STRUCTURED QUERY LANGUAGE (SQL) PROJECT

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Table of Contents

1. Introduction

1.1 Background

1.2 Objectives

1.3 Scope

2. Step 1: Database Setup & Data Import

3. Step 2: Data Cleaning

4. Step 3: Exploratory Data Analysis (EDA)

5. Step 4: Business Analysis (Project Questions i–x)

6. Step 5: Reporting & Insights

7. Appendix: Full One-Click Project Script

1.Introduction

1.1 Background

In the fiercely competitive retail sector, companies depend on precise data analysis to assess performance, predict trends, and comprehend customer behavior. A multitude of data is produced by sales transactions, including revenue, product categories, volumes, and consumer demographics. However, before it can be turned into useful insights, this raw data frequently needs to be cleaned and organized.

By using **SQL Server**, businesses can create a centralized sales database that not only stores data securely but also enables robust querying for business intelligence. This project demonstrates the complete workflow — from database creation and data cleaning to analysis and reporting.

1.2 Objective

The main objectives of this project are:

- **Database Setup** – Create a retail sales database in SQL Server and import raw transactional data.
- **Data Cleaning** – Identify and handle missing or invalid values to ensure data integrity.
- **Exploratory Data Analysis (EDA)** – Use SQL queries to uncover initial trends and distributions.
- **Business Analysis** – Answer specific business questions related to sales performance, customer behavior, and product categories.
- **Comprehensive Reporting** – Document the entire process and findings in a structured format suitable for professional portfolio use.

1.3 Scope

This project covers the end-to-end lifecycle of a retail sales analytics task:

- Designing the sales database schema.
- Importing CSV data into SQL Server using the **Import Flat File**.
- Performing data cleaning operations to remove null values and ensure consistency.
- Executing SQL queries to answer ten business-focused questions such as:
    - Which product categories generate the highest revenue?
    - Who are the top 5 customers based on spending?
    - What is the sales distribution across different shifts (morning, afternoon, evening)?
    - Which month in each year is the best-performing?
- Preparing a professional report of all steps, queries, and insights.

By the end of this project, the dataset evolves from **raw CSV files** to a **cleaned, structured, and analyzed sales database**, showcasing practical SQL skills applicable to real-world retail analytics. Below is the step-by-step process of data setup and import.

2 Step 1: Database Setup & Data Import

1.a Creating the Database

The first step in this project is to set up the retail sales database where all the transactional data will be stored. Using SQL Server Management Studio (SSMS), we begin by creating a dedicated database using syntax named retail_sales.

```
-- Create database

CREATE DATABASE retail_sales;
```

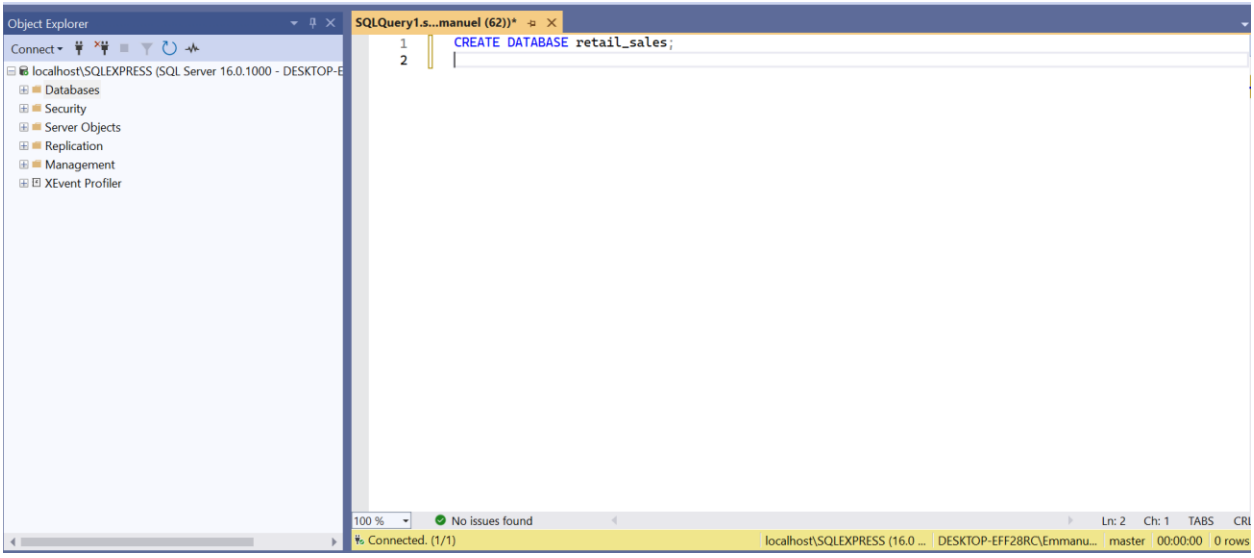


Fig i

In Fig i, this ensures that all subsequent operations — table creation, data import, and analysis are organized within a single structured database environment.

1.b Importing the Data

The dataset provided is in CSV format. In SQL Server Management Studio, the Import Flat File is used to bring the data. The process is as follows:

- Right-click the database (retail_sales) in Object Explorer.
- Select Tasks > Import Flat File.
- Browse and select the CSV file (e.g., *SQL - retail_sales TB.csv*).
- Preview the data and ensure columns are mapped correctly to the schema.
- Complete the wizard to import the data.

