

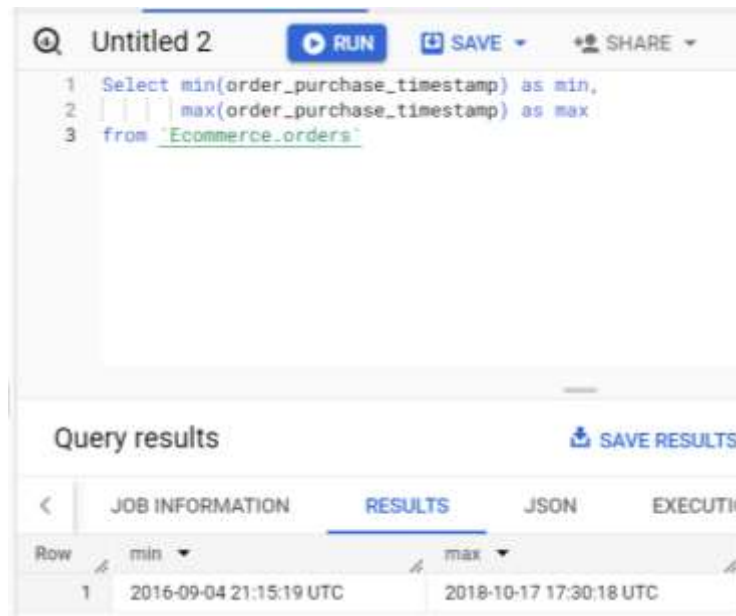
BUSINESS CASE

1. Import the dataset and do usual exploratory analysis steps like checking the structure & characteristics of the dataset:

1. Data type of all columns in the "customers" table.

| Field name | Type |
|--|---------|
| customer_id | STRING |
| customer_unique_id | STRING |
| customer_zip_code_prefix | INTEGER |
| customer_city | STRING |
| customer_state | STRING |

2. Get the time range between which the orders were placed.



Untitled 2 [RUN] [SAVE] [SHARE]

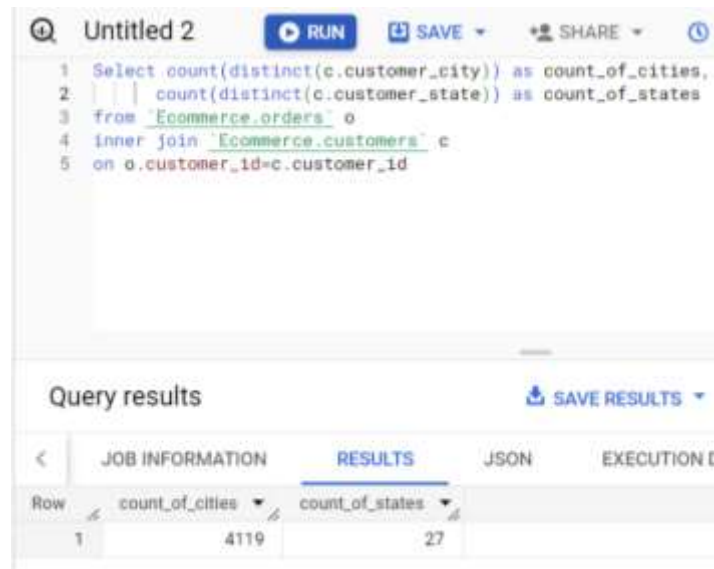
```
1 Select min(order_purchase_timestamp) as min,
2     max(order_purchase_timestamp) as max
3 from `Ecommerce.orders`
```

Query results [SAVE RESULTS]

| Row | min | max |
|-----|-------------------------|-------------------------|
| 1 | 2016-09-04 21:15:19 UTC | 2018-10-17 17:30:18 UTC |

From the above result, we can conclude that the **first** purchase of the order was made on **4th September 2016 (9:15 pm)** and the **last** purchase was made on **17th October 2018 (5:30 pm)**

