

# Retail Analytics Case Study

This case study is about using data analytics to solve problems in a company.

## Key Concepts:

- **Product Sales Analysis:** Finding out which products sell well and which don't.
  - **Example:** If Product A sold 500 units and Product B sold only 10, focus marketing on Product A and consider removing Product B.
- **Customer Segmentation:** Group customers based on how much they buy.
  - **Example:**
  - No Orders: Customer bought 0 items.
  - Low: Bought 1-10 items.
  - Mid: Bought 10-30 items.
  - High Value: Bought more than 30 items.  
This helps target promotions (e.g., special offers for high-value customers).
- **Customer Behavior Analysis:** Study patterns like repeat purchases and loyalty.
  - **Example:** If a customer buys every month, offer them a loyalty discount to keep them coming back.
- **Data Cleaning:** Use SQL to clean data and explore trends.
  - **Example:** Remove duplicate sales records, check for missing values, and calculate total sales per product.

## Datasets Used:

- **Sales Transaction (who bought what, when, and how much)**
- **Customer Profiles (customer details)**
- **Product Inventory (product details and stock)**

**Primary Goal:** Using these analyses to improve marketing, manage inventory better, and increase customer satisfaction.

## Task 1: Identifying and Removing Duplicate Records

```
1 select transactionid, count(*)
2 from sales_transaction
3 group by 1
4 having count(*) >1;
5
6 create table st as
7 select distinct *
8 from sales_transaction;
9
10 drop table sales_transaction;
11
12 alter table st
13 rename to sales_transaction;
14
15 select *
16 from sales_transaction;
```

## Task 2: Correcting Product Price Discrepancies

```
1 select a.transactionid, a.price as transactionprice, b.price as inventoryprice
2 from sales_transaction a
3 join product_inventory b
4 on a.productid = b.productid and a.price<>b.price;
5
6 update sales_transaction a
7 set a.price = (select b.price from product_inventory b where a.productid = b.productid)
8 where a.productid in (select productid from product_inventory b where a.price<>b.price);
9
10 select *
11 from sales_transaction;
```

## Task 3: Identifying & Handling NULL Values

```
1 select
2 sum(case when age is null then 1 else 0 end) as age_null,
3 sum(case when gender is null then 1 else 0 end) as gender_null,
4 sum(case when location is null then 1 else 0 end) as location_null,
5 sum(case when joindate is null then 1 else 0 end) as date_null
6 from customer_profiles;
```

```
1 -- select
2 -- sum(case when age is null then 1 else 0 end) as age_null,
3 -- sum(case when gender is null then 1 else 0 end) as gender_null,
4 -- sum(case when location is null then 1 else 0 end) as location_null,
5 -- sum(case when joindate is null then 1 else 0 end) as date_null
6 -- from customer_profiles;
7
8 select count(*)
9 from customer_profiles
10 where location is null;
```

```
8 select count(*)
9 from customer_profiles
10 where location is null;
11
12 update customer_profiles
13 set location = "unknown"
14 where location is null;
15
16 select *
17 from customer_profiles;
```

## Task 4: Cleaning and Converting Date Formats

```

1 select a.transactionid, a.price as transactionprice, b.price as inventoryprice
2 from sales_transaction a
3 join product_inventory b
4 on a.productid = b.productid and a.price <> b.price;
5
6 update sales_transaction a           I
7 set a.price = (select b.price from product_inventory b where a.productid = b.productid)
8 where a.productid in (select productid from product_inventory b where a.price <> b.price);
9
10 select *
11 from sales_transaction;

```

### Task 5: Summarising Total Sales and Quantity Per Product

```

1 select productid,
2 sum(quantitypurchased) as totalunitssold,
3 sum(quantitypurchased*price) as totalsales
4 from sales_transaction
5 group by 1
6 order by 3 desc;

```

### Taks 6: Analyzing Customer Purchase Frequency by Counting Transactions

```

1 select customerid, count(*) as numberoftransactions
2 from sales_transaction
3 group by 1
4 order by 2 desc;

```

### Task 7: Evaluating Product Category Performance using JOINS

```

1
2 select b.category, sum(a.quantitypurchased) as totalunitssold,
3 sum(a.quantitypurchased*a.price) as totalsales
4 from sales_transaction a
5 join product_inventory b
6 on a.productid = b.productid
7 group by 1
8 order by 3 desc;

```

### Task 8: Identifying Top 10 Products by Revenue

```

1 select productid, sum(quantitypurchased*price) as totalrevenue
2 from sales_transaction
3 group by 1
4 order by 2 desc
5 limit 10;

```

### Task 9: Filtering Aggregates: Understanding WHERE vs. HAVING

```

1 select productid, sum(quantitypurchased) as totalunitssold
2 from sales_transaction
3 group by 1
4 having totalunitssold>0
5 order by 2
6 limit 10;

```

### Taks 10: Calculating Daily Sales Metrics

```
1 select transactiondate_updated as datetrans,
2 count(*) as transaction_count,
3 sum(quantitypurchased) as totalunitssold,
4 sum(quantitypurchased*price) as totalsales
5 from sales_transaction
6 group by 1
7 order by 1 desc;
8
9 --select *
10 -- from sales_transaction;
```

### Task 11: Calculating Month-on-Month Sales Growth with Window Functions

```
1 with cte1 as (
2 select month(transactiondate_updated) as `month`,
3 round(sum(quantitypurchased*price),2) as total_sales
4 from sales_transaction
5 group by 1),
6
7 cte2 as
8 (select *, log(total_sales) over(order by `month`) as previous_month_sales
9 from cte1)
10
11 select *, round(100*((total_sales - previous_month_sales)/previous_month_sales),2) as mom_growth_percentage
12 from cte2;
```

### Task 12: Identifying High-Value Customers with Multiple Conditions

```
1 select customerid, count(*) as numberoftransactions,
2 sum(quantitypurchased*price) as totalspent
3 from sales_transaction
4 group by 1
5 having numberoftransactions>10 and totalspent>1000
6 order by 3 desc;
```