KULLIYYAH OF INFORMATION AND COMMUNICATION TECHNOLOGY

DEPARTMENT OF INFORMATION SYSTEMS

INFO 4311 – Data Warehousing

Semester 2 2021/2022

Data Profiling and High-Level Data Warehouse Design Project

Section: 1

Team Name: Group C

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1. **INTRODUCTION**

Large retail & commerce stores have long been in a fierce competition among one another especially with the on-growing demand of household goods, food items, clothing, and a wide variety of other industries. With such tremendous amount of data available within their reach, it's important to analyse it and to develop association methods along with it to further understand which products best suit their location, categories, and customer segments. By doing so, losses could be avoided, and a better comprehension of the customer purchasing behaviour is achieved.

The objective of this project is to comprehend what exactly the data set is about, identify key relationships between each attribute, and to perform complex SQL Queries on-demand. Such process of examining, analysing, and crafting useful summaries of data is known as data profiling. By profiling data for the superstore used in our example here, we'd be able to yield a high-level overview which can eliminate costly errors such as data quality issues, risks, and values outside the normal range. All these factors are made to align with the business's standards & goals, and in turn produces critical insights into data that the superstore can leverage to their advantage.

1. **DATA SETS**

1. Description

- The sample data taken is from a superstore giant's data set. The data set include the details of every individual orders.
- The metadata for the dataset include the following columns:

COLUMNS	METADATA
RowID	The unique ID for each row
OrderID	The unique ID for each order
Order Date	Order Date of the product.
Ship Date	The product's shipping date.
Ship Mode	Customer-specified shipping method
Customer ID	Unique ID to identify each Customer.
Customer Name	Name Of the Customer
Segment	The segment where the Customer belongs.
Country	Customer's country of residence.
City	The Customer's city of residence.
State	Customer's state of residence
Postal Code	Every customer's postal code.
Region	Region where the Customer belong.
Product ID	Unique ID of the Product.
Category	Category of the product ordered.
Sub-Category	Sub-Category of the product ordered.
Product Name	Name of the Product
Sales	Sales of the Product.
Quantity	Quantity of the Product.
Discount	Discount provided.
Profit	Profit/Loss that was made.

1. Reference to Data Set

- View Data Orders Table
 - Superstore Dataset
- External Link Sourced from Kaggle
 - o https://www.kaggle.com/datasets/vivek468/superstore-dataset-final

1. Data Set Tables

 Data Source
 Number of Rows
 Natural Key(s)
 Each row represents

 Superstore Dataset
 9994 rows, 20 columns
 OrderID, CustomerID, Product ID
 An individual order

1. **DATA PROFILING**

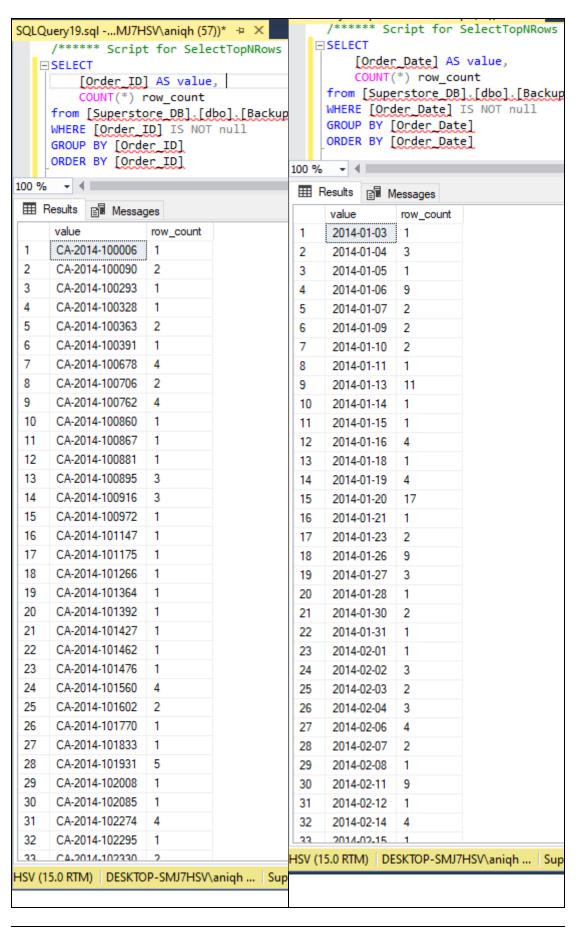
1.

Use SQL queries to run a profile on your data set (for each column) following the statements below. Show the SQL queries and the results. You may compile the results for the columns in the table(s) and present them in the best manner possible.

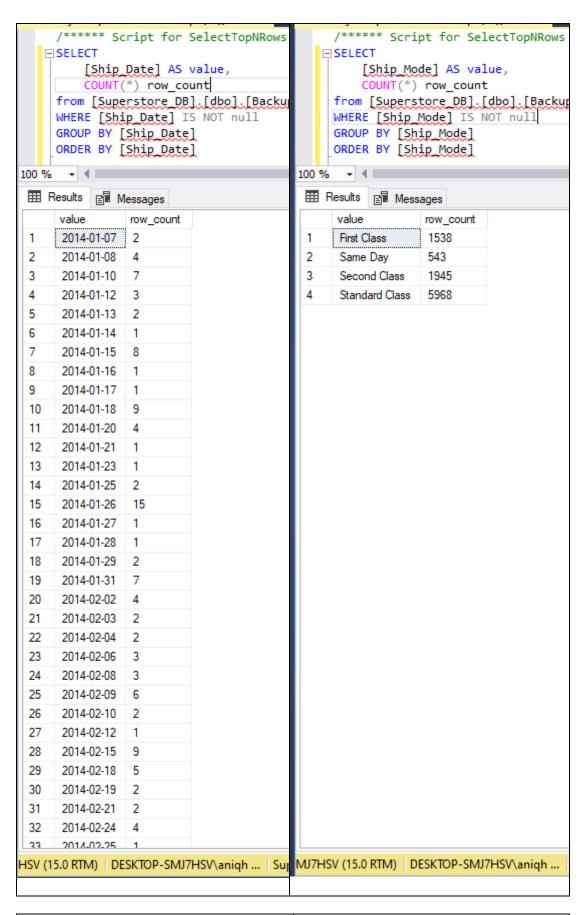
- 1. **Value Distribution -** Find the number of times each value in the column occurs (also show the value). This profile helps you identify problems in your data, such as incorrect number of distinct values in a column. For example, you profile a column that is supposed to contain states in Malaysia and discover more than 13 distinct values. (*3 points*)
 - SQL Statement Template:

```
column AS value,
    coUNT(*) row_count
from table
WHERE column IS NOT null
GROUP BY column
ORDER BY column;
```

