drop down (the Data Type: button also displays the data type for the current selection).



3. Expand the Order_Details table

The Orders table contains a reference to a Details table, which contains the individual products that were included in each Order. When you connect to data sources with multiples tables (such as a relational database) you can use these references to build up your query

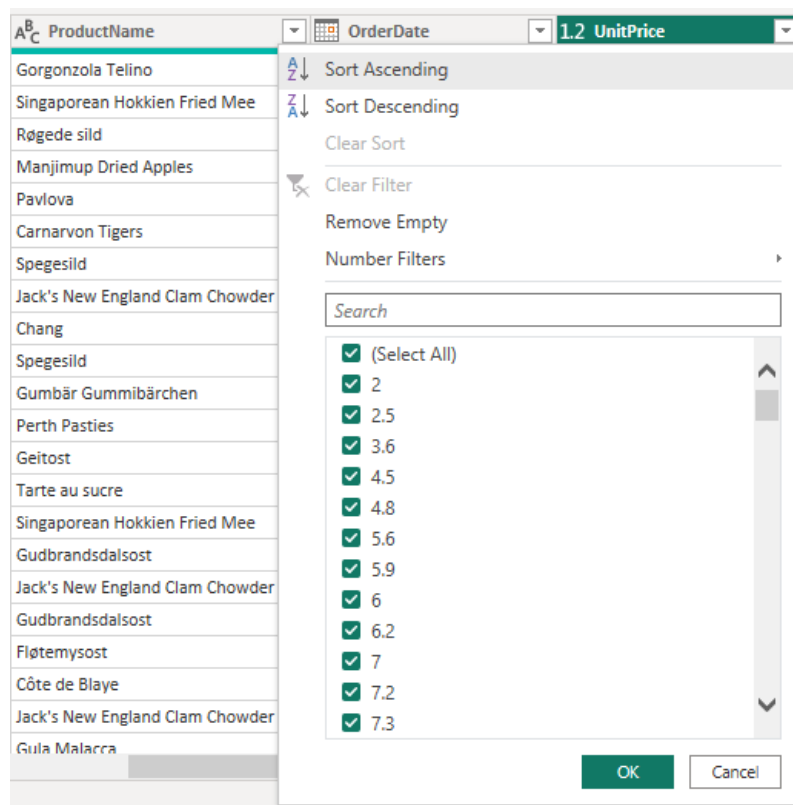
In this step, you expand the Order_Details table that is related to the Orders table, to combine the ProductID, UnitPrice, and Quantity columns from Order_Details into the Orders table.

This is a representation of the data in these tables:

The Expand operation combines columns from a related table into a subject table. When the query runs, rows from the related table (Order_Details) are combined into rows from the subject table (Orders).

After you expand the Order_Details table, three new columns and additional rows are added to the Orders table, one for each row in the nested or related table.

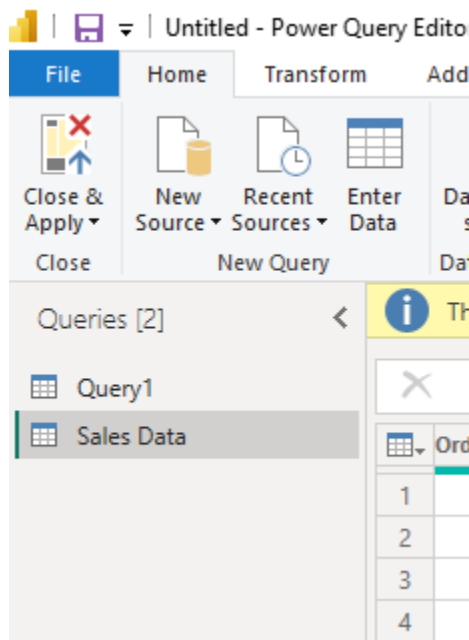
1. In the Query View, scroll to the Order_Details column.
2. In the Order_Details column, select the expand icon ().
3. In the Expand drop-down:
 - a. Select (Select All Columns) to clear all columns.
 - b. Select ProductID, UnitPrice, and Quantity.
 - c. Click OK.



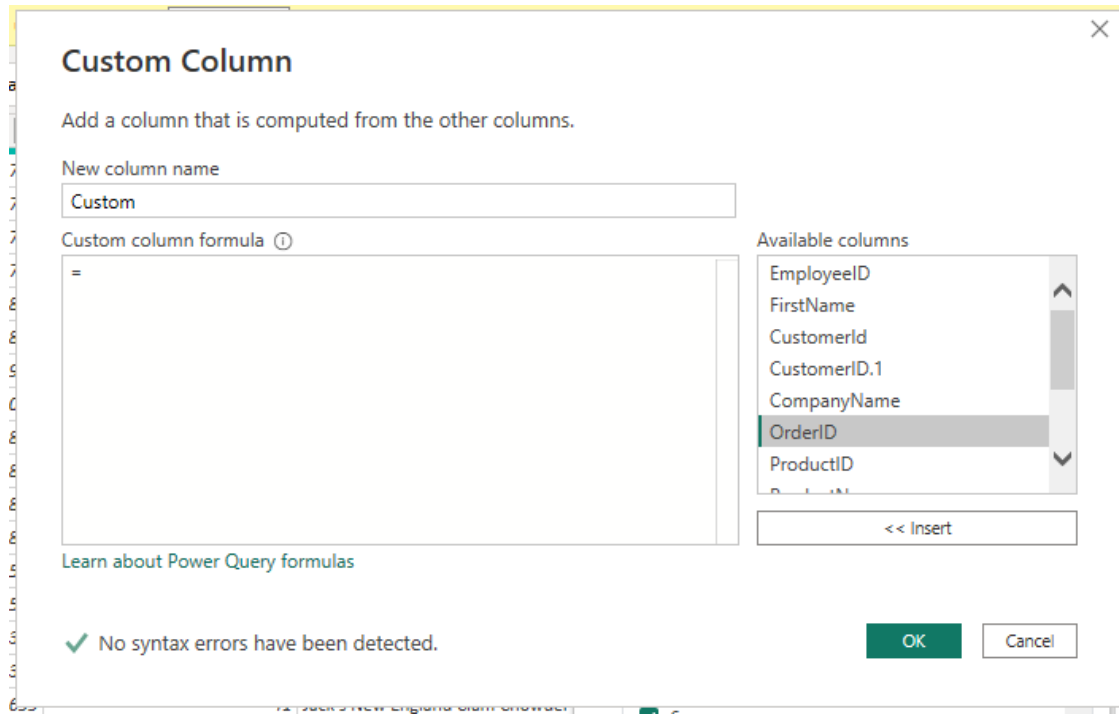
4. Calculate the line total for each Order_Details row Power BI Desktop lets you to create calculations based on the columns you are importing, so you can enrich the data that you connect to.

In this step, you create a Custom Column to calculate the line total for each Order_Details row. Calculate the line total for each Order_Details row:

1. In the Add Column ribbon tab, click Add Custom Column.

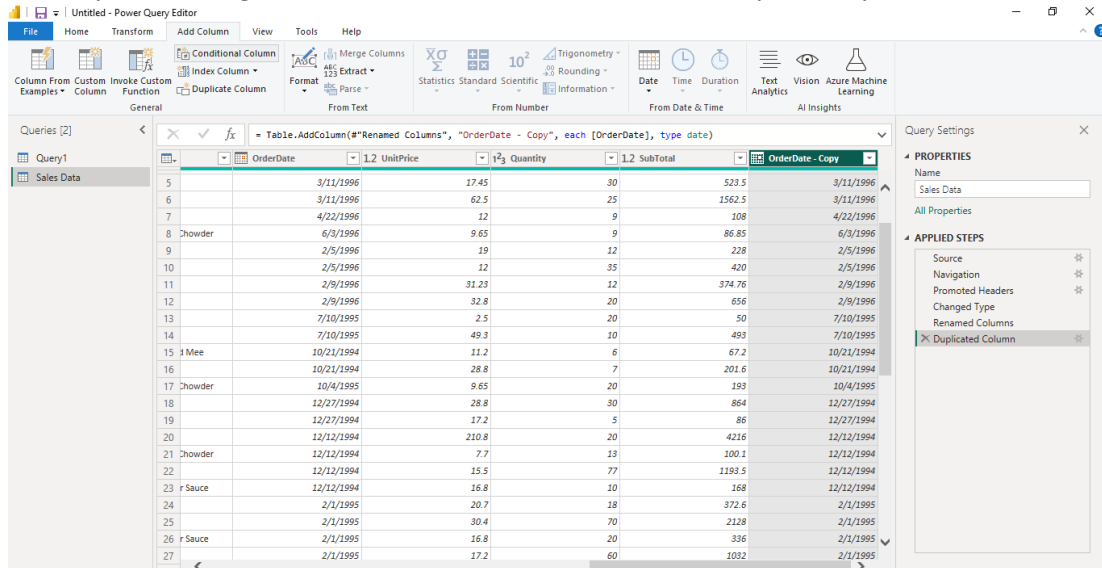


2. In the Add Custom Column dialog box, in the Custom Column Formula textbox, enter [Order_Details.UnitPrice] * [Order_Details.Quantity].
3. In the New column name textbox, enter LineTotal.
4. Click OK.



5. Rename and reorder columns in the query In this step you finish making the model easy to work with when creating reports, by renaming the final columns and changing their order.

1. In Query Editor, drag the Line Total column to the left, after Ship Country.



2. Remove the Order_Details. prefix from the Order_Details.ProductID, Order_Details.UnitPrice and Order_Details.Quantity columns, by double-clicking on each column header, and then deleting that text from the column name.

6. Combine the Products and Total Sales queries

Power BI Desktop does not require you to combine queries to report on them. Instead, you can create Relationships between datasets. These relationships can be created on any column that is common to your datasets we have Orders and Products data that share a common

'ProductID' field, so we need to ensure there's a relationship between them in the model we're using with Power BI Desktop. Simply specify in Power BI Desktop that the columns from each table are related (i.e. columns that have the same values).

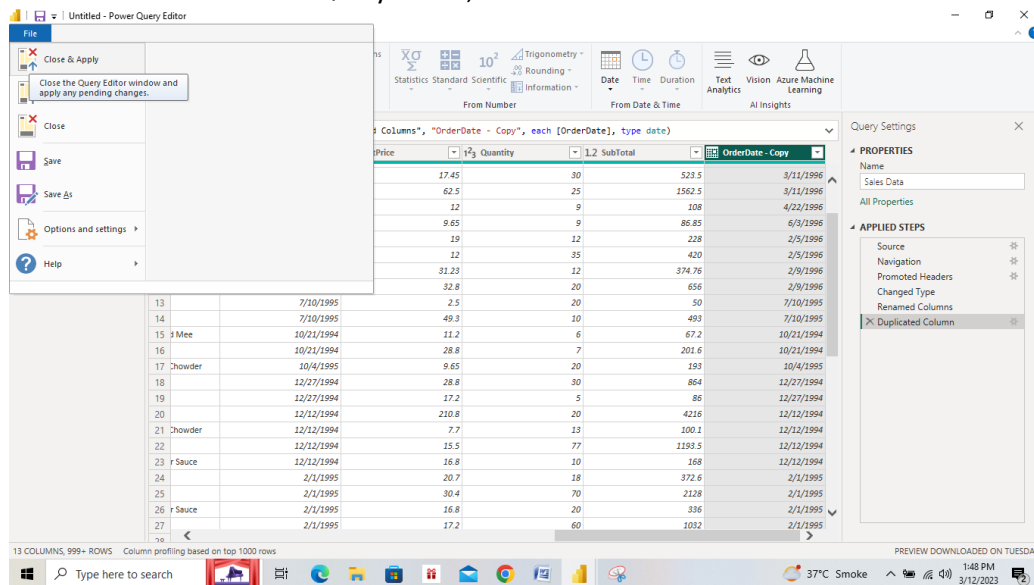
Power BI Desktop works out the direction and cardinality of the relationship for you.

In some cases, it will even detect the relationships automatically.

In this task, you confirm that a relationship is established in Power BI Desktop between the Products and Total Sales queries

Step 1: Confirm the relationship between Products and Total Sales

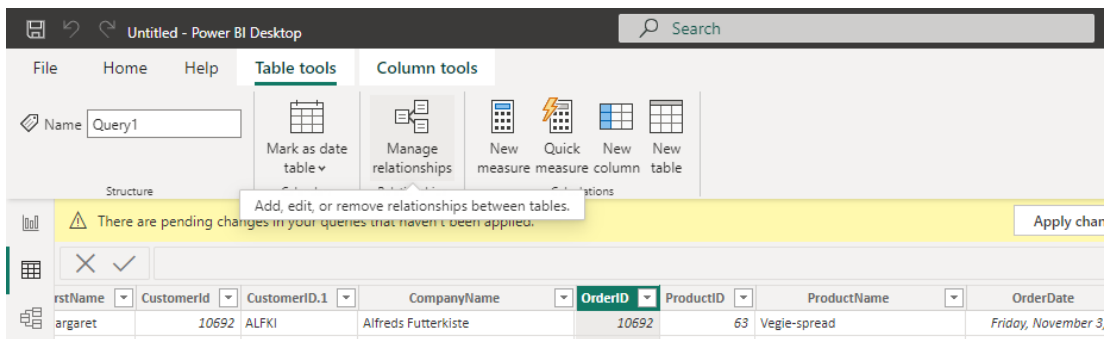
1. First, we need to load the model that we created in Query Editor into Power BI Desktop. From the Home ribbon of Query Editor, select Close & Load.



