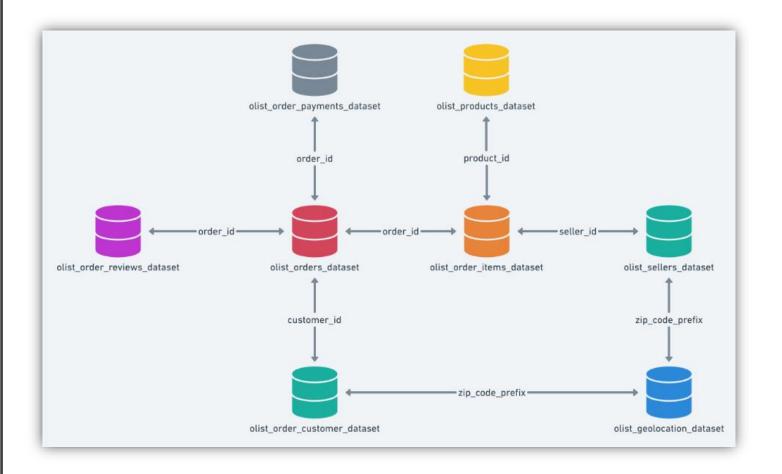
Customer Segmentation

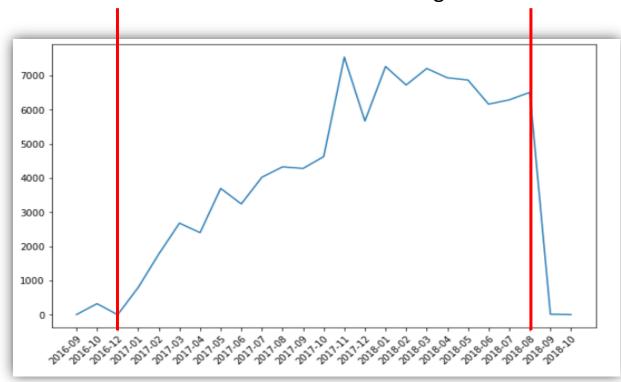


Datasets

- **Customers**: This dataset has information about the customer and their location.
- **Geolocation**: This dataset has information about the Brazilian zip codes and their latitude/longitude coordinates.
- Order Items: This dataset has information about the items purchased within each order.
- **Order Payments**: This dataset has information about the order payment options.
- Order Reviews: This dataset has information about the reviews made by the customers.
- Orders: This dataset has information about all customer orders.
- Products: This dataset has information about all the products sold by Olist.
- **Sellers**: This dataset has information about the sellers that fulfilled the orders made at Olist.
- Category Name Translation: This dataset has
 English/Portuguese translations for all products sold at Olist.



