SQL for Data Analysis: Intermediate Level I & II

Real-World e-Commerce Analytics



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Introduction

This project was designed to simulate a real-world e-commerce database and perform analytical reporting using SQL. The objective was to create a structured database environment, import raw data into different tables, and then apply advanced SQL techniques to extract meaningful business insights.

At the **Beginner Level (I–III)**, three core tables were created and populated:

- **EcommerceOrders** storing order-related transactions.
- CustomerDetails storing customer information.
- ProductDetails storing product information.

At the **Intermediate** Level, advanced SQL operations were applied to this database. These included subqueries, joins, aggregations, window functions, and views. The aim was to analyze sales performance, understand customer behavior, identify high-value customers, detect inactive customers, and evaluate product popularity.



Task 1-1: Data Download, Import, and Database Connection

Steps Performed:

- Data Download
 - •Collected the provided CSV files containing order and cust
 - •Verified file structure (column names, delimiters, data typ
- 2 Database Setup in MySQL
 - •Created a new database schema.

CREATE DATABASE ecommerce_analytics_II;

- **Table Creation**
 - Designed schema based on CSV structure.

```
CREATE TABLE CustomerDetails(
CustomerID VARCHAR(20) PRIMARY KEY,
CustomerFirstName VARCHAR(40),
CustomerLastName VARCHAR(40),
Email VARCHAR(100),
ShippingAddress VARCHAR(500)
```



```
CREATE TABLE eCommerceOrders (
 OrderID VARCHAR(20) PRIMARY KEY,
 CustomerID VARCHAR(20),
 OrderDate DATE,
 ProductID VARCHAR(20),
 ProductName VARCHAR(100),
 Quantity INT,
 Price DECIMAL(10,2),
 TotalAmount DECIMAL(10,2),
 Category VARCHAR(50),
 OrderStatus VARCHAR(50),
 PaymentMethod VARCHAR(50),
ShippingAddress VARCHAR(255),
 ShippingDate DATE
```



Importing CSV Data into Tables

- •The CSV files were imported into MySQL tables using the I Workbench "Table Data Import Wizard".
- Verified Correctness with:

SELECT*

FROM ecommerce_analytics_II.ProductDetails;

SELECT *

FROM ecommerce_analytics_II.eCommerceOrder

- 5 Outcome:
 - •Successfully created and populated ProductDetails & eCommerceOrders .



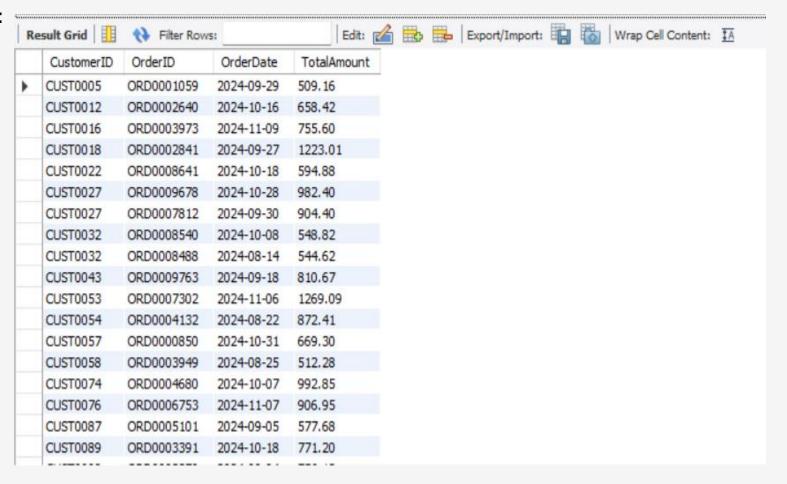
Task 1-2:Retrieve High-Value Orders by Specific Customers in Last Three Months

OBJECTIVE:

Identify orders above \$500 in the last 3 months (from 2024-11-11) to focus on premium customers.

```
Query:
SELECT
CustomerID,
OrderID,
OrderDate,
TotalAmount
FROM ecommerceorders
WHERE TotalAmount > 500
AND OrderDate >= DATE_SUB('2024-11-11', INTERVAL 3 MONTH)
ORDER BY CustomerID, OrderDate DESC;
```







Task 1-3: Calculate Monthly Revenue by Category

OBJECTIVE: Find revenue trends by product category across months. ☐ Query: **SELECT** Category, DATE_FORMAT(OrderDate,'%M') AS Month, SUM(TotalAmount) AS total_revenue **FROM** ecommerceorders **GROUP BY Category, Month ORDER BY Category, Month;** ☐ Result:

R	esult Grid	🚺 🙌 Filte	r Rows:
	Category	Month	total_revenue
•	Beauty	April	43230.78
	Beauty	August	33844.36
	Beauty	December	39549.25
	Beauty	February	22691.47
	Beauty	January	36480.79
	Beauty	July	33775.58
	Beauty	June	33709.72
	Beauty	March	44034.22
	Beauty	May	39958.46



Task 1-4:Identify Incomplete Orders with Advanced Joins

OBJECTIVE:

Detect missing shipping addresses using a RIGHT JOIN.