- Count the Cities & States of customers who ordered during the given period.



```
1 Select count(distinct(c.customer_city)) as count_of_cities,
2 | count(distinct(c.customer_state)) as count_of_states
3 from 'Ecommerce.orders' o
4 inner join 'Ecommerce.customers' c
5 on o.customer_id=c.customer_id
```

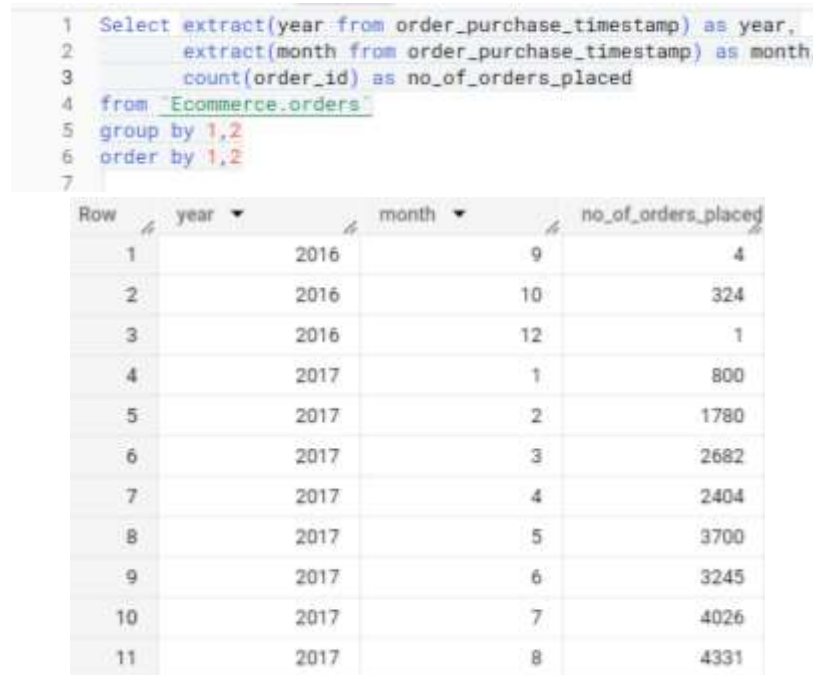
Query results

| Row | count_of_cities | count_of_states |
|-----|-----------------|-----------------|
| 1 | 4119 | 27 |

Customers from **4119 cities** across **27 states** have purchased in the given period

2. In-depth Exploration:

- Is there a growing trend in the no. of orders placed over the past years?



```
1 Select extract(year from order_purchase_timestamp) as year,
2 | extract(month from order_purchase_timestamp) as month,
3 count(order_id) as no_of_orders_placed
4 from 'Ecommerce.orders'
5 group by 1,2
6 order by 1,2
7
```

| Row | year | month | no_of_orders_placed |
|-----|------|-------|---------------------|
| 1 | 2016 | 9 | 4 |
| 2 | 2016 | 10 | 324 |
| 3 | 2016 | 12 | 1 |
| 4 | 2017 | 1 | 800 |
| 5 | 2017 | 2 | 1780 |
| 6 | 2017 | 3 | 2682 |
| 7 | 2017 | 4 | 2404 |
| 8 | 2017 | 5 | 3700 |
| 9 | 2017 | 6 | 3245 |
| 10 | 2017 | 7 | 4026 |
| 11 | 2017 | 8 | 4331 |

From the above result we can see that, during initial months of the Target setup there were lesser number of orders placed and later on it's increasing, but not exponentially. We observe that number of orders gets **stable around midyear of 2017**, we also observe that during the month of **November, January number of orders** are the **most** which may be due to

Black Friday sales and holiday season and the orders **decreased** drastically around **October 2018** which may indicate that Target was wrapping up their operations.

