

STRUCTURED QUERY LANGUAGE (SQL) PROJECT

BY OKOLI KOSISOCHUKWU ANASTESIA

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.

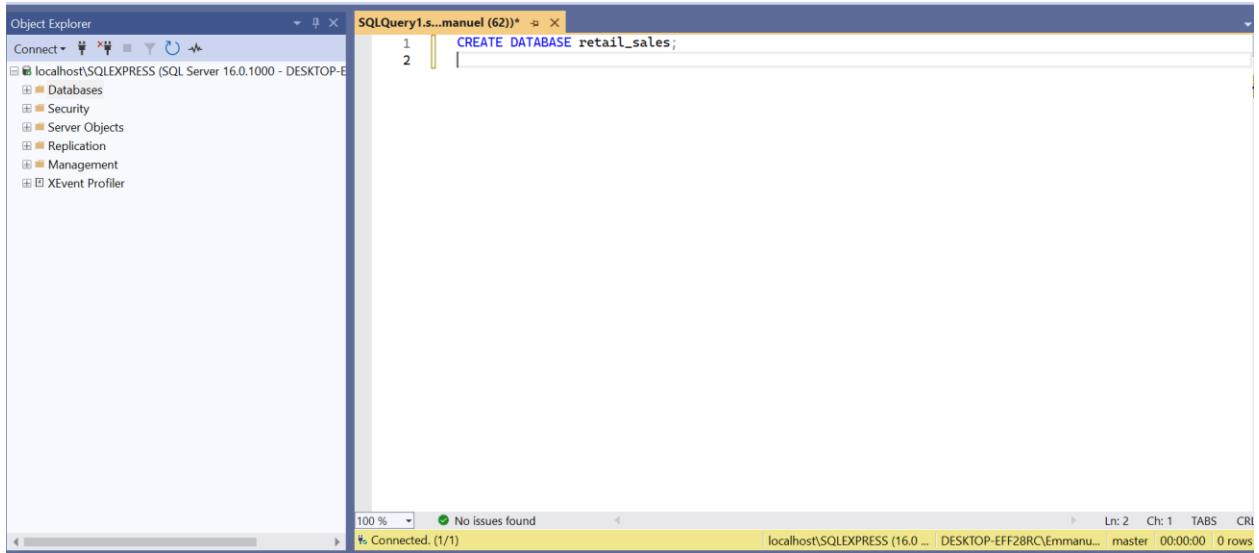
2 Step 1: Database Setup & Data Import

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;
```



The screenshot shows the SQL Server Management Studio interface. On the left, the Object Explorer pane displays a connection to 'localhost\SQLEXPRESS' (SQL Server 16.0.1000 - DESKTOP-EFF28RC) with nodes for Databases, Security, Server Objects, Replication, Management, and XEvent Profiler. The central pane is titled 'SQLQuery1.s...manuel (62)*' and contains the following SQL code:

```
CREATE DATABASE retail_sales;
```

The status bar at the bottom indicates 'Connected. (1/1)', the server as 'localhost\SQLEXPRESS (16.0 ... DESKTOP-EFF28RC\Emmanuel...', the database as 'master', and the duration as '00:00:00 | 0 rows'.

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.

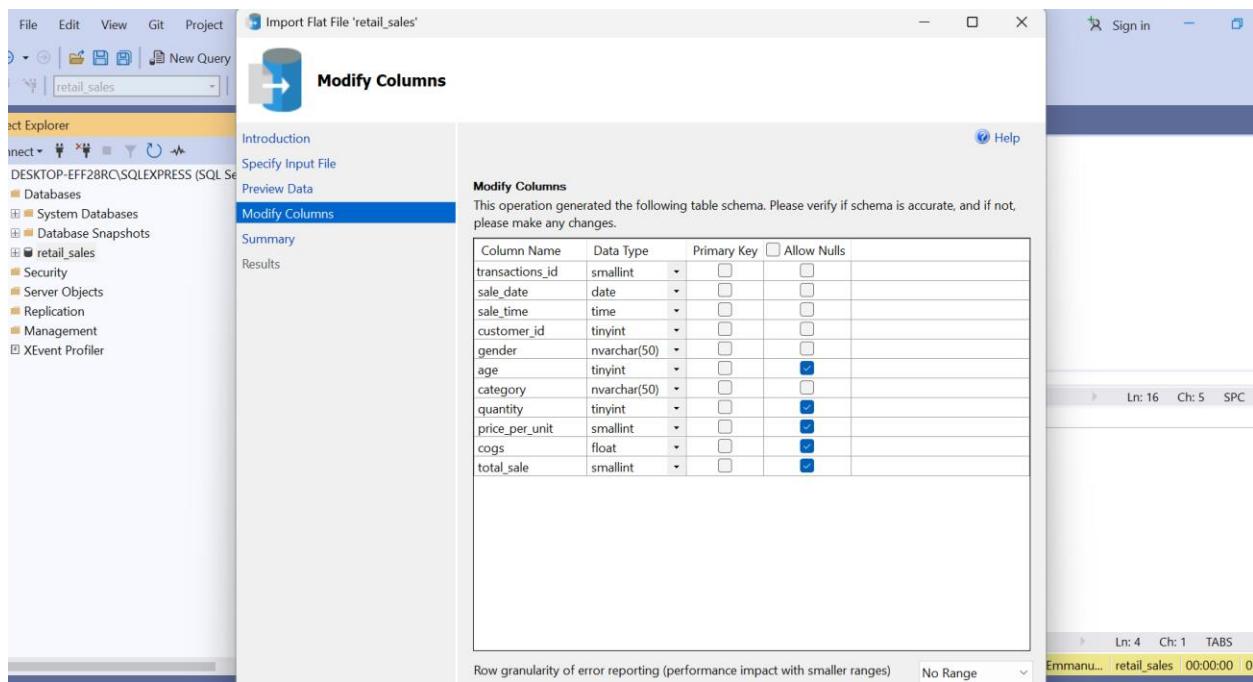


Fig ii

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.

