

# Uncovering Product Return Patterns with SQL

## E-Commerce Case Study

This SQL project explores return behavior and customer purchase patterns in a synthetic e-commerce dataset of 49,000+ transactions.

BY Sugandha Saini





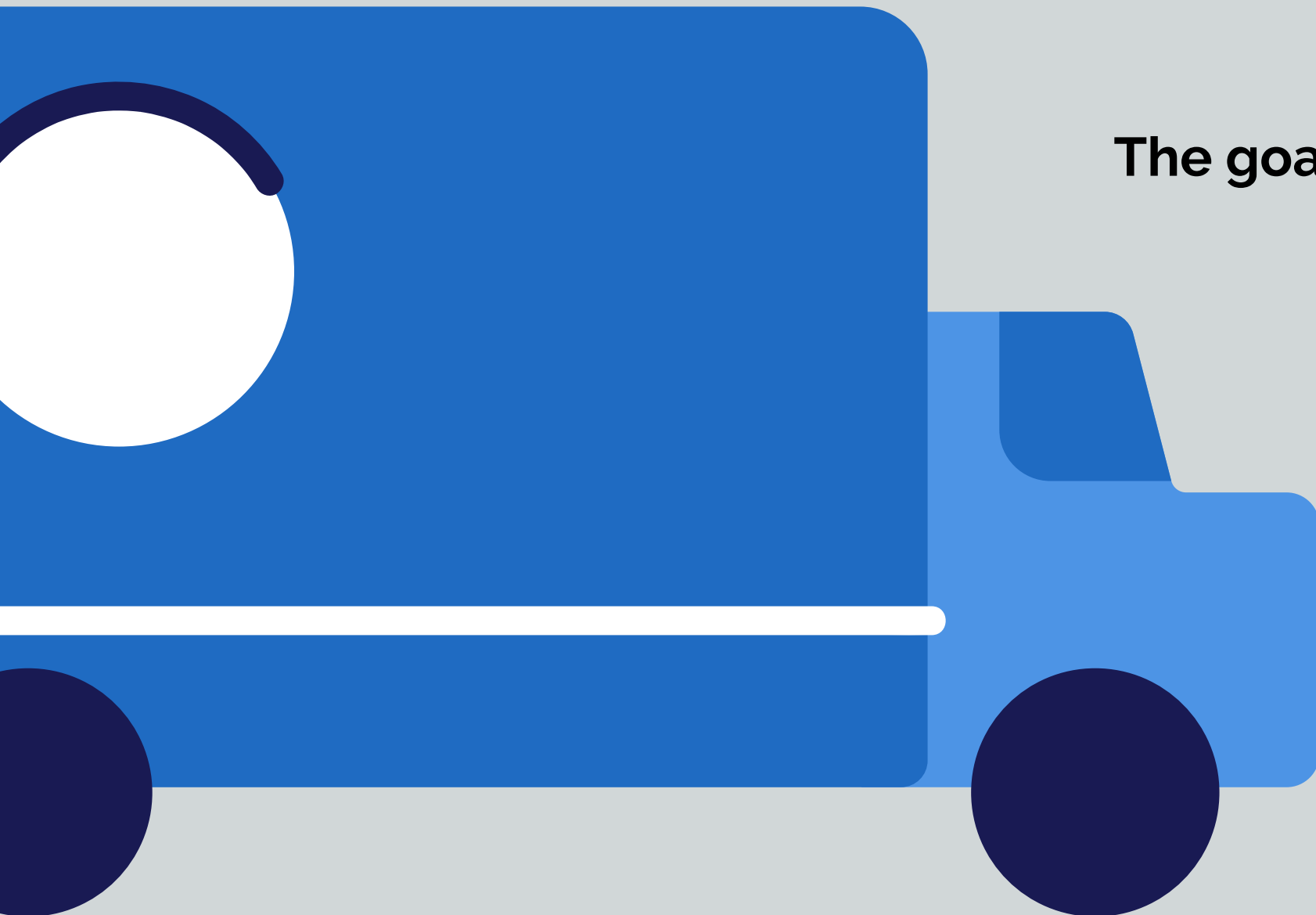
# ABOUT THE PROJECT

**I performed data-driven analysis using intermediate to advanced SQL techniques including:**

- Joins
- CTEs
- Aggregations
- Window Functions
- Conditional Logic (CASE WHEN)