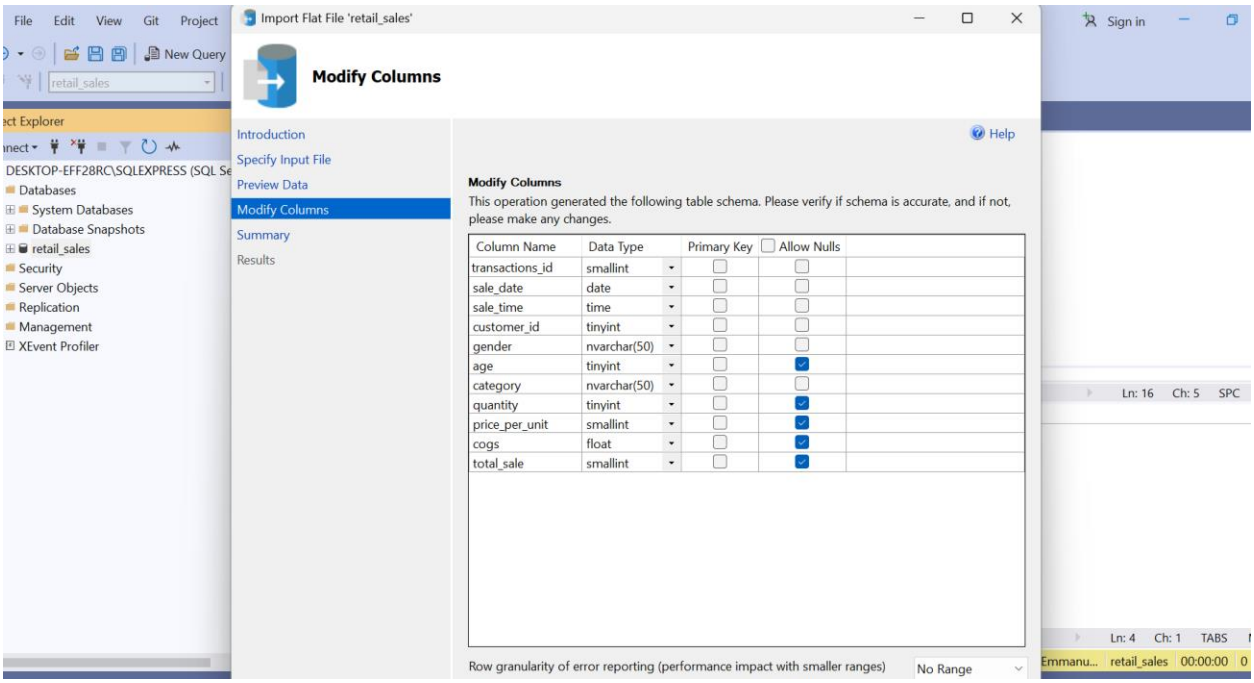


Fig ii

In Fig ii, transaction_id was checked as the primary key, no allow null. While allow null was checked for other columns.

1.c Verifying Data Import

Once the data is imported, it is essential to verify that the records have been loaded successfully. This is done by querying the first few rows of the table.

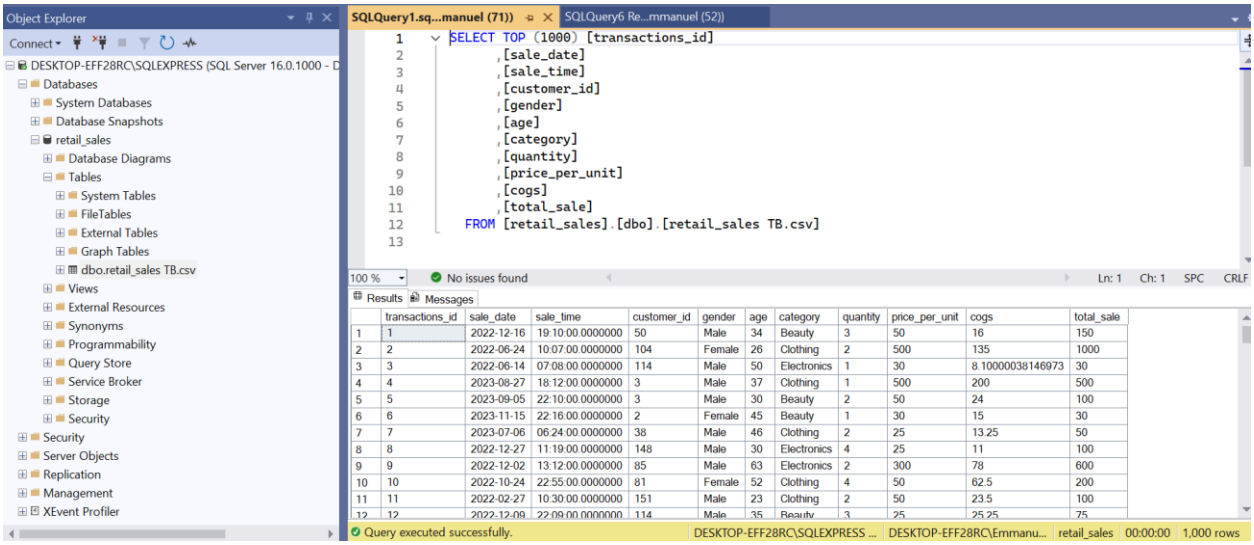


Fig iii

At the retail_sales TB.csv table, right click, Select Top 1000 rows.

Fig iii shows a representation showing the dataset. At this point, the database is fully set up and populated with raw sales data. The next step will involve Data Cleaning (Step 2), where we will identify and remove null or invalid values to prepare the dataset for analysis.

3 Step 2: Data Cleaning

Once the database was successfully created and populated with the raw retail sales data, the next step involved cleaning the dataset to ensure accuracy and reliability of insights.

2.a Identifying Null Values

To check for missing or null values across critical columns such as age, quantity, price_per_unit, cogs, and total_sale, the following SQL query was executed:

```
SELECT
    SUM (CASE WHEN age IS NULL THEN 1 ELSE 0 END) AS null_age,
    SUM (CASE WHEN quantity IS NULL THEN 1 ELSE 0 END) AS null_quantity,
    SUM (CASE WHEN price_per_unit IS NULL THEN 1 ELSE 0 END) AS null_price,
    SUM (CASE WHEN cogs IS NULL THEN 1 ELSE 0 END) AS null_cogs,
    SUM (CASE WHEN total_sale IS NULL THEN 1 ELSE 0 END) AS null_total_sale
FROM retail_sales TB.csv;
```

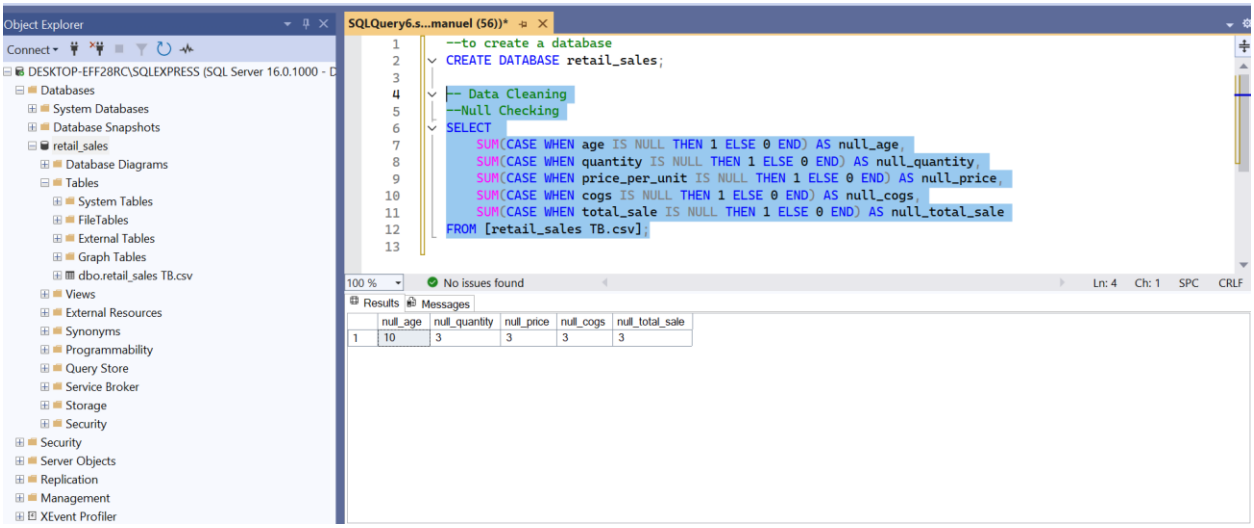


Fig iv

In Fig iv, this query provided a quick summary of the number of null values in each column.