2. Can we see some kind of monthly seasonality in terms of the no. of orders being placed?

| Row | year | month | no_of_orders_placed |
|-----|------|-------|---------------------|
| 5 | 2017 | 2 | 1780 |
| 6 | 2017 | 3 | 2682 |
| 7 | 2017 | 4 | 2404 |
| 8 | 2017 | 5 | 3700 |
| 9 | 2017 | 6 | 3245 |
| 10 | 2017 | 7 | 4026 |
| 11 | 2017 | 8 | 4331 |
| 12 | 2017 | 9 | 4285 |
| 13 | 2017 | 10 | 4631 |
| 14 | 2017 | 11 | 7544 |
| 15 | 2017 | 12 | 5673 |
| 16 | 2018 | 1 | 7269 |
| 17 | 2018 | 2 | 6728 |

We see some kind of monthly seasonality where customers placed the more number of orders in **November 2017** which may be due to **black Friday sales** and in **January** due to **holiday season**. We also observe that number of orders are being more or less stable during the months of **july and august** of both years **2017 & 2018**

3. During what time of the day, do the Brazilian customers mostly place their orders? (Dawn, Morning, Afternoon or Night)
- 0-6 hrs : Dawn
 - 7-12 hrs : Mornings
 - 13-18 hrs : Afternoon
 - 19-23 hrs : Night

```

1 select case when hour between 0 and 6 then "Dawn"
2             when hour between 7 and 12 then "Mornings"
3             when hour between 13 and 18 then "Afternoon"
4             else "Night"
5             end as time_of_the_day,
6             count(order_id) as no_of_orders_placed
7 from (select extract(hour from order_purchase_timestamp) as hour, order_id
8 from `Ecommerce.orders`)
9 group by 1
10 order by 2

```

Query results [SAVE RESULTS](#) [EXPLORE DATA](#)

| < | JOB INFORMATION | RESULTS | JSON | EXECUTION DETAILS | CHAI |
|-----|-----------------|---------------------|------|-------------------|------|
| Row | time_of_the_day | no_of_orders_placed | | | |
| 1 | Dawn | 5242 | | | |
| 2 | Mornings | 27733 | | | |
| 3 | Night | 28331 | | | |
| 4 | Afternoon | 38135 | | | |

From the above result we observe that most of the orders were placed during **afternoon**(i.e,13-18hrs) time of the day

3. Evolution of E-commerce orders in the Brazil region:

1. Get the month on month no. of orders placed in each state.

```

1 select c.customer_state,
2        extract(year from o.order_purchase_timestamp) as year,
3        extract(month from o.order_purchase_timestamp) as month,
4        count(o.order_id) as no_of_orders_placed
5 from `Ecommerce.orders` o
6 inner join `Ecommerce.customers` c
7 on o.customer_id=c.customer_id
8 group by 1,2,3
9 order by 2,3,4 desc

```

| Row | customer_state | year | month | no_of_orders_placed |
|-----|----------------|------|-------|---------------------|
| 1 | SP | 2016 | 9 | 2 |
| 2 | RR | 2016 | 9 | 1 |
| 3 | RS | 2016 | 9 | 1 |
| 4 | SP | 2016 | 10 | 113 |
| 5 | RJ | 2016 | 10 | 56 |
| 6 | MG | 2016 | 10 | 40 |
| 7 | RS | 2016 | 10 | 24 |
| 8 | PR | 2016 | 10 | 19 |
| 9 | SC | 2016 | 10 | 11 |
| 10 | GO | 2016 | 10 | 9 |
| 11 | CE | 2016 | 10 | 8 |
| 12 | PE | 2016 | 10 | 7 |
| 13 | DF | 2016 | 10 | 6 |
| 14 | BA | 2016 | 10 | 4 |

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Thus, the number of orders placed across each state on month on month are given in the result. We also observe that number of orders placed in the **state SP** are **more** which provides an incentive to the Target to enhance their operations in a manner where the customers base can be retained and further increased in all the cities in the State SP by providing more discounts.

