# **Business Case: Target SQL**

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- Setup: Big Query

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

# 1. Data type of columns in a table

Here I am returning all the columns of the **orders** table as it is more important table in our dataset.

### Query:

```
SELECT column_name, data_type
FROM `divine-bonbon-381109`.Project_1.INFORMATION_SCHEMA.COLUMNS
WHERE TABLE_NAME = "orders"
```

| Row | column_name                   | data_type |
|-----|-------------------------------|-----------|
| 1   | order_id                      | STRING    |
| 2   | customer_id                   | STRING    |
| 3   | order_status                  | STRING    |
| 4   | order_purchase_timestamp      | TIMESTAMP |
| 5   | order_approved_at             | TIMESTAMP |
| 6   | order_delivered_carrier_date  | TIMESTAMP |
| 7   | order_delivered_customer_date | TIMESTAMP |
| 8   | order_estimated_delivery_date | TIMESTAMP |

1. There are only 2 types of datatypes in orders table they are **STRING** and **TIMESTAMP**.

# 2. Time period for which the data is given

We can get the time period of the data by using **orders** table and column **order\_purchase\_timestamp**, **order\_delivered\_carrier\_date**, **order\_delivered\_customer\_date**.

```
SELECT MIN(o.order_purchase_timestamp)
min_order_purchase_timestamp,
MAX(o.order_purchase_timestamp)
max_order_purchase_timestamp,

MIN(o.order_delivered_carrier_date)
min_delivered_carrier_date,
MAX(o.order_delivered_carrier_date)
max_delivered_carrier_date,

MIN(o.order_delivered_customer_date)
min_delivered_customer_purchase_date,
MAX(o.order_delivered_customer_date)
max_delivered_customer_purchase_date

FROM `Project_1.orders` o
```

| Row | min_order_purchase_timestamp    | max_order_purchase_timestamp    |
|-----|---------------------------------|---------------------------------|
| 1   | 2016-09-04 21:15:19 UTC         | 2018-10-17 17:30:18 UTC         |
|     |                                 |                                 |
| Row | min_delivered_carrier_date      | max_delivered_carrier_date      |
| 1   | 2016-10-08 10:34:01 UTC         | 2018-09-11 19:48:28 UTC         |
|     |                                 |                                 |
|     |                                 |                                 |
| Row | min_delivered_customer_purchase | max_delivered_customer_purchase |
| 1   | 2016-10-11 13:46:32 UTC         | 2018-10-17 13:22:46 UTC         |

#### Observations:

- 1. We can find min date in the dataset is 2016-09-04.
- 2. We can find max date in the dataset is 2018-10-17
- 3. The date difference between two dates is **774** days, which is almost equal to two years.

# 3. Cities and States of customers ordered during the given period

From the schema I can see that orders table, geolocation is connected by customers table. So we need three tables in the query.

#### Query:

```
SELECT count(distinct geo.geolocation_city) as count_of_cities,
count(distinct geo.geolocation_state) as count_of_states

FROM `Project_1.orders` ord

JOIN `Project_1.customers` cus ON ord.customer_id =
cus.customer_id

JOIN `Project_1.geolocation` geo ON
geo.geolocation_zip_code_prefix = cus.customer_zip_code_prefix
```



```
1. Total cities = 5812
2. Total states = 27
```

# 2. In-depth Exploration:

1. Is there a growing trend on e-commerce in Brazil? How can we describe a complete scenario? Can we see some seasonality with peaks at specific month

We have to find number of years for each month and finds insights from the data.

#### Query:

```
SELECT order_final.year_num year, order_final.month_num month, count(*) no_of_orders

FROM

(SELECT

ord.order_purchase_timestamp, EXTRACT(MONTH FROM (CAST(ord.order_purchase_timestamp AS DATE))) month_num, EXTRACT(YEAR FROM (CAST(ord.order_purchase_timestamp AS DATE))) year_num

FROM Project_1.orders ord) AS order_final

group by order_final.year_num, order_final.month_num

order by order_final.year_num, order_final.month_num
```

| Row | year // | month // | no_of_orders |
|-----|---------|----------|--------------|
| 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         |
| 12  | 2017    | 9        | 4285         |
| 13  | 2017    | 10       | 4631         |
| 14  | 2017    | 11       | 7544         |
| 15  | 2017    | 12       | 5673         |
| 16  | 2018    | 1        | 7269         |
| 17  | 2018    | 2        | 6728         |
| 18  | 2018    | 3        | 7211         |
| 19  | 2018    | 4        | 6939         |
| 20  | 2018    | 5        | 6873         |
|     |         |          |              |

- 1. There is no trend in the orders sometimes they are increasing sometimes they are decreasing.
- 2 What time do Brazilian customers tend to buy (Dawn, Morning, Afternoon or Night)?