2.b Removing Records with Missing Values

Since null values can distort sales totals and averages, rows containing them were removed using:

```
DELETE FROM [retail_sales TB.csv];
```

```
WHERE age IS NULL
   OR quantity IS NULL
   OR price_per_unit IS NULL
   OR cogs IS NULL
   OR total_sale IS NULL;
```

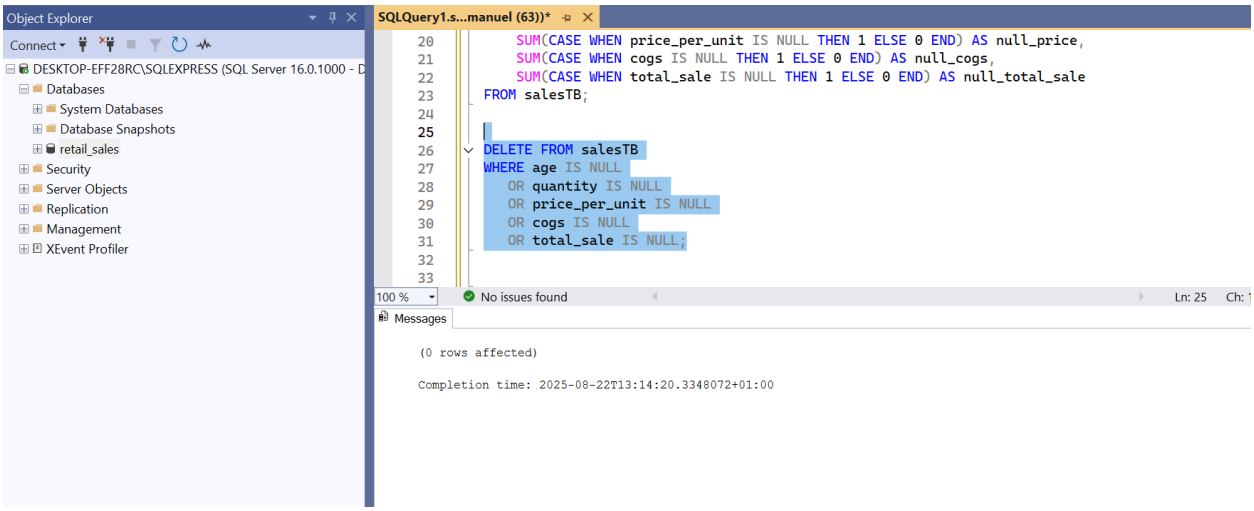


Fig v

In Fig v, the result returned **0**, confirming that the dataset was now clean and ready for analysis.

4 Step 3: Exploratory Data Analysis (EDA)

With the cleaned dataset in place, the next step was to perform **exploratory data analysis (EDA)**. The goal of this phase was to better understand the structure of the sales data, identify patterns, and prepare for deeper business analysis.

3.a Checking Total Records

The first step was to confirm the number of transactions available after cleaning:

SELECT COUNT (*) AS total_transactions

FROM [retail_sales TB.csv];

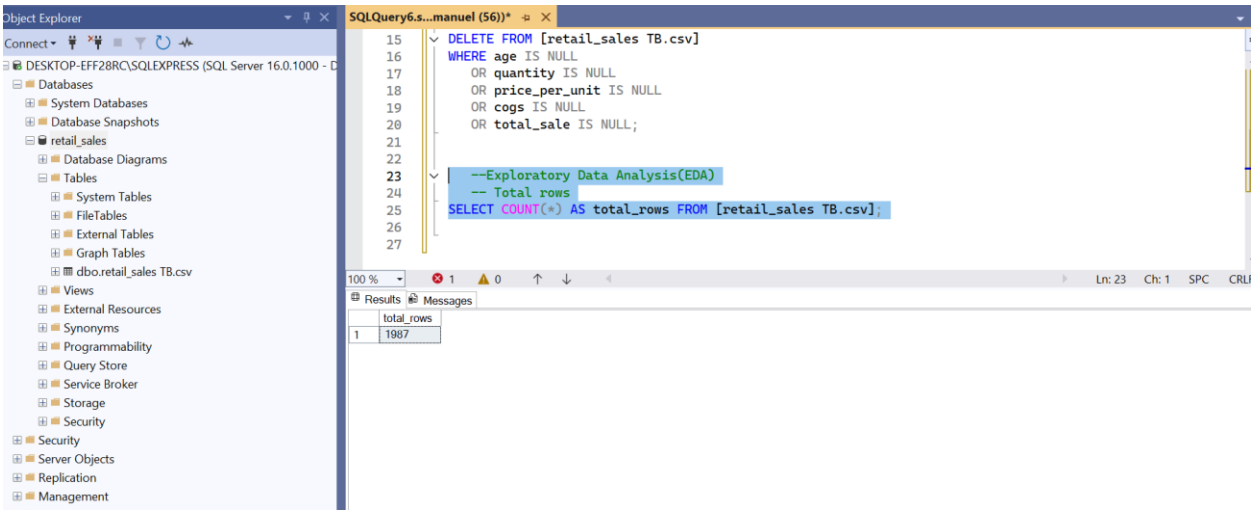


Fig vi

In Fig vi, this query returned the total count of rows in the dataset, representing all valid transactions.

3.b Date Range of Sales

To understand the time period covered by the dataset, the following query was run:

SELECT

MIN (sale_date) AS start_date,

MAX (sale_date) AS end_date

FROM [retail_sales TB.csv];

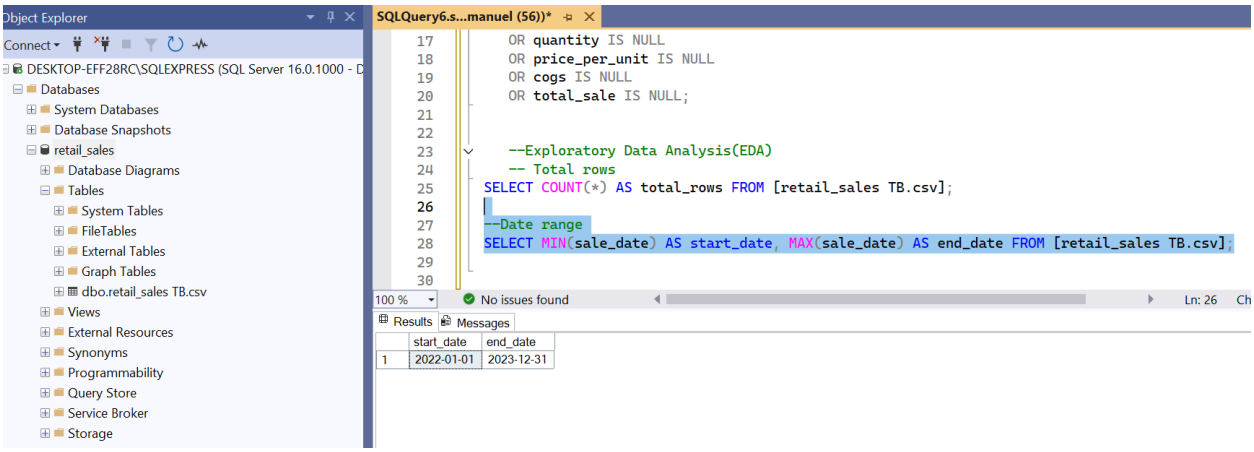


Fig vii

In Fig vii, this revealed the earliest and latest sales dates captured in the database.