2. How are the customers distributed across all the states?

```

1 select customer_state,
2        count(customer_id) as no_of_customers
3 from `Ecommerce.customers`
4 group by 1
5 order by 2 desc

```

| Row | customer_state | no_of_customers |
|-----|----------------|-----------------|
| 1 | SP | 41746 |
| 2 | RJ | 12852 |
| 3 | MG | 11635 |
| 4 | RS | 5466 |
| 5 | PR | 5045 |
| 6 | SC | 3637 |
| 7 | BA | 3390 |
| 8 | DF | 2140 |
| 9 | ES | 2033 |
| 10 | GO | 2020 |
| 11 | PE | 1652 |
| 12 | CE | 1336 |
| 13 | PA | 975 |

Load more

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We observe that there are **more** number of customers from the states of **SP, RJ, MG, RS** and **less** number of customers from the states of **RR, AP, AC, AM**. With this given insights

we can enhance Target operations to increase the customer base across all the states by **providing better customer service and support**. It also helps us to find **targeted consumers**.

4. Impact on Economy: Analyze the money movement by e-commerce by looking at order prices, freight and others.

1. Get the % increase in the cost of orders from year 2017 to 2018 (include months between Jan to Aug only).

You can use the "payment value" column in the payments table to get the cost of orders.

```

1 select t.order_id,
2      current_price as cost_of_order,
3      round(((new_price-current_price)/current_price)*100,2) as
      increasing_cost_percentage
4 from (select order_id,
5      payment_value as current_price,
6      lead(payment_value,1) over(order by payment_value) as new_price
7 from `Ecommerce.payments`
8 where payment_value>0) t
9 inner join `Ecommerce.orders` o
10 on t.order_id=o.order_id
11 where extract(year from order_purchase_timestamp)>2016
12      and extract(month from order_purchase_timestamp) between 1 and 8
13 order by 2,3

```

| | JOB INFORMATION | RESULTS | JSON | EXECUTION DETAIL |
|-----|-------------------------------|---------------|----------------------|------------------|
| Row | order_id | cost_of_order | increasing_cost_perc | |
| 1 | 7db5f2eb8f5f54db9f9e71ba42... | 0.01 | 0.0 | |
| 2 | fb4de3600d359f84927517e78f... | 0.01 | 0.0 | |
| 3 | ca4b9f3ce6fc19e8533501cf8c... | 0.01 | 0.0 | |
| 4 | 25b5b0ea53b7d5a2d5712a0d... | 0.01 | 200.0 | |
| 5 | 4884bd0f5624b0b7919209656... | 0.03 | 0.0 | |
| 6 | d8c4293616878e2f015351d8f... | 0.05 | 0.0 | |
| 7 | 2a4cf847d8e23123b4421ef7f7... | 0.05 | 40.0 | |
| 8 | 874904b7cdcd377a298d82e30... | 0.08 | 12.5 | |
| 9 | 2122719f6fc72fba16b0b3001d... | 0.09 | 11.11 | |
| 10 | c96a303a722b89bc14a23b7a2... | 0.1 | 10.0 | |
| 11 | 33e70b1305003580be0b3447... | 0.11 | 0.0 | |
| 12 | 4fb76fa13b108a0d047848342... | 0.13 | 0.0 | |
| 13 | 9cd6015fc7c3321b8cfc4459af... | 0.13 | 7.69 | |
| 14 | dfc5e1ae34c20f47a8a88a2a5... | 0.14 | 0.0 | |

We observe that increasing percentage in the cost of orders in the time period between 2017 and 2018(jan to aug)

2. Calculate the Total & Average value of order price for each state.

```

1  select c.customer_state,
2         round(sum(o1.price),2) as total_order_price,
3         round(avg(o1.price),2) as average_order_price
4  from `Ecommerce.customers` c
5  inner join `Ecommerce.orders` o
6  on c.customer_id=o.customer_id
7  inner join `Ecommerce.order_items` o1
8  on o.order_id= o1.order_id
9  group by 1
10 order by 2 desc, 3 desc