```

1 1  SELECT TOP (1000) [transactions_id]
2 , [sale_date]
3 , [sale_time]
4 , [customer_id]
5 , [gender]
6 , [age]
7 , [category]
8 , [quantity]
9 , [price_per_unit]
10 , [cogs]
11 , [total_sale]
12 FROM [retail_sales].[dbo].[retail_sales TB.csv]
13

```

Results

transactions_id	sale_date	sale_time	customer_id	gender	age	category	quantity	price_per_unit	cogs	total_sale
1	2022-12-16	19:10:00.0000000	50	Male	34	Beauty	3	50	16	150
2	2022-06-24	10:07:00.0000000	104	Female	26	Clothing	2	500	135	1000
3	2022-06-14	07:08:00.0000000	114	Male	50	Electronics	1	30	8.10000038146973	30
4	2023-08-27	18:12:00.0000000	3	Male	37	Clothing	1	500	200	500
5	2023-09-05	22:10:00.0000000	3	Male	30	Beauty	2	50	24	100
6	2023-11-15	22:16:00.0000000	2	Female	45	Beauty	1	30	15	30
7	2023-07-06	06:24:00.0000000	38	Male	46	Clothing	2	25	13.25	50
8	2022-12-27	11:19:00.0000000	148	Male	30	Electronics	4	25	11	100
9	2022-12-02	13:12:00.0000000	85	Male	63	Electronics	2	300	78	600
10	2022-10-24	22:55:00.0000000	81	Female	52	Clothing	4	50	62.5	200
11	2022-02-27	10:30:00.0000000	151	Male	23	Clothing	2	50	23.5	100
12	2022-12-09	22:09:00.0000000	114	Male	35	Beauty	3	25	25.25	75

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;

```

The screenshot shows the SSMS interface with the Object Explorer on the left and a query results window on the right. The query results window displays the following data:

	null_age	null_quantity	null_price	null_cogs	null_total_sale
1	10	3	3	3	3

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;

```

  20   SUM(CASE WHEN price_per_unit IS NULL THEN 1 ELSE 0 END) AS null_price,
  21   SUM(CASE WHEN cogs IS NULL THEN 1 ELSE 0 END) AS null_cogs,
  22   SUM(CASE WHEN total_sale IS NULL THEN 1 ELSE 0 END) AS null_total_sale
  23
  24
  25   FROM salesTB;
  26
  27   DELETE FROM salesTB
  28   WHERE age IS NULL
  29   OR quantity IS NULL
  30   OR price_per_unit IS NULL
  31   OR cogs IS NULL
  32   OR total_sale IS NULL;
  33
  
```

(0 rows affected)