We have to divide hours into Dawn, Morning, Afternoon and Night, based on the hour they purchased time. Then we should count number of customers bought in Dawn, Morning, Afternoon and Night.

```
SELECT orders_final.moment,
count(orders_final.order_purchase_timestamp) count_of_orders
```

```
FROM
     (SELECT
     order_sub.order_purchase_timestamp, order_sub.hour,
     CASE
         WHEN hour >= 0 AND hour <= 7 THEN 'dawn'
         WHEN hour > 7 AND hour <= 12 THEN 'morning'
         WHEN hour > 12 AND hour <= 17 THEN 'after'
         WHEN hour > 17
                                        THEN 'night'
     END moment
     FROM
     (SELECT
     ord.order_purchase_timestamp, EXTRACT(HOUR FROM
(CAST(ord.order_purchase_timestamp AS DATETIME))) hour
     FROM Project_1.orders ord) AS order_sub) AS orders_final
     GROUP BY orders final.moment
```

| Row | moment  | // | count_of_orders |
|-----|---------|----|-----------------|
| 1   | morning |    | 26502           |
| 2   | dawn    |    | 6473            |
| 3   | after   |    | 32366           |
| 4   | night   |    | 34100           |

- 1. Most of the orders are in the Night.
- 2. I think we can decrease the employee force at Dawn.

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

# 1. Get month on month orders by states

First we have to extract month from timestamp given, then we should combine tables orders, customers, geolocation and we should count the order by state and monthe.

Query:

```
SELECT geolocation_state, month_num, count(order_id) orders
FROM

(SELECT ord.order_id, EXTRACT(MONTH FROM
    (CAST(ord.order_purchase_timestamp AS DATE))) month_num,
    geo.geolocation_state

FROM `Project_1.orders` ord

JOIN `Project_1.customers` cus ON ord.customer_id =
    cus.customer_id

JOIN `Project_1.geolocation` geo ON
    geo.geolocation_zip_code_prefix = cus.customer_zip_code_prefix)
    as final_table

group by 1, 2
    order by orders DESC
```

| Row | geolocation_state | month_num | orders |
|-----|-------------------|-----------|--------|
| 1   | SP                | 8         | 660764 |
| 2   | SP                | 5         | 617152 |
| 3   | SP                | 7         | 596542 |
| 4   | SP                | 6         | 550147 |
| 5   | SP                | 3         | 540963 |
| 6   | SP                | 4         | 522895 |
| 7   | SP                | 2         | 460265 |
| 8   | SP                | 1         | 453515 |
| 9   | SP                | 11        | 412132 |
| 10  | SP                | 12        | 325198 |
| 11  | RJ                | 5         | 310453 |
| 12  | RJ                | 8         | 308365 |
| 13  | MG                | 3         | 307366 |
| 14  | RJ                | 3         | 301419 |
| 15  | RJ                | 7         | 298006 |
| 16  | MG                | 8         | 293873 |
| 17  | MG                | 5         | 292339 |

- 1. Most of the orders are from SP state.
- 2. So we should not lose customers from that state, we should keep running customer retention process for this state.

#### 2. Distribution of customers across the states in Brazil

To get this, first we should combine tables customers and geolocation tables ang get count of customers for each state.

#### Query:

#### SELECT

```
geo.geolocation_state, count(cus.customer_id) count_of_customers
FROM `Project_1.customers` cus
```

JOIN Project\_1.geolocation geo ON cus.customer\_zip\_code\_prefix =
geo.geolocation\_zip\_code\_prefix

GROUP BY geo.geolocation\_state

ORDER BY count\_of\_customers DESC

# Output:

| Row | geolocation_state | count_of_customers |
|-----|-------------------|--------------------|
| 1   | SP                | 5620430            |
| 2   | RJ                | 3015690            |
| 3   | MG                | 2878728            |
| 4   | RS                | 805370             |
| 5   | PR                | 626021             |
| 6   | SC                | 538638             |
| 7   | BA                | 365875             |
| 8   | ES                | 316654             |
| 9   | GO                | 133146             |
| 10  | MT                | 122395             |
| 11  | PE                | 114588             |
| 12  | DF                | 93309              |
| 13  | PA                | 83554              |
| 14  | CE                | 63507              |
| 15  | MS                | 61473              |
| 16  | MA                | 53383              |
| 17  | AL                | 34861              |
| 18  | PB                | 27714              |