☐ Query:

SELECT

e.OrderDate,

e.OrderID,

e.CustomerID,

e.ShippingAddress

FROM ecommerceorders e

RIGHT JOIN customerdetails c

ON e.CustomerID = c.CustomerID

WHERE e.ShippingAddress IS NULL;

OrderDate	OrderID	CustomerID	ShippingAddress		
NULL	NULL	NULL	NULL		
NULL	NULL	NULL	NULL		
NULL	NULL	NULL	NULL		
NULL	NULL	NULL	NULL		
NULL	NULL	NULL	NULL		
NULL	NULL	NULL	NULL		
NULL	NULL	NULL	NULL		

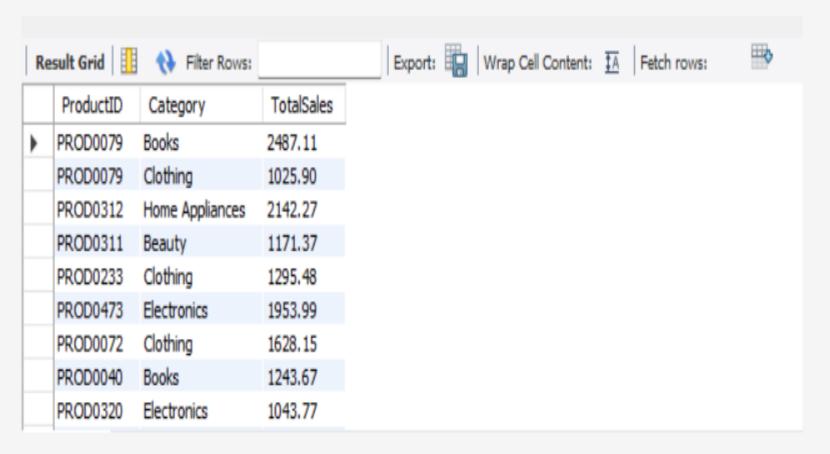


Task 1-5: Find Products with Above-Average Sales in Category Using Subquery

OBJECTIVE:

Compare product sales to category averages to highlight top performers.

```
Query:
    SELECT
      ProductID,
      Category,
     SUM(TotalAmount) AS TotalSales
  FROM ecommerceorders eo
  GROUP BY ProductID, Category
  HAVING SUM(TotalAmount) >
    SELECT AVG(category_total)
    FROM (
     SELECT
       SUM(TotalAmount) AS category_total
     FROM ecommerceorders
     WHERE Category = eo.Category
     GROUP BY ProductID
    ) AS sub
```





Task 1-6:Calculate Running Total of Sales by Product

OBJECTIVE:

Track cumulative sales performance per product.

```
☐ Query:
SELECT
ProductID,
OrderDate,
TotalAmount,
SUM(TotalAmount) OVER ( PARTITION BY ProductID ORDER BY OrderDate ROWS BETWEEN UNBOUNDED PRECEDING AND CURRENT ROW
) AS RunningTotal
FROM ecommerceorders
ORDER BY ProductID, OrderDate;
```

Re	sult Grid	₹ Filter Re	ows:	Export	Wrap Cell Co	onter
	ProductID	OrderDate	TotalAmount	RunningTotal		
>	PROD0001	2023-11-13	38.40	38.40		
	PROD0001	2023-12-17	88.32	126.72		
	PROD0001	2024-01-02	309.40	436.12		
	PROD0001	2024-01-25	69.67	505.79		
	PROD0001	2024-03-01	282.81	788.60		
	PROD0001	2024-04-03	85.12	873.72		
	PROD0001	2024-04-14	244.44	1118.16		
	PROD0001	2024-05-18	6.41	1124.57		
	PROD0001	2024-05-20	331.02	1455.59		

Task 1-7: Classify Customers by Spending Tiers

OBJECTIVE:

