

Shop-Performance-Analysis

Section A Google Sheets / Excel

1. Import the dataset into Excel.

2. Created the following summaries:

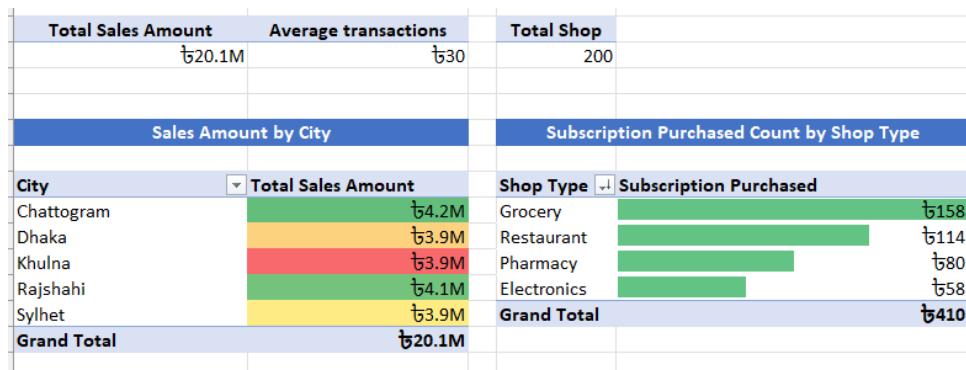
- Total sales amount by date
- Total sales amount by shop type
- Average transactions per shop

Total Sales Amount	Average transactions					
Total Sales Amount by Date		Total Sales Amount by Shop Type		Average Transactions per shop		
Date	Total Sales Amount	Date	Total Sales Amount	Date	Average Transactions	
01 January 2024	241973	Electronics	2850278	1000	726	
02 January 2024	176197	Grocery	8001731	1001	733	
03 January 2024	269268	Pharmacy	3759647	1002	725	
04 January 2024	271056	Restaurant	5471903	1003	734	
05 January 2024	196218			1004	726	
06 January 2024	238615			1005	729	
07 January 2024	256331			1006	726	
08 January 2024	207249			1007	730	
09 January 2024	234248			1008	728	
10 January 2024	230591			1009	728	
11 January 2024	205899			1010	729	
12 January 2024	167465			1011	724	
13 January 2024	226794			1012	721	
14 January 2024	290105			1013	731	
15 January 2024	263865			1014	732	
16 January 2024	316404			1015	730	
17 January 2024	268961			1016	731	
18 January 2024	225794			1017	721	
19 January 2024	193308			1018	734	
20 January 2024	203684			1019	738	
21 January 2024	258802			1020	730	
22 January 2024	265600			1021	736	
23 January 2024	195924			1022	731	
24 January 2024	229961			1023	729	
25 January 2024	251298			1024	731	
26 January 2024	278012			1025	732	
27 January 2024	245561			1026	727	
28 January 2024	230358			1027	733	
29 January 2024	231731			1028	735	
30 Januarv 2024	274048			1029	735	

< > Task_dataset **Summaries** Two Pivot Tables +

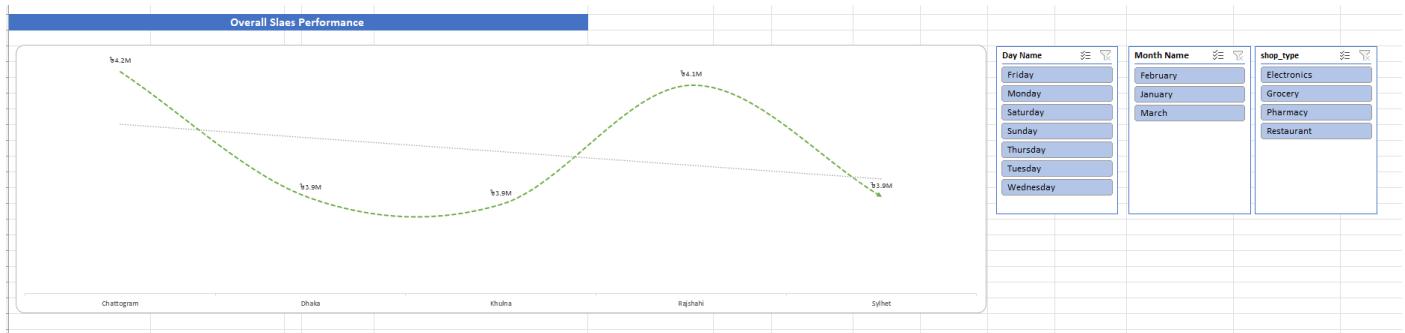
3. Create at least two Pivot Tables:

- Sales amount by city
- Subscription purchased count by shop type



4. Create one chart (bar/line/column your choice) that best explains overall business performance

(I also used interactive filter for charts will show data dynamically)



Section B SQL

Q1. Daily Performance Classification

- Show date
- Total sales amount
- Total transactions
- Classify activity type as good or bad or worst
- I define my own logic for classification based on sales or transactions.

