1.1 Data Types of columns in a table

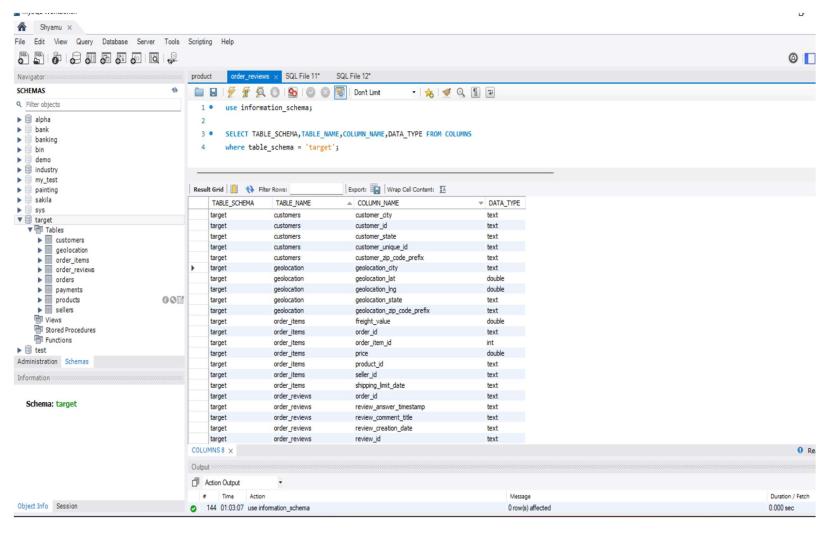
Use information schema;

SELECT TABLE SCHEMA, TABLE NAME, COLUMN NAME, DATA TYPE

FROM COLUMNS

WHERE TABLE SCHEMA = 'target'

NOTE: target is name of the database where all the files are stored



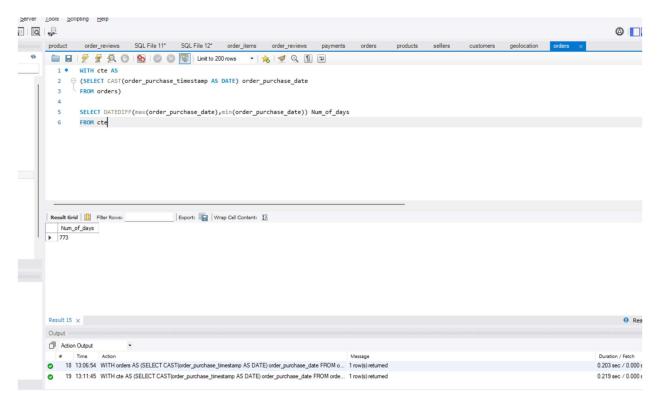
1.2 Time period for which the data is given

WITH cte AS

(SELECT CAST(order_purchase_timestamp AS DATE) order_purchase_date FROM orders)

SELECT DATEDIFF(max(order_purchase_date),min(order_purchase_date))
Num_of_days

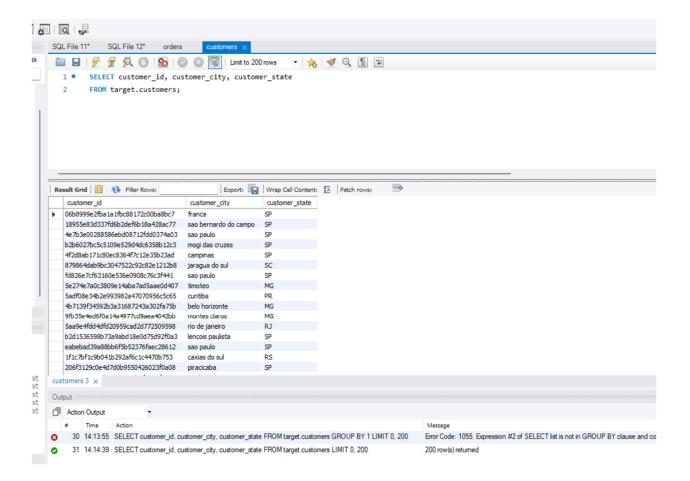
FROM cte



Finding: From above results we can see that the data is given of nearly more than 2 years. Or more precisely of 773 days.