## Observations:

1. Most of the customers are also from SP state, so we should consider this state as important state for our business.

- 4. Impact on Economy: Analyze the money movement by e-commerce by looking at order prices, freight and others.
- 1 Get % increase in cost of orders from 2017 to 2018 (include months between Jan to Aug only) You can use "payment\_value" column in payments table

First get sum of payment\_value in 2017  $\rightarrow$  1 and get sum of payment\_value in 2018  $\rightarrow$  2. Divide 2 by 1 and multiply by 100 to get percentage increase in cost of orders from 2017 to 2018.

```
SELECT ROUND((SUM(payments_2018) / SUM(payments_2017)) * 100, 2)
percentage_increase_from_2017_to_2018
FROM
(SELECT
CASE
 WHEN EXTRACT(YEAR FROM (CAST(ord.order_purchase_timestamp AS
DATETIME))) = 2017
       AND EXTRACT(MONTH FROM (CAST(ord.order_purchase_timestamp
AS DATETIME))) <= 7
 THEN pay.payment_value
 ELSE 0
END AS payments_2017,
CASE
 WHEN EXTRACT(YEAR FROM (CAST(ord.order_purchase_timestamp AS
DATETIME))) = 2018
       AND EXTRACT(MONTH FROM (CAST(ord.order_purchase_timestamp
AS DATETIME))) <= 7
 THEN pay.payment_value
```

```
Row percentage_increase_from_2017_to_2018

1 256.2
```

#### Observations:

- 1. There is 256% increase in sales compared to 2017, this is a good sign for business, so we should continue it.
- 2. Mean & Sum of price and freight value by customer state

Combine tables customers, geolocation, order\_items get price, freight from customer state. Mean and sum them group by state.

# Query:

#### SELECT

```
geo.geolocation_state state, ROUND(SUM(ord_ite.price), 2)
sum_price, ROUND(SUM(ord_ite.freight_value), 2) sum_freight,
ROUND(SUM(ord_ite.price) / count(cus.customer_id), 2)
average_price, ROUND(SUM(ord_ite.freight_value) /
count(cus.customer_id), 2) average_freight
FROM `Project_1.customers` cus

JOIN `Project_1.geolocation` geo ON
geo.geolocation_zip_code_prefix = customer_zip_code_prefix

JOIN Project_1.orders ord ON ord.customer_id = cus.customer_id
```

```
JOIN `Project_1.order_items` ord_ite ON ord.order_id =
ord_ite.order_id
```

GROUP BY geo.geolocation\_state

ORDER BY sum\_price DESC, sum\_freight DESC, average\_price DESC, average\_freight DESC

## Output:

| Row | state // | sum_price   | sum_freight // | average_price | average_freight |
|-----|----------|-------------|----------------|---------------|-----------------|
| 1   | SP       | 711838740   | 98574572.43    | 111.28        | 15.41           |
| 2   | RJ       | 440142503   | 71966793.75    | 127.81        | 20.9            |
| 3   | MG       | 397190155   | 67058347.09    | 121.18        | 20.46           |
| 4   | RS       | 111183139   | 19910834.35    | 120.18        | 21.52           |
| 5   | PR       | 85392469.28 | 14432159.77    | 119.21        | 20.15           |
| 6   | SC       | 79666423.29 | 13472314.62    | 127.41        | 21.55           |
| 7   | BA       | 62377311.67 | 11345094.0     | 149.64        | 27.22           |
| 8   | ES       | 43634878.56 | 7799979.09     | 123.36        | 22.05           |
| 9   | MT       | 22777072.82 | 4177068.03     | 156.63        | 28.72           |
| 10  | G0       | 20860945.92 | 3590268.56     | 134.62        | 23.17           |
| 11  | PE       | 17545068.94 | 4195977.72     | 137.42        | 32.87           |
| 12  | PA       | 15586180.17 | 3409472.09     | 166.98        | 36.53           |
| 13  | DF       | 13141649.62 | 2214955.55     | 124.66        | 21.01           |
| 14  | CE       | 10819201.81 | 2306600.06     | 151.32        | 32.26           |
| 15  | MS       | 9891112.52  | 1698977.52     | 139.1         | 23.89           |
| 16  | MA       | 9020091.01  | 2275191.86     | 150.95        | 38.08           |
| 17  | AL       | 7191886.1   | 1237356.22     | 196.64        | 33.83           |
| 18  | PB       | 6278650.25  | 1350462.24     | 198.86        | 42.77           |

#### Observations:

1. We should decrease the price and freight from these values, we should come up with some logic, so that this should get decrease.