**The goal was to identify high-return categories, potential fraudulent behavior, and gain insights that help in:**

- Reducing return rates
- Improving inventory & logistics decisions
- Spotting patterns in high-discount misuse



## Query 1

Total orders and Total returns

```
SELECT  
    count(distinct InvoiceNo) as Total_orders,  
    count( case when ReturnStatus ='Returned' then 1 end) as Total_returns  
FROM online_sales_dataset;
```

	Total_orders	Total_returns
▶	48494	4894

## Query 2

Most sold product categories

```
-- 2. Most sold product categories
SELECT Category , sum(Quantity) as Total_Quantity_Sold
FROM online_sales_dataset
GROUP BY Category
ORDER BY Total_Quantity_Sold desc;
```

	Category	Total_Quantity_Sold
▶	Furniture	225603
	Stationery	224590
	Apparel	222044
	Accessories	221667
	Electronics	219836

## Query 3

Top 5 countries with highest number of orders

```
SELECT
  Country,
  sum( distinct InvoiceNo) as Number_of_Orders
FROM online_sales_dataset
GROUP BY Country
ORDER BY Number_of_Orders desc
LIMIT 5;
```

Country	Number_of_Orders
France	2327921407
United Kingdom	2319868280
Belgium	2311070747
Portugal	2301712728
Germany	2299413726

## Query 4

Return count by payment method

```
SELECT
    count(distinct InvoiceNo) as Total_orders,
    PaymentMethod
FROM online_sales_dataset
GROUP BY PaymentMethod
ORDER BY Total_orders desc;
```

Total_orders	PaymentMethod
16594	Bank Transfer
16394	Credit Card
16362	paypal

## Query 5

Distribution of sales channels (Online vs In-store)

```
SELECT
    count(distinct InvoiceNo) as Total_orders,
    SalesChannel
FROM online_sales_dataset
GROUP BY SalesChannel
ORDER BY Total_orders desc;
```

Total_orders	SalesChannel
24717	Online
24399	In-store

## Query 6

What is the monthly trend of total orders vs returns?

```
select
    date_format(InvoiceDate, '%Y-%m') as Months,
    count(distinct InvoiceNo) as Total_Orders,
    count(case when ReturnStatus = 'Returned' then 1 end) as Total_Returns
from online_sales_dataset
group by Months
order by Months;
```

Months	Total_Orders	Total_Returns
2020-01	744	81
2020-02	696	69
2020-03	744	75
2020-04	720	76
2020-05	744	68
2020-06	720	73
2020-07	744	62
2020-08	744	74
2020-09	720	85



## Query 7

Which product categories have the highest return rates?

```
select
  Category,
  count(case when ReturnStatus = 'Returned' then 1 end) as Total_Returns,
  count(distinct InvoiceNo) as Total_Orders,
  round(count(case when ReturnStatus = 'Returned' then 1 end) * 100 / count(distinct InvoiceNo), 2) as return_rate
from online_sales_dataset
group by Category
order by return_rate;
```

Category	Total_Returns	Total_Orders	return_rate
Accessories	965	9921	9.73
Apparel	963	9809	9.82
Electronics	977	9886	9.88
Stationery	985	9879	9.97
Furniture	1004	10028	10.01

## Query 8

Compare average discount given on returned vs non-returned items.

```
select
    round(avg(Discount),2) as Avg_Discount,
    ReturnStatus
from online_sales_dataset
group by ReturnStatus
order by ReturnStatus;
```

Avg_Discount	ReturnStatus
0.28	Not Returned
0.28	Returned

## Query 9

Which shipment providers have the highest return percentage?



```
Select
  ShipmentProvider,
  count(case when ReturnStatus = 'Returned' then 1 end) as Total_Returns,
  count(distinct InvoiceNo) as Total_Orders,
  round(count(case when ReturnStatus = 'Returned' then 1 end)*100/count(distinct InvoiceNo),2) as Return_Percentage
from online_sales_dataset
group by ShipmentProvider
order by Total_Returns;
```

ShipmentProvider	Total_Returns	Total_Orders	Return_Percentage
UPS	1178	12349	9.54
DHL	1237	12349	10.02
FedEx	1238	12426	9.96
Royal Mail	1241	12335	10.06