1.3 Cities and States of customers ordered during the given period

SELECT customer_id, customer_city, customer_state FROM target.customers;



2.1 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 months?

WITH cte AS

 $(SELECT\ year, total_orders,\ ABS((COALESCE(lead(total_orders)\ over(), 0)\ -\ total_orders))\\ YoY_Change$

FROM

(SELECT EXTRACT(year FROM order_purchase_timestamp) as year, COUNT(*) total_orders

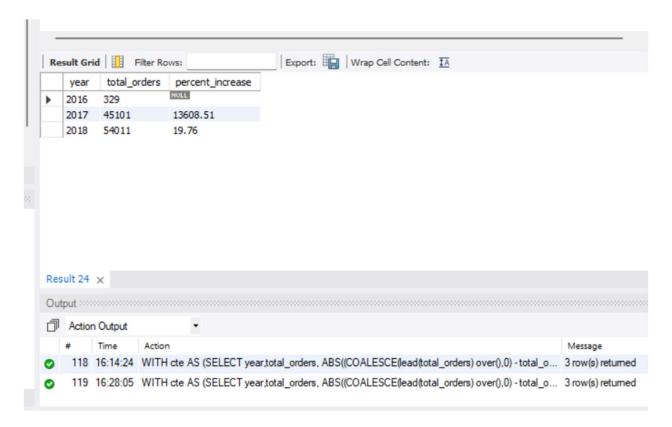
FROM orders

GROUP BY 1

order BY 2) order table)

SELECT year, total_orders, ROUND(lag((YoY_Change/total_orders)*100) over(),2) AS percent_increase

FROM cte



Findings:-

- From the above picture we can infer that there is nearly 13608 percent of increment from 2016 to 2017. This huge increase is due to insufficient data from 2016 as data of only 3 months is available.
- There is 19.76 percent increment from 2017 to 2018 which show numbers of orders are increasing at quite high rate indicating the positive sign in e-commerce activities.

Part 2

WITH cte AS

(SELECT EXTRACT(month FROM order_purchase_timestamp) month, DATE_FORMAT(order_purchase_timestamp, '%M') as order_purchase_month, order_id

FROM orders

```
)
```

SELECT CASE

WHEN month IN (1,2,3) THEN 'Summer'

WHEN month IN (4,5,6) THEN 'Autumn'

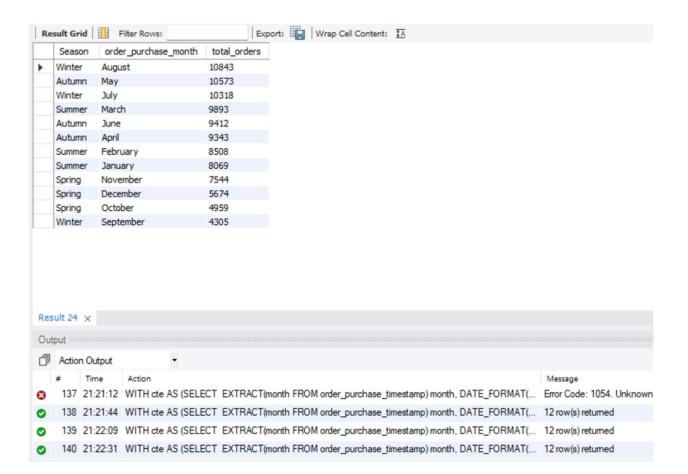
WHEN month IN (7,8,9) THEN 'Winter'

WHEN month IN (10,11,12) THEN 'Spring'

END Season, order_purchase_month, COUNT(order_id) total_orders

FROM cte

GROUP BY 1,2



F	i	n	d	i	n	g	S	•
•	•	••	u	•		a	•	•

ORDER BY 2 DESC

• The number of orders placed in Winter and Autumn is greater than any other seasons in Brazil. August month shows the peak of the orders.