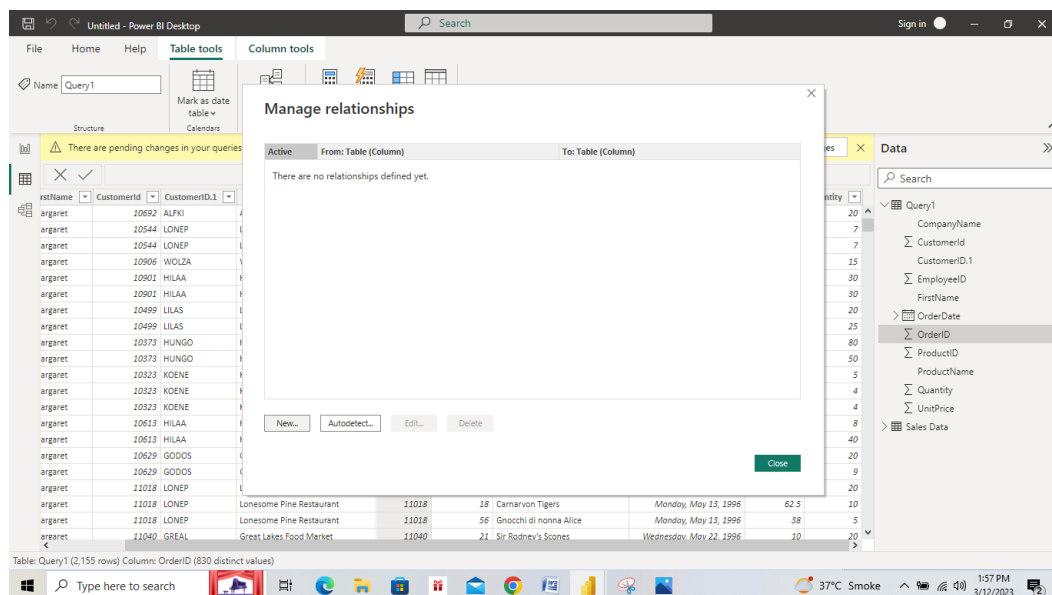
2. Power BI Desktop loads the data from the two queries.



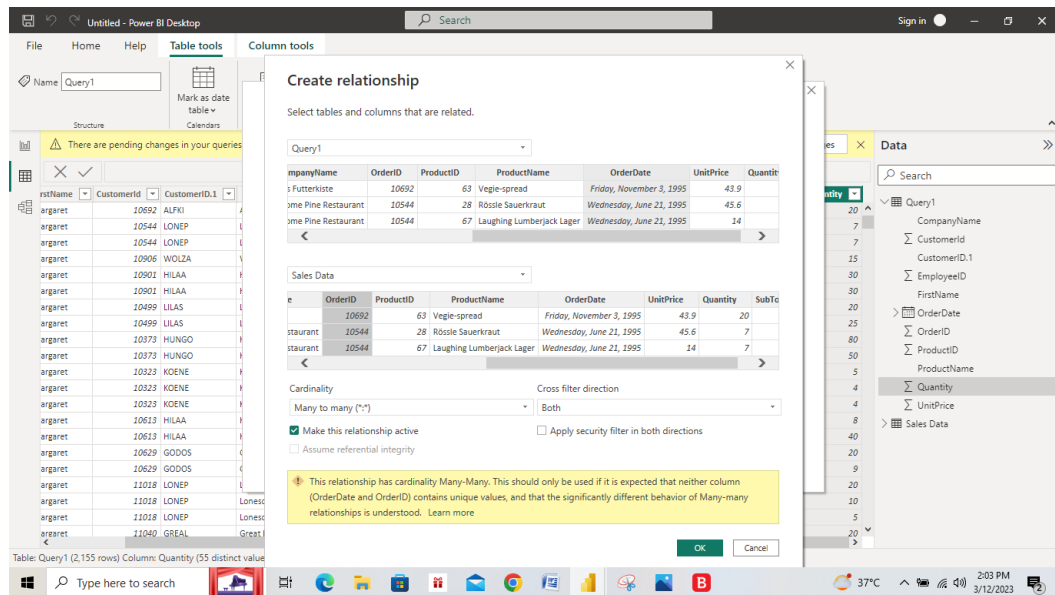
3. Once the data is loaded, select the Manage Relationships button Home ribbon.



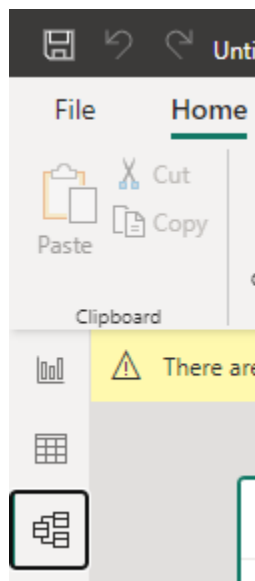
4. Select new button



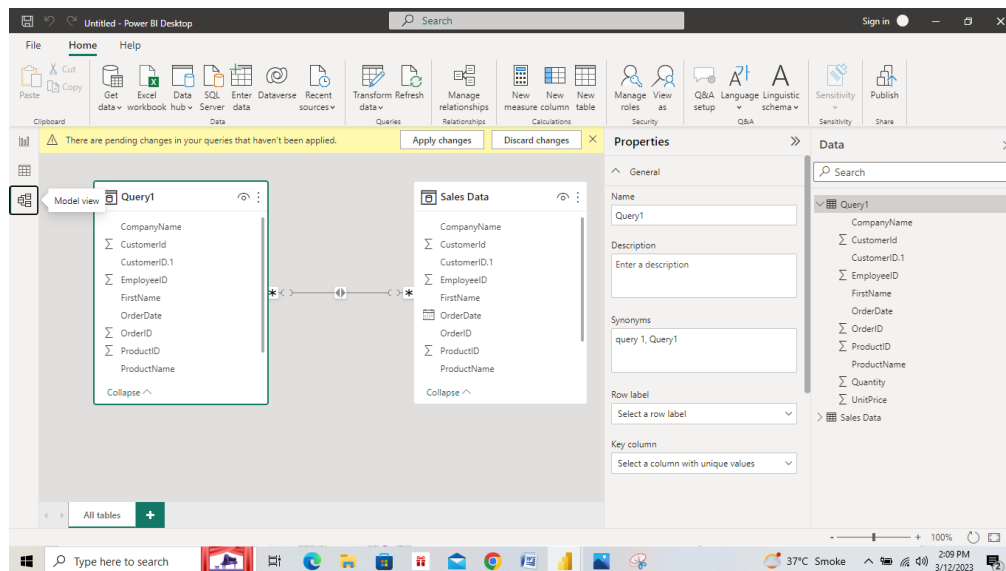
5. When we attempt to create the relationship, we see that one already exists! As shown in the Create Relationship dialog (by the shaded columns), the ProductsID fields in each query already have an established relationship.



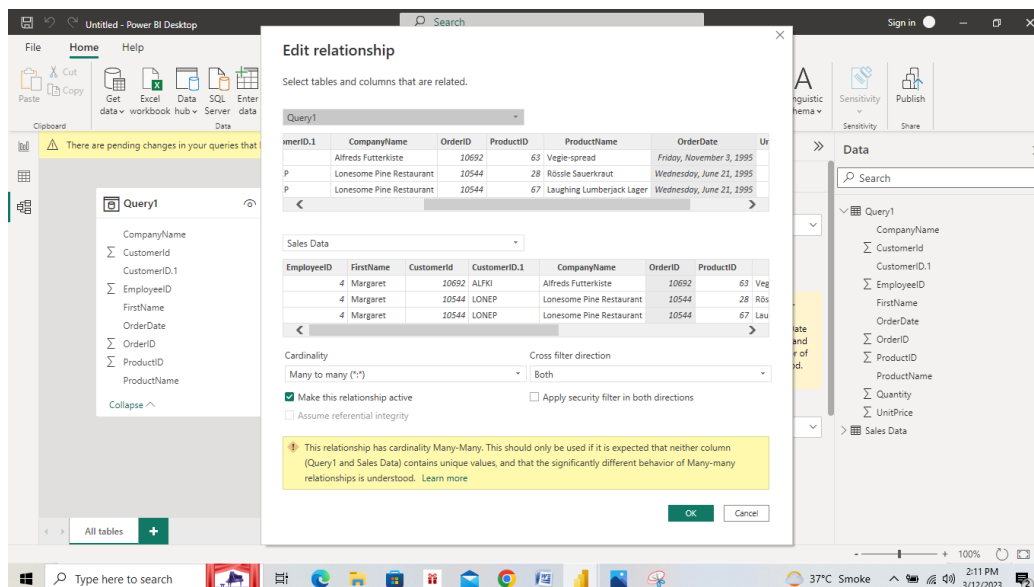
5. Select cancel , and then select relationship view in power BI desktop



6. We see the following, which visualizes the relationship between the queries.



7. When you double-click the arrow on the line that connects the two queries, an Edit Relationship dialog appears



8. No need to make any changes, so we'll just select Cancel to close the Edit Relationship dialog.

Practical 11: Data Modelling and Analytics with Pivot Table in Excel

Data Model is used for building a model where data from various sources can be combined by creating relationships among the data sources. A Data Model integrates the tables, enabling extensive analysis using PivotTables, Power Pivot, and Power View.

A Data Model is created automatically when you import two or more tables simultaneously from a database.

The existing database relationships between those tables is used to create the Data Model in Excel.

Step 1 – Open a new blank Workbook in Excel.

Step 2 – Click on the DATA tab.

Step 3 – In the Get External Data group, click on the option From Access. The Select Data Source dialog box opens.

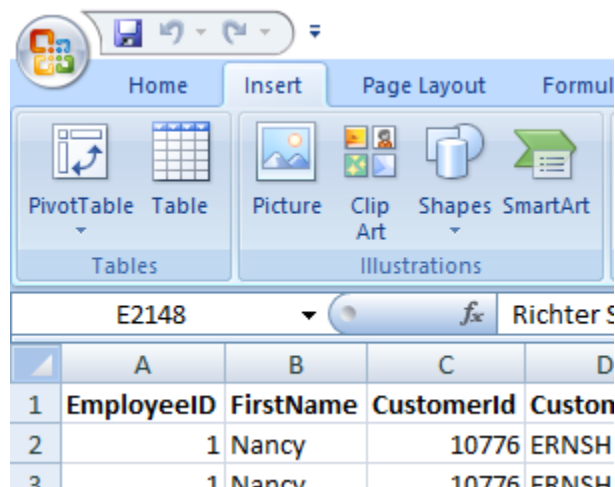
Step 4 – Select Events.accdb, Events Access Database file.

Step 5 – The Select Table window, displaying all the tables found in the database, appears.