Completion time: 2025-08-22T13:14:20.3348072+01:00

Fig v

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

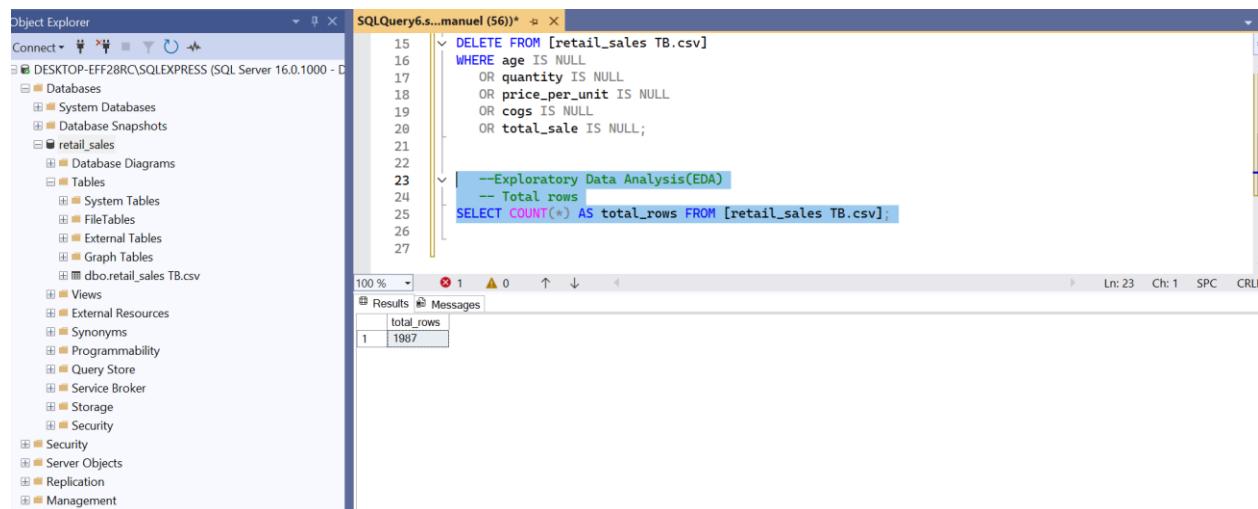
4 Step 3: Exploratory Data Analysis (EDA)

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];
```



The screenshot shows the SSMS interface. On the left is the Object Explorer pane, which lists the database structure including databases, tables, and security objects. The central area is the SQL Query window titled 'SQLQuery6.s...manuel (56)*'. It contains the following code:

```
15  |  ✓ DELETE FROM [retail_sales TB.csv]  
16  |  WHERE age IS NULL  
17  |  OR quantity IS NULL  
18  |  OR price_per_unit IS NULL  
19  |  OR cogs IS NULL  
20  |  OR total_sale IS NULL;  
21  |  
22  |  ✓ --Exploratory Data Analysis(EDA)  
23  |  --- Total rows  
24  |  
25  |  SELECT COUNT(*) AS total_rows FROM [retail_sales TB.csv];  
26  |  
27  |
```

The results pane at the bottom shows a single row of data:

total_rows
1 1987

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];
```

```
17 OR quantity IS NULL  
18 OR price_per_unit IS NULL  
19 OR cogs IS NULL  
20 OR total_sale IS NULL;  
21  
22  
23 --Exploratory Data Analysis(EDA)  
24 -- Total rows  
25 SELECT COUNT(*) AS total_rows FROM [retail_sales TB.csv];  
26  
27 --Date range  
28 SELECT MIN(sale_date) AS start_date, MAX(sale_date) AS end_date FROM [retail_sales TB.csv];  
29  
30  
31  
32  
33
```