2.2 What time do Brazilian customers tend to buy (Dawn, Morning, Afternoon or Night)?

```
WITH time as

(SELECT_DATE_FORMAT(order_purchase_timestamp,'%H') times, order_id as Num_of_orders

FROM orders
)

SELECT CASE

WHEN times between 5 and 6 THEN 'Dawn'

WHEN times between 7 and 11 THEN 'Morning'

WHEN times between 12 and 20 THEN 'Afternoon'

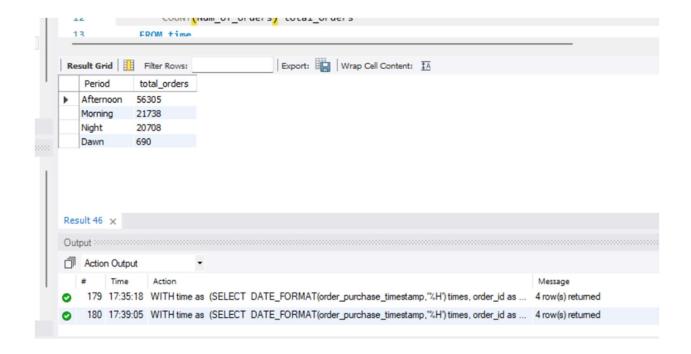
ELSE 'Night'

END as Period,

COUNT(Num_of_orders) total_orders

FROM time

GROUP BY 1
```



Note :- In the above case to find the time period, we categorized the hours of a day in certain periods as follows

Dawn: Between 5 and 6
 Morning: Between 7 and 11
 Afternoon: Between 12 and 20

4) Night:- Remaining left hours are considered as night.

Findings:-

• From above result we can see that the tendency of Brazilians to buy during Afternoon is significantly more than other time

3.1 Get month on month orders by states

WITH cte as

(SELECT customer_state, DATE_FORMAT(order_purchase_timestamp,'%M') as months, DATE_FORMAT(order_purchase_timestamp,'%m') m_num, count(*) total_orders

FROM orders o

JOIN customers c

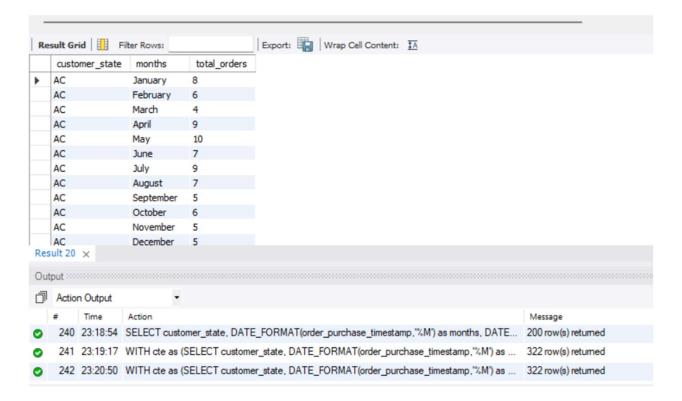
ON o.customer_id = c.customer_id

GROUP BY 1,2,3

ORDER BY 1,3)

SELECT customer_state, months, total_orders

FROM cte



3.2 Distribution of customers across the states in Brazil

SELECT customer_state,count(*) Customer_distribution

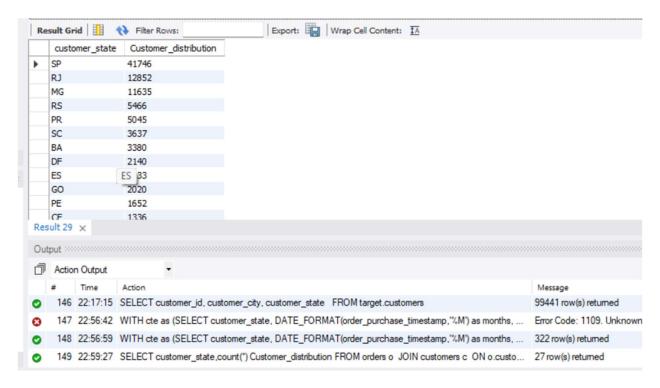
FROM orders o

JOIN customers c

ON o.customer_id = c.customer_id

GROUP BY 1

ORDER BY 2 DESC;