3.c Distinct Product Categories

To examine the variety of products sold, distinct categories were identified:

SELECT DISTINCT category

FROM [retail_sales TB.csv];

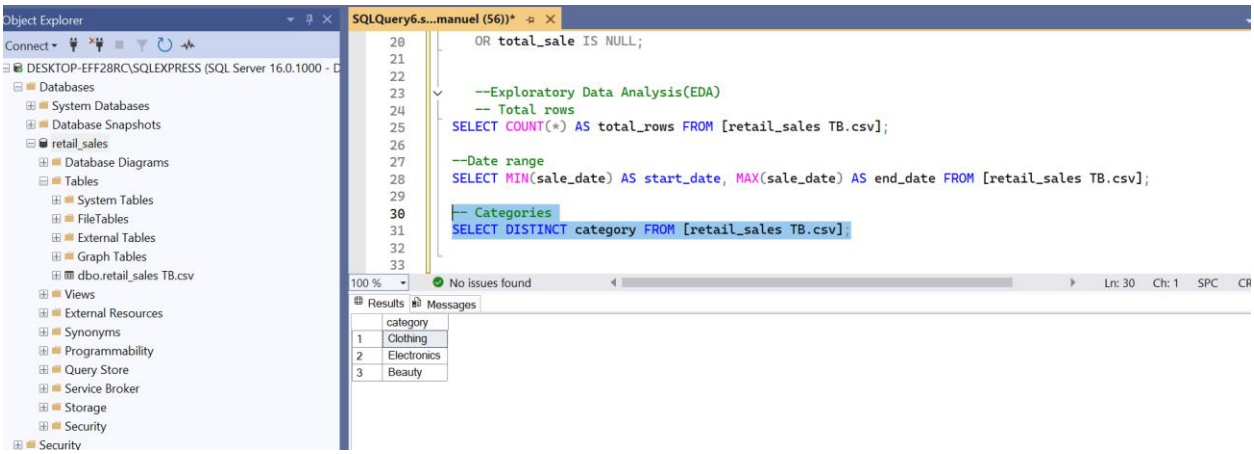


Fig viii

In Fig viii, this provided a list of all product categories available in the dataset (e.g., Clothing, Beauty, Electronics, etc.).

3.d Gender Distribution of Customers

Customer demographics were checked by analyzing gender distribution:

SELECT gender, COUNT (*) AS transactions

FROM [retail_sales TB.csv];

GROUP BY gender;

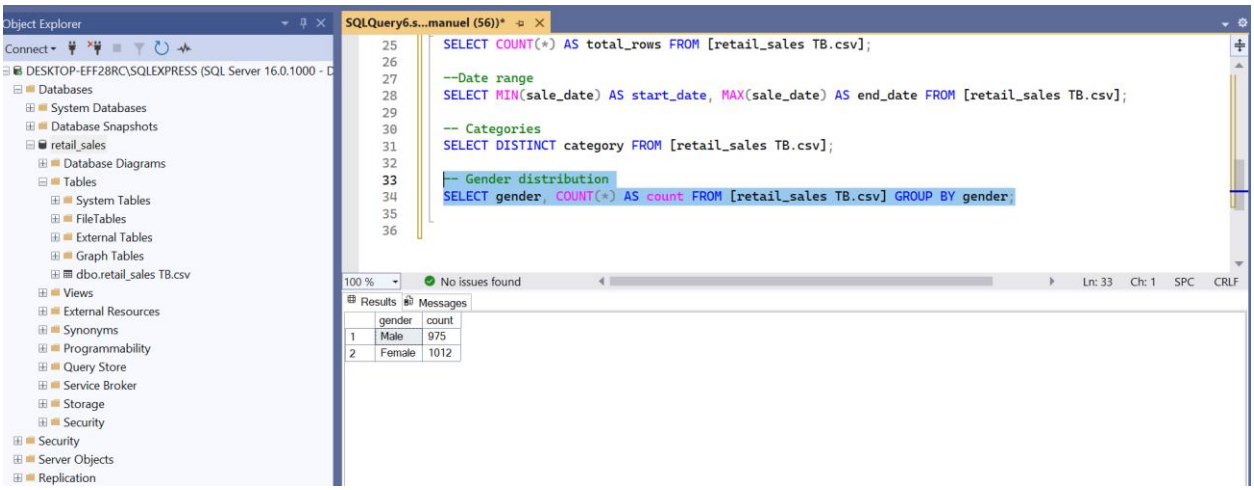


Fig ix

In Fig ix, this showed how many transactions were made by male vs. female customers.

3.e Age Range of Customers

The age profile of customers was explored using:

SELECT

MIN (age) AS youngest_customer,

MAX (age) AS oldest_customer,

AVG (age) AS avg_customer_age

FROM [retail_sales TB.csv];

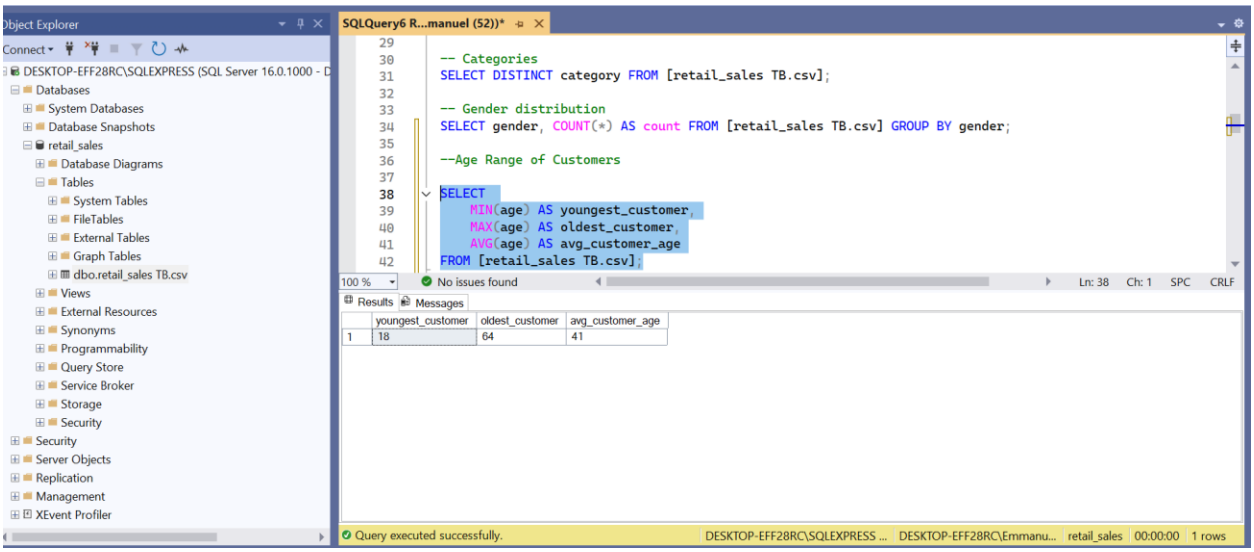


Fig x

In Fig x, this gave insights into the customer base in terms of age.