```

| Row | customer_state | total_order_price | average_order_price |
|-----|----------------|-------------------|---------------------|
| 1 | SP | 5202955.05 | 109.65 |
| 2 | RJ | 1824092.67 | 125.12 |
| 3 | MG | 1585308.03 | 120.75 |
| 4 | RS | 750304.02 | 120.34 |
| 5 | PR | 683083.76 | 119.0 |
| 6 | SC | 520553.34 | 124.65 |
| 7 | BA | 511349.99 | 134.6 |
| 8 | DF | 302603.94 | 125.77 |
| 9 | GO | 294591.95 | 126.27 |
| 10 | ES | 275037.31 | 121.91 |
| 11 | PE | 262788.03 | 145.51 |
| 12 | CE | 227254.71 | 153.76 |
| 13 | PA | 178947.81 | 165.69 |

Load more

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From the above result, we observe that states like **SP, RJ, MG** have highest total order price as there were huge number of orders placed in these states. It gives huge opportunity for Target to enhance their operations in these states to increase their revenue.

3. Calculate the Total & Average value of order freight for each state.

```

1  select c.customer_state,
2         round(sum(o1.freight_value),2) as total_freight_value,
3         round(avg(o1.freight_value),2) as average_freight_value
4  from `Ecommerce.customers` c
5  inner join `Ecommerce.orders` o
6  on c.customer_id=o.customer_id
7  inner join `Ecommerce.order_items` o1
8  on o.order_id= o1.order_id
9  group by 1
10 order by 2 desc, 3 desc

```

| Row | customer_state | total_freight_value | average_freight_value |
|-----|----------------|---------------------|-----------------------|
| 1 | SP | 718723.07 | 15.15 |
| 2 | RJ | 305589.31 | 20.96 |
| 3 | MG | 270853.46 | 20.63 |
| 4 | RS | 135522.74 | 21.74 |
| 5 | PR | 117851.68 | 20.53 |
| 6 | BA | 100156.68 | 26.36 |
| 7 | SC | 89660.26 | 21.47 |
| 8 | PE | 59449.66 | 32.92 |
| 9 | GO | 53114.98 | 22.77 |
| 10 | DF | 50625.5 | 21.04 |
| 11 | ES | 49764.6 | 22.06 |
| 12 | CE | 48351.59 | 32.71 |
| 13 | PA | 38699.3 | 35.83 |

Load more

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From the above result, we observe, as the orders are more from states like SP, RJ, MG...their freight value is on top. This provides us to **optimize logistics cost**.

5. Analysis based on sales, freight and delivery time.

1. Find the no. of days taken to deliver each order from the order's purchase date as delivery time.

Also, calculate the difference (in days) between the estimated & actual delivery date of an order.

Do this in a single query.

You can calculate the delivery time and the difference between the estimated & actual delivery date using the given formula:

- **time_to_deliver** = order_delivered_customer_date - order_purchase_timestamp
- **diff_estimated_delivery** = order_estimated_delivery_date - order_delivered_customer_date

```

1 select order_id,
2        date_diff(order_delivered_customer_date,order_purchase_timestamp,day) as
   delivery_time,
3        date_diff(order_estimated_delivery_date,order_delivered_customer_date,day) as
   diff_estimated_delivery
4 from 'Ecommerce.orders'
5 order by 1,2,3
6