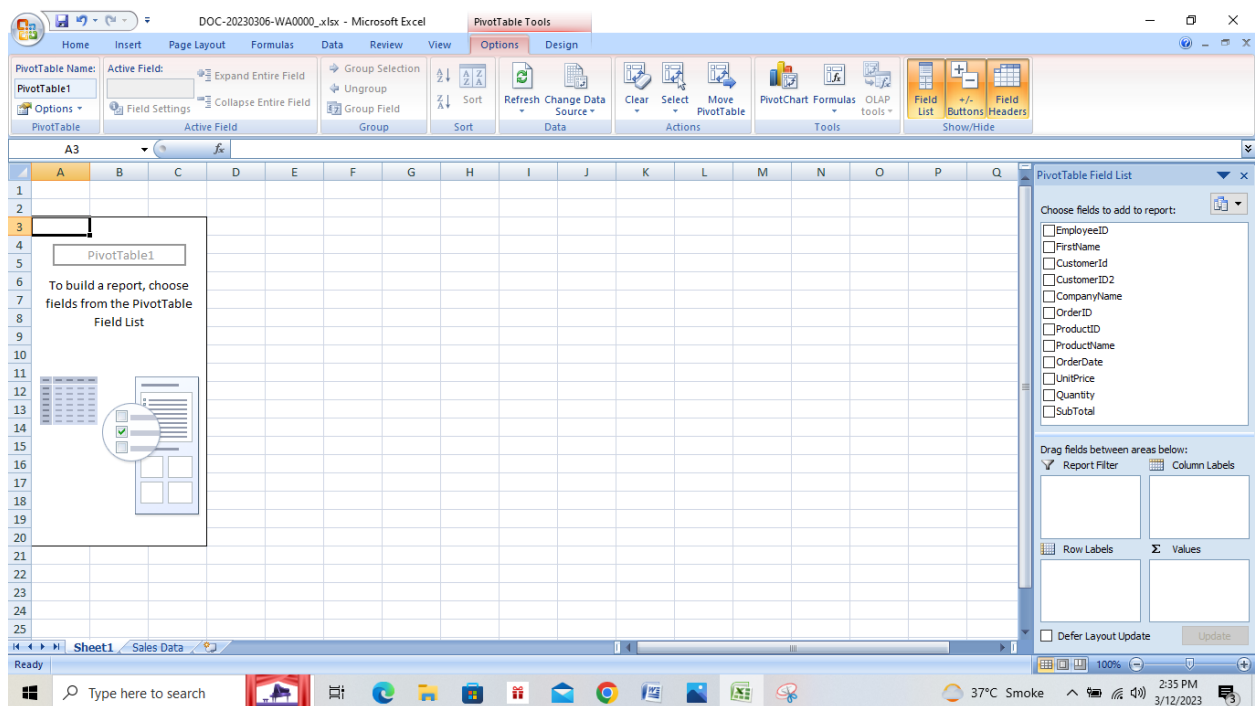
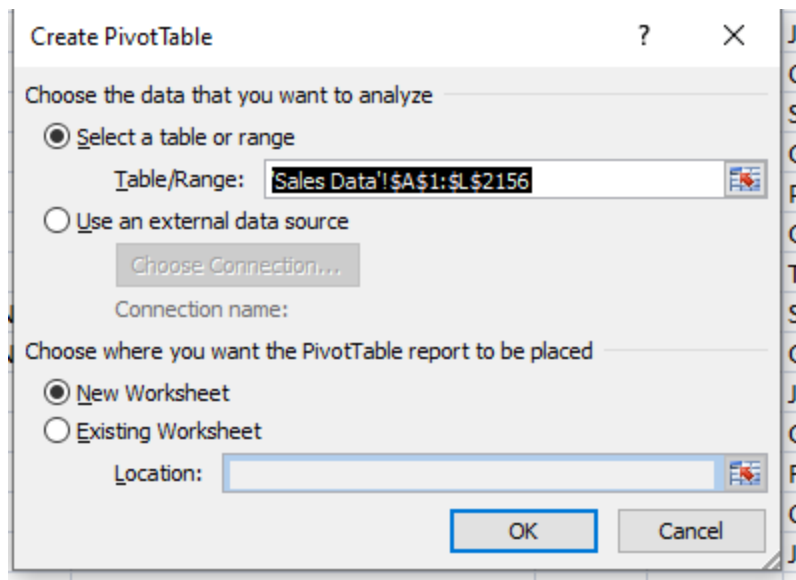
Step 6 – Tables in a database are similar to the tables in Excel. Check the 'Enable selection of multiple tables' box, and select all the tables. Then click OK.

Step 7 – The Import Data window appears. Select the PivotTable Report option. This option imports the tables into Excel and prepares a PivotTable for analyzing the imported tables. Notice that the checkbox at the bottom of the window - 'Add this data to the Data Model' is selected and disabled.

Step 8 – The data is imported, and a PivotTable is created using the imported tables.



	A	B	C	D
1	EmployeeID	FirstName	CustomerId	Custom
2	1	Nancy	10776	ERNSH
3	1	Nancy	10776	ERNSH



Explore Data Using PivotTable

Step 1 – You know how to add fields to PivotTable and drag fields across areas. Even if you are not sure of the final report that you want, you can play with the data and choose the bestsuited report.

In PivotTable Fields, click on the arrow beside the table - Medals to expand it to show the fields in that table. Drag the NOC_CountryRegion field in the Medals table to the COLUMNS area.

Step 2 – Drag Discipline from the Disciplines table to the ROWS area.

Step 3 – Filter Discipline to display only five sports: Archery, Diving, Fencing, Figure Skating, and Speed Skating. This can be done either in PivotTable Fields area, or from the Row Labels filter in the PivotTable itself.

Step 4 – In PivotTable Fields, from the Medals table, drag Medal to the VALUES area.

Step 5 – From the Medals table, select Medal again and drag it into the FILTERS area.

The screenshot shows the Microsoft Excel interface with a PivotTable and the PivotTable Field List task pane. The PivotTable is located in the range A3:D25. The PivotTable Fields task pane is open on the right side of the screen.

PivotTable Data:

Row Labels	Sum of EmployeeID	Sum of ProductID	Sum of UnitPrice
8/4/1994	15	125	58.6
8/5/1994	12	65	61
8/8/1994	21	301	116.1
8/9/1994	12	113	94
8/10/1994	9	119	40.4
8/11/1994	15	153	30.8
8/12/1994	36	113	88.3
8/15/1994	6	130	36.6
8/16/1994	12	143	59.9
8/17/1994	3	39	57.8
8/18/1994	8	58	28.8
8/19/1994	24	286	97.1
8/22/1994	24	68	71.4
8/23/1994	36	144	46.2
8/24/1994	12	43	22.9
8/25/1994	4	87	43.2
8/26/1994	3	12	30.4
8/29/1994	12	175	73.1
8/30/1994	16	101	126.8
8/31/1994	10	105	29.8
9/1/1994	8	112	54

PivotTable Field List:

- Choose fields to add to report:
 - ☒ EmployeeID
 - ☐ FirstName
 - ☐ CustomerID
 - ☐ CustomerID2
 - ☐ CompanyName
 - ☐ OrderID
 - ☒ ProductID
 - ☐ ProductName
 - ☒ OrderDate
 - ☒ UnitPrice
 - ☐ Quantity
 - ☐ SubTotal
- Drag fields between areas below:
 - Report Filter: (Empty)
 - Column Labels: (Empty)
 - Row Labels: OrderDate
 - Values:
 - Sum of EmployeeID
 - Sum of ProductID
 - Sum of UnitPrice

Step 6 – click dropdown list to the right of the column labels

Step 7 – select value filter and then select greater than

Step 8 – click ok

DOC-20230306-WA0000.xlsx - Microsoft Excel

PivotTable Tools: Options, Design

PivotTable Name: PivotTable1

Active Field: Sum of UnitPrice

Field Settings: Active Field

Value Filters

Greater Than

Row Labels	Sum of EmployeeID	Sum of ProductID	Sum of UnitPrice
Sort Oldest to Newest	15	125	58.6
Sort Newest to Oldest	12	65	61
More Sort Options...	21	301	116.1
Clear Filter From "OrderDate"	12	113	94
Date Filters	9	119	40.4
	15	152	30.8
			88.3
			36.6
			59.9
			57.8
			28.8
			97.1
			71.4
			46.2
			22.9
			43.2
			30.4
			73.1
8/30/1994	16	101	126.8
8/31/1994	10	105	29.8
9/1/1994	8	112	54

PivotTable Field List

Choose fields to add to report:

- ☒ EmployeeID
- ☐ FirstName
- ☐ CustomerID
- ☐ CustomerID2
- ☐ CompanyName
- ☐ OrderID
- ☒ ProductID
- ☐ ProductName
- ☒ OrderDate
- ☒ UnitPrice
- ☐ Quantity
- ☐ SubTotal

Drag fields between areas below:

Report Filter

Column Labels

Values

Row Labels

OrderDate

Sum of Emplo...

Sum of Produ...

Sum of UnitPrice

Defer Layout Update

Update

The Value Filters dialog box for the count of Medals is greater than appears.

Step 9 – Type 80 in the Right Field.

Step 10 – Click OK.

DOC-20230306-WA0000.xlsx - Microsoft Excel

PivotTable Tools: Options, Design

PivotTable Name: PivotTable1

Active Field: Sum of UnitPrice

Field Settings: Active Field

Value Filter (OrderDate)

Show items for which

Sum of EmployeeID

is greater than

80

OK

Cancel

Row Labels	Sum of EmployeeID	Sum of ProductID	Sum of UnitPrice
9/7/1995	52	214	253.79
12/28/1995	56	410	281.55
2/13/1996	70	314	533
2/16/1996	71	414	177.65
2/21/1996	52	295	161.5
3/6/1996	51	299	200.73
3/20/1996	60	447	337.83
4/5/1996	68	570	287.25
4/10/1996			
4/17/1996			
4/18/1996			
4/25/1996			
5/3/1996			
5/8/1996			
5/13/1996	51	313	250.5
5/17/1996	78	523	708.69
5/28/1996	68	379	195.5
6/5/1996	68	997	698.25
Grand Total	1173	7659	5536.17

PivotTable Field List

Choose fields to add to report:

- ☒ EmployeeID
- ☐ FirstName
- ☐ CustomerID
- ☐ CustomerID2
- ☐ CompanyName
- ☐ OrderID
- ☒ ProductID
- ☐ ProductName
- ☒ OrderDate
- ☒ UnitPrice
- ☐ Quantity
- ☐ SubTotal

Drag fields between areas below:

Report Filter

Column Labels

Values

Row Labels

OrderDate

Sum of Emplo...

Sum of Produ...

Sum of UnitPrice

Defer Layout Update

Update

DOC-20230306-WA0000.xlsx - Microsoft Excel

PivotTable Tools

Home Insert Page Layout Formulas Data Review View Options Design

Clipboard Font Alignment Number Styles Cells Editing

Calibri 11

General

AutoSum Fill Clear Sort & Find & Filter

D12 287.25

Row Labels	Sum of EmployeeID	Sum of ProductID	Sum of UnitPrice
9/7/1995	52	214	253.79
12/28/1995	56	410	281.55
2/13/1996	70	314	533
2/16/1996	71	414	177.65
2/21/1996	52	295	161.5
3/6/1996	299	200.73	
3/20/1996	447	337.83	
4/5/1996	570	287.25	
4/10/1996	383	204.1	
4/17/1996	51	288	171.04
4/18/1996	71	471	195.95
4/25/1996	116	477	385.05
5/3/1996	74	533	177.4
5/8/1996	51	332	316.39
5/13/1996	51	313	250.5
5/17/1996	78	523	708.69
5/28/1996	68	379	195.5
6/5/1996	68	997	698.25
Grand Total	1173	7659	5536.17

Sum of EmployeeID
Value: 52
Row: 2/21/1996
Column: Sum of EmployeeID

PivotTable Field List

Choose fields to add to report:

- ☒ EmployeeID
- ☐ FirstName
- ☐ CustomerId
- ☐ CustomerID2
- ☐ CompanyName
- ☐ OrderID
- ☒ ProductID
- ☐ ProductName
- ☒ OrderDate
- ☒ UnitPrice
- ☐ Quantity
- ☐ SubTotal

Drag fields between areas below:

Report Filter

Column Labels

Row Labels

OrderDate

Sum of EmployeeID

Sum of ProductID

Sum of UnitPrice

Defer Layout Update

Update

Sheet1 Sales Data

Ready

Type here to search

35°C Smoke

2:48 PM 3/12/2023

Practical 5: Apply the what – if Analysis for data visualization. Design and generate necessary reports based on the data warehouse data.

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.

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

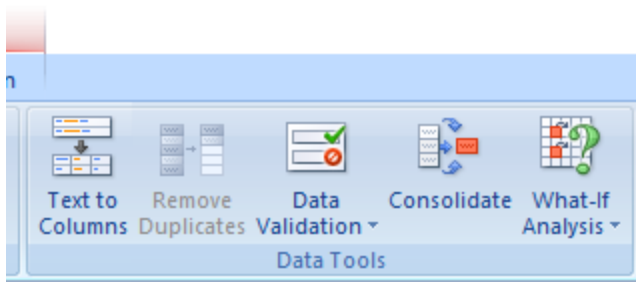
Create Different Scenarios But what if you sell 70% for the highest price? And what if you sell 80% for the highest price? Or 90%, or even 100%? Each different percentage is a different scenario. You can use the Scenario Manager to create these scenarios.

Note: You can simply type in a different percentage into cell C4 to see the corresponding result of a scenario in cell D10.