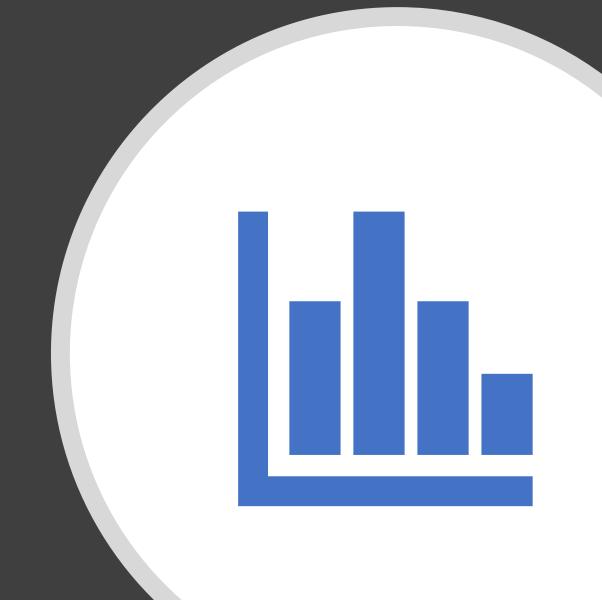
Orders between Jan 2017 - Aug 2018

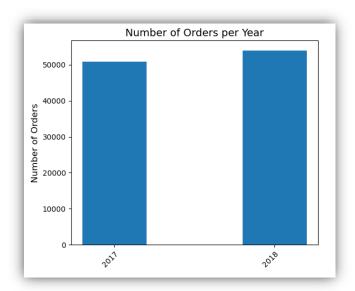


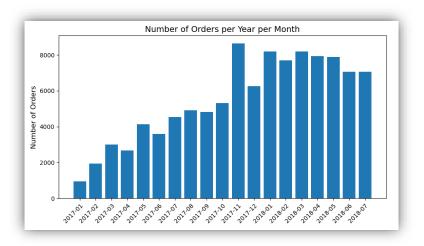
Data Cleaning: Orders

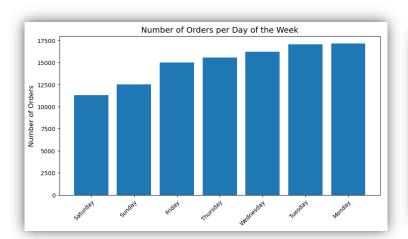
- Dropped orders from 2016 and after August of 2018 as there were very small numbers
- Empty Values:
 - Order Approved At: Used the Order Purchase
 Timestamp for those orders that were delivered
 - Order Delivered Carrier Date: Used the Order Approved Date for those orders that were delivered
 - Order Delivered Customer Date: Used the difference between the Order Delivered Customer Date and the Order Delivered Carrier Date for all orders and got the median number of days. Added the days to the Order Delivered Carrier Date

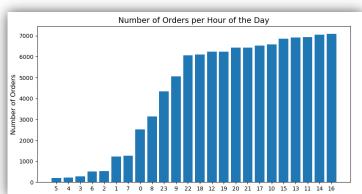
Exploratory Data Analysis









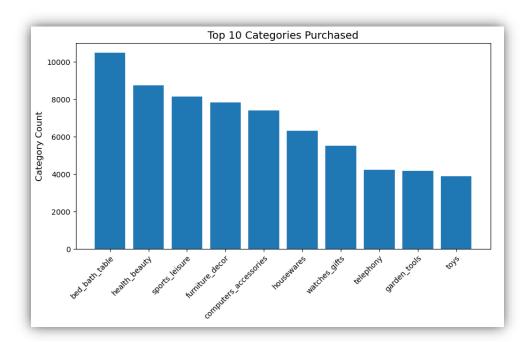


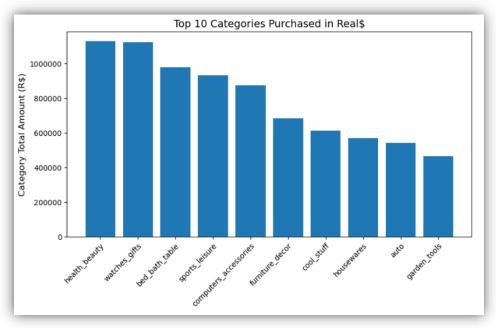
Order Purchases Timeframes

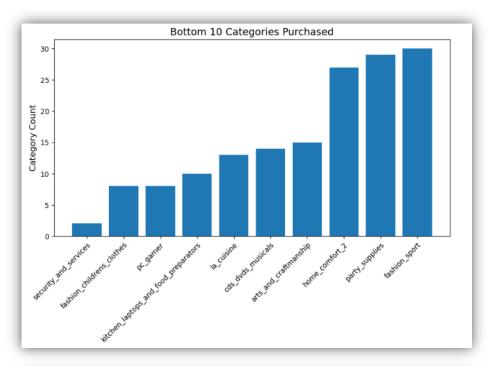
- The number of orders grew between 2017 to 2018.
- November and January had the highest number of orders probably due to the holidays.
- March, April, and May were also good purchasing months.
- Weekdays were also more popular purchasing days as the beginning of the week was best probably due to items needed from the weekend and early evenings.