5 Step 4: Business Analysis (Project Questions)

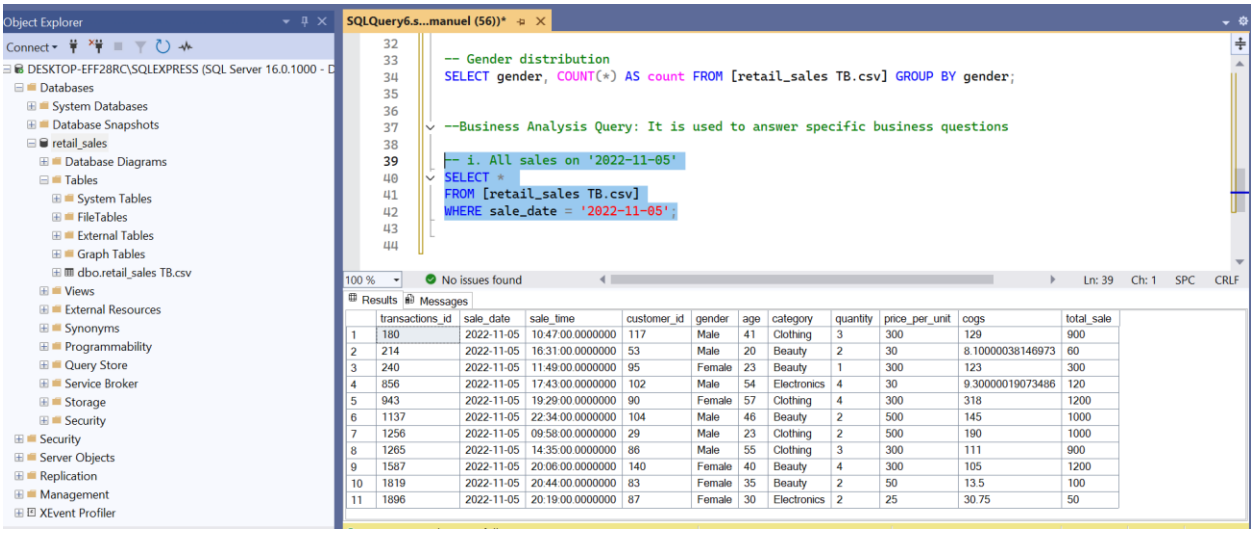
In this phase, SQL was used to answer specific business questions based on the cleaned retail sales dataset.

4.a Retrieve all columns for sales made on *2022-11-05*

SELECT *

FROM [retail_sales TB.csv]

WHERE sale_date = '2022-11-05';



	transactions_id	sale_date	sale_time	customer_id	gender	age	category	quantity	price_per_unit	cogs	total_sale
1	180	2022-11-05	10:47:00.0000000	117	Male	41	Clothing	3	300	129	900
2	214	2022-11-05	16:31:00.0000000	53	Male	20	Beauty	2	30	8.10000038146973	60
3	240	2022-11-05	11:49:00.0000000	95	Female	23	Beauty	1	300	123	300
4	856	2022-11-05	17:43:00.0000000	102	Male	54	Electronics	4	30	9.30000019073486	120
5	943	2022-11-05	19:29:00.0000000	90	Female	57	Clothing	4	300	318	1200
6	1137	2022-11-05	22:34:00.0000000	104	Male	46	Beauty	2	500	145	1000
7	1256	2022-11-05	09:58:00.0000000	29	Male	23	Clothing	2	500	190	1000
8	1265	2022-11-05	14:35:00.0000000	86	Male	55	Clothing	3	300	111	900
9	1587	2022-11-05	20:06:00.0000000	140	Female	40	Beauty	4	300	105	1200
10	1819	2022-11-05	20:44:00.0000000	83	Female	35	Beauty	2	50	13.5	100
11	1896	2022-11-05	20:19:00.0000000	87	Female	30	Electronics	2	25	30.75	50

Fig xi

In Fig xi, this returns all transactions recorded on November 5, 2022.

4.b Retrieve transactions where the category is *Clothing* and quantity sold is more than 4 in November 2022

```
SELECT *
FROM [retail_sales TB.csv]
WHERE category = 'Clothing'
  AND quantity > 4
  AND sale_date BETWEEN '2022-11-01' AND '2022-11-30';
```

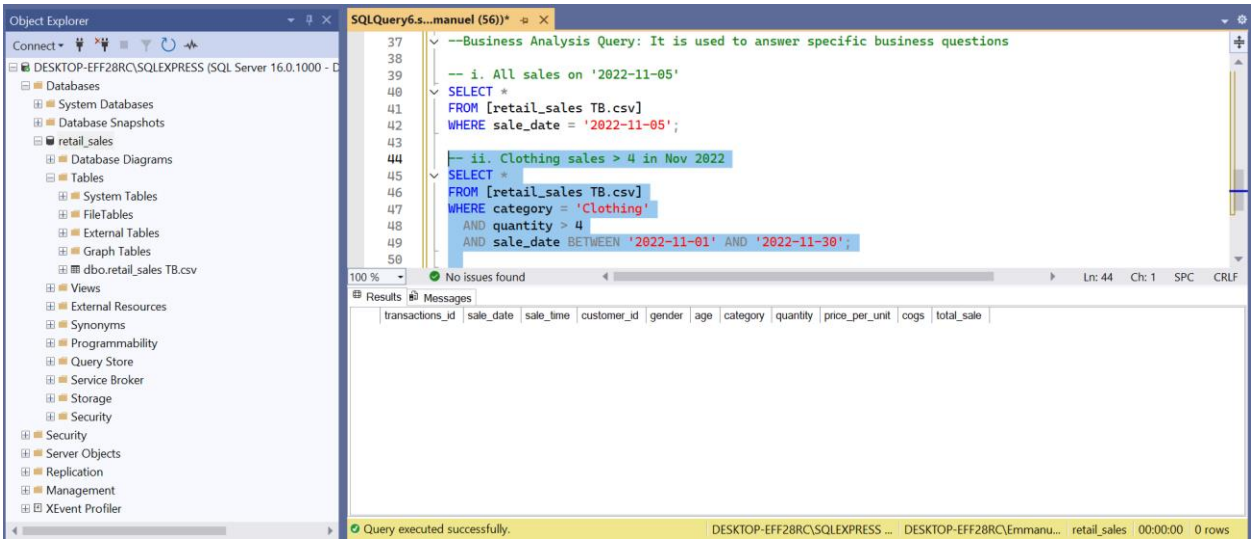


Fig xii

In Fig xii, it filters transactions for the Clothing category with bulk orders in Nov-2022.