4.1 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

with cte1 as

(SELECT DATE_FORMAT(order_purchase_timestamp,'%Y') as year, DATE_FORMAT(order_purchase_timestamp,'%m') as date1, SUM(payment_value) value

FROM orders o

JOIN payments p using(order_id)

GROUP BY 1,2),

cte2 as

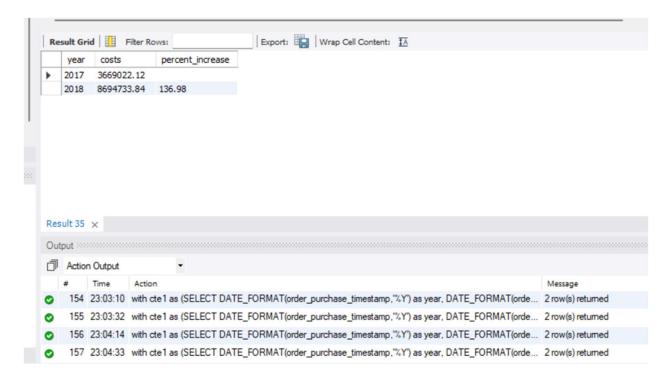
(SELECT year, sum(value) cost, lead(sum(value)) over(order by year)-sum(value) c increment FROM cte1

WHERE date1 between 1 and 8 and year <> 2016

group by 1)

SELECT year, ROUND(cost,2) costs, COALESCE(lag(ROUND((c_increment/cost)*100,2)) over(),'') as percent_increase

FROM cte2



Findings:-

There is more than 135 % increase in cost of orders from 2017 to 2018, indicating tremendous growth in terms of cost of orders

4.2 Mean & Sum of price and freight value by customer state

SELECT customer_state, ROUND(avg(price)) Mean_of_Price, ROUND(sum(price)) Sum_of_Price,

ROUND(avg(freight_value)) Mean_of_freight, ROUND(sum(freight_value)) Sum_of_freight

FROM order_items oi

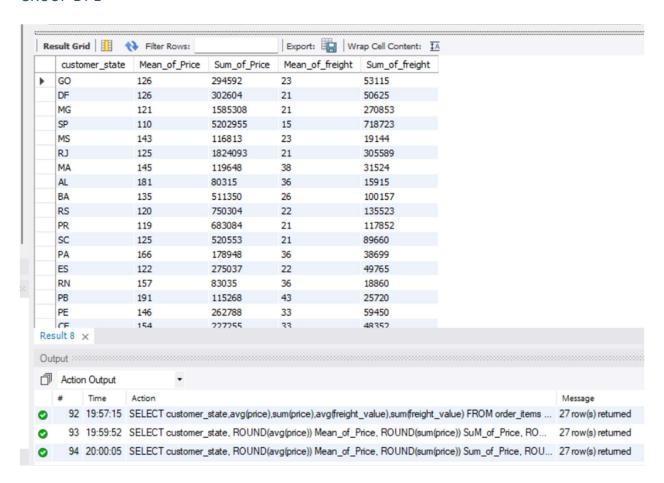
JOIN orders o

using(order_id)

JOIN customers c

using(customer_id)

GROUP BY 1



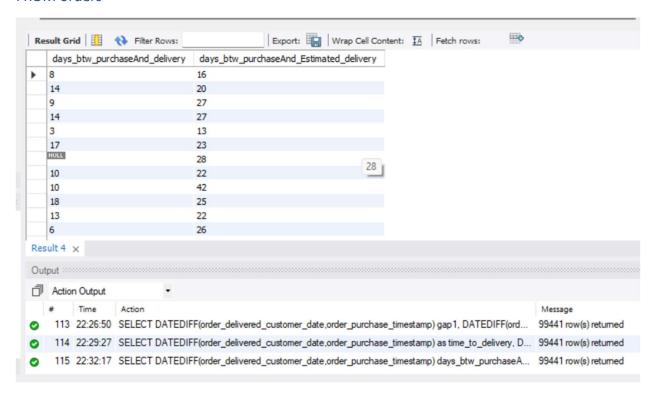
5.1 Calculate days between purchasing, delivering and estimated delivery