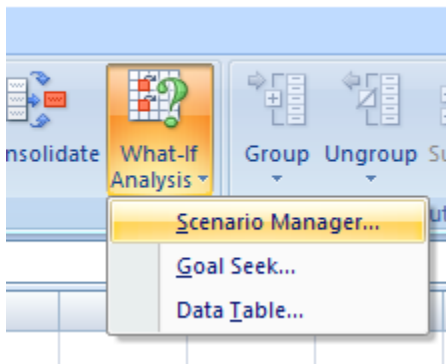
C8 $=B4*(1-C4)$				
	A	B	C	D
1	Book Store			
2				
3		total number of books	% sold for the highest price	
4		100	60%	
5				
6			number of books	unit profit
7		highest price	60	\$50
8		lower price	40	\$20
9				
10			total profit	\$3,800
11				

However, what-if analysis enables you to easily compare the results of different scenarios. Read on.

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

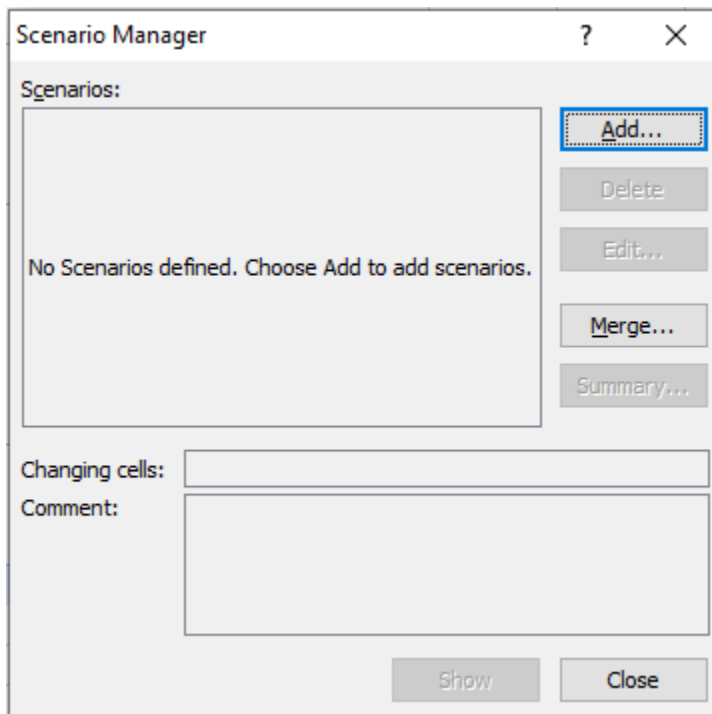


2. Click Scenario Manager.



The Scenario Manager dialog box appears.

3. Add a scenario by clicking on Add.
4. Type a name (60% highest), select cell C4 (% sold for the highest price) for the Changing cells and click on OK.



Add Scenario ? X

Scenario name:
60% highest

Changing cells:
SC\$4

Ctrl+click cells to select non-adjacent changing cells.

Comment:
Created by excel-easy.com on 1/24/2020

Protection

☒ Prevent changes

☐ Hide

OK Cancel

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

Scenario Values ? X

Enter values for each of the changing cells.

1: SC\$4 0.6

Add OK Cancel

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

Scenario Manager ? X

Scenarios:

- 60% highest
- 70% highest
- 80% highest
- 90% highest
- 100% highest

Add... Delete Edit... Merge... Summary...

Changing cells: SC\$4

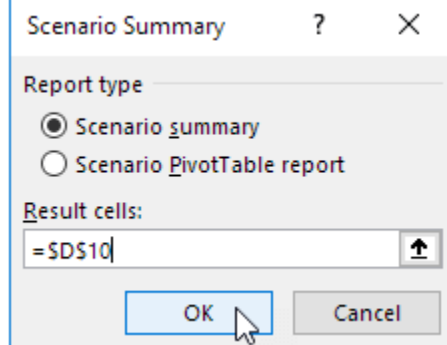
Comment: Created by excel-easy.com on 1/24/2020

Show Close

Scenario Summary

To easily compare the results of these scenarios, execute the following steps.

1. Click the Summary button in the Scenario Manager.
2. Next, select cell D10 (total profit) for the result cell and click on OK



Scenario Summary ? X

Report type

☒ Scenario summary

☐ Scenario PivotTable report

Result cells:

=\$D\$10

OK Cancel

Scenario Summary						
Current Values: 60% highest 70% highest 80% highest 90% highest 100% highest						
Changing Cells:						
\$C\$4	60%	60%	70%	80%	90%	100%
Result Cells:						
\$D\$10	\$3,800	\$3,800	\$4,100	\$4,400	\$4,700	\$5,000

Practical 6: Implementation of Classification algorithm in R Programming.

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)

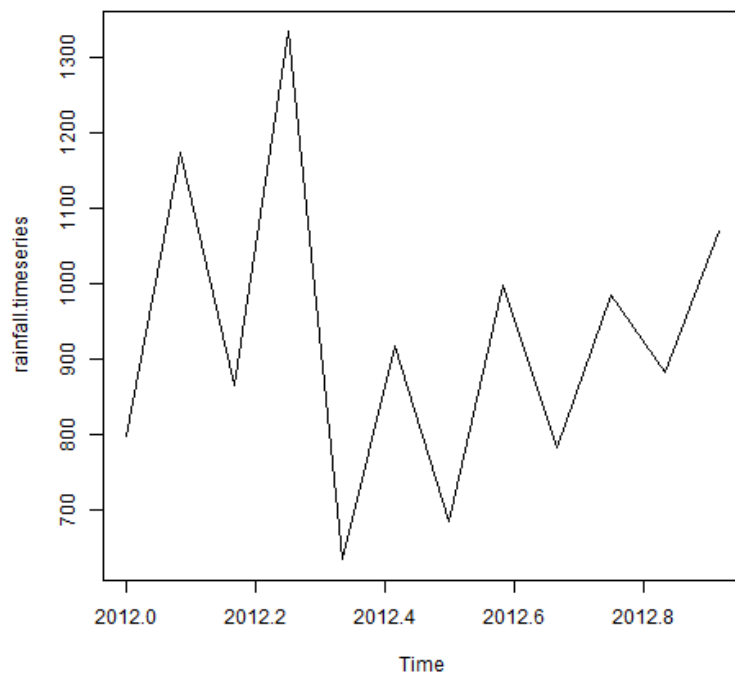
# Convert it to a time series object.
rainfall.timeseries <- ts(rainfall,start = c(2012,1),frequency = 12)

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

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

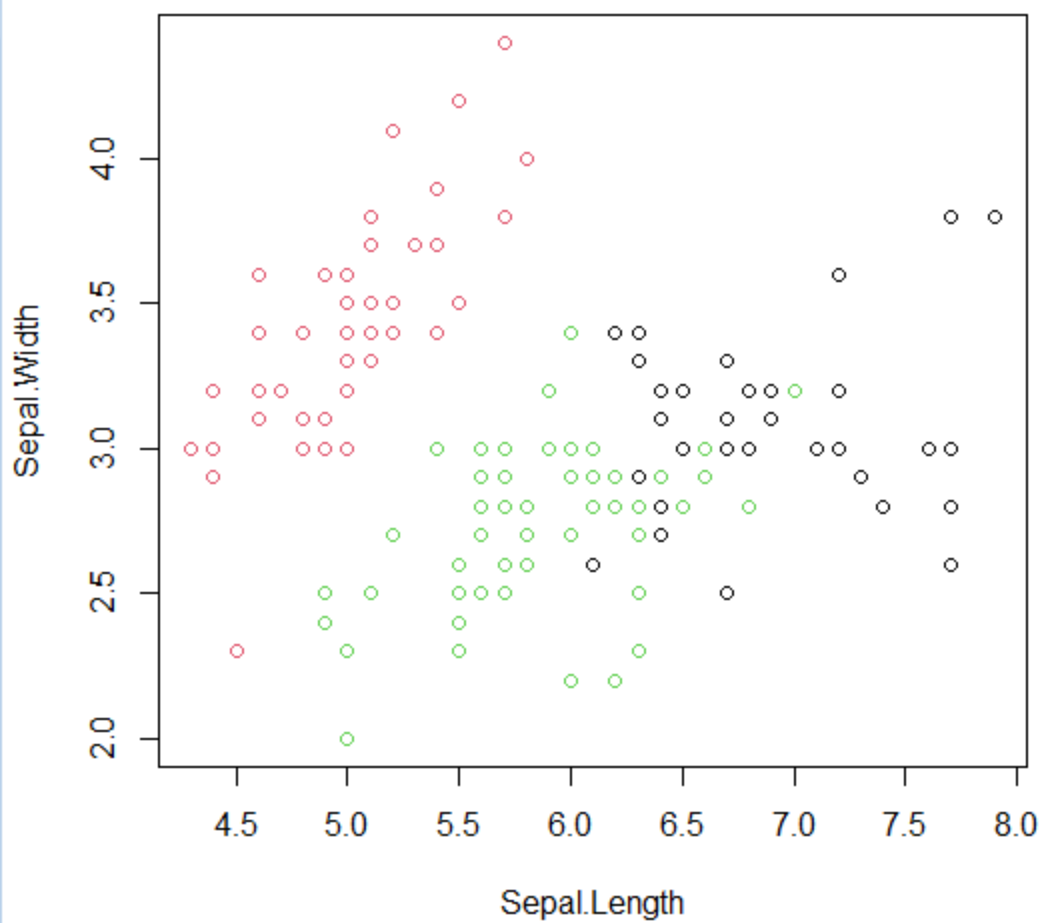
# Plot a graph of the time series.
plot(rainfall.timeseries)

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



[illegible]

R Graphics: Device 2 (ACTIVE)



Practical 4: Creating a Cube in SQL server 2012

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.

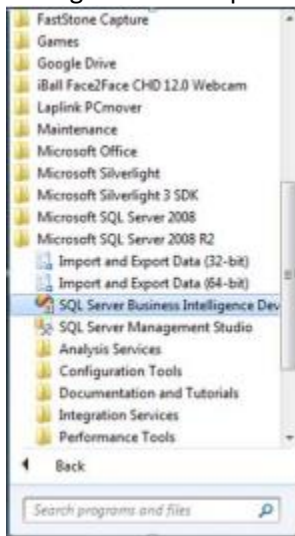
Download T-SQL script attached with this article for creation of Sales Data Warehouse or download from this article "Create First Data Warehouse" and run it in your SQL Server.

Follow the given steps to run the query in SSMS (SQL Server Management Studio).

1. Open SQL Server Management Studio 2008
 2. Connect Database Engine
 3. 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
- Developing an OLAP Cube
For creation of OLAP Cube in Microsoft BIDS Environment, follow the 10 easy steps given below

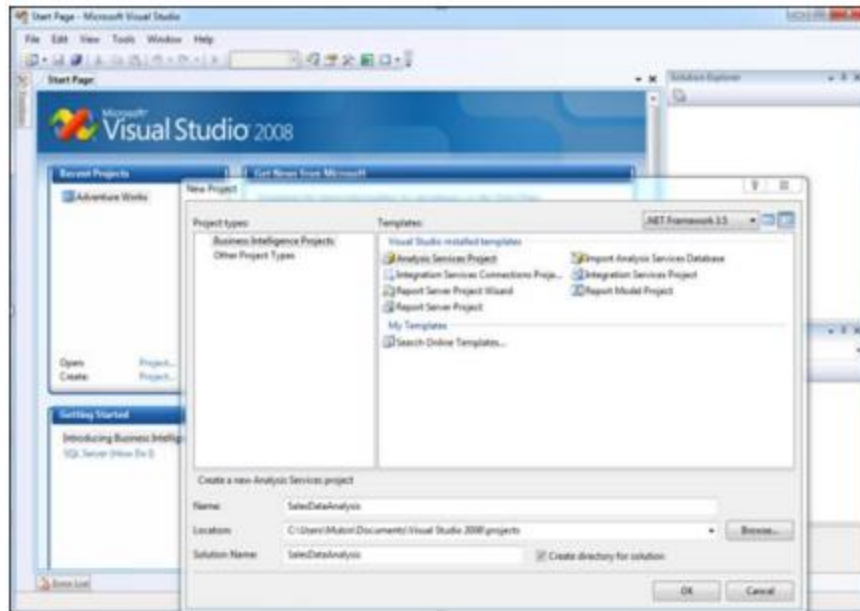
Step 1: Start BIDS Environment

Click on Start Menu -> Microsoft SQL Server 2008 R2 -> Click SQL Server Business Intelligence Development Studio