4.c Calculate the total sales for each category

SELECT category, SUM (total_sale) AS total_sales

FROM [retail_sales TB.csv]

GROUP BY category;

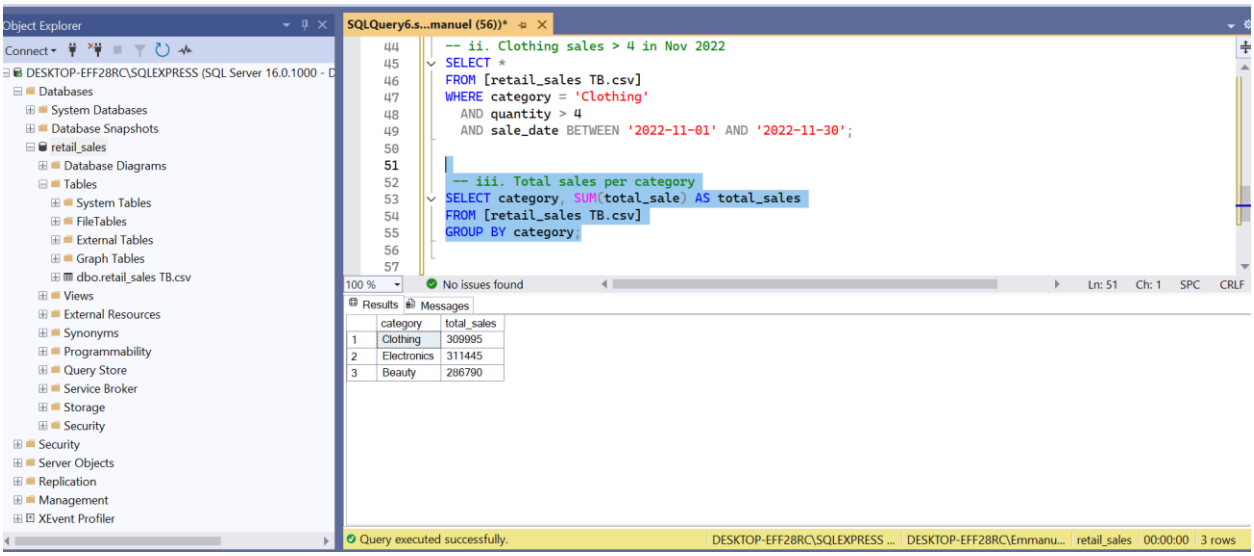


Fig xiii

In Fig xiii, it provides sales contribution of each category.

4.d Find the average age of customers who purchased items from the *Beauty* category

SELECT AVG (age) AS avg_age_beauty_customers

FROM [retail_sales TB.csv]

WHERE category = 'Beauty';

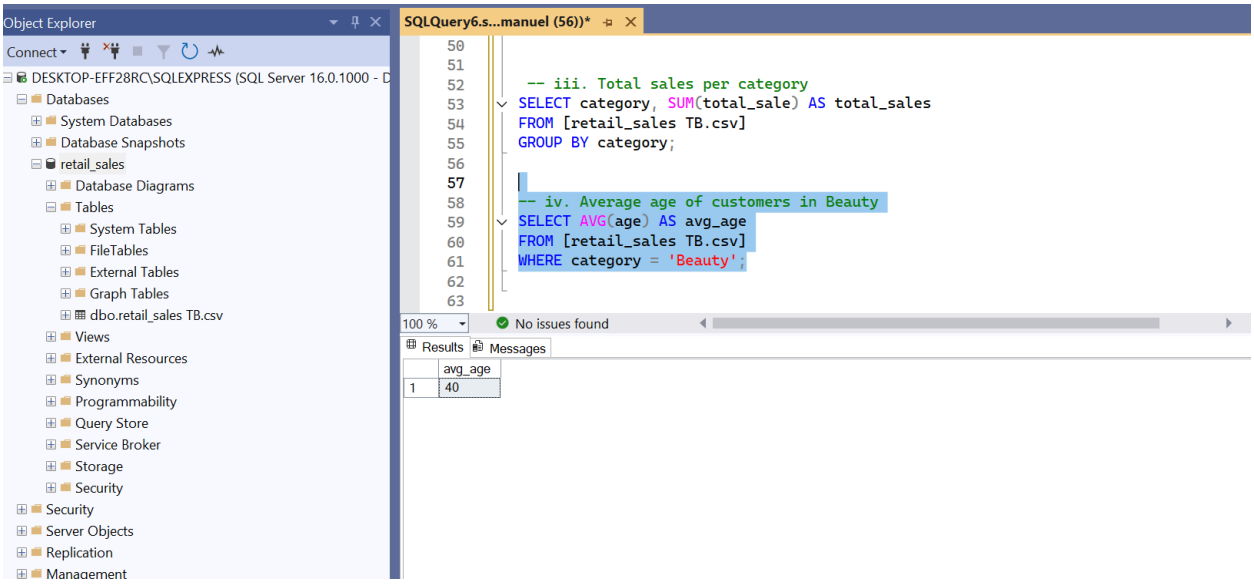


Fig xiv

In Fig xiv, it shows demographic insights for Beauty category buyers.

4.e Find all transactions where total_sale is greater than 1000

```
SELECT *

FROM [retail_sales TB.csv]

WHERE total_sale > 1000;
```

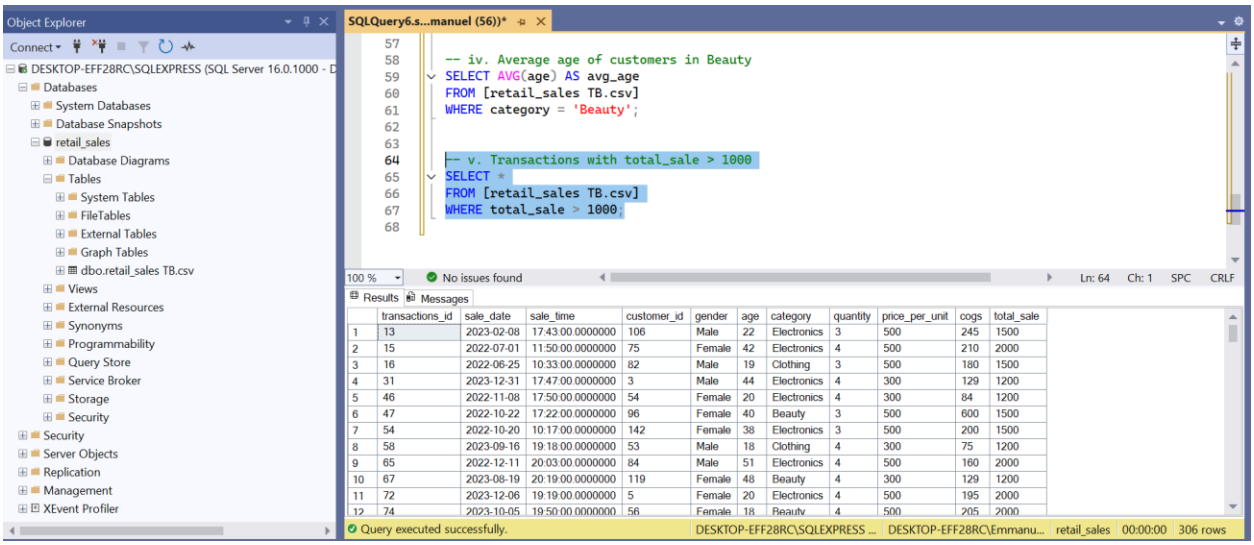


Fig xv

In Fig xv, it identifies high-value transactions.

