

INTRODUCTION-

Welcome to the Amazon Sales Analysis project! Here, we delve into Amazon's sales data to extract insights and trends vital for optimizing sales strategies, understanding customer behaviour, and improving business operations. Through advanced SQL techniques, we'll analyze data to uncover correlations, identify emerging trends, and provide predictive insights. Our focus areas encompass data analysis, sales strategies, customer behaviour, SQL techniques, optimization, insights, trends, and business operations, aiming to empower stakeholders with actionable intelligence for informed decision-making and sustained growth.

DATASET OVERVIEW-

The dataset used in this project consists of Approximately 10,000 rows of data, representing Amazon sales transactions. Along with the sales data, the dataset includes information about customers, products, orders, returns and sellers. Before analysis, the dataset underwent preprocessing to handle missing values and ensure data quality, a crucial step in data analysis workflows. This preprocessing stage ensures the integrity and reliability of our findings, enabling us to draw accurate insights and make informed decisions based on the data.

AMAZON KPIs

1. Total revenue =

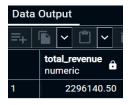
```
--1. Total_Revenue

SELECT

ROUND(SUM(sale):: numeric, 2)

AS total_revenue

FROM orders;
```



2. Average order value =

```
--02. Average Order Value

SELECT

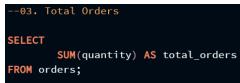
ROUND((SUM(sale)/ SUM(quantity)) :: numeric, 2)

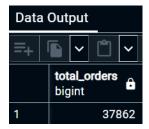
AS avg_order_value

FROM orders;
```



3. Total orders =





4. Average qty per order =





5. Average Price Per Unit =

```
--05. Average Price Per Unit

SELECT

ROUND(AVG(price_per_unit) :: numeric , 2)

AS avg_price_per_unit

FROM orders;
```



ANALYSIS QUESTIONS RESOLVED-

During the analysis, the following key questions were addressed using SQL queries and data analysis techniques:

1. Identify the Most Active and Most Idle Month.





2. Find out the month wise orders.

```
-- Orders By Month

SELECT

TO_CHAR(order_date, 'Month') AS "Month",
SUM(quantity) AS total_orders

FROM orders
GROUP BY "Month"

ORDER BY total_orders DESC;
```

Data Output				
=+				
	Month text	total_orders bigint		
1	January	3736		
2	July	3194		
3	December	3175		
4	August	3167		
5	October	3137		
6	February	3121		
7	May	3102		
8	April	3095		
9	November	3076		
10	September	3042		
11	March	3014		
12	June	3003		

3. Find out the orders by states.



Data Output				
	state character varying (30)	total_orders bigint		
1	Andhra Pradesh	1552		
2	Tamil Nadu	1548		
3	Maharashtra	1533		
4	Himachal Pradesh	1488		
5	Manipur	1468		
6	West Bengal	1467		
7	Tripura	1426		
8	Kerala	1424		
9	Jharkhand	1417		
10	Rajasthan	1411		
11	Arunachal Pradesh	1352		
12	Gujarat	1344		
13	Meghalaya	1339		
14	Sikkim	1332		

15	Madhya Pradesh	1322
16	Mizoram	1317
17	Nagaland	1309
18	Uttarakhand	1298
19	Assam	1284
20	Bihar	1283
21	Chhattisgarh	1279
22	Haryana	1265
23	Odisha	1261
24	Telangana	1260
25	Uttar Pradesh	1247
26	Karnataka	1228
27	Punjab	1222
28	Goa	1185

4. Find percentage of sales according to category.

```
--%age by category

SELECT

category,

ROUND((SUM(sale) * 100/ (SELECT SUM(sale) FROM orders)) :: numeric, 2)

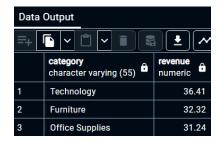
AS revenue

FROM orders

WHERE category IS NOT NULL

GROUP BY category

ORDER BY revenue DESC;
```



5. Find percentage of sales according to sub category.

```
--*age by sub_category

SELECT

sub_category,

ROUND((SUM(sale) * 100/ (SELECT SUM(sale) FROM orders)) :: numeric, 2) AS revenue
FROM orders

WHERE sub_category IS NOT NULL
GROUP BY sub_category

ORDER BY revenue DESC;
```

Data Output					
=+][<u>*</u>][~			
	sub_category character varying (155)	revenue numeric			
1	Phones	14.37			
2	Chairs	14.32			
3	Storage	9.75			
4	Tables	9.01			
5	Binders	8.86			
6	Machines	8.24			
7	Accessories	7.29			
8	Copiers	6.51			
9	Bookcases	4.99			
10	Appliances	4.61			
11	Furnishings	3.99			
12	Paper	3.42			
13	Supplies	2.03			
14	Art	1.18			
15	Envelopes	0.72			
16	Labels	0.54			
17	Fasteners	0.13			