SQL Query:

```

SELECT
    DATE(date) AS date,
    SUM(total_sales_amount) AS total_sales_amount,
    SUM(transactions_count) AS total_transactions,
    CASE
        WHEN SUM(total_sales_amount) >= 100000 AND SUM(transactions_count) >= 100 THEN 'Good'
        WHEN SUM(total_sales_amount) BETWEEN 50000 AND 99999 OR SUM(transactions_count) BETWEEN 50 AND 99 THEN 'Bad'
        ELSE 'Worst'
    END AS activity_type
FROM
    task_dataset
GROUP BY
    DATE(date)
ORDER BY
    2 desc;

```

Result Grid					Filter Rows:	Export:	Wrap Cell Content:
	date	total_sales_amount	total_transactions	activity_type			
▶	2024-03-11	318011	587	Good			
	2024-01-16	316404	813	Good			
	2024-03-15	308285	671	Good			
	2024-03-04	301413	639	Good			
	2024-01-14	290105	729	Good			
	2024-02-23	286493	790	Good			
	2024-03-05	282655	835	Good			
	2024-03-14	279381	621	Good			
Result 32							

Q2. New vs Existing User Analysis

- Writing a SQL query to compare new user's vs existing users in terms of:
- Total sales amount
- Number of shops
- Subscription purchases

SQL Query:

```
SELECT
CASE
WHEN new_user = 1 THEN 'New User'
ELSE 'Existing User'
END AS user_type,
FORMAT(SUM(total_sales_amount), 0) AS total_sales_amount,
COUNT(DISTINCT shop_id) AS number_of_shops,
SUM(subscription_purchased) AS subscription_purchases
FROM task_dataset
GROUP BY new_user;
select * from task_dataset;
-- Q3. Subscription Conversion by Shop Type
-- Writting a SQL query to calculate:
-- Subscription conversion rate by shop type
SELECT
shop_type,
COUNT(DISTINCT shop_id) AS total_shops,
SUM(subscription_purchased) AS subscription_purchased,
ROUND(
    (SUM(total_sales_amount)/ (SELECT SUM(total_sales_amount) FROM task_dataset)) * 100,
    2
) as sales_percentage,
ROUND(
    (SUM(subscription_purchased) / COUNT(DISTINCT shop_id)) * 100,
    2
) AS subscription_conversion_rate_pct
FROM task_dataset
GROUP BY shop_type;
```

Result Grid				
Filter Rows: <input type="text"/>				
	user_type	total_sales_amount	number_of_shops	subscription_purchases
▶	Existing User	14,313,584	200	271
	New User	5,769,975	189	139

Q3. Subscription Conversion by Shop Type

- Writing a SQL query to calculate:
- Subscription conversion rate by shop type

SQL Query:

```
SELECT
    shop_type,
    COUNT(DISTINCT shop_id) AS total_shops,
    SUM(subscription_purchased) AS subscription_purchased,
    ROUND(
        (SUM(total_sales_amount) / (SELECT SUM(total_sales_amount) FROM task_dataset)) * 100,
        2
    ) AS sales_percentage,
    ROUND(
        (SUM(subscription_purchased) / COUNT(DISTINCT shop_id)) * 100,
        2
    ) AS subscription_conversion_rate_pct
FROM task_dataset
GROUP BY shop_type;
```

Result Grid					
	shop_type	total_shops	subscription_purchased	sales_percentage	subscription_conversion_rate_pct
▶	Electronics	159	58	14.19	36.48
	Grocery	194	158	39.84	81.44
	Pharmacy	163	80	18.72	49.08
	Restaurant	184	114	27.25	61.96

Q4. City Performance

- Writing a SQL query to:
- Find the top 5 cities by total sales
- Only include cities where at least one subscription was purchased

```
SELECT
    city,
    FORMAT(SUM(total_sales_amount), 0) AS total_sales_amount,
    SUM(subscription_purchased) AS subscription_purchased
FROM
    task_dataset
GROUP BY 1
HAVING SUM(COALESCE(subscription_purchased, 0)) >= 1
ORDER BY 3 DESC
LIMIT 5;
```

Result Grid			
	city	total_sales_amount	subscription_purchased
▶	Chattogram	4,166,972	99
	Sylhet	3,930,634	88
	Dhaka	3,927,997	78
	Khulna	3,916,410	74
	Rajshahi	4,141,546	71

Section C Business Insights

1. Which shop type appears to be the most valuable for the business? Why?

Answers: The most valuable segment is definitely **Grocery** shop.

It's the clear winner because it's pulling the most weight for the company. Not only does it drive the biggest chunk of sales (nearly 40%), but the shop owners there are incredibly loyal—over 81% of them are actually signing up for subscriptions.

shop_type	total_shops	subscription_purchased	sales_percentage	subscription_conversion_rate_pct
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Restaurant	184	114	27.25	61.96

2. Do new users or existing users contribute more to total sales? What does this imply?

Answers: Existing users contribute way more to **total sales**.

It is implied that:

- **They spend much more:** Even though the number of shops is nearly the same, existing users bring in over **14 million**, while new users bring in less than **6 million**.
- **They are more loyal:** Existing users are buying twice as many **subscriptions** as the new guys.
- **The takeaway:** This implies that the business is great at keeping people around once they join, but **new users aren't yet convinced** to spend big or commit to a subscription.

user_type	total_sales_amount	number_of_shops	subscription_purchases
Existing User	14,313,584	200	271
New User	5,769,975	189	139

3. Suggest one business or product improvement based on the data.

We should be to turn **new users** into **regulars** faster.

Recommendations:

Offer a first-month subscription discount: This encourages the 189 new shops to sign up immediately, following the high subscription habit of existing users.