SELECT

DATEDIFF(order_delivered_customer_date,order_purchase_timestamp) days btw purchaseAnd delivery,

DATEDIFF(order_estimated_delivery_date,order_purchase_timestamp) days btw purchaseAnd Estimated delivery

FROM orders



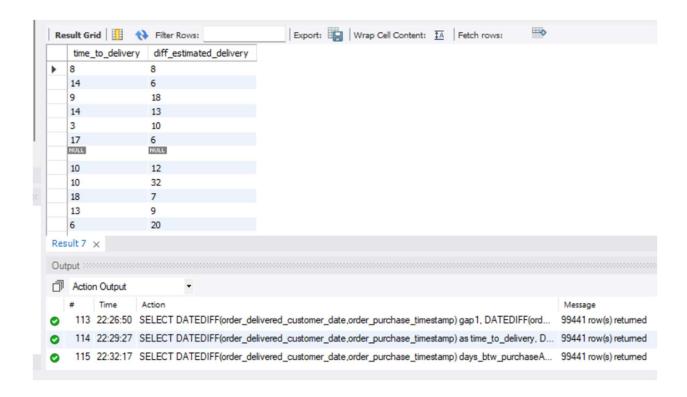
5.2 Find time_to_delivery & diff_estimated_delivery. Formula for the same given below:

- time_to_delivery = order_purchase_timestamporder_delivered_customer_date
- diff_estimated_delivery = order_estimated_delivery_dateorder_delivered_customer_date

SELECT DATEDIFF(order_delivered_customer_date,order_purchase_timestamp) as time to delivery,

DATEDIFF(order_estimated_delivery_date,order_delivered_customer_date) as diff_estimated_delivery

FROM orders



5.3 Group data by state, take mean of freight_value, time_to_delivery, diff_estimated_delivery

WITH cte AS

(SELECT DATEDIFF(order_delivered_customer_date,order_purchase_timestamp) as time_to_delivery,

DATEDIFF(order_estimated_delivery_date,order_delivered_customer_date) as diff_estimated_delivery, customer_id, order_id

FROM orders)

SELECT customer_state, ROUND(avg(freight_value),2) as freight_value_mean, ROUND(avg(time_to_delivery),2) time_to_delivery_mean,

ROUND(avg(diff_estimated_delivery),2) diff_estimated_delivery_mean

FROM customers c

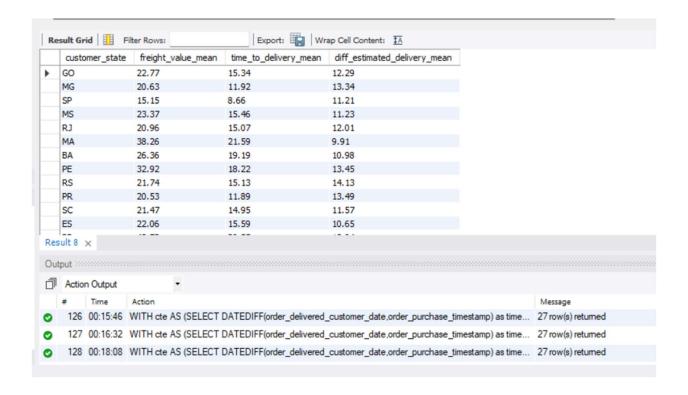
JOIN cte ct

using (customer_id)

JOIN order_items o

using (order_id)