# 5. Analysis on sales, freight and delivery time

1. Calculate days between purchasing, delivering and estimated delivery

Here I am going to calculate average purchasing, delivery and estimated delivery time as question is not making a lot of sense, <- This I am going to do using orders table.

Note: Here I am considering order\_delivered\_customer\_date as delivering date.

#### Query:

```
SELECT ROUND(SUM(date_diff_delivery) / COUNT(order_id), 2) as avg_delivery_time, ROUND(SUM(date_diff_estd_delivery) / COUNT(order_id), 2) avg_estd_delivery_time

FROM

(SELECT ord.order_id, ord.order_purchase_timestamp, ord.order_delivered_customer_date, ord.order_estimated_delivery_date,

DATE_DIFF(CAST(ord.order_delivered_customer_date as DATETIME), CAST(ord.order_purchase_timestamp as DATETIME), DAY) date_diff_delivery,

DATE_DIFF(CAST(ord.order_estimated_delivery_date as DATETIME), CAST(ord.order_delivered_customer_date as DATETIME), CAST(ord.order_delivered_customer_date as DATETIME), DAY) date_diff_estd_delivery

FROM Project_1.orders ord) AS final_table
```

### Output:



- 2. Find time\_to\_delivery & diff\_estimated\_delivery. Formula for the same given below:
  - time\_to\_delivery = order\_purchase\_timestamp-order\_delivered\_customer\_date
  - diff\_estimated\_delivery = order\_estimated\_delivery\_date-order\_delivered\_customer\_date

```
SELECT ord.order_id, ord.order_purchase_timestamp,
ord.order_delivered_customer_date,
ord.order_estimated_delivery_date,

DATE_DIFF(CAST(ord.order_delivered_customer_date as DATETIME),
CAST(ord.order_purchase_timestamp as DATETIME), DAY)
time_to_delivery ,

DATE_DIFF(CAST(ord.order_estimated_delivery_date as DATETIME),
CAST(ord.order_delivered_customer_date as DATETIME), DAY)
diff_estimated_delivery

FROM Project_1.orders ord

ORDER BY time_to_delivery DESC, diff_estimated_delivery DESC
```

| Row | order_id //                      | order_purchase_timestamp | order_delivered_customer_date | order_estimated_delivery_date | time_to_delivery | diff_estimated_delivery |
|-----|----------------------------------|--------------------------|-------------------------------|-------------------------------|------------------|-------------------------|
| 1   | ca07593549f1816d26a572e06dc1eab6 | 2017-02-21 23:31:27 UTC  | 2017-09-19 14:36:39 UTC       | 2017-03-22 00:00:00 UTC       | 210              | -181                    |
| 2   | 1b3190b2dfa9d789e1f14c05b647a14a | 2018-02-23 14:57:35 UTC  | 2018-09-19 23:24:07 UTC       | 2018-03-15 00:00:00 UTC       | 208              | -188                    |
| 3   | 440d0d17af552815d15a9e41abe49359 | 2017-03-07 23:59:51 UTC  | 2017-09-19 15:12:50 UTC       | 2017-04-07 00:00:00 UTC       | 196              | -165                    |
| 4   | 2fb597c2f772eca01b1f5c561bf6cc7b | 2017-03-08 18:09:02 UTC  | 2017-09-19 14:33:17 UTC       | 2017-04-17 00:00:00 UTC       | 195              | -155                    |
| 5   | 285ab9426d6982034523a855f55a885e | 2017-03-08 22:47:40 UTC  | 2017-09-19 14:00:04 UTC       | 2017-04-06 00:00:00 UTC       | 195              | -166                    |
| 6   | 0f4519c5f1c541ddec9f21b3bddd533a | 2017-03-09 13:26:57 UTC  | 2017-09-19 14:38:21 UTC       | 2017-04-11 00:00:00 UTC       | 194              | -161                    |
| 7   | 47b40429ed8cce3aee9199792275433f | 2018-01-03 09:44:01 UTC  | 2018-07-13 20:51:31 UTC       | 2018-01-19 00:00:00 UTC       | 191              | -175                    |
| 8   | 2fe324febf907e3ea3f2aa9650869fa5 | 2017-03-13 20:17:10 UTC  | 2017-09-19 17:00:07 UTC       | 2017-04-05 00:00:00 UTC       | 190              | -167                    |
| 9   | 2d7561026d542c8dbd8f0daeadf67a43 | 2017-03-15 11:24:27 UTC  | 2017-09-19 14:38:18 UTC       | 2017-04-13 00:00:00 UTC       | 188              | -159                    |