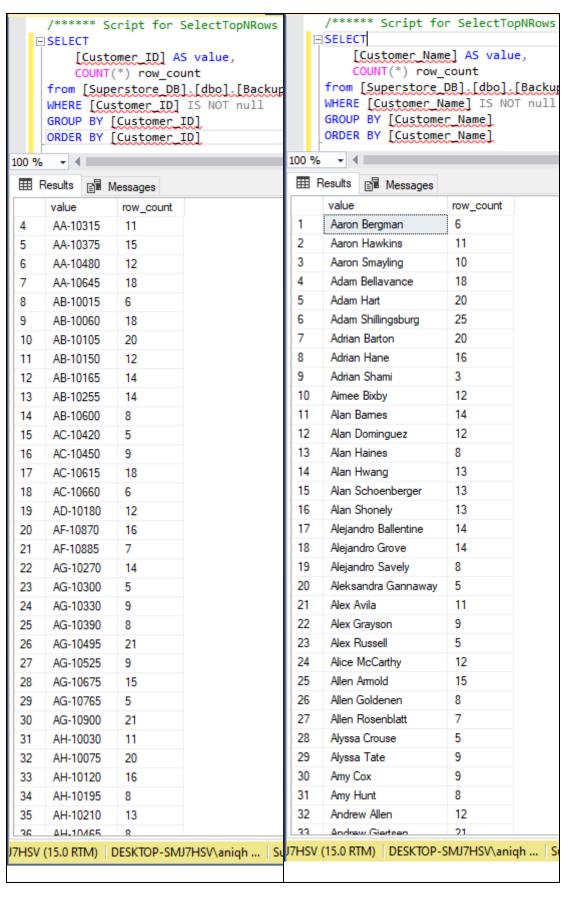
Order_ID							Order_Date	•				•
	•	5,009	rows	sel	ected	-	•	•	1,237	rows	selected	_
	Ind	licates	the nui	mber	of it	tems		Ind	dicates	the	number	of
	bo	ught in	the same	e orde	er.				ders/iten y(date).	ns mad	e during	that



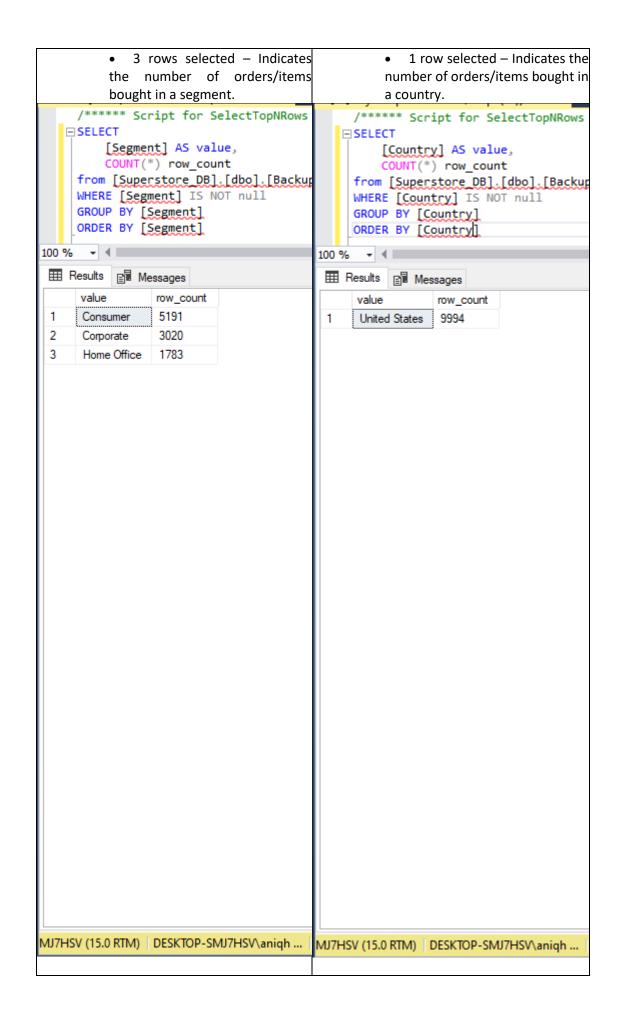
- 1,334 rows selected Indicates the number of orders/items shipped out on a certain day (date).
- 4 rows selected Indicates the number of orders/items shipped out using a specific shipping mode.



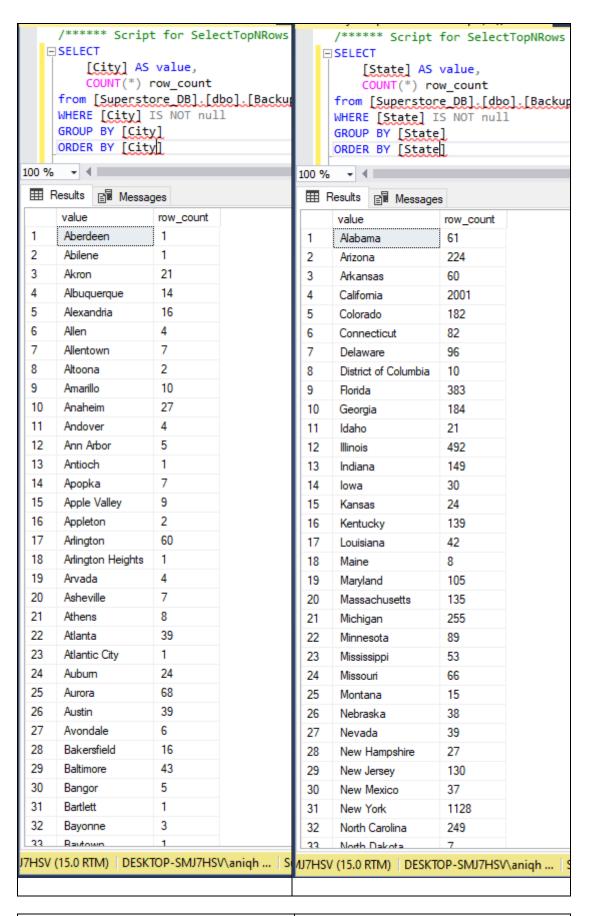
- 793 rows selected Indicates the number of orders/items bought by a customer.
- 793 rows selected Indicates the number of orders/items bought by a customer specified by his/her name.