GROUP BY customer_state



5.4 Sort the data to get the following:

5.5 Top 5 states with highest/lowest average freight value - sort in desc/asc limit 5

Part 1

WITH cte AS

(SELECT DATEDIFF(order_delivered_customer_date,order_purchase_timestamp) as time_to_delivery,

DATEDIFF(order_estimated_delivery_date,order_delivered_customer_date) as diff_estimated_delivery, customer_id, order_id

FROM orders)

SELECT customer state, ROUND(avg(freight value),2) as freight value mean

FROM customers c

JOIN cte ct

using (customer id)

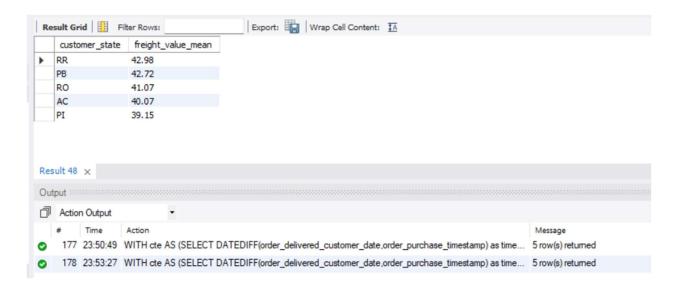
JOIN order_items o

using (order_id)

GROUP BY customer state

ORDER BY 2 DESC

LIMIT 5



Findings:- The above query gives the top 5 states having most freight_value_mean

Part 2

WITH cte AS

(SELECT DATEDIFF(order_delivered_customer_date,order_purchase_timestamp) as time_to_delivery,

 $\label{linear_delivery_date} DATEDIFF (order_estimated_delivery_date, order_delivered_customer_date) \ as \ diff_estimated_delivery, \ customer_id, \ order_id$

FROM orders)

SELECT customer_state, ROUND(avg(freight_value),2) as freight_value_mean,

FROM customers c

JOIN cte ct

using (customer_id)

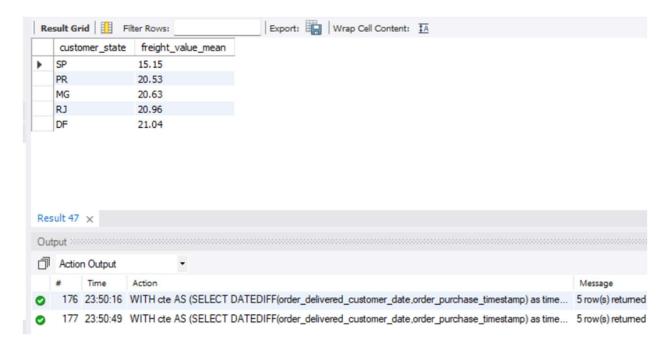
JOIN order_items o

using (order_id)

GROUP BY customer_state

ORDER BY 2

LIMIT 5



Findings:- The above query gives the 5 state with lowest freight_value_mean

5.6 Top 5 states with highest/lowest average time to delivery

Part 1

WITH cte AS

(SELECT DATEDIFF(order_delivered_customer_date,order_purchase_timestamp) as time_to_delivery,

DATEDIFF(order_estimated_delivery_date,order_delivered_customer_date) as diff_estimated_delivery, customer_id, order_id

FROM orders)

SELECT customer_state, ROUND(avg(time_to_delivery),2) time_to_delivery_mean

FROM customers c

JOIN cte ct

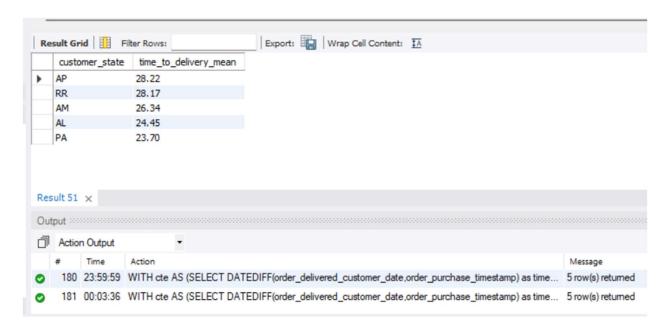
using (customer_id)

JOIN order_items o

using (order_id)

GROUP BY customer_state

ORDER BY 2 DESC LIMIT 5



Findings:- From above result, we can say that these are the 5 states where delivery is taking very long times as compared to other states.