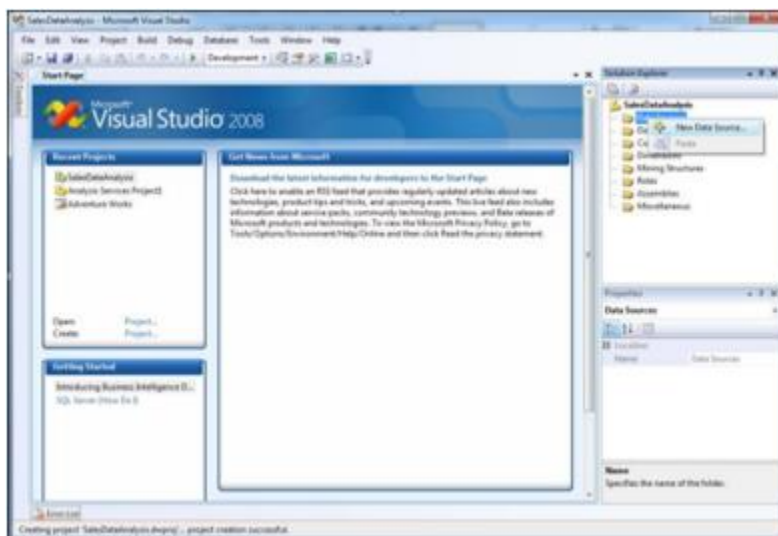
Step 2: Start Analysis Services Project

Click File -> New -> Project -> Business Intelligence Projects -> select Analysis Services Project -> Assign Project Name -> Click OK

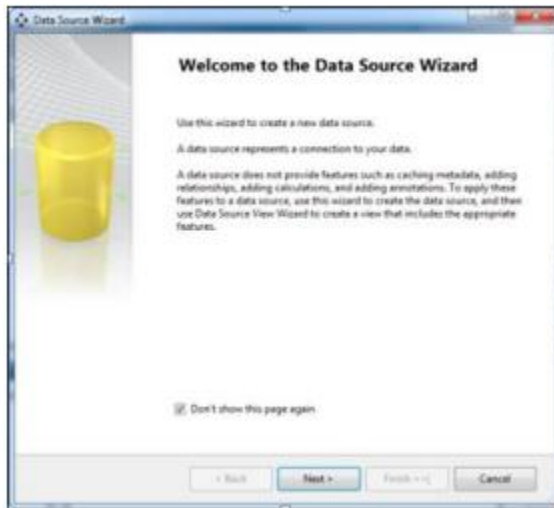


Step 3: Creating New Data Source

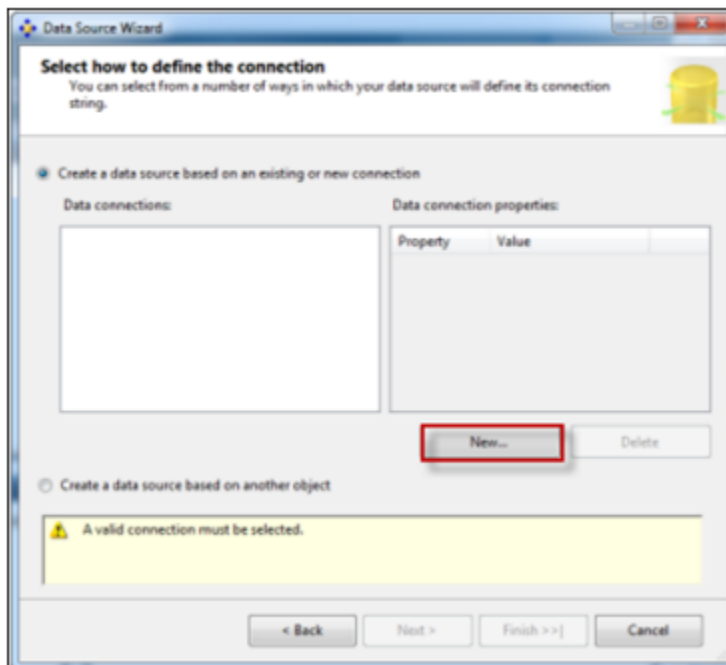
3.1 In Solution Explorer, Right click on Data Source -> Click New Data Source



3.2 Click on Next

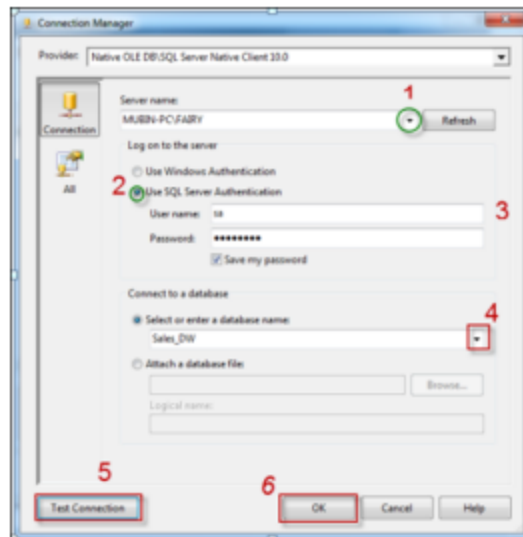


3.3 Click on New Button

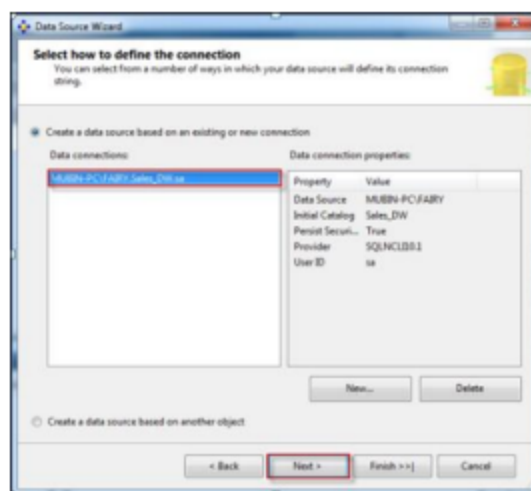


3.4 Creating New connection

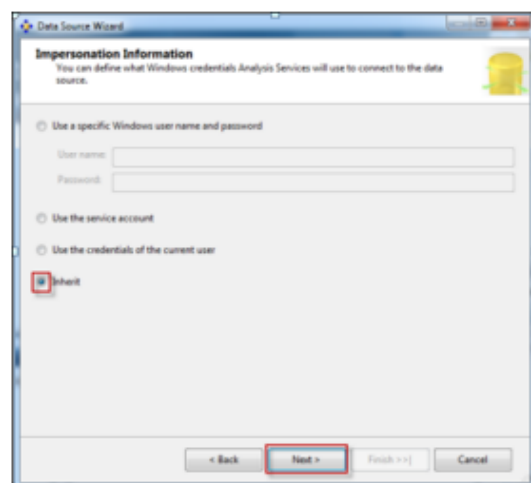
1. Specify Your SQL Server Name where your Data Warehouse was created
2. Select Radio Button according to your SQL Server Authentication mode
3. Specify your Credentials using which you can connect to your SQL Server
4. Select database Sales_DW.
5. Click on Test Connection and verify for its success
6. Click OK



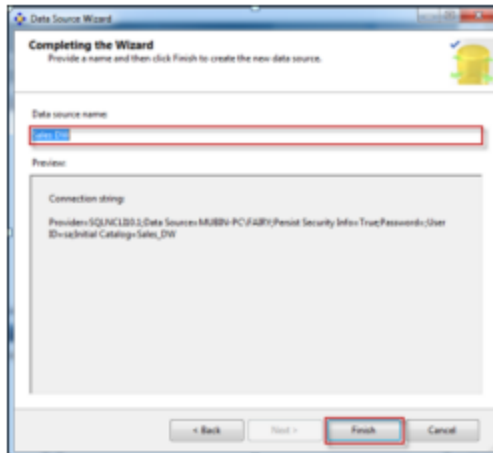
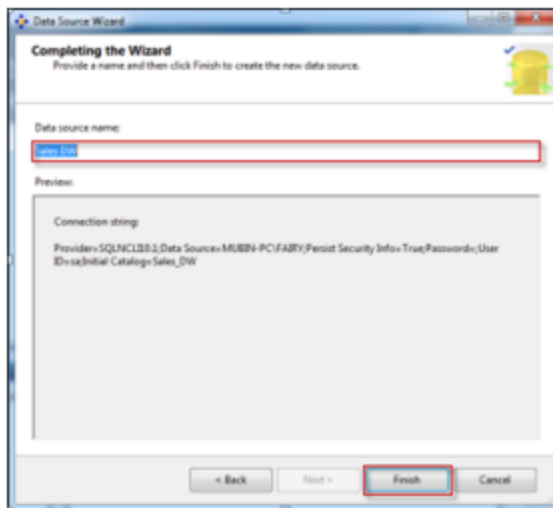
3.5 Select Connection created in **Data Connections**-> Click Next



3.6 Select Option **Inherit**

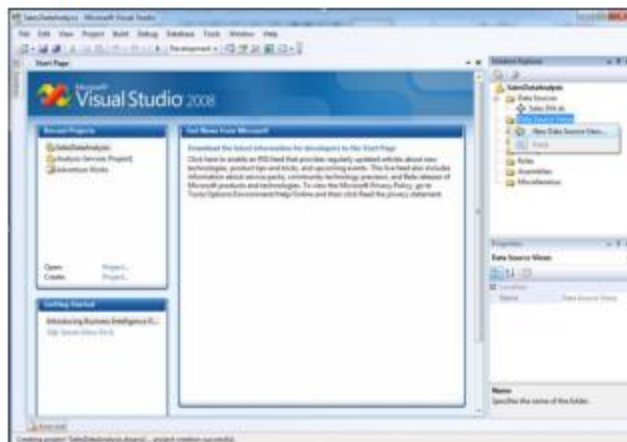


3.7 Assign Data Source Name -> Click Finish



Step 4: Creating New Data Source View

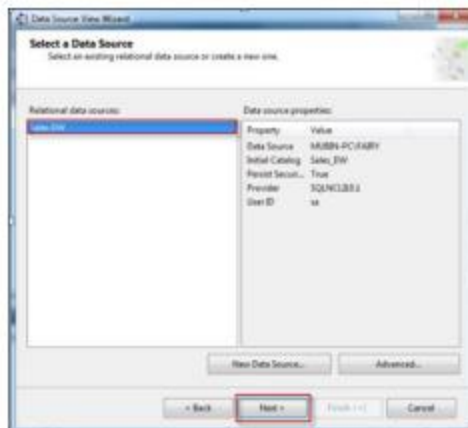
4.1 In the Solution Explorer, Right Click on **Data Source View** -> Click on **New Data Source View**



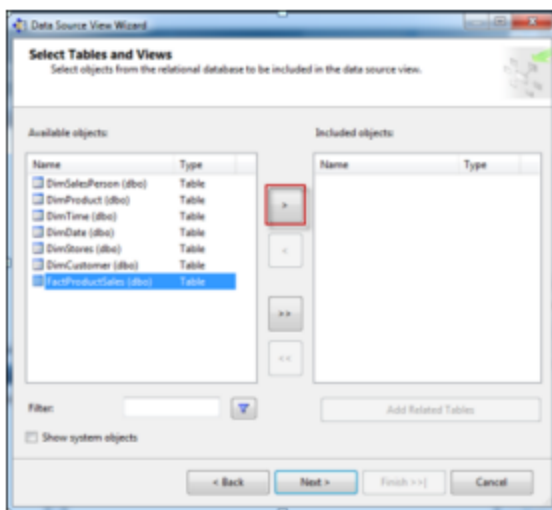
4.2 Click Next



4.3 Select Relational Data Source we have created previously (Sales_DW)-> Click Next



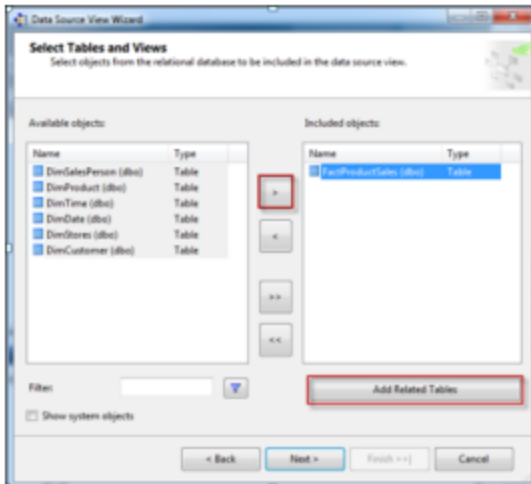
4.4 First move your **Fact Table** to the right side to include in object list.



Select FactProductSales Table -> Click on Arrow Button to move the selected object to Right Pane.