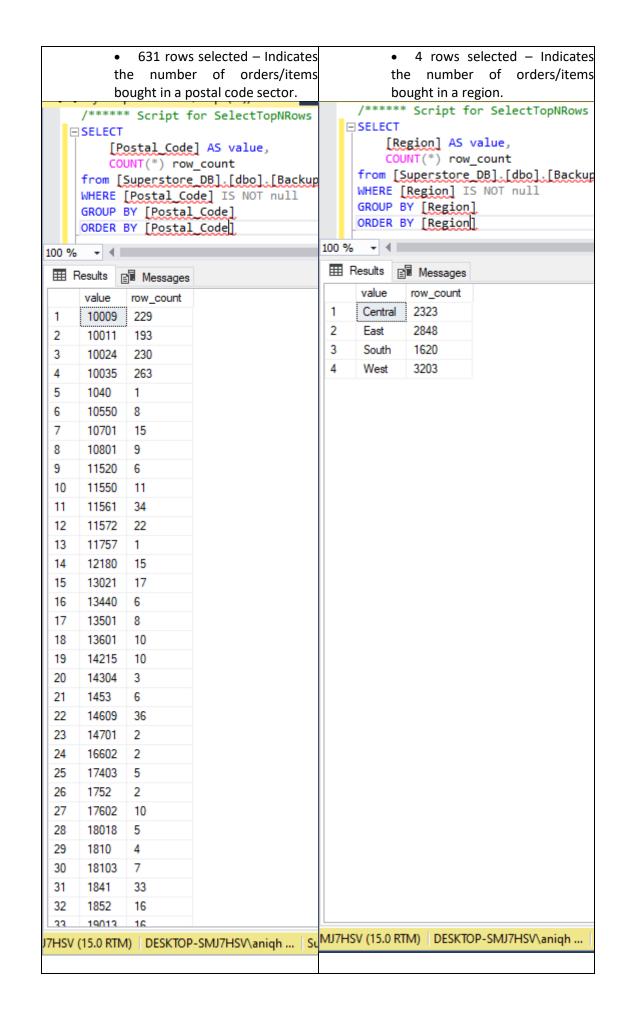


Segment Country

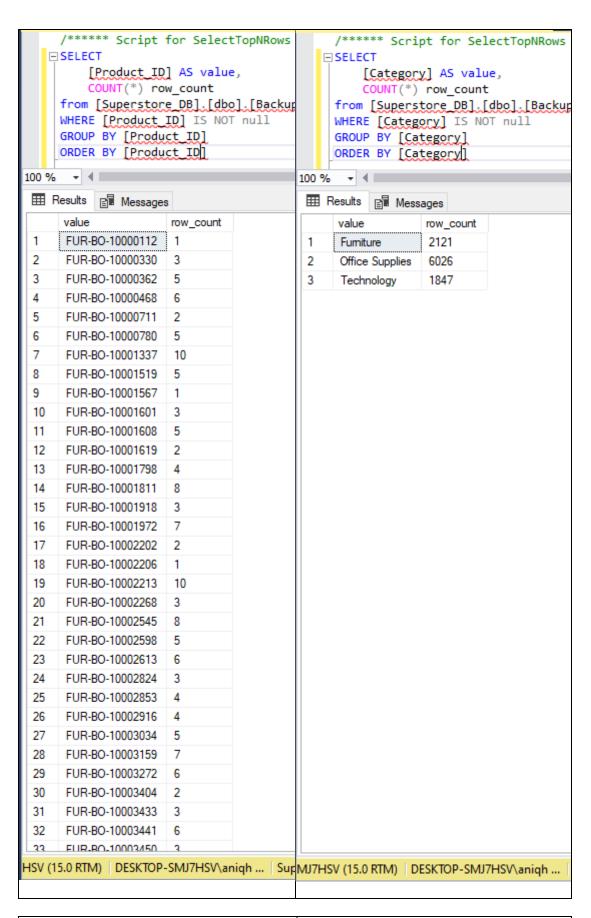


City	State	
	 531 rows selected – Indicates 	 49 rows selected – Indicates
	the number of orders/items	the number of orders/items
	bought in a city.	bought in a state.