Part 2

WITH cte AS

(SELECT DATEDIFF(order_delivered_customer_date,order_purchase_timestamp) as time_to_delivery,

DATEDIFF(order_estimated_delivery_date,order_delivered_customer_date) as diff_estimated_delivery, customer_id, order_id

FROM orders)

SELECT customer_state, ROUND(avg(time_to_delivery),2) time_to_delivery_mean

FROM customers c

JOIN cte ct

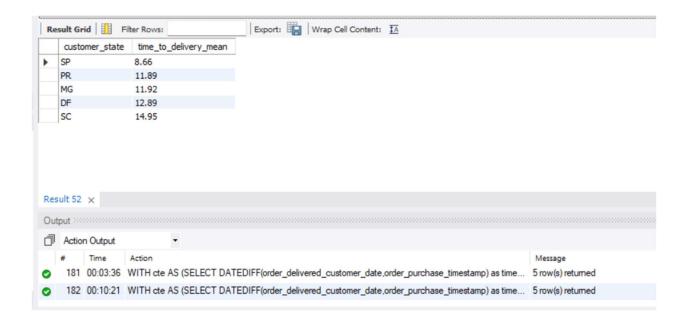
using (customer_id)

JOIN order_items o

using (order_id)

GROUP BY customer_state

ORDER BY 2 LIMIT 5



Finding: - From above we can infer that the delivery in above 5 states is very fast as compared to other states.

5.7 Top 5 states where delivery is really fast/ not so fast compared to estimated date

Part 1

WITH CTE AS

(SELECT customer_id, DATEDIFF(order_delivered_customer_date,order_purchase_timestamp) delivery_days,

FROM orders)

SELECT DISTINCT customer_state, delivery_days

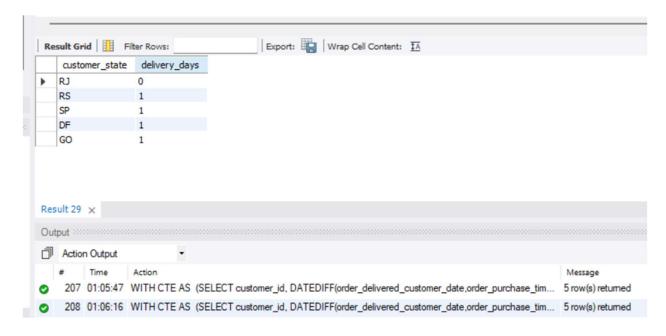
FROM customers

JOIN CTE using(customer_id)

WHERE delivery_days is not null

ORDER BY 2

LIMIT 5



Part 2

WITH CTE AS

(SELECT customer_id, DATEDIFF(order_delivered_customer_date,order_purchase_timestamp) delivery_days,

DATEDIFF(order_estimated_delivery_date,order_purchase_timestamp) Estimated_delivery_days

FROM orders)

SELECT DISTINCT customer_state, delivery_days, Estimated_delivery_days

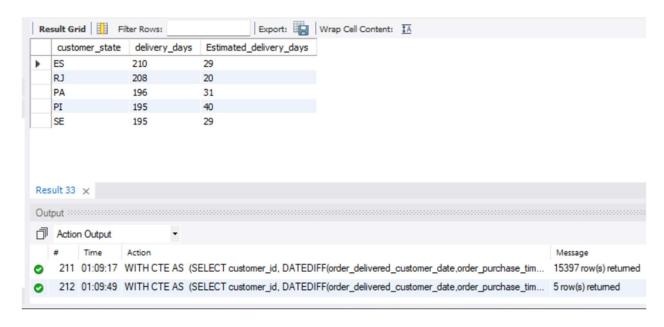
FROM customers

JOIN CTE using(customer_id)