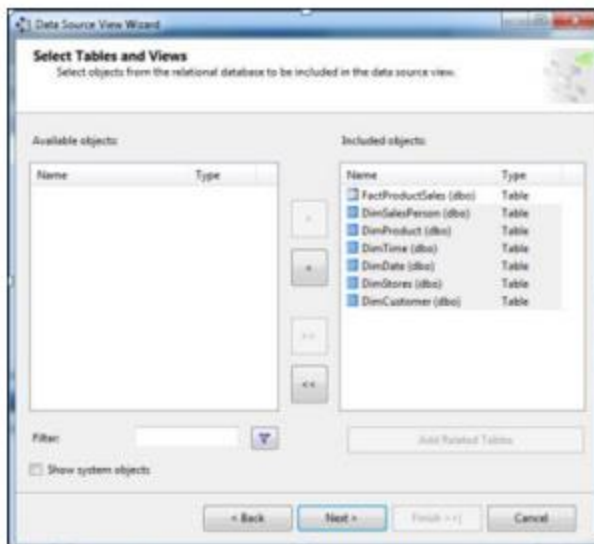
4.5 Now to **add dimensions** which are **related** to your **Fact Table**, follow the given steps:

Select **Fact Table** in Right Pane (Fact product Sales) -> Click On **Add Related Tables**

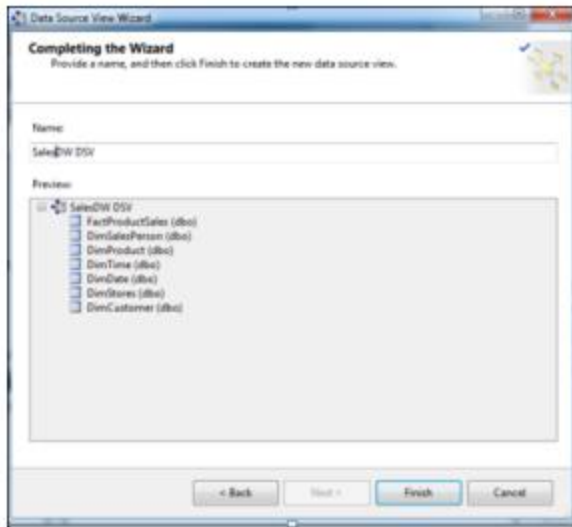


4.6 It will add all associated dimensions to your Fact table as per relationship specified in your SQL DW (Sales_DW).

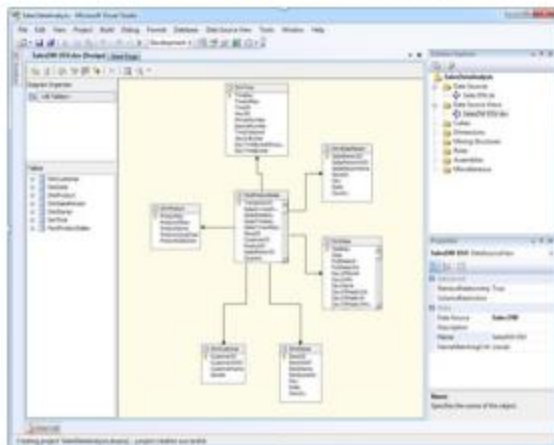
Click **Next**.



4.7 Assign Name (SalesDW DSV)-> Click **Finish**

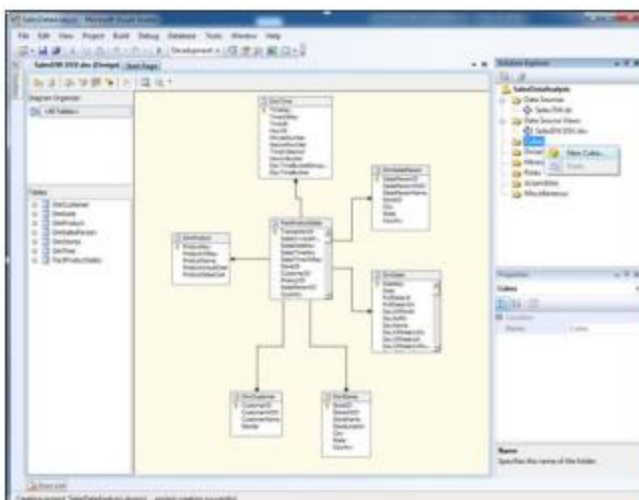


4.8 Now Data Source View is ready to use.



Step 5: Creating New Cube

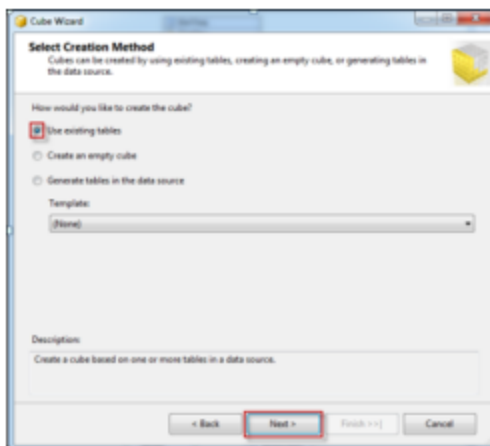
5.1 In Solution Explorer -> Right Click on **Cube-> Click **New Cube****



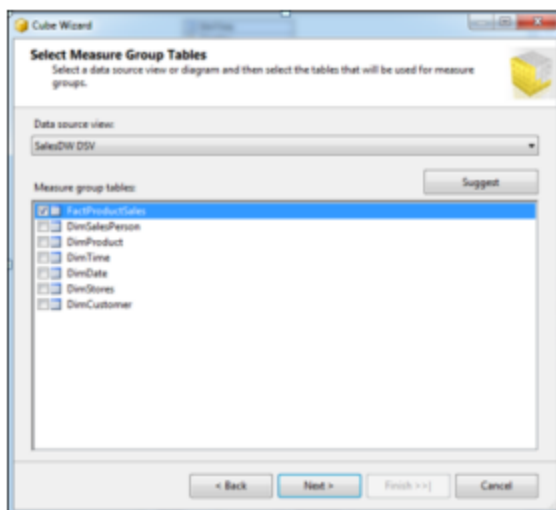
5.2 Click Next



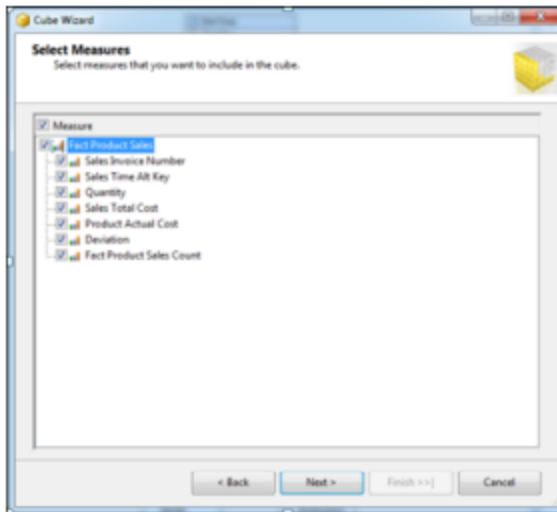
5.3 Select Option Use existing Tables -> Click Next



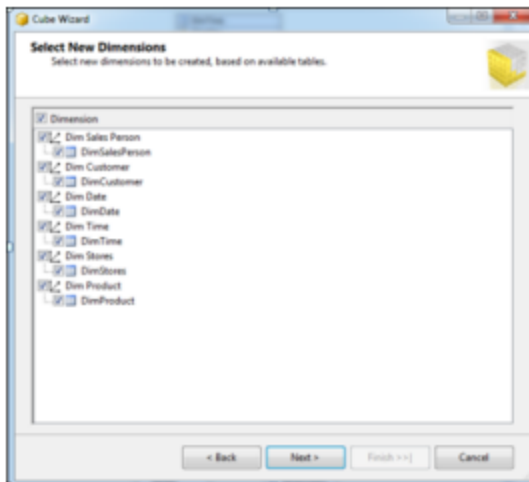
5.4 Select Fact Table Name from Measure Group Tables (FactProductSales) -> Click Next



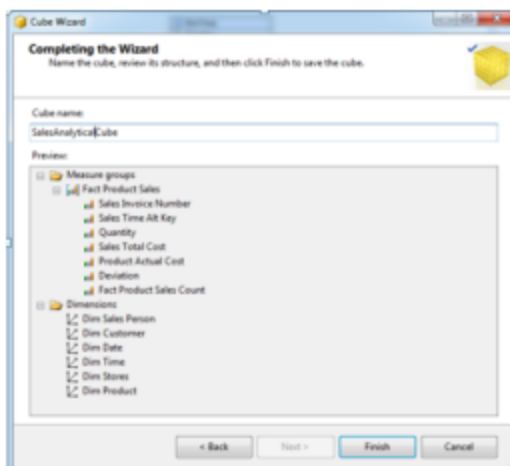
5.5 Choose Measures from the List which you want to place in your Cube --> Click Next



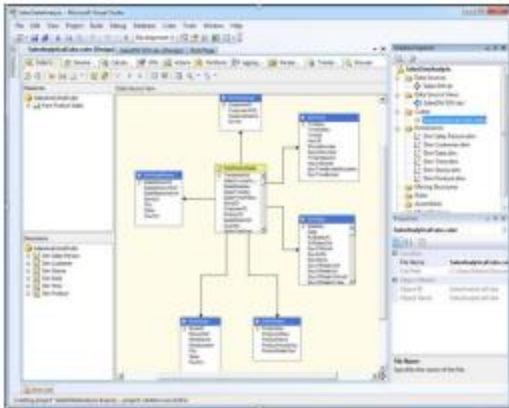
5.6 Select All **Dimensions** here which are associated with your Fact Table-> Click **Next**



5.7 Assign **Cube Name** (SalesAnalyticalCube) -> Click **Finish**

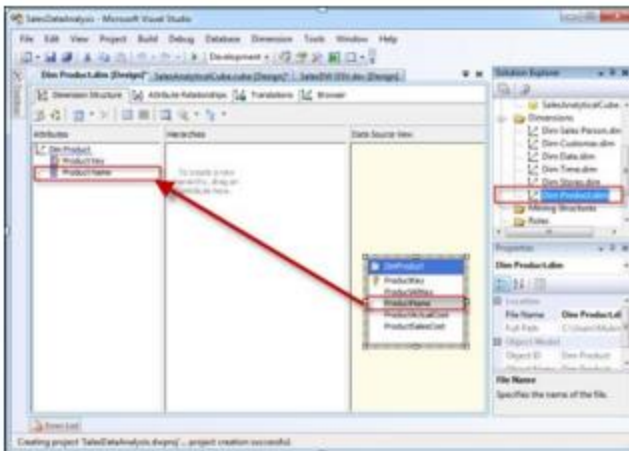


5.8 Now your Cube is ready, you can see the newly created cube and dimensions added in your solution explorer.



Step 6: Dimension Modification

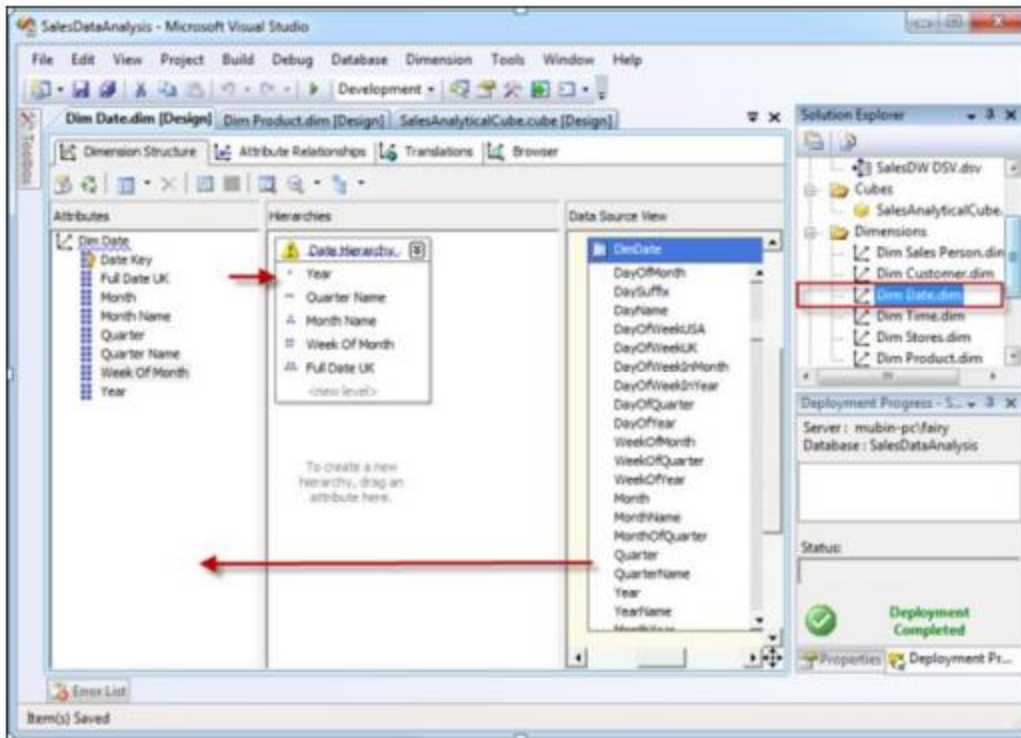
In Solution Explorer, double click on dimension **Dim Product** -> Drag and Drop Product Name from Table in Data Source View and Add in Attribute Pane at left side.