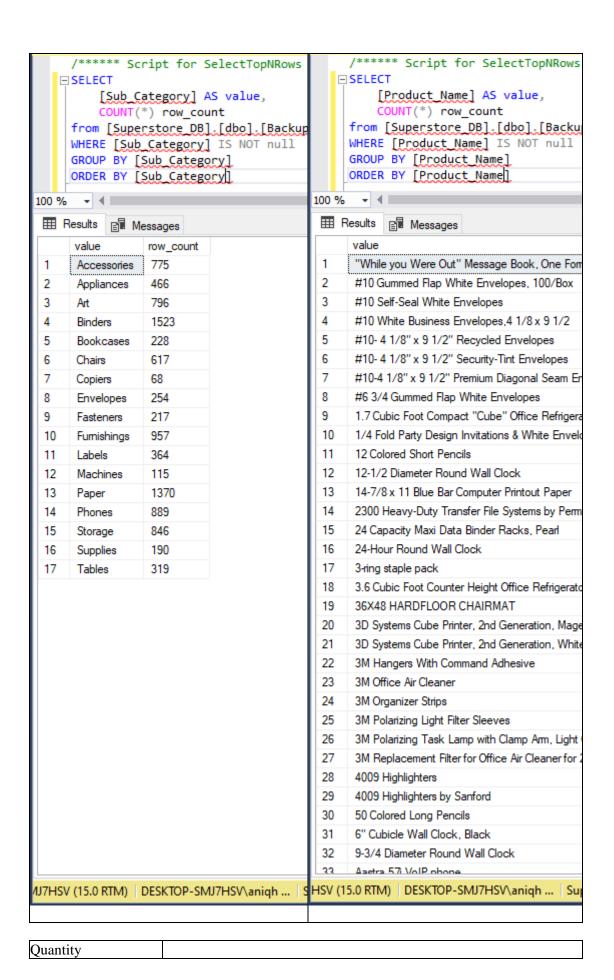


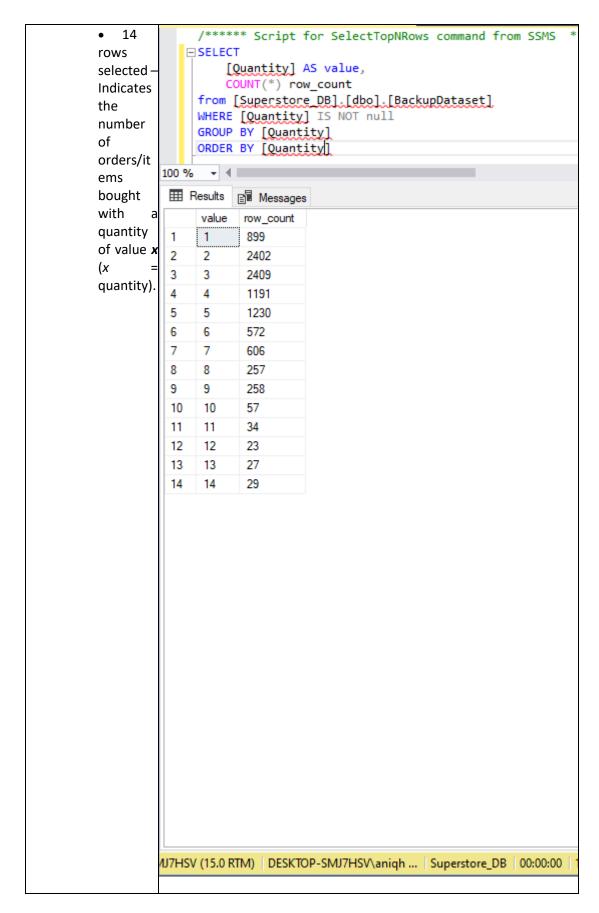


Product_ID	Category
• 1,862 rows selected –	 3 rows selected – Indicates
Indicates the number of times the	the number of orders/items
item was included in an order.	bought in a category.



- 17 rows selected Indicates the number of orders/items bought in a sub-category.
- 1,850 rows selected Indicates the number of times the item was included in an order specified by the product name.





1. **Statistics** – Find the minimum, maximum, average, standard deviation and variance for numeric columns; minimum and maximum for **datetime** columns. This profile helps

you identify problems in your data, such as dates that are not valid. For example, you profile a column of historical dates and discover a maximum date that is in the future. (3 points)

- Our number columns:
 - o Int/Decimal/Float: Sales, Profit, Quantity, Discount
 - o Date: Order_Date, Ship_Date
 - SQL Statement Template:

```
SELECT MIN(Column) AS Minimum,

MAX(Column) AS Maximum,

AVG(Column) AS Mean,

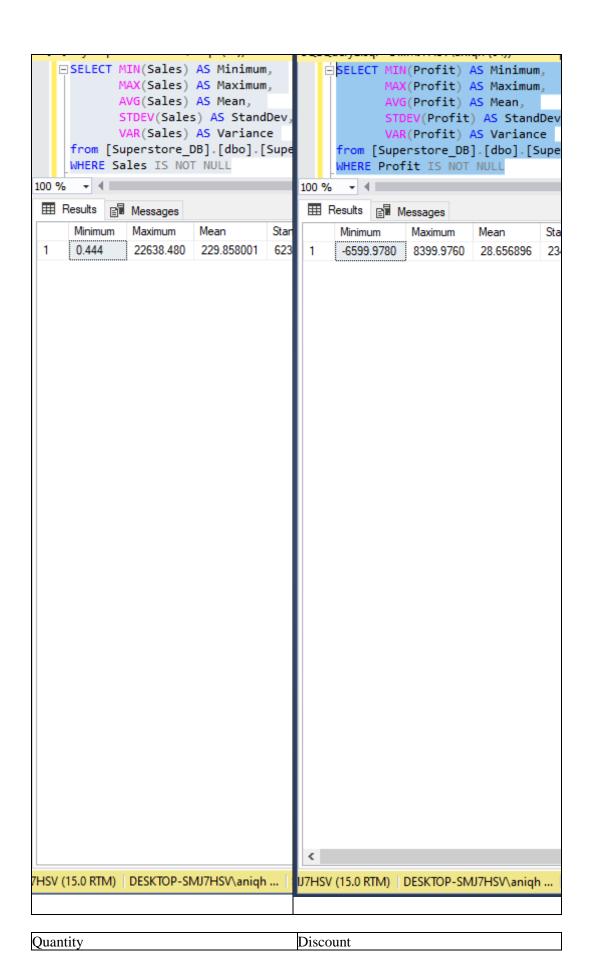
STDEV(Column) AS StandDev,

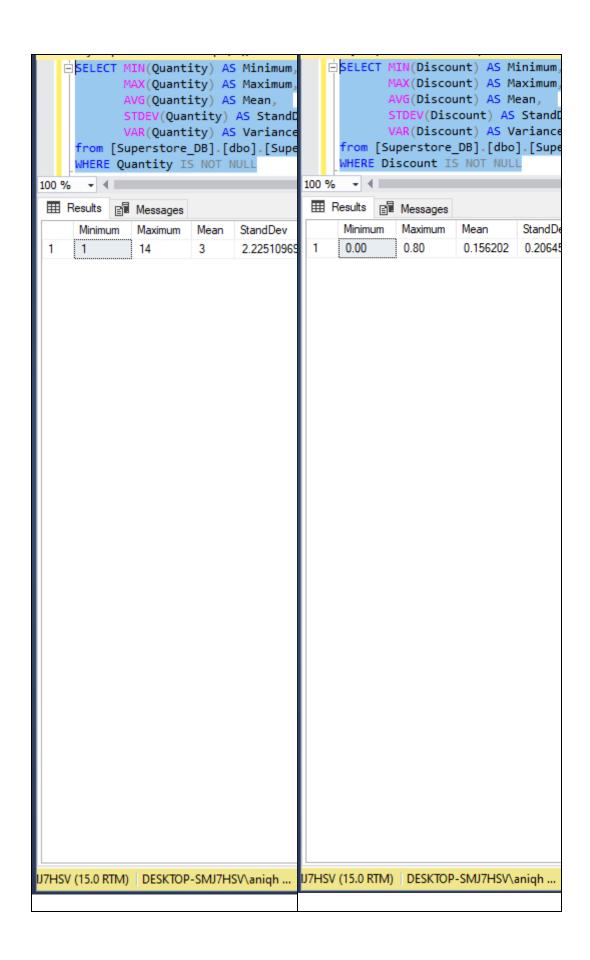
VAR(Column) AS Variance

from table

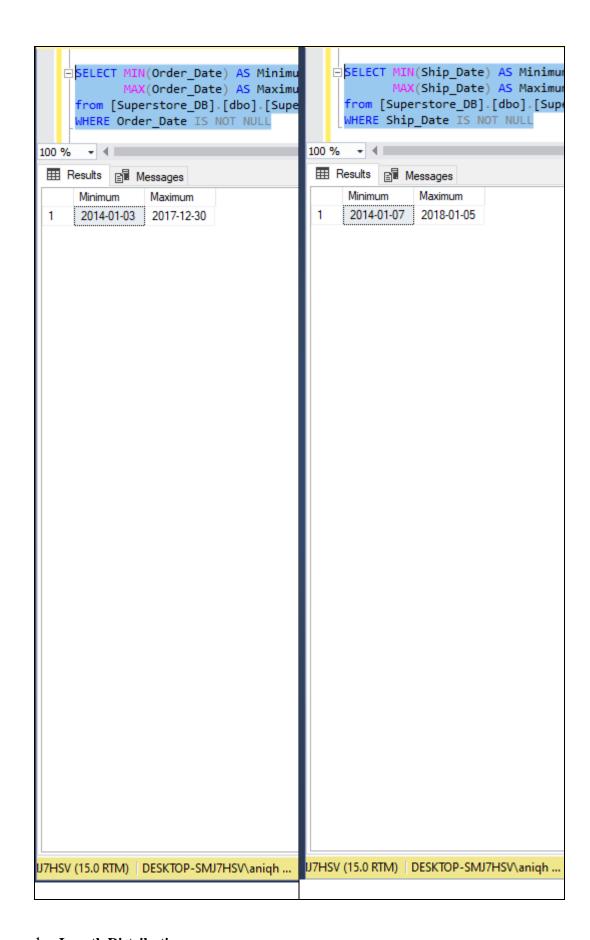
WHERE Column IS NOT NULL
```