```


| Row | order_id | days_to_deliver | diff_estimated_deliv |
|-----|-------------------------------|-----------------|----------------------|
| 1 | 00010242fe8c5a6d1ba2dd792... | 7 | 8 |
| 2 | 00018f77f2f0320c557190d7a1... | 16 | 2 |
| 3 | 000229ec398224ef6ca0657da... | 7 | 13 |
| 4 | 00024acbcd0a6daa1e931b03... | 6 | 5 |
| 5 | 00042b26cf59d7ce69dfabb4e... | 25 | 15 |
| 6 | 00048cc3ae777c65dbb7d2a06... | 6 | 14 |
| 7 | 00054e8431b9d7675808bcb8... | 8 | 16 |
| 8 | 000576fe39319847cbb9d288c... | 5 | 15 |
| 9 | 0005a1a1728c9d785b8e2b08b... | 9 | 0 |
| 10 | 0005f50442cb953dcd1d21e1f... | 2 | 18 |

From the above result, it provides us with an insight to reduce delivery time on orders where it takes more number of days to deliver, which in turn increase the sales of Target provided with better customer service and satisfaction.

- Find out the top 5 states with the highest & lowest average freight value.

```

1 select c.customer_state,
2        round(avg(oi.freight_value),2) as average_freight_value
3 from   Ecommerce_customers c
4 inner join Ecommerce_orders o
5 on c.customer_id=o.customer_id
6 inner join Ecommerce_order_items oi
7 on o.order_id= oi.order_id
8 group by 1
9 order by 2 desc
10 limit 5

```

| Row | customer_state | average_freight_valu |
|-----|----------------|----------------------|
| 1 | RR | 42.98 |
| 2 | PB | 42.72 |
| 3 | RO | 41.07 |
| 4 | AC | 40.07 |
| 5 | PI | 39.15 |

| Row | customer_state | average_freight_valu |
|-----|----------------|----------------------|
| 1 | SP | 15.15 |
| 2 | PR | 20.53 |
| 3 | MG | 20.63 |
| 4 | RJ | 20.96 |
| 5 | DF | 21.04 |

It provides us with states having highest and lowest average freight value. It helps to focus on these states to enhance their operations by optimizing the freight cost.

3. Find out the top 5 states with the highest & lowest average delivery time.

```
1 select c.customer_state,
2        round(avg(date_diff(order_delivered_customer_date,order_purchase_timestamp,day)),
3              2) as avg_delivery_time
4 from 'Ecommerce.orders' o
5 inner join 'Ecommerce.customers' c
6 on o.customer_id=c.customer_id
7 group by 1
8 order by 2 desc
9 limit 5
10
11
```

Query results [SAVE RESULTS](#) [EXPLORE DATA](#)

| Row | customer_state | avg_delivery_time |
|-----|----------------|-------------------|
| 1 | RR | 28.98 |
| 2 | AP | 26.73 |
| 3 | AM | 25.99 |
| 4 | AL | 24.04 |
| 5 | PA | 23.32 |

```
1 select c.customer_state,
2        round(avg(date_diff(order_delivered_customer_date,order_purchase_timestamp,day)),
3              2) as avg_delivery_time
4 from 'Ecommerce.orders' o
5 inner join 'Ecommerce.customers' c
6 on o.customer_id=c.customer_id
7 group by 1
8 order by 2
9 limit 5
10
11
```

Query results [SAVE RESULTS](#) [EXPLORE DATA](#)

| Row | customer_state | avg_delivery_time |
|-----|----------------|-------------------|
| 1 | SP | 8.3 |
| 2 | PR | 11.53 |
| 3 | MG | 11.54 |
| 4 | DF | 12.51 |
| 5 | SC | 14.48 |

It provides us with states having highest and lowest delivery time, which helps Target to focus on states with higher delivery time, where it can either **localize warehousing** near these states or **outsource it to third party for better service**.