3 Group data by state, take mean of freight\_value, time\_to\_delivery, diff\_estimated\_delivery

We can do this easily this if we have 2nd query, we just have to take mean of freight value, time\_to\_delivery, diff\_estimated\_delivery by group by of state.

#### Query:

#### SELECT

```
geo.geolocation_state state, ROUND(SUM(ord_ite.freight_value)/
COUNT(ord.order_id), 2) mean_freight,

ROUND(SUM(ord.time_to_delivery)/ COUNT(ord.order_id), 2)
mean_time_to_delivery,
```

```
ROUND(SUM(ord.diff_estimated_delivery)/ COUNT(ord.order_id), 2)
mean_diff_estimated_delivery
FROM
(SELECT ord.order_id, ord.customer_id,
DATE_DIFF(CAST(ord.order_delivered_customer_date as DATETIME),
CAST(ord.order_purchase_timestamp as DATETIME), DAY)
time_to_delivery ,
DATE_DIFF(CAST(ord.order_estimated_delivery_date as DATETIME),
CAST(ord.order_delivered_customer_date as DATETIME), DAY)
diff_estimated_delivery
FROM Project_1.orders ord) as ord
JOIN Project_1.order_items ord_ite ON ord_ite.order_id =
ord.order_id
JOIN `Project_1.customers` cus ON cus.customer_id =
ord.customer_id
JOIN Project_1.geolocation geo ON cus.customer_zip_code_prefix =
geo.geolocation_zip_code_prefix
GROUP BY geo.geolocation_state
ORDER BY mean_freight DESC, mean_time_to_delivery DESC,
mean_diff_estimated_delivery DESC
```

| Row | state | mean_freight // | mean_time_to_delivery | mean_diff_estimated_delivery |
|-----|-------|-----------------|-----------------------|------------------------------|
| 1   | PB    | 42.77           | 19.83                 | 13.1                         |
| 2   | RR    | 42.47           | 20.77                 | 18.59                        |
| 3   | PI    | 39.48           | 17.7                  | 12.0                         |
| 4   | AC    | 39.1            | 19.98                 | 19.14                        |
| 5   | MA    | 38.08           | 20.7                  | 9.53                         |
| 6   | RO    | 37.43           | 18.52                 | 19.56                        |
| 7   | ТО    | 37.36           | 16.3                  | 12.13                        |
| 8   | PA    | 36.53           | 22.58                 | 14.12                        |
| 9   | AP    | 35.66           | 30.11                 | 16.23                        |
| 10  | SE    | 34.67           | 21.12                 | 9.37                         |
| 11  | AM    | 34.62           | 24.62                 | 21.41                        |
| 12  | RN    | 34.07           | 18.56                 | 13.89                        |

4 Sort the data to get the following:

5 Top 5 states with highest/lowest average freight value - sort in desc/asc limit 5 If we have solution 3, just order by average freight value and use desc/asc

DESC/ Highest average freight values top 5:

#### Query:

#### SELECT

```
geo.geolocation_state state, ROUND(SUM(ord_ite.freight_value)/
COUNT(ord.order_id), 2) mean_freight,

ROUND(SUM(ord.time_to_delivery)/ COUNT(ord.order_id), 2)
mean_time_to_delivery,

ROUND(SUM(ord.diff_estimated_delivery)/ COUNT(ord.order_id), 2)
mean_diff_estimated_delivery

FROM

(SELECT_ord.order_id, ord.customer_id,
```

```
DATE_DIFF(CAST(ord.order_delivered_customer_date as DATETIME),
CAST(ord.order_purchase_timestamp as DATETIME), DAY)
time_to_delivery ,
DATE_DIFF(CAST(ord.order_estimated_delivery_date as DATETIME),
CAST(ord.order_delivered_customer_date as DATETIME), DAY)
diff_estimated_delivery
FROM Project_1.orders ord) as ord
JOIN Project_1.order_items ord_ite ON ord_ite.order_id =
ord.order_id
JOIN `Project_1.customers` cus ON cus.customer_id =
ord.customer_id
JOIN Project_1.geolocation geo ON cus.customer_zip_code_prefix =
geo.geolocation_zip_code_prefix
GROUP BY geo.geolocation_state
ORDER BY mean_freight DESC
LIMIT 5
```