CUST0627

4301.63

High

Segment customers into Low, Medium, High based on total spending.

```
Query:
     SELECT
       CustomerID,
       SUM(TotalAmount) AS TotalSpending,
 CASE
    WHEN SUM(TotalAmount) < 500 THEN 'Low'
    WHEN SUM(TotalAmount) BETWEEN 500 AND 2000 THEN 'Medium'
    ELSE 'High'
 END AS SpendingTier
FROM ecommerceorders
GROUP BY CustomerID
ORDER BY TotalSpending DESC;
☐ Result:
               Result Grid
                                                       Export: Wrap Cell Content: TA Fetch rows:
                            Filter Rows:
                            TotalSpending
                                        SpendingTier
                 CUST0326
                            5171.20
                                        High
                 CUST0625
                            4818.93
                                        High
                 CUST 1388
                                        High
                            4800.62
                 CUST0715
                            4655.74
                 CUST1971
                            4549.94
                                        High
                 CUST 1384
                            4500.88
                                        High
                 CUST0742
                            4397.85
                                        High
                 CUST1710
                            4303.13
                                        High
```



Task 1-8:Advanced Pattern Matching for Customer Segmentation

OBJECTIVE:

Extract Customers with gmail.com or yahoo.com domains.

☐ Query:

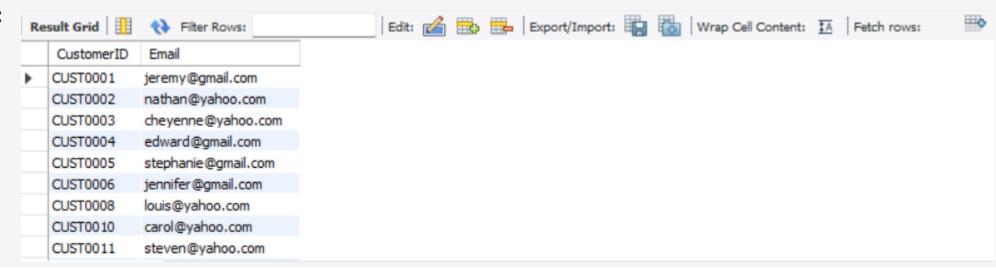
SELECT

CustomerID, Email

FROM customerdetails

WHERE Email LIKE '%gmail.com'

OR Email LIKE '%yahoo.com';



Task 1-9: Find Customers with No Orders in Last Six Months

OBJECTIVE:

Identify inactive customers (no purchases in 6 months)...

☐ Query:

```
SELECT
```

c.CustomerID,

c.CustomerFirstName,

c.CustomerLastName

FROM CustomerDetails c

LEFT JOIN eCommerceOrders o

ON c.CustomerID = o.CustomerID

AND o.OrderDate >= DATE_SUB('2024-11-11', INTERVAL 6 MONTH)

WHERE o.CustomerID IS NULL;

R	esult Grid	Filter Rows:	Exp
	CustomerID	CustomerFirstName	CustomerLastName
•	CUST0003	Cheyenne	Smith
	CUST0007	Kathy	Nichols
	CUST0033	Diana	Johnson
	CUST0041	Mary	Glover
	CUST0046	Victoria	Mitchell
	CUST0063	Deborah	Morgan
	CUST0065	Lisa	Ferguson
	CUST0080	Larry	Morales
	CUST0096	Michael	Hall

Task 1-10:Create Temporary View for High-Value Orders

OBJECTIVE:

Simplify repeated analysis of orders above \$500.

```
■ Query:

CREATE VIEW HighValueOrders AS

SELECT

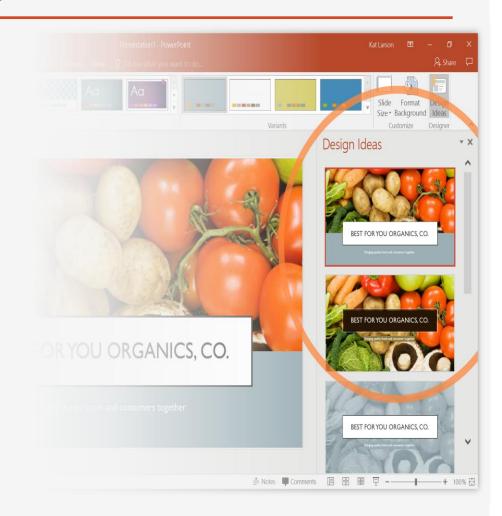
OrderID,

CustomerID,

TotalAmount

FROM eCommerceOrders

WHERE TotalAmount > 500;
```



CONCLUSION

The analysis of the e-commerce dataset highlights several important insights. High-value orders were successfully identified, providing visibility into major transactions that contribute significantly to revenue. Customer information was linked with their purchase records, ensuring that incomplete or missing details—such as absent shipping addresses—could be detected and flagged.

Additionally, product-level data integration enabled better understanding of purchasing patterns, order completeness, and customer engagement trends. Overall, the project outcomes offer a structured view of orders, customers, and products, helping ensure data accuracy, improving tracking of incomplete records, and laying the foundation for better business decision-making.