Step 7: Creating Attribute Hierarchy In Date Dimension

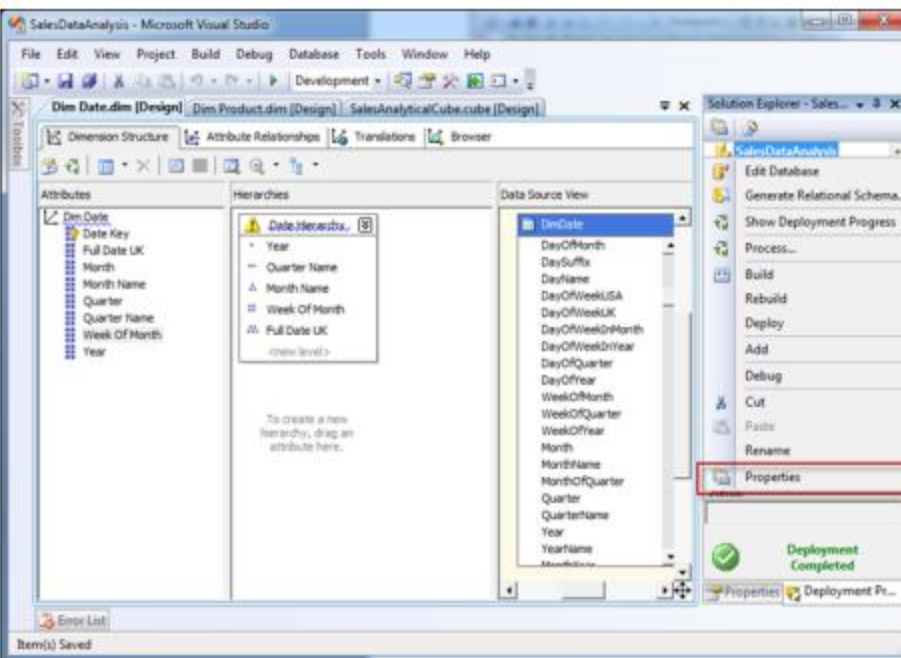
Double click On **Dim Date** dimension -> Drag and Drop Fields from Table shown in Data Source View to Attributes-> Drag and Drop attributes from leftmost pane of attributes to middle pane of Hierarchy.

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



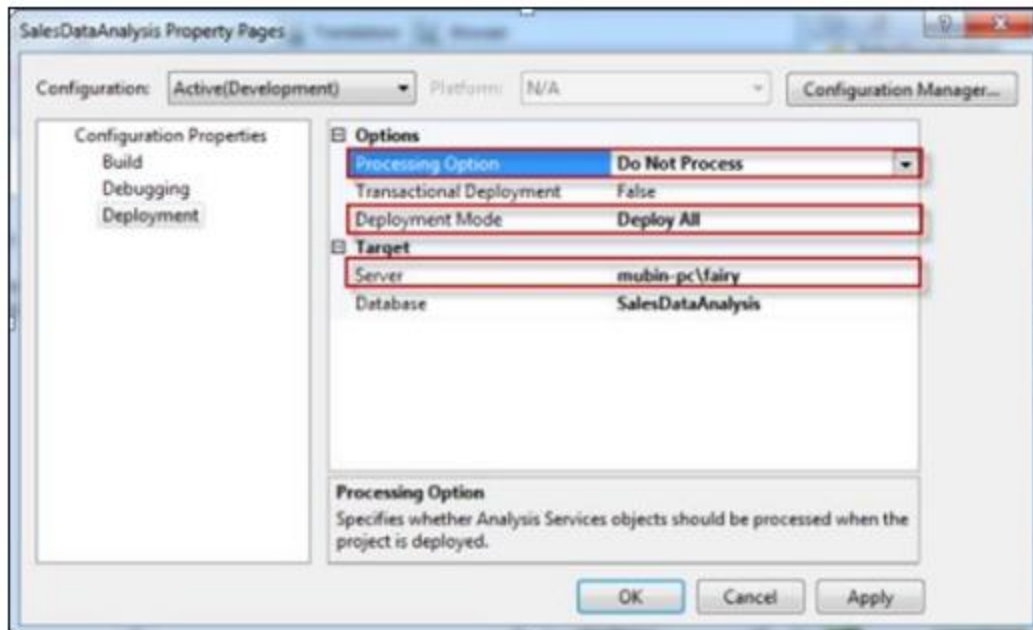
Step 8: Deploy the Cube

8.1 In Solution Explorer, right click on Project Name (SalesDataAnalysis) --> Click Properties

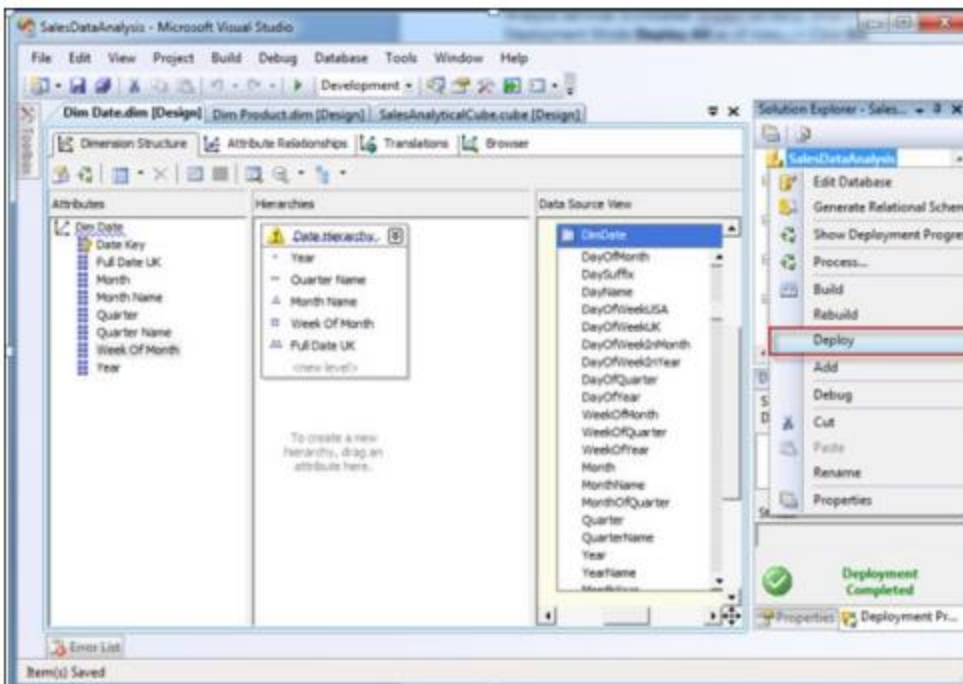


8.2 Set Deployment Properties First

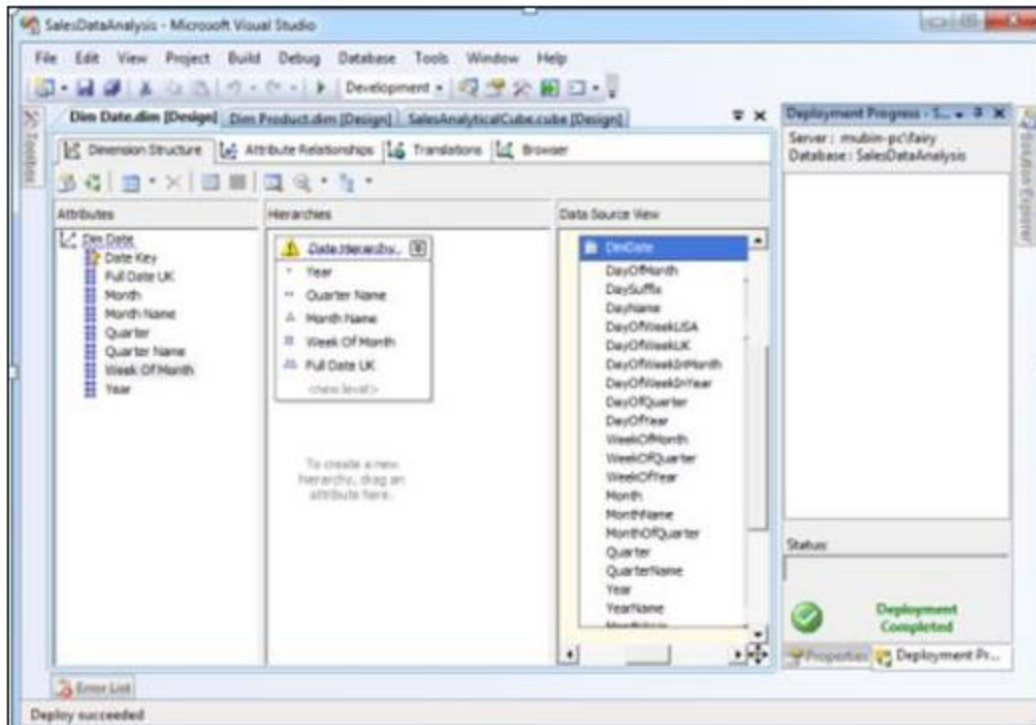
In Configuration Properties, Select Deployment-> Assign Your SQL Server Instance Name Where Analysis Services Is Installed (*mubin-pc\fairly*) (*Machine Name\Instance Name*) -> Choose Deployment Mode **Deploy All** as of now ->Select Processing Option **Do Not Process** -> Click **OK**



8.3 In Solution Explorer, right click on **Project Name** (SalesDataAnalysis) --> Click **Deploy**

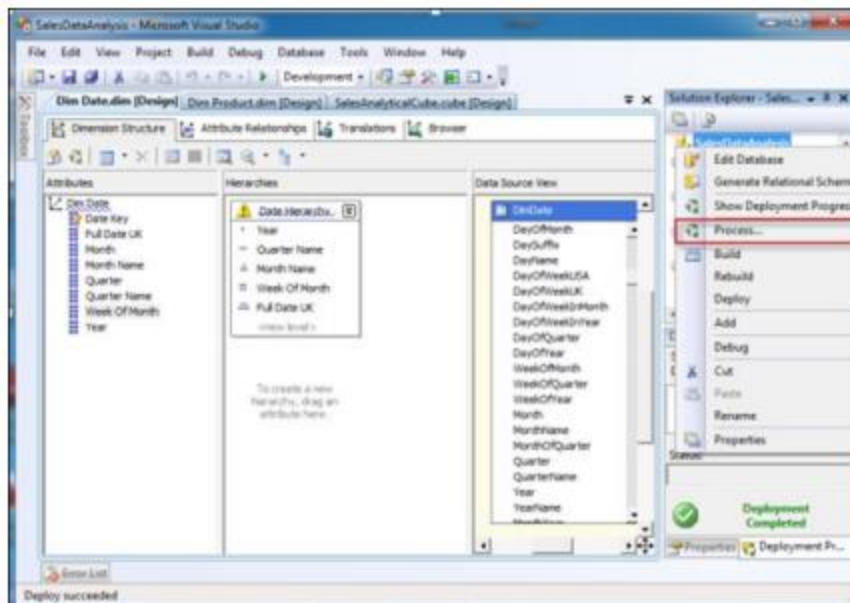


8.4 Once Deployment will finish, you can see the message **Deployment Completed** in deployment Properties.

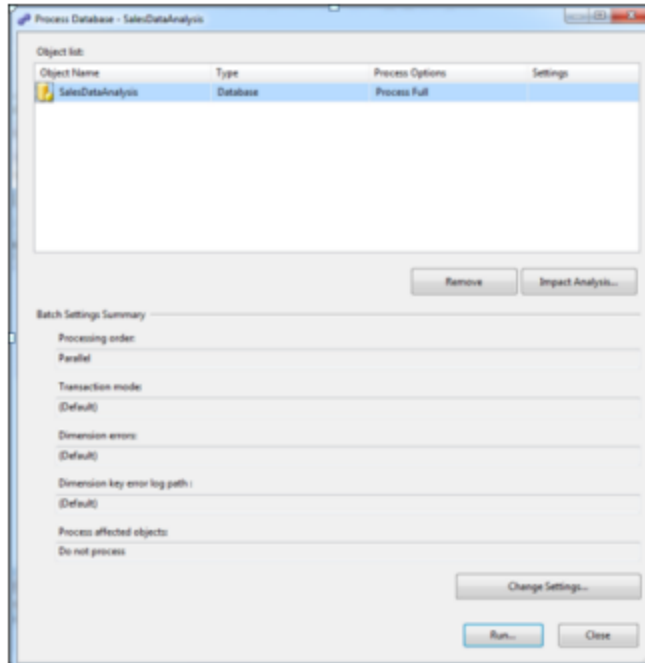


Step 9: Process the Cube

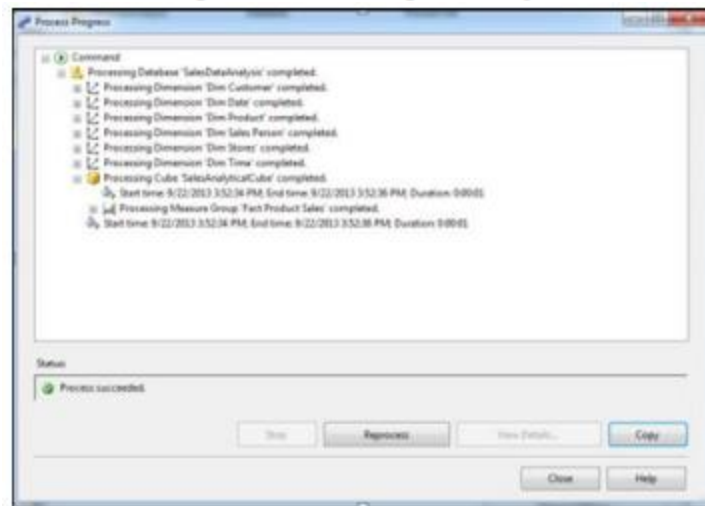
9.1 In Solution Explorer, right click on Project Name (SalesDataAnalysis) -- > Click **Process**