| Row | state // | mean_freight // | mean_time_to_delivery | mean_diff_estimated_delivery |
|-----|----------|-----------------|-----------------------|------------------------------|
| 1   | PB       | 42.77           | 19.83                 | 13.1                         |
| 2   | RR       | 42.47           | 20.77                 | 18.59                        |
| 3   | PI       | 39.48           | 17.7                  | 12.0                         |
| 4   | AC       | 39.1            | 19.98                 | 19.14                        |
| 5   | MA       | 38.08           | 20.7                  | 9.53                         |

#### ASC/ Lowest average freight values top 5:

```
SELECT
```

```
geo.geolocation_state state, ROUND(SUM(ord_ite.freight_value)/
COUNT(ord.order_id), 2) mean_freight,

ROUND(SUM(ord.time_to_delivery)/ COUNT(ord.order_id), 2)
mean_time_to_delivery,
```

```
ROUND(SUM(ord.diff_estimated_delivery)/ COUNT(ord.order_id), 2)
mean_diff_estimated_delivery
FROM
(SELECT ord.order_id, ord.customer_id,
DATE_DIFF(CAST(ord.order_delivered_customer_date as DATETIME),
CAST(ord.order_purchase_timestamp as DATETIME), DAY)
time_to_delivery ,
DATE_DIFF(CAST(ord.order_estimated_delivery_date as DATETIME),
CAST(ord.order_delivered_customer_date as DATETIME), DAY)
diff_estimated_delivery
FROM Project_1.orders ord) as ord
JOIN Project_1.order_items ord_ite ON ord_ite.order_id =
ord.order_id
JOIN `Project_1.customers` cus ON cus.customer_id =
ord.customer_id
JOIN Project_1.geolocation geo ON cus.customer_zip_code_prefix =
geo.geolocation_zip_code_prefix
GROUP BY geo.geolocation_state
ORDER BY mean_freight ASC
LIMIT 5
```

| Row | state | mean_freight // | mean_time_to_delivery | mean_diff_estimated_delivery |
|-----|-------|-----------------|-----------------------|------------------------------|
| 1   | SP    | 15.41           | 8.66                  | 11.07                        |
| 2   | PR    | 20.15           | 11.21                 | 13.42                        |
| 3   | MG    | 20.46           | 11.57                 | 13.21                        |
| 4   | RJ    | 20.9            | 14.39                 | 12.06                        |
| 5   | DF    | 21.01           | 12.65                 | 12.25                        |

# 6. Top 5 states with highest/lowest average time to delivery

If we have solution 3, just order by average time to delivery and use desc/asc

Lowest AVG delivery time:

#### Query:

**SELECT** 

```
geo.geolocation_state state, ROUND(SUM(ord_ite.freight_value)/
COUNT(ord.order_id), 2) mean_freight,
ROUND(SUM(ord.time_to_delivery)/ COUNT(ord.order_id), 2)
mean_time_to_delivery,
ROUND(SUM(ord.diff_estimated_delivery)/ COUNT(ord.order_id), 2)
mean_diff_estimated_delivery
FROM
(SELECT ord.order_id, ord.customer_id,
DATE_DIFF(CAST(ord.order_delivered_customer_date as DATETIME),
CAST(ord.order_purchase_timestamp as DATETIME), DAY)
time_to_delivery ,
DATE_DIFF(CAST(ord.order_estimated_delivery_date as DATETIME),
CAST(ord.order_delivered_customer_date as DATETIME), DAY)
diff_estimated_delivery
FROM Project_1.orders ord) as ord
JOIN Project_1.order_items ord_ite ON ord_ite.order_id =
ord.order id
JOIN `Project_1.customers` cus ON cus.customer_id =
ord.customer_id
JOIN Project_1.geolocation geo ON cus.customer_zip_code_prefix =
geo.geolocation_zip_code_prefix
GROUP BY geo.geolocation_state
ORDER BY mean_time_to_delivery ASC
LIMIT 5
```

| Row | state // | mean_freight | mean_time_to_delivery | mean_diff_estimated_delivery_ |
|-----|----------|--------------|-----------------------|-------------------------------|
| 1   | SP       | 15.41        | 8.66                  | 11.07                         |
| 2   | PR       | 20.15        | 11.21                 | 13.42                         |
| 3   | MG       | 20.46        | 11.57                 | 13.21                         |
| 4   | DF       | 21.01        | 12.65                 | 12.25                         |
| 5   | RJ       | 20.9         | 14.39                 | 12.06                         |