4. Find out the top 5 states where the order delivery is really fast as compared to the estimated date of delivery.

You can use the difference between the averages of actual & estimated delivery date to figure out how fast the delivery was for each state.

```
1 select c.customer_state,
2 round(avg(date_diff(order_estimated_delivery_date,order_delivered_customer_date,
3 day)),2) as avg_diff_estimated_delivery
4 from `Ecommerce.orders` o
5 inner join `Ecommerce.customers` c
6 on o.customer_id=c.customer_id
7 group by 1
8 order by 2 desc
9 limit 5
```

| Row | customer_state | avg_diff_estimated_delivery |
|-----|----------------|-----------------------------|
| 1 | AC | 19.76 |
| 2 | RO | 19.13 |
| 3 | AP | 18.73 |
| 4 | AM | 18.61 |
| 5 | RR | 16.41 |

In these states, delivery is fast compared to estimated date of delivery.

6. Analysis based on the payments:

1. Find the month on month no. of orders placed using different payment types.

```
1 select p.payment_type,
2 extract(year from o.order_purchase_timestamp) as year,
3 extract(month from o.order_purchase_timestamp) as month,
4 count(o.order_id) as no_of_orders_placed
5 from `Ecommerce.orders` o
6 inner join `Ecommerce.payments` p
7 on o.order_id=p.order_id
8 group by 1,2,3
9 order by 1,2,3
```

| Row | payment_type | year | month | no_of_orders_placed |
|-----|--------------|------|-------|---------------------|
| 1 | UPI | 2016 | 10 | 63 |
| 2 | UPI | 2017 | 1 | 197 |
| 3 | UPI | 2017 | 2 | 398 |
| 4 | UPI | 2017 | 3 | 590 |
| 5 | UPI | 2017 | 4 | 496 |
| 6 | UPI | 2017 | 5 | 772 |
| 7 | UPI | 2017 | 6 | 707 |
| 8 | UPI | 2017 | 7 | 845 |
| 9 | UPI | 2017 | 8 | 938 |
| 10 | UPI | 2017 | 9 | 903 |
| 11 | UPI | 2017 | 10 | 993 |
| 12 | UPI | 2017 | 11 | 1509 |
| 13 | UPI | 2017 | 12 | 1160 |
| 14 | UPI | 2018 | 1 | 1518 |

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From the above result, it provides us with number of orders placed using different payments mode like debit card, credit card, UPI, and voucher. We observe that huge number of orders placed using UPI on the months of November, January.

2. Find the no. of orders placed on the basis of the payment installments that have been paid.

```

select p.payment_installments,
       count(o.order_id) as no_of_orders_placed
from `Ecommerce.orders` o
inner join `Ecommerce.payments` p
on o.order_id = p.order_id
where p.payment_installments >=1
group by 1
order by 1

```

| Row | payment_installment | no_of_orders_placed |
|-----|---------------------|---------------------|
| 1 | 1 | 52546 |
| 2 | 2 | 12413 |
| 3 | 3 | 10461 |
| 4 | 4 | 7098 |
| 5 | 5 | 5239 |
| 6 | 6 | 3920 |
| 7 | 7 | 1626 |
| 8 | 8 | 4268 |
| 9 | 9 | 644 |
| 10 | 10 | 5328 |
| 11 | 11 | 23 |
| 12 | 12 | 133 |
| 13 | 13 | 16 |

Load more

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These are the number of orders placed on the basis of the payment installments that have been paid. We see that huge number of orders are placed where customers have paid on first payment installment

7.Actionable Insights & Recommendations

The following recommendations are given to optimize the sales of Target:

- Data analysis have provided with better opportunity to know who the customers and from which location which helps us to develop buyer persona.
- To improve the brand awareness of Target, focus on quality of products and improve marketing tactics to increase the customer base, where the insights provided in this business case can help to know our targeted consumers
- Delivery time of products can be optimized by providing warehouses near certain locations where the customer base is more. It can also be outsourced for efficiency.
- Providing better customer support service can make them re-engaged to buy even more products.
- Can make use of social media (digital marketing) to drive new visitors to the store and make them into paying consumers in the states where there are lower number of orders placed
- Freight cost can be optimized by outsourcing transportation