9.2 Click on **Run** button to process the Cube

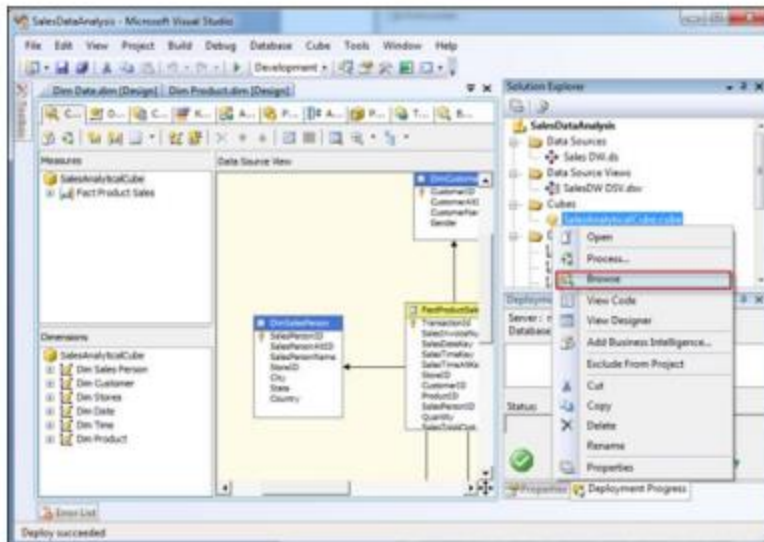


9.3 Once processing is complete, you can see **Status** as **Process Succeeded** --> Click **Close** to close both the open windows for processing one after the other.



Step 10: Browse the Cube for Analysis

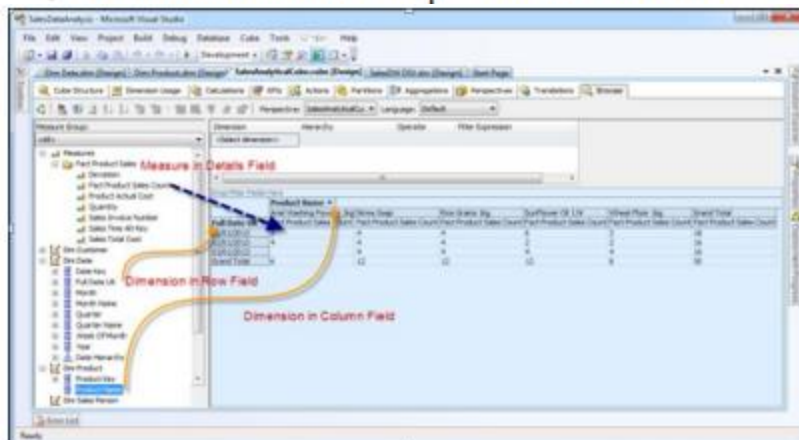
10.1 In Solution Explorer, right click on Cube Name (SalesDataAnalysisCube) -- > Click **Browse**



10.2 Drag and drop measures in to Detail fields, & Drag and Drop Dimension Attributes in Row Field or Column fields.

Now to **Browse Our Cube**

1. Product Name Drag & Drop into Column
2. Full Date UK Drag & Drop into Row Field
3. FactProductSalesCount Drop this measure in Detail area



Practical 7: Practical Implementation of Decision Tree using R Tool

```
install.packages("party")
```

The package "party" has the function `ctree()` which is used to create and analyze decision tree.
Syntax

The basic syntax for creating a decision tree in R is –
`ctree(formula, data)`

Input Data

We will use the R in-built data set named `readingSkills` to create a decision tree. It describes the score of someone's readingSkills if we know the variables "age", "shoeSize", "score" and whether the person is a native speaker or not.

Here is the sample data.

```
# Load the party package. It will automatically load other
# dependent packages.
library(party)
# Print some records from data set readingSkills.
print(head(readingSkills))
```

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

```
  nativeSpeaker  age shoeSize  score
1         yes    5  24.83189 32.29385
2         yes    6  25.95238 36.63105
3         no   11  30.42170 49.60593
4         yes    7  28.66450 40.28456
5         yes   11  31.88207 55.46085
6         yes   10  30.07843 52.83124
Loading required package: methods
Loading required package: grid
.....
.....
```

We will use the `ctree()` function to create the decision tree and see its graph.

```
# Load the party package. It will automatically load other
# dependent packages.
library(party)
# Create the input data frame.
input.dat <- readingSkills[c(1:105),]
# Give the chart file a name.
png(file = "decision_tree.png")
# Create the tree.
output.tree <- ctree(
  nativeSpeaker ~ age + shoeSize + score,
  data = input.dat)
# Plot the tree.
plot(output.tree)
# Save the file.
dev.off()
```


Output:-

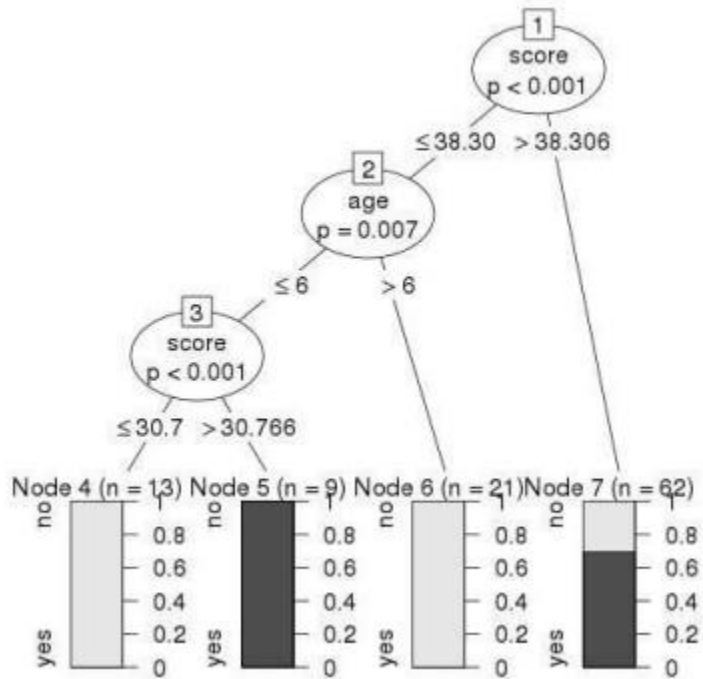
```
null device
  1
Loading required package: methods
Loading required package: grid
Loading required package: mvtnorm
Loading required package: modeltools
Loading required package: stats4
Loading required package: strucchange
Loading required package: zoo
```

Attaching package: 'zoo'

The following objects are masked from 'package:base':

as.Date, as.Date.numeric

Loading required package: sandwich



Practical 9: Prediction Using Linear Regression

In Linear Regression these two variables are related through an equation, where exponent (power) of both these variables is 1. Mathematically a linear relationship represents a straight line when plotted as a graph. A non-linear relationship where the exponent of any variable is not equal to 1 creates a curve.

$y = ax + b$ is an equation for linear regression.

Where, y is the response variable, x is the predictor variable and a and b are constants which are called the coefficients.

A simple example of regression is predicting weight of a person when his height is known.

To do this we need to have the relationship between height and weight of a person.

The steps to create the relationship is –

- Carry out the experiment of gathering a sample of observed values of height and corresponding weight.
- Create a relationship model using the `lm()` functions in R.
- Find the coefficients from the model created and create the mathematical equation using these
- Get a summary of the relationship model to know the average error in prediction. Also called residuals.
- To predict the weight of new persons, use the `predict()` function in R.

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 resposne 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

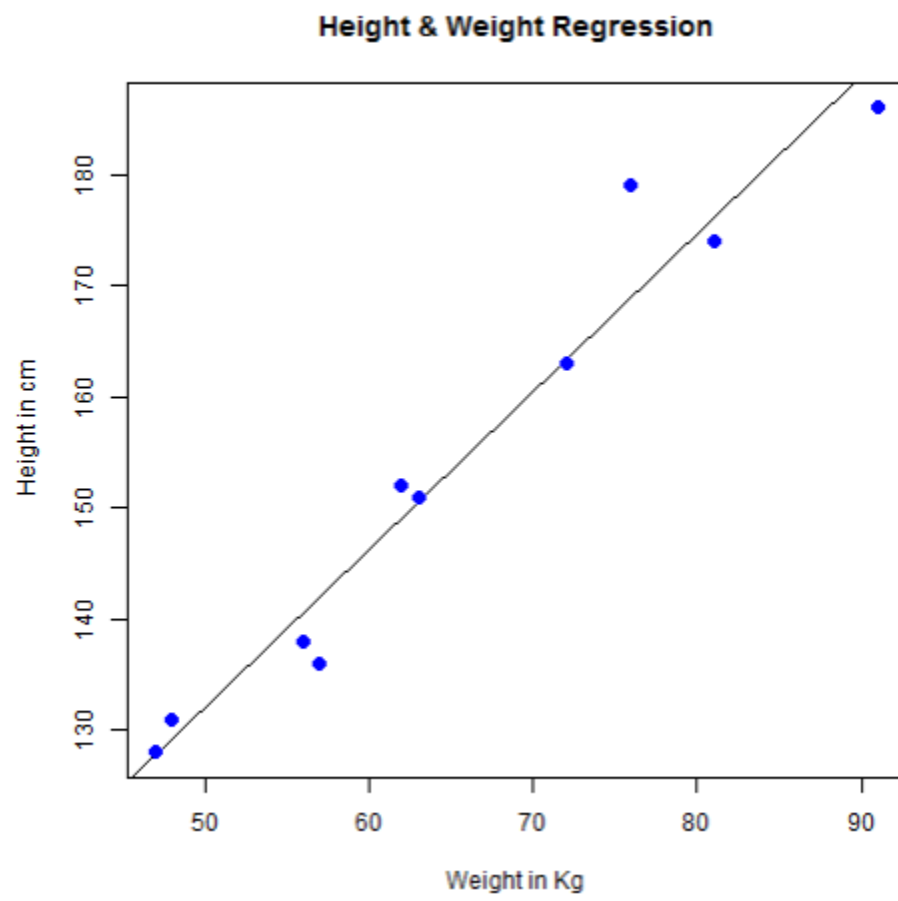
```
# Create the predictor and response variable.
x <- c(151, 174, 138, 186, 128, 136, 179, 163, 152, 131)
y <- c(63, 81, 56, 91, 47, 57, 76, 72, 62, 48)
relation <- lm(y~x)

# 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 10: Data Analysis using Time Series Analysis

Time series is a series of data points in which each data point is associated with a timestamp. A simple example is the price of a stock in the stock market at different points of time on a given day. Another example is the amount of rainfall in a region at different months of the year. R language uses many functions to create, manipulate and plot the time series data. The data for the time series is stored in an R object called time-series object. It is also a R data object like a vector or data frame.

The time series object is created by using the `ts()` function.

Syntax

The basic syntax for `ts()` function in time series analysis is –
`timeseries.object.name <- ts(data, start, end, frequency)`

Following is the description of the parameters used –

- data is a vector or matrix containing the values used in the time series.
- start specifies the start time for the first observation in time series.
- end specifies the end time for the last observation in time series.
- frequency specifies the number of observations per unit time.

Except the parameter "data" all other parameters are optional.

Example

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

Jan	Feb	Mar	Apr	May	Jun	Jul	Aug	Sep	
2012	799.0	1174.8	865.1	1334.6	635.4	918.5	685.5	998.6	784.2
	Oct	Nov	Dec						
2012	985.0	882.8	1071.0						