#### DESC OR Highest Avg delivery time:

```
SELECT
geo.geolocation_state state, ROUND(SUM(ord_ite.freight_value)/
COUNT(ord.order_id), 2) mean_freight,
ROUND(SUM(ord.time_to_delivery)/ COUNT(ord.order_id), 2)
mean_time_to_delivery,
ROUND(SUM(ord.diff_estimated_delivery)/ COUNT(ord.order_id), 2)
mean_diff_estimated_delivery
FROM
(SELECT ord.order_id, ord.customer_id,
DATE_DIFF(CAST(ord.order_delivered_customer_date as DATETIME),
CAST(ord.order_purchase_timestamp as DATETIME), DAY)
time_to_delivery ,
DATE_DIFF(CAST(ord.order_estimated_delivery_date as DATETIME),
CAST(ord.order_delivered_customer_date as DATETIME), DAY)
diff_estimated_delivery
FROM Project_1.orders ord) as ord
JOIN Project_1.order_items ord_ite ON ord_ite.order_id =
ord.order id
JOIN `Project_1.customers` cus ON cus.customer_id =
ord.customer id
JOIN Project_1.geolocation geo ON cus.customer_zip_code_prefix =
geo.geolocation_zip_code_prefix
```

```
GROUP BY geo.geolocation_state

ORDER BY mean_time_to_delivery DESC

LIMIT 5
```

| Row | state | mean_freight | mean_time_to_delivery | mean_diff_estimated_delivery |
|-----|-------|--------------|-----------------------|------------------------------|
| 1   | AP    | 35.66        | 30.11                 | 16.23                        |
| 2   | AM    | 34.62        | 24.62                 | 21.41                        |
| 3   | AL    | 33.83        | 22.74                 | 9.05                         |
| 4   | PA    | 36.53        | 22.58                 | 14.12                        |
| 5   | SE    | 34.67        | 21.12                 | 9.37                         |

#### 7. Top 5 states where delivery is really fast/ not so fast compared to estimated date

If we have solution 3, just order by mean\_diff\_estimated\_delivery and use desc/asc

Fast delivery compared to estimated Date:

```
SELECT
geo.geolocation_state state, ROUND(SUM(ord_ite.freight_value)/
COUNT(ord.order_id), 2) mean_freight,
ROUND(SUM(ord.time_to_delivery)/ COUNT(ord.order_id), 2)
mean_time_to_delivery,
ROUND(SUM(ord.diff_estimated_delivery)/ COUNT(ord.order_id), 2)
mean_diff_estimated_delivery
FROM
(SELECT ord.order_id, ord.customer_id,
DATE_DIFF(CAST(ord.order_delivered_customer_date as DATETIME),
CAST(ord.order_purchase_timestamp as DATETIME), DAY)
time_to_delivery,

DATE_DIFF(CAST(ord.order_estimated_delivery_date as DATETIME),
CAST(ord.order_delivered_customer_date as DATETIME),
CAST(ord.order_delivered_customer_date as DATETIME),
DAY)
diff_estimated_delivery
```

```
FROM Project_1.orders ord) as ord

JOIN Project_1.order_items ord_ite ON ord_ite.order_id = ord.order_id

JOIN `Project_1.customers` cus ON cus.customer_id = ord.customer_id

JOIN Project_1.geolocation geo ON cus.customer_zip_code_prefix = geo.geolocation_zip_code_prefix

GROUP BY geo.geolocation_state

ORDER BY mean_diff_estimated_delivery ASC

LIMIT 5
```

| Row | state // | mean_freight // | mean_time_to_delivery | mean_diff_estimated_delivery_ |
|-----|----------|-----------------|-----------------------|-------------------------------|
| 1   | AL       | 33.83           | 22.74                 | 9.05                          |
| 2   | SE       | 34.67           | 21.12                 | 9.37                          |
| 3   | MA       | 38.08           | 20.7                  | 9.53                          |
| 4   | CE       | 32.26           | 20.54                 | 10.51                         |
| 5   | ES       | 22.05           | 14.87                 | 10.83                         |