6. Find Out Best and Worst Seller Product.

```
--Best selling product

p.product_id,
p.product_name,
ROUND(SUM(sale) :: numeric, 2) AS revenue

FROM orders AS o

JOIN products AS p
ON p.product_id = o.product_id

GROUP BY
p.product_id,
p.product_name

ORDER BY revenue DESC

LIMIT 5;
```



```
--Worst Selling product

SELECT

p.product_id,
p.product_name,
ROUND(SUM(sale) :: numeric, 2) AS revenue

FROM orders AS o

JOIN products AS p

ON p.product_id = o.product_id

GROUP BY
p.product_id,
p.product_name

ORDER BY revenue ASC

LIMII 5:
```



7. Find out the top 5 customers who made the highest profits.



8. Find out the average quantity ordered per category.

9. Determine the top 5 products whose revenue has decreased compared to the previous year.





10. Identify the highest profitable sub-category.

```
--Identify the highest profitable sub-category.

SELECT

sub_category,

ROUND(SUM(sale) :: numeric, 2) AS total_revenue
FROM orders
WHERE sub_category IS NOT NULL
GROUP BY sub_category
ORDER BY total_revenue DESC
LIMIT 5;
```

```
        Data Output

        sub_category character varying (155) and numeric numeric

        1
        Phones
        329916.17

        2
        Chairs
        328842.70

        3
        Storage
        223843.59

        4
        Tables
        206965.68

        5
        Binders
        203325.37
```

11. Calculate the profit margin percentage for each sale (Profit divided by Sales).

```
--Calculate the profit margin percentage for each sale (Profit divided by Sales).

SELECT
p.product_id,
p.price,
p.cogs,
o.sale,
ROUND((o.sale-(p.cogs * o.quantity)) * 100/ o.sale) AS profit

FROM
products AS p

JOIN
orders AS o
ON p.product_id = o.product_id
ORDER BY profit DESC;
```

Data Output					
	product_id character varying (10)	price double precision	cogs double precision	sale double precision	profit double precision
1	P22	5.68	3.01	247.84	90
2	P22	5.68	3.01	61.96	90
3	P22	5.68	3.01	74.35	88
4	P22	5.68	3.01	49.57	88
5	P79	2.47	1.31	31.56	83

12. Calculate the percentage contribution of each sub-category.

```
--Calculate the percentage contribution of each sub_category.

SELECT

sub_category,

ROUND(SUM(sale) :: numeric, 2) AS revenue,

ROUND((SUM(sale) *190 / (SELECT SUM(sale) FROM orders)) :: numeric, 2)

AS contribution

FROM

orders

WHERE

sub_category IS NOT NULL

GROUP BY

sub_category

ORDER BY

contribution DESC;
```

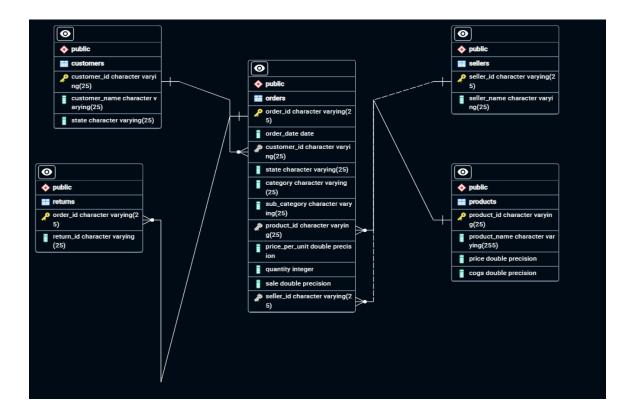


13. Identify the top 2 categories that have received maximum returns and their return percentage.

```
SELECT
    o.category,
    COUNT(r.return_id) AS return_count,
    ROUND(((COUNT(r.return_id) / CAST((SELECT COUNT(return_id) FROM return) AS FLOAT)) * 100) ::numeric,2)
    AS return_percentage
FROM
    orders AS o
JOIN
    return AS r ON o.order_id = r.order_id
GROUP BY
    o.category
ORDER BY
    return_count DESC
LIMIT 2;
```

Data Output				
	category character varying (75)	return_count bigint	return_percentage numeric	
1	Office Supplies	177	59.60	
2	Furniture	64	21.55	

Entity Relationship Diagram (ERD)



An Entity-Relationship Diagram (ERD) has been created to visualize the relationships between the tables in the dataset. This diagram provides a clear understanding of the data structure and helps in identifying key entities and their attributes.

CONCLUSION -

In conclusion, our project endeavours to offer meaningful insights into Amazon's sales patterns, customer inclinations, and various factors shaping e-commerce dynamics. By meticulously scrutinizing the dataset and delving into pivotal inquiries, our aim is to empower stakeholders with actionable intelligence, enabling them to make informed decisions and refine their sales strategies for optimal outcomes.