4.f Find the total number of transactions made by each gender in each category

SELECT gender, category, COUNT (transaction_id) AS total_transactions

FROM [retail_sales TB.csv]

GROUP BY gender, category

ORDER BY category, gender;

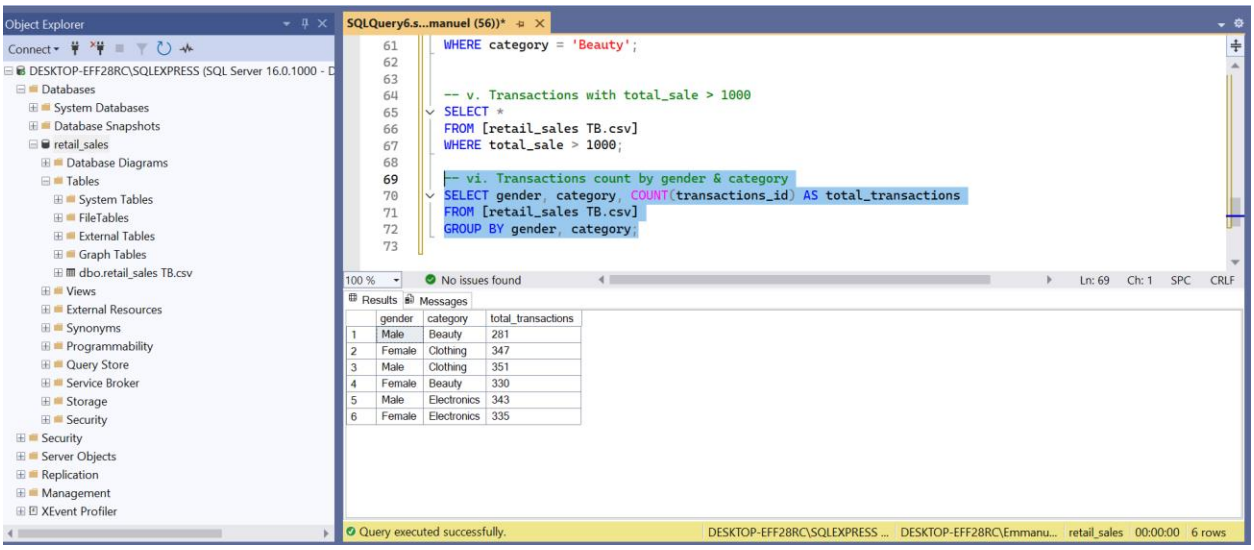


Fig xvi

In Fig xvi, it breaks down transactions by gender within each category.

4.g Calculate the average sale for each month and find the best-selling month in each year

```
-- Monthly average sales
SELECT
    YEAR (sale_date) AS year,
    MONTH (sale_date) AS month,
    AVG (total_sale) AS avg_monthly_sale
FROM [retail_sales TB.csv]
GROUP BY YEAR (sale_date), MONTH (sale_date)
ORDER BY year, month;
```

4.h Best-selling month per year

```
SELECT year, month, avg_monthly_sale
```

```
FROM (
    SELECT
        YEAR (sale_date) AS year,
        MONTH (sale_date) AS month,
        AVG (total_sale) AS avg_monthly_sale,
        RANK () OVER (PARTITION BY YEAR (sale_date) ORDER BY AVG (total_sale) DESC) AS rnk
    FROM [retail_sales TB.csv]
    GROUP BY YEAR (sale_date), MONTH (sale_date)
) ranked
WHERE rnk = 1;
```

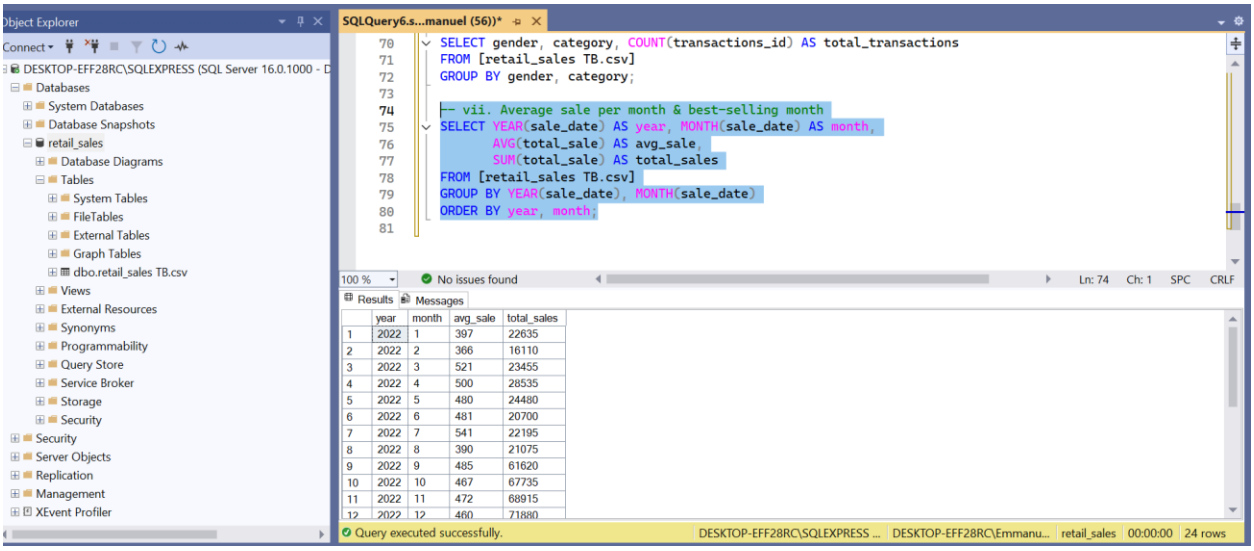


Fig xvii

In Fig xvii, first query shows monthly trends, second highlights the best month each year.

4.i Find the top 5 customers based on highest total sales

SELECT TOP 5 customer_id, SUM (total_sale) AS total_sales

FROM [retail_sales TB.csv]

GROUP BY customer_id

ORDER BY total_sales DESC;

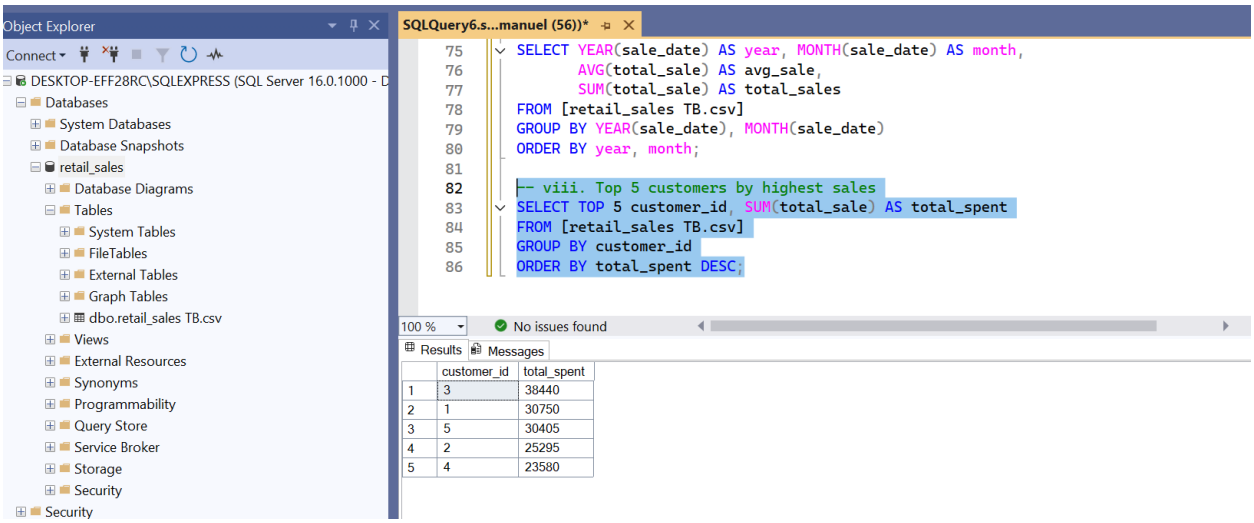


Fig xviii

In Fig xviii, it Identifies the most valuable customers.