No issues found

Results Messages

	start_date	end_date
1	2022-01-01	2023-12-31

Ln: 26 Ch

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];
```

```
20 OR total_sale IS NULL;  
21  
22  
23 --Exploratory Data Analysis(EDA)  
24 -- Total rows  
25 SELECT COUNT(*) AS total_rows FROM [retail_sales TB.csv];  
26  
27 --Date range  
28 SELECT MIN(sale_date) AS start_date, MAX(sale_date) AS end_date FROM [retail_sales TB.csv];  
29  
30 -- Categories  
31 SELECT DISTINCT category FROM [retail_sales TB.csv];  
32  
33
```

No issues found

Results Messages

	category
1	Clothing
2	Electronics
3	Beauty

Ln: 30 Ch: 1 SPC CR

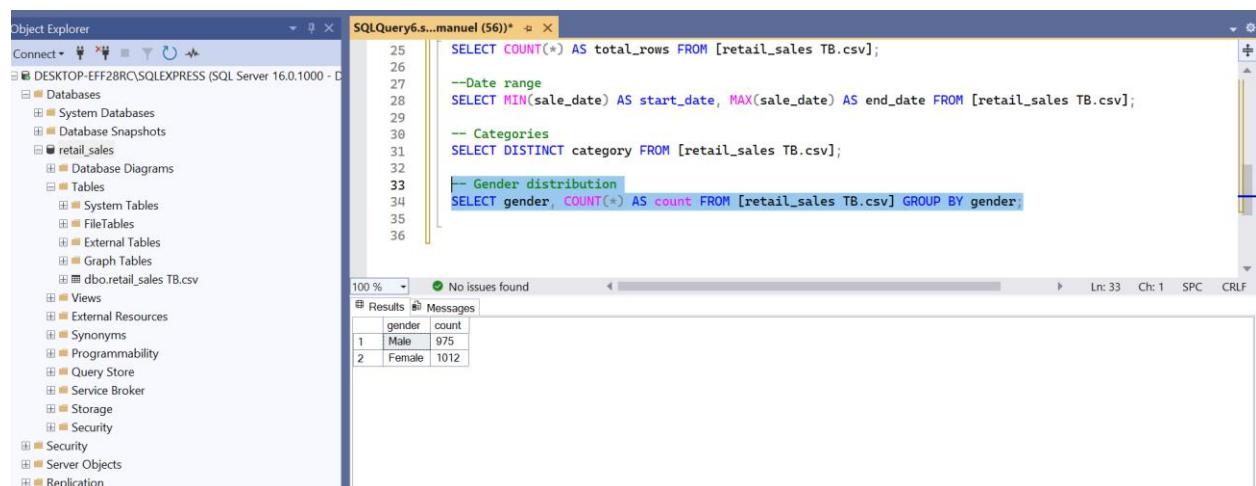
Fig viii

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;
```



The screenshot shows the Object Explorer on the left, connected to 'DESKTOP-EFF28RC\SQLEXPRESS (SQL Server 16.0.1000 - D)'. Under the 'retail_sales' database, the 'Tables' node is expanded, showing 'dbo.retail_sales TB.csv'. The main window displays a query in the SQLQuery6...manuel (56)* tab:

```
25  
26  
27  
28  
29  
30  
31  
32  
33  
34  
35  
36  
  
--Date range  
SELECT MIN(sale_date) AS start_date, MAX(sale_date) AS end_date FROM [retail_sales TB.csv];  
  
-- Categories  
SELECT DISTINCT category FROM [retail_sales TB.csv];  
  
-- Gender distribution  
SELECT gender, COUNT(*) AS count FROM [retail_sales TB.csv] GROUP BY gender;
```

The results pane shows a table with two rows:

gender	count
Male	975
Female	1012

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];

```

```

-- Categories
SELECT DISTINCT category FROM [retail_sales TB.csv];
-- Gender distribution
SELECT gender, COUNT(*) AS count FROM [retail_sales TB.csv] GROUP BY gender;
--Age Range of Customers
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';

```

The screenshot shows the SSMS interface. On the left is the Object Explorer pane, which lists the database structure for 'DESKTOP-EFF28RC\SQLEXPRESS (SQL Server 16.0.1000 - D)'. It includes sections for Databases, Tables, Views, and other system objects. In the center is the SQLQuery6.s...manuel (56)* window containing the following T-SQL code:

```

32
33
34
35
36
37
38
39
40
41
42
43
44
-- Gender distribution
SELECT gender, COUNT(*) AS count FROM [retail_sales TB.csv] GROUP BY gender;
--Business Analysis Query: It is used to answer specific business questions
-- i. All sales on '2022-11-05'
SELECT *
FROM [retail_sales TB.csv]
WHERE sale_date = '2022-11-05';

```

The results grid below the code shows 11 rows of transaction data from November 5, 2022. The columns are: transactions_id, sale_date, sale_time, customer_id, gender, age, category, quantity, price_per_unit, cogs, and total_sale.

	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';

```

```

--Business Analysis Query: It is used to answer specific business questions
-- i. All sales on '2022-11-05'
SELECT *
FROM [retail_sales TB.csv]
WHERE sale_date = '2022-11-05';

-- ii. Clothing sales > 4 in Nov 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;`

```

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

-- iii. Total sales per category
SELECT category, SUM(total_sale) AS total_sales
FROM [retail_sales TB.csv]
GROUP BY category;

```

No issues found

category	total_sales
1 Clothing	309995
2 Electronics	311445
3 Beauty	286790

Query executed successfully.

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';