#### Slow delivery compared to estimated Date:

#### Query:

```
SELECT
```

FROM

```
geo.geolocation_state state, ROUND(SUM(ord_ite.freight_value)/
COUNT(ord.order_id), 2) mean_freight,

ROUND(SUM(ord.time_to_delivery)/ COUNT(ord.order_id), 2)
mean_time_to_delivery,

ROUND(SUM(ord.diff_estimated_delivery)/ COUNT(ord.order_id), 2)
mean_diff_estimated_delivery
```

```
(SELECT ord.order_id, ord.customer_id,
DATE_DIFF(CAST(ord.order_delivered_customer_date as DATETIME),
CAST(ord.order_purchase_timestamp as DATETIME), DAY)
time_to_delivery ,
DATE_DIFF(CAST(ord.order_estimated_delivery_date as DATETIME),
CAST(ord.order_delivered_customer_date as DATETIME), DAY)
diff_estimated_delivery
FROM Project_1.orders ord) as ord
JOIN Project_1.order_items ord_ite ON ord_ite.order_id =
ord.order_id
JOIN `Project_1.customers` cus ON cus.customer_id =
ord.customer id
JOIN Project_1.geolocation geo ON cus.customer_zip_code_prefix =
geo.geolocation_zip_code_prefix
GROUP BY geo.geolocation_state
ORDER BY mean_diff_estimated_delivery DESC
LIMIT 5
```

| Row | state | mean_freight // | mean_time_to_delivery | mean_diff_estimated_delivery |
|-----|-------|-----------------|-----------------------|------------------------------|
| 1   | AM    | 34.62           | 24.62                 | 21.41                        |
| 2   | RO    | 37.43           | 18.52                 | 19.56                        |
| 3   | AC    | 39.1            | 19.98                 | 19.14                        |
| 4   | RR    | 42.47           | 20.77                 | 18.59                        |
| 5   | AP    | 35.66           | 30.11                 | 16.23                        |

# 6. Payment Analysis

1 Month over Month count of orders for different payment types

We have to combine tables orders, payments and group by month, payment\_type and count the orders for that month, payment\_type using group by.

### Query:

```
SELECT ord.month, pay.payment_type, count(ord.order_id)
count_of_orders

FROM

(SELECT ord.order_id, ord.customer_id,

EXTRACT(MONTH FROM CAST(ord.order_delivered_customer_date as
DATETIME)) month

FROM Project_1.orders ord) as ord

JOIN Project_1.payments pay ON pay.order_id = ord.order_id

WHERE ord.month is not null

GROUP BY ord.month, pay.payment_type

ORDER BY ord.month, pay.payment_type, count_of_orders DESC
```

| Row | month // | payment_type | count_of_orders |
|-----|----------|--------------|-----------------|
| 1   | 1        | UPI "        | 1454            |
| 2   | 1        | credit_card  | 5211            |
| 3   | 1        | debit_card   | 107             |
| 4   | 1        | voucher      | 385             |
| 5   | 2        | UPI          | 1425            |
| 6   | 2        | credit_card  | 5609            |
| 7   | 2        | debit_card   | 79              |
| 8   | 2        | voucher      | 412             |
| 9   | 3        | UPI          | 1899            |
| 10  | 3        | credit_card  | 7086            |

1. From the top rows I can observe most of the orders are using UPI and credit card.

# 2 Count of orders based on the no. of payment installments

We should join tables ord and payments and group by payment\_installments and count the orders.

# Query:

#### SELECT

```
pay.payment_installments, COUNT(ord.order_id) count_of_orders
```

#### FROM

```
`Project_1.orders` ord
```

JOIN Project\_1.payments pay ON ord.order\_id = pay.order\_id

GROUP BY pay.payment\_installments

ORDER BY count\_of\_orders DESC

| Row | payment_installments // | count_of_orders |
|-----|-------------------------|-----------------|
| 1   | 1                       | 52546           |
| 2   | 2                       | 12413           |
| 3   | 3                       | 10461           |
| 4   | 4                       | 7098            |
| 5   | 10                      | 5328            |
| 6   | 5                       | 5239            |
| 7   | 8                       | 4268            |
| 8   | 6                       | 3920            |
| 9   | 7                       | 1626            |
| 10  | 9                       | 644             |
| 11  | 12                      | 133             |

1. Most of the orders are with low payment\_installments.  $\rightarrow$  means people are not willing to buy things with installments.