Sales	Profit
-------	--------





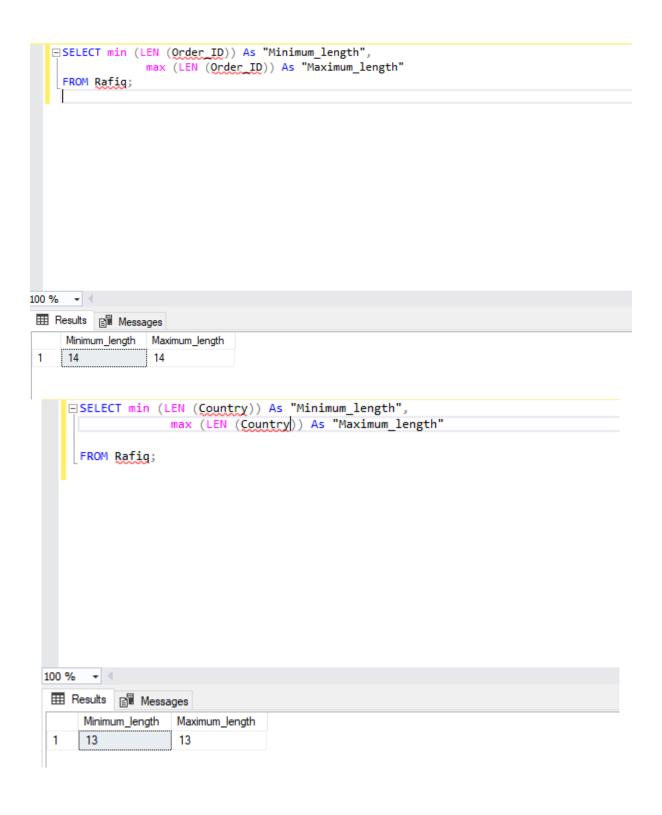
O 1 D - 4 -	Ship Date
lirder 11916	INID LISTA
Order Date	Ship Date