WHERE delivery_days is not null

ORDER BY 2 DESC

LIMIT 5



6.1 Month over Month count of orders for different payment types

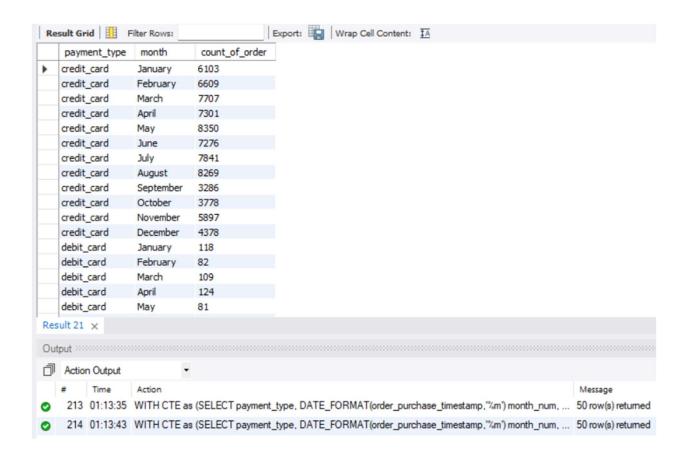
WITH CTE as

order by payment type, month num

```
(SELECT payment_type, DATE_FORMAT(order_purchase_timestamp,'%m') month_num,
DATE_FORMAT(order_purchase_timestamp,'%M') month, COUNT(*) count_Of_Order
FROM orders o

JOIN payments p
ON o.order_id = p.order_id
GROUP BY 1,2,3
)

SELECT payment_type, month, count_of_order
FROM cte
```



6.2 Count of orders based on the no. of payment installments

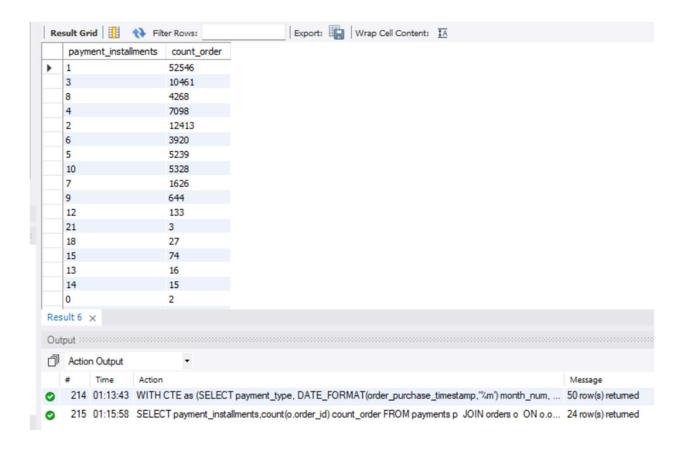
SELECT payment_installments,count(o.order_id) count_order

FROM payments p

JOIN orders o

ON o.order_id = p.order_id

GROUP BY 1



Actionable Insights

- 1) The number of orders increased more than 19.75% from 2017 to 2018
- 2) Highest number of orders are placed in the winter season having August month as its peak.
- 3) Customers generally buy more during afternoon period as compared to others.
- 4) Highest number of customers are present in SP state of Brazil.
- 5) From 2017 to 2018, there is more than 135% YoY increase in cost of orders.
- 6) States like RR, PB, RO, AC and PI have average freight value more than 40.
- 7) States like AP, RR, AM, AL, and PA have average delivery time nearly of 25 days.

- 8) In States like ES, RJ, PA, PI and SE some deliveries are taking nearly 200 days to complete.
- 9) Highest number of orders are made using credit card.
- 10) Count of order is significantly higher when there are just 1 payment instalments.

Recommendations

- 1) Advertise in the state having low number of customers to increase the reach.
- 2) Reduce the average number of freight days in state where it is taking too much time.
- 3) Find the reason and solve the problem of states where some products are taking drastically large number of days to deliver as compared to expected date of delivery.
- 4) Try to give some offers on payment methods which is not used so frequently.