4.j Find the number of unique customers who purchased items from each category

SELECT category, COUNT (DISTINCT customer_id) AS unique_customers

FROM [retail_sales TB.csv]

GROUP BY category;

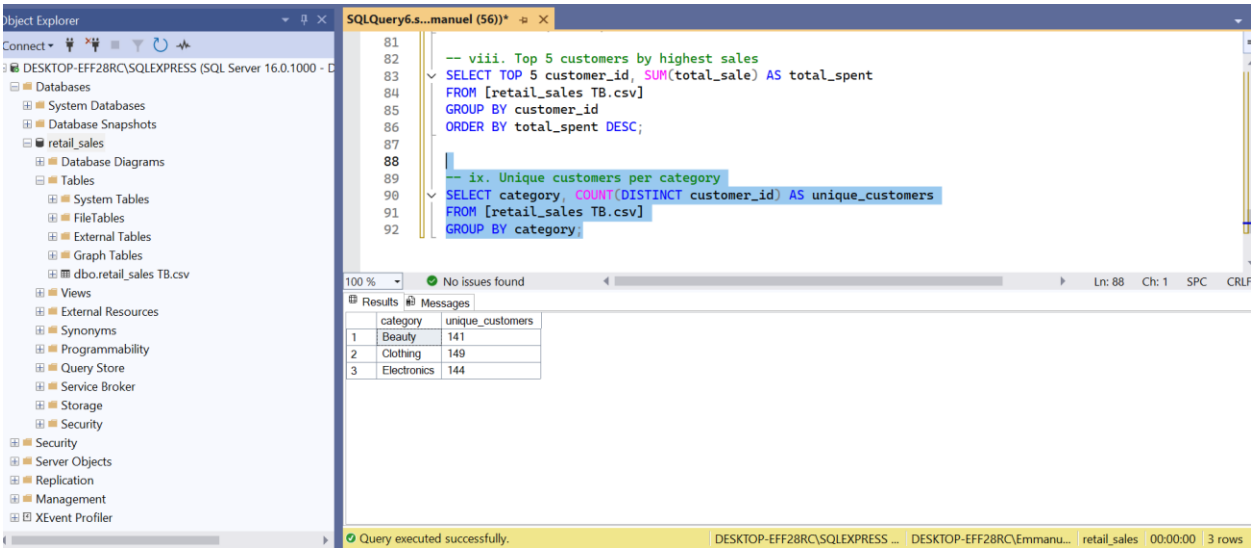


Fig xix

In Fig xix, it measures customer reach per product category.

4.k Create shifts and count number of orders per shift

```
SELECT
    CASE
        WHEN sale_time < '12:00:00' THEN 'Morning'
        WHEN sale_time BETWEEN '12:00:00' AND '17:00:00' THEN 'Afternoon'
        ELSE 'Evening'
    END AS shift,
    COUNT (transaction_id) AS total_orders
FROM [retail_sales TB.csv]
GROUP BY
    CASE
```

```
    WHEN sale_time < '12:00:00' THEN 'Morning'
    WHEN sale_time BETWEEN '12:00:00' AND '17:00:00' THEN 'Afternoon'
    ELSE 'Evening'
  END
ORDER BY total_orders DESC;
```

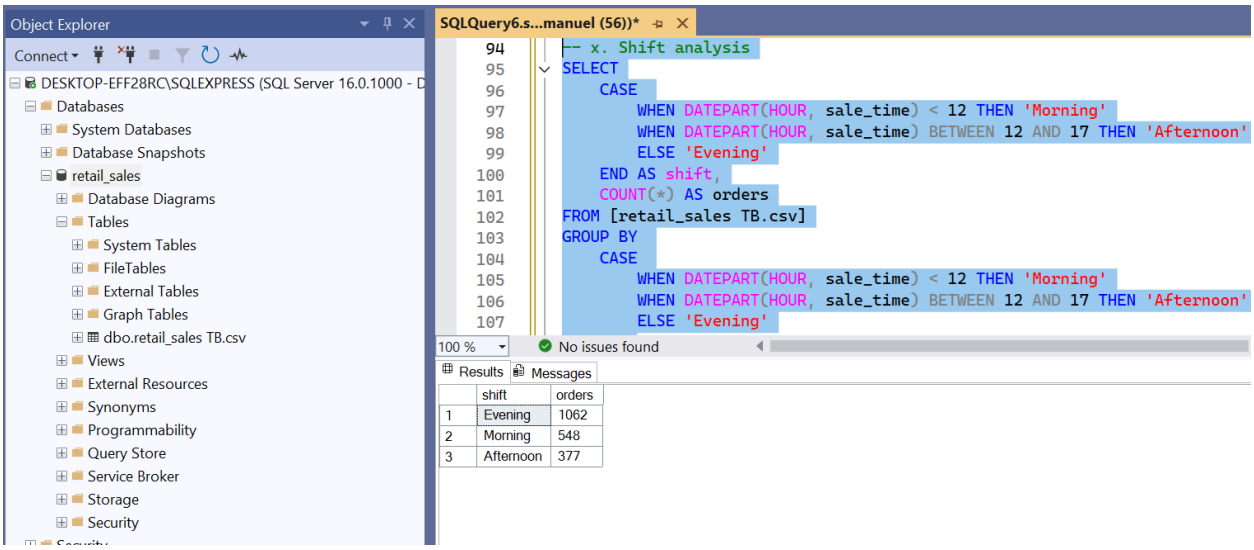


Fig xx

In Fig xx, it categorizes transactions into shifts and shows order volume per time of day.

6 Step 5: Reporting & Insights

The study yields information that directs operations, marketing, and retailing. Examples and practical suggestions are provided below;

- **Category Performance:** Priority funding for marketing and inventories should go to the categories with the highest overall sales.

• **Customer Demographics:** If younger customers are drawn to beauty, focus your campaigns on platforms that appeal to them.

• **High-Value Orders:** The majority of transactions over 1000 fall into particular categories; take into account protection plans and premium packages.

• **Seasonality:** Peak months, which are typically November through December, call for more staff, inventory, and promotions.

• **Top Customers:** Incorporate targeted incentives and loyalty benefits to increase lifetime value and retention.

• **Shift Patterns:** Adjusting staffing and in-store promotions for the store window is advised due to afternoon traffic peaks.

Explore the full project in the link below

https://drive.google.com/file/d/1BG0bSIaOiXNTV4fhXBEuFXRPExaRDaMV/view?usp=drive_link