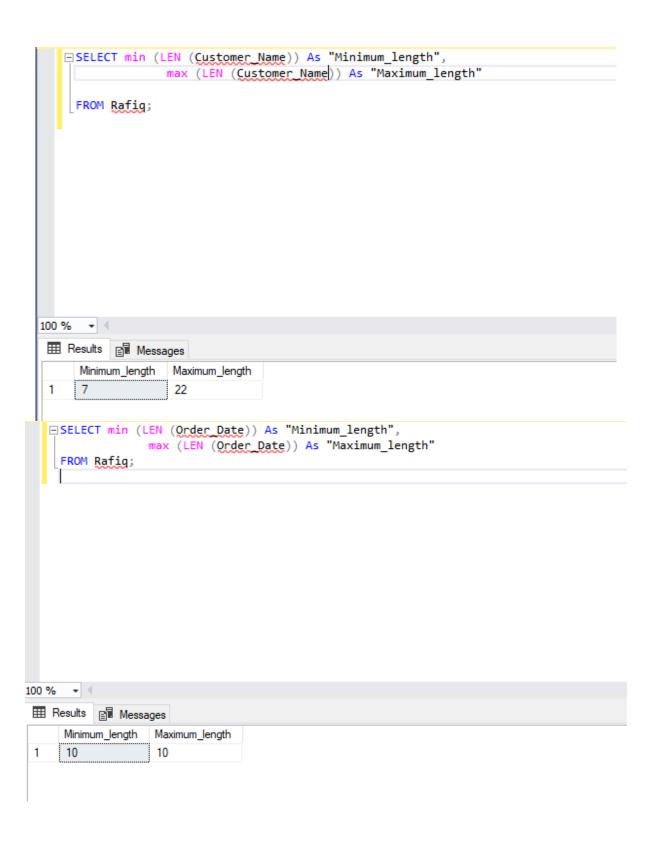


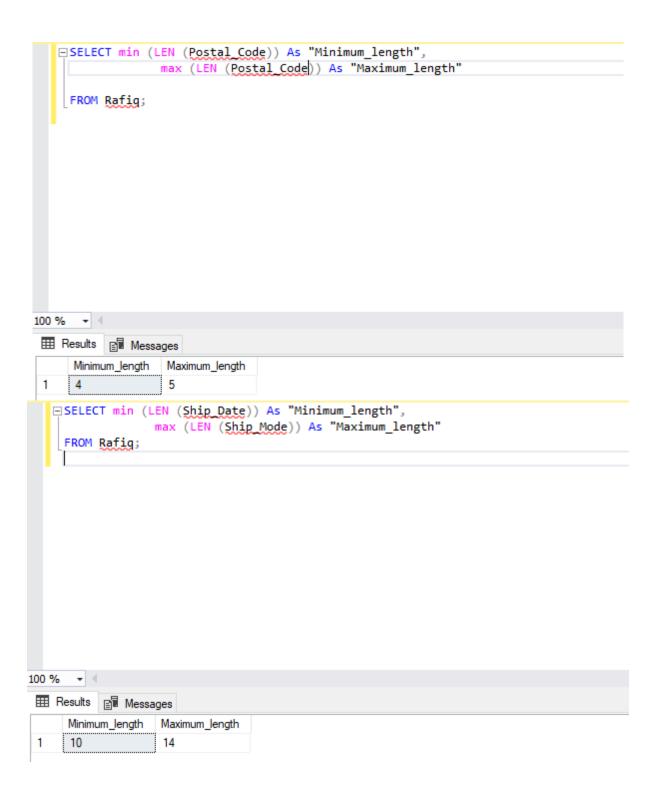
1. Length Distribution –

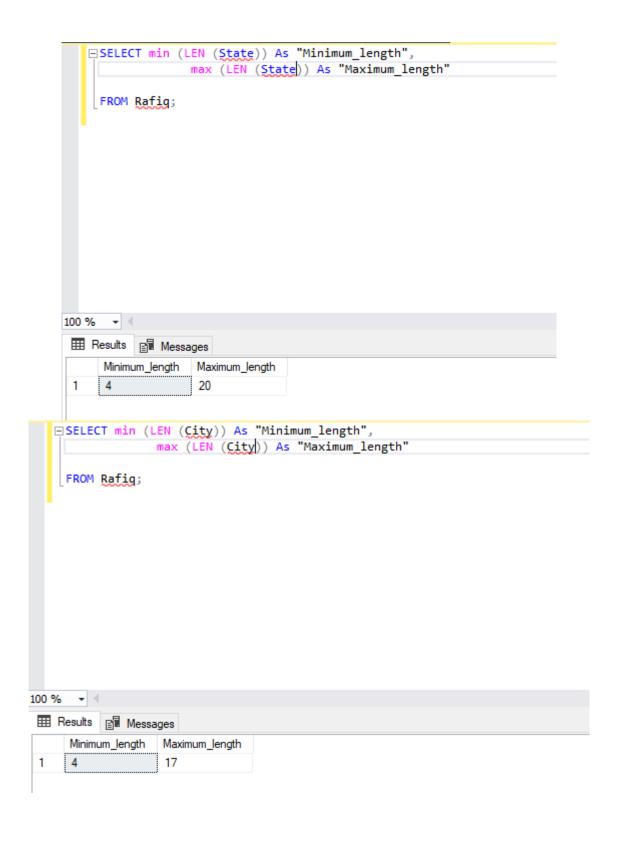
Length Distribution – Find the minimum and maximum lengths of string values in the selected columns. Show, also, the minimum and maximum string values. This profile helps you identify problems in your data, such as values that are valid. For example, you profile a column of Malaysia states codes that should be two characters and discover values longer than two characters. (3 points)

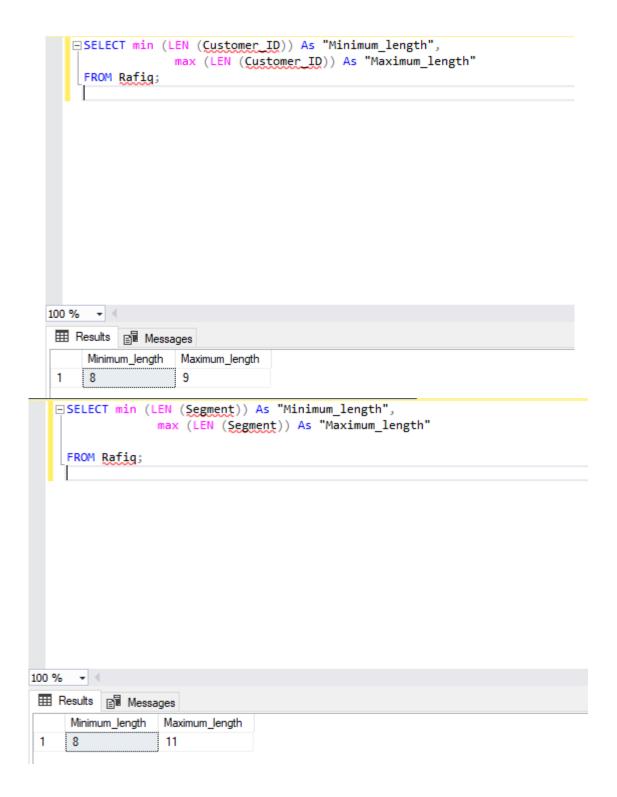
*This method is similar to value distribution, except that it only reports the lengths of string values in selected columns and the percentage of rows in the table corresponding to each value. This can assist us in filtering out invalid data.











1. **Null Ratio** – Find the number of null values in the columns and reports the percentage of null values in the columns. This profile helps you identify problems in your data, such as an unexpectedly high ratio of null values in a column. For example, you profile a postcode column and discover an unacceptably high percentage of missing codes. (*3 points*)

*** This method reports the percentage of null values available in a specific column. If we have a high ratio of null values in our dataset, it will affect the decision making. We have to

make sure the null ratio in our dataset is within the acceptable limit. The less null values we have the better.

Our column :Order_Date,Ship_Mode,Ship_Date,Customer_ID, Product_ID,Category, Postal_code,Sales,Quantity,Discount,Profit

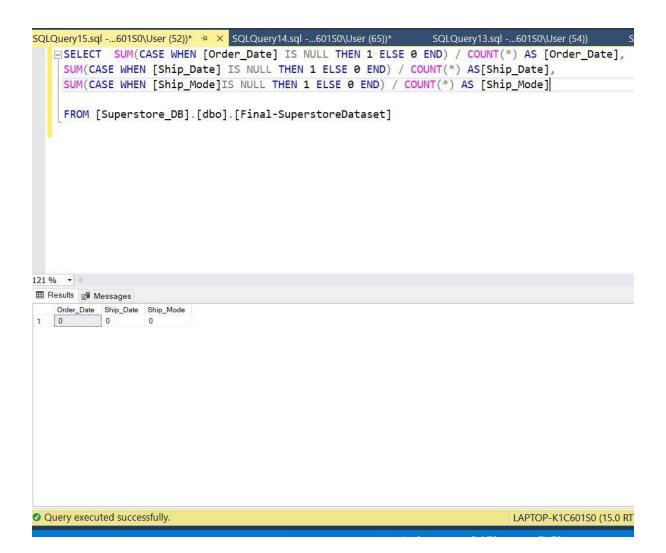
SQL Statement Template:

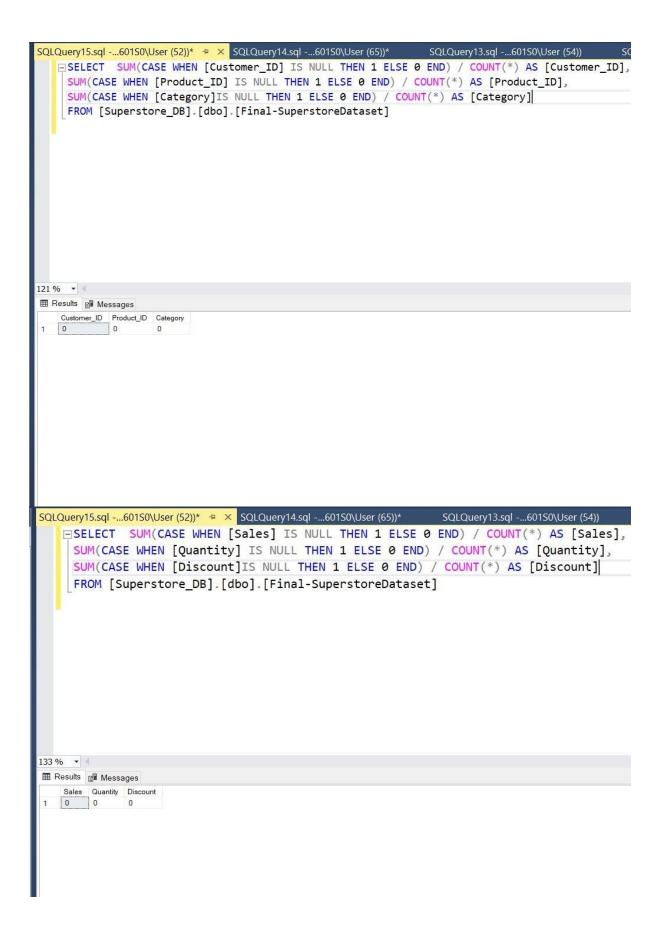
SELECT SUM(CASE WHEN Column1 IS NULL THEN 1 ELSE 0 END) / COUNT(*) AS Column1,

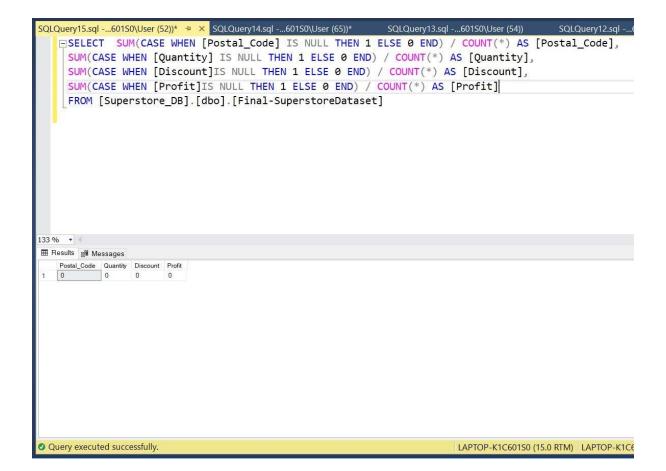
SUM(CASE WHEN Column2 IS NULL THEN 1 ELSE 0 END) / COUNT(*) AS Column2,

SUM(CASE WHEN Column3 IS NULL THEN 1 ELSE 0 END) / COUNT(*) AS Column3

FROM table;





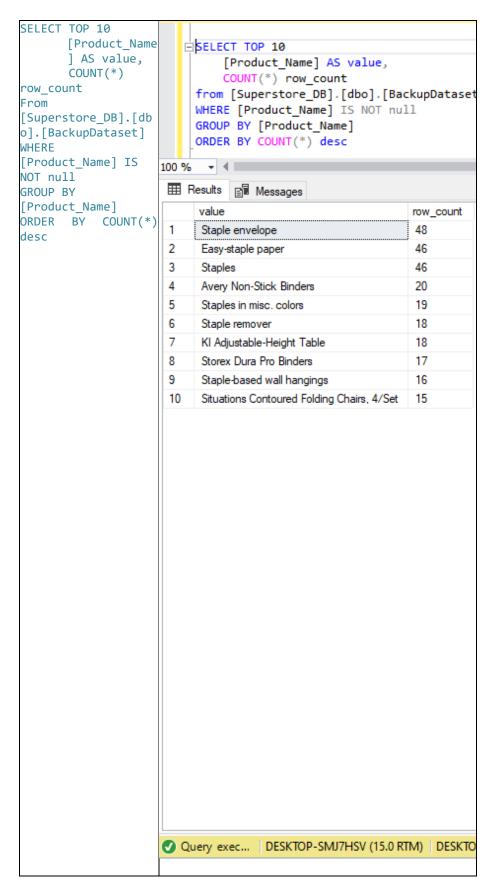


1. Use your own creativity to produce ONE (1) SQL statement and the corresponding English statement to describe the query. You must use any element indicated below. You may also

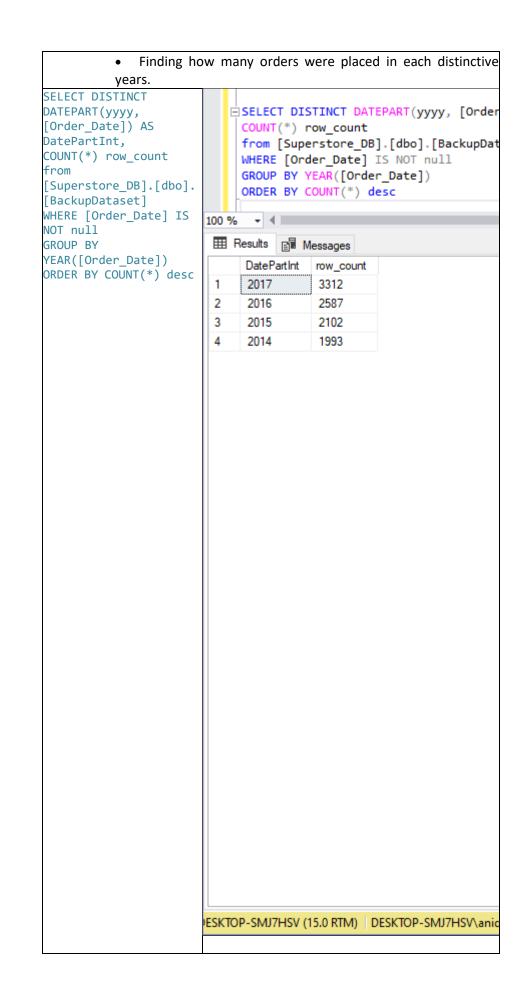
include other additional functions.