```

Object Explorer
Connect ▾ X X X X X
DESKTOP-EFF28RC\SQLEXPRESS (SQL Server 16.0.1000 - D
  □ Databases
    □ System Databases
    □ Database Snapshots
  □ retail_sales
    □ Database Diagrams
    □ Tables
      □ System Tables
      □ FileTables
      □ External Tables
      □ Graph Tables
    □ dbo.retail_sales TB.csv
  □ Views
  □ External Resources
  □ Synonyms
  □ Programmability
  □ Query Store
  □ Service Broker
  □ Storage
  □ Security
  □ Security
  □ Server Objects
  □ Replication
  □ Management

SQLQuery6.s...manuel (56)*
50
51
52
53 -- iii. Total sales per category
54 SELECT category, SUM(total_sale) AS total_sales
55 FROM [retail_sales TB.csv]
56 GROUP BY category;
57
58 -- iv. Average age of customers in Beauty
59 SELECT AVG(age) AS avg_age
60 FROM [retail_sales TB.csv]
61 WHERE category = 'Beauty';
62
63
100 % No issues found
Results Messages
avg_age
1 40

```

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;

```

```

57 -- iv. Average age of customers in Beauty
58
59     SELECT AVG(age) AS avg_age
60     FROM [retail_sales TB.csv]
61     WHERE category = 'Beauty';
62
63
64 -- v. Transactions with total_sale > 1000
65
66     SELECT *
67     FROM [retail_sales TB.csv]
68     WHERE total_sale > 1000;

```

Results

	transaction_id	sale_date	sale_time	customer_id	gender	age	category	quantity	price_per_unit	cogs	total_sale
1	13	2023-02-08	17:43:00.0000000	106	Male	22	Electronics	3	500	245	1500
2	15	2022-07-01	11:50:00.0000000	75	Female	42	Electronics	4	500	210	2000
3	16	2022-06-25	10:33:00.0000000	82	Male	19	Clothing	3	500	180	1500
4	31	2023-12-31	17:47:00.0000000	3	Male	44	Electronics	4	300	129	1200
5	46	2022-11-06	17:50:00.0000000	54	Female	20	Electronics	4	300	84	1200
6	47	2022-10-22	17:22:00.0000000	96	Female	40	Beauty	3	500	600	1500
7	54	2022-10-20	10:17:00.0000000	142	Female	38	Electronics	3	500	200	1500
8	58	2023-09-16	19:18:00.0000000	53	Male	18	Clothing	4	300	75	1200
9	65	2022-12-11	20:03:00.0000000	84	Male	51	Electronics	4	500	160	2000
10	67	2023-08-10	20:19:00.0000000	119	Female	48	Beauty	4	300	129	1200
11	72	2023-12-06	19:19:00.0000000	5	Female	20	Electronics	4	500	195	2000
12	74	2023-10-05	19:50:00.0000000	56	Female	18	Beauty	4	500	205	2000

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;

```

Object Explorer
Connect ▾ DESKTOP-EFF28RC\SQLEXPRESS (SQL Server 16.0.1000 - D)
  □ Databases
    □ System Databases
    □ Database Snapshots
  □ retail_sales
    □ Database Diagrams
  □ Tables
    □ System Tables
    □ FileTables
    □ External Tables
    □ Graph Tables
    □ dbo.retail_sales TB.csv
  □ Views
  □ External Resources
  □ Synonyms
  □ Programmability
  □ Query Store
  □ Service Broker
  □ Storage
  □ Security
  □ Security
  □ Server Objects
  □ Replication
  □ Management
  □ XEvent Profiler

SQLQuery6.s...manuel (56)*
61 WHERE category = 'Beauty';
62
63
64
65 -- v. Transactions with total_sale > 1000
66 SELECT *
67   FROM [retail_sales TB.csv]
68   WHERE total_sale > 1000;
69
70 -- vi. Transactions count by gender & category
71 SELECT gender, category, COUNT(transactions_id) AS total_transactions
72   FROM [retail_sales TB.csv]
73   GROUP BY gender, category;

Results 100 % No issues found
gender category total_transactions
1 Male Beauty 281
2 Female Clothing 347
3 Male Clothing 351
4 Female Beauty 330
5 Male Electronics 343
6 Female Electronics 335

Query executed successfully. DESKTOP-EFF28RC\SQLEXPRESS ... DESKTOP-EFF28RC\Emmanu... retail_sales 00:00:00 6 rows