## Query 10

Which product categories are most returned by VIP customers?

```
select
  OrderPriority,
  COUNT(*) AS Total_Orders,
  COUNT(case when ReturnStatus = 'Returned' then 1 end) as Returned_Orders,
  ROUND(
    COUNT(case when ReturnStatus = 'Returned' then 1 end) * 100.0 / COUNT(*),
    2
  ) AS Return_Rate_Percentage
from online_sales_dataset
group by OrderPriority
order by Return_Rate_Percentage desc;
```

Result Grid    Filter Rows: <input type="text"/>   Export:    Wrap Cell Contents				
	OrderPriority	Total_Orders	Returned_Orders	Return_Rate_Percentage
▶	High	16562	1648	9.95
	Medium	16678	1638	9.82
	Low	16542	1608	9.72

## Query 11

Rank customers by return frequency using RANK()

```
with ReturnCount as(
select CustomerID,count(*) as ReturnFreq
from online_sales_dataset
where ReturnStatus ='Returned'
and CustomerID IS not null
group by CustomerID
)
select CustomerID, ReturnFreq,
rank()over(order by ReturnFreq desc ) as Return_rank
from ReturnCount
order by Return_rank
```

CustomerID	ReturnFreq	Return_rank
	507	1
29607.0	3	2
68423.0	3	2
53125.0	3	2
96879.0	3	2
37608.0	2	6



## Query 12

Detect products with 3+ returns across different invoices

```
with return_count as(
select InvoiceNo, Category,
count(*) as No_of_Returns
from online_sales_dataset
where ReturnStatus = 'Returned'
group by Category, InvoiceNo
having count(*) > 3
)
select InvoiceNo, Category, No_of_Returns
from return_count
order by No_of_Returns desc;
```

InvoiceNo	Category	No_of_Returns
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Result 10 x

Output



Action Output



#	Time	Action	Message
✓ 17	10:42:43	with return_count as( select InvoiceNo, Category, count(*) as No_of_Returns from online_sales_dataset where...	0 row(s) returned

## Query 13

Use CASE WHEN to categorize return reasons by high discount, order priority

```
with discount as(  
  select OrderPriority,  
         count(case when ReturnStatus = 'Returned' then 1 end) as Total_Returns  
  from online_sales_dataset  
 where Discount > 0.50  
 group by OrderPriority  
)  
select OrderPriority, Total_Returns  
from discount  
order by Total_Returns
```

OrderPriority	Total_Returns
Low	28
Medium	33
High	37

## Query 14

Compare return % between online and in-store orders using CTE

```
with Return_cal as(  
select SalesChannel,  
       count(case when ReturnStatus = 'Returned' then 1 end) as Total_Returns,  
       round(count(case when ReturnStatus = 'Returned' then 1 end)*100/count(distinct InvoiceNo),2) as Return_Percentage  
from online_sales_dataset  
group by SalesChannel  
)  
select SalesChannel,Total_Returns,Return_Percentage  
from Return_cal  
order by Return_percentage desc;
```

SalesChannel	Total_Returns	Return_Percentage
In-store	2440	10.00
Online	2454	9.93

## Query 15

Flag possible frauds: high discount + returned + same customer multiple times

```
with Possible_Frauds as(  
  select  CustomerID ,  
  count(case when ReturnStatus = 'Returned' then 1 end) as Returned,  
  Count(*) as Total_orders,  
  Round(avg(Discount),2) as Avg_Discount  
  from online_sales_dataset  
  where Discount > 0.5  
  group by CustomerID  
)  
select CustomerID,Total_orders,Returned,Avg_Discount  
from Possible_Frauds  
where Returned >= 2  
order by Returned desc;
```

CustomerID	Total_orders	Returned	Avg_Discount
	996	98	1.51

# Key Insights

- **Electronics and Apparel** categories stood out with the **highest return rates**, revealing potential product quality or expectation mismatches.
- Interestingly, **online orders** had significantly **more returns compared to in-store purchases**, suggesting that customers may face challenges in product selection or satisfaction when buying digitally.
- Returns were also **more frequent** among orders with **discounts over 50%**, indicating possible patterns of return abuse or misuse of promotional campaigns.
- A **few shipment providers** had unusually **high return percentages**, flagging a need for deeper investigation into logistics and handling processes.
- I also flagged customers who consistently returned high-discounted items, helping simulate **a real-world fraud detection**