(3 points)

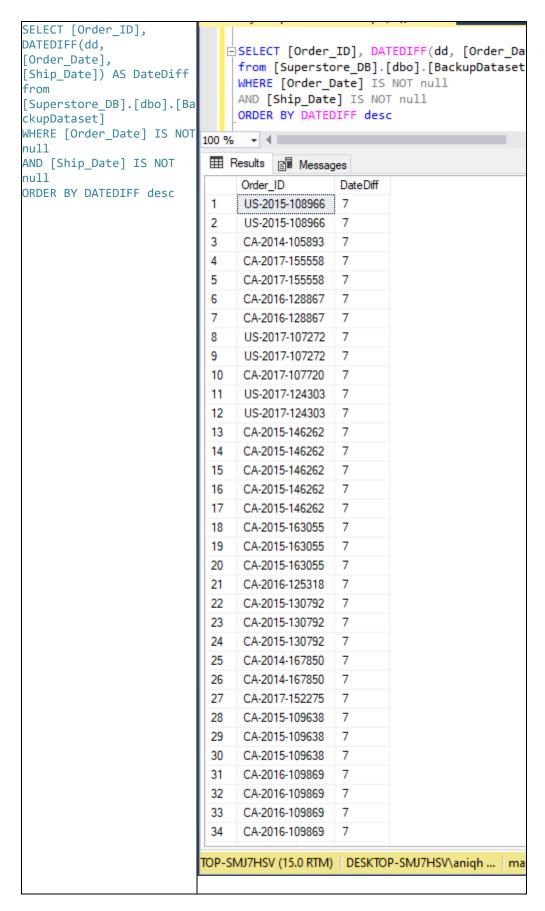
- a. **TOP**() function. **TOP**() function returns the top specified number of rows. Use the function in conjunction with the **ORDER BY** clause.
 - Finding the top 10 most bought items.



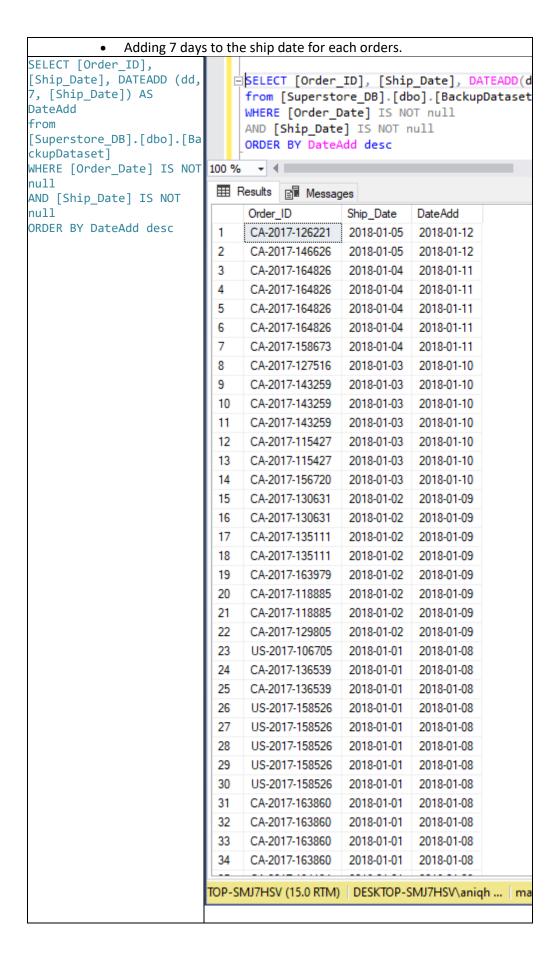
a. ${\bf DATEPART}()$ function. ${\bf DATEPART}()$ function returns individual parts of a date.



- a. **DATEDIFF**(). **DATEDIFF**() function returns the date difference in days, months, year etc.
 - Finding the difference between shipping date and the time the order was placed (order date).



a. **DATEADD**(). **DATEADD**() adds or subtracts a specified time interval from a date.

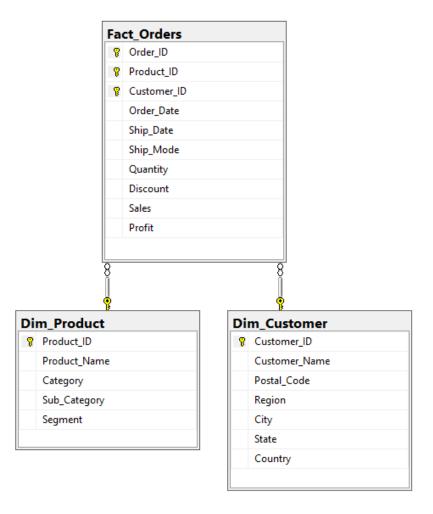


1. DATA WAREHOUSE BUS MATRIX

Use the BUS MATRIX (Must be accompanied with a brief explanation of your proposal),

1. To propose a business process that can be formed from the data set sourced for the data warehouse to be created and to explain the business process. (3 points)

		a arra to explain t		le perme	
Business Process	Fact Table	Granularity	Facts	Dim Customer	Dim Product
Busiliess Process	ract rable	Granularity	Facts Order ID,	Dim_customer	Dilli_Product
		One row per	Order_Date,		
Order Creation	Fact_Orders	new order entry	Customer_ID	X	x
		One row per	Order ID,		
Superstore Sales	Fact_Orders	Order Sale	Sales	X	X
		One row per			
		Order to be	Order ID,		
Order Fulfillment	Fact_Orders	shipped out	Ship_Date	×	x



1. To identify and explain the proposed fact and dimensions tables, their granularities, and facts. (*4 points*)

The measurement of the business process is contained in the fact table, as is FK for the dimensional tables. On the other hand, dimensional tables contain measurement attributes stored in fact tables. As you can see, the following are the facts that define our company. (Order_ID, Product_ID, Customer_ID, Order_Date, Ship_Date, Ship_Mode, Quantity, Discount, Sales, and Profit).

Alternatively, all of the FK are contain dimensional tables. It also contains properties of measurement in dimensional tables. The primary keys to our fact table are Dim Product and Dim Customers. So, the fact table will contain to our all-primary key. Product_ID, Product_Name, Category, Sub Category, Segment, Prostal Code, Region, City, State, and Country all are contain to our FK.

Importing data:

Dimensional tables must load first because they contain primary keys that are unique numbers and all of the details, whereas fact tables contain foreign keys. As a result, we cannot load the fact table without a foreign key.