```

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;

```

The screenshot shows the SSMS interface with two windows:

- Object Explorer:** Shows the database structure for 'DESKTOP-EFF28RC\SQLEXPRESS'. It includes databases, tables (e.g., 'retail_sales'), and views.
- SQLQuery6.s...manuel (56)*:** This window contains two queries. The first query counts transactions by gender and category. The second query calculates average sales per month and highlights the best-selling month for each year.
- Results:** The results of the second query are displayed in a table. The table has columns: year, month, avg_sale, and total_sales. The data is as follows:

	year	month	avg_sale	total_sales
1	2022	1	397	22635
2	2022	2	366	16110
3	2022	3	521	23455
4	2022	4	500	28535
5	2022	5	480	24480
6	2022	6	481	20700
7	2022	7	541	22195
8	2022	8	390	21075
9	2022	9	485	61620
10	2022	10	467	67735
11	2022	11	472	68915
12	2022	12	460	71880

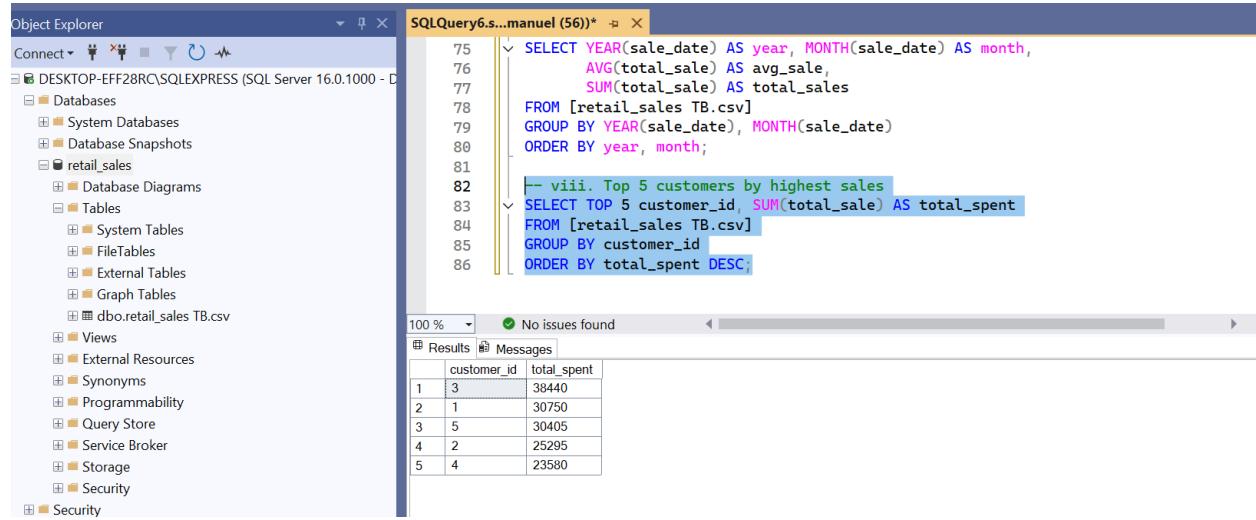
At the bottom of the Results window, it says 'Query executed successfully.'

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;
```



The screenshot shows the SQL Server Management Studio interface. On the left is the Object Explorer pane, which lists the database structure including databases, tables, and views. The central area is the SQL Query Editor window titled "SQLQuery6.s...manuel (56)*". It contains two queries. The first query calculates yearly average and total sales. The second query, starting at line 82, retrieves the top 5 customers by total sales. The results pane at the bottom shows a table with 5 rows of data.

customer_id	total_spent
3	38440
1	30750
5	30405
2	25295
4	23580

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;
```

category	unique_customers
Beauty	141
Clothing	149
Electronics	144

Status bar at the bottom: 'Query executed successfully.'"/>

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;

```

```

-- x. Shift analysis
SELECT
    CASE
        WHEN DATEPART(HOUR, sale_time) < 12 THEN 'Morning'
        WHEN DATEPART(HOUR, sale_time) BETWEEN 12 AND 17 THEN 'Afternoon'
        ELSE 'Evening'
    END AS shift,
    COUNT(*) AS orders
FROM [retail_sales TB.csv]
GROUP BY
    CASE
        WHEN DATEPART(HOUR, sale_time) < 12 THEN 'Morning'
        WHEN DATEPART(HOUR, sale_time) BETWEEN 12 AND 17 THEN 'Afternoon'
        ELSE 'Evening'
    END

```

shift	orders
Evening	1062
Morning	548
Afternoon	377

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/1BG0bSlaoIXNTV4fhXBEuFXRPExaRDaMV/view?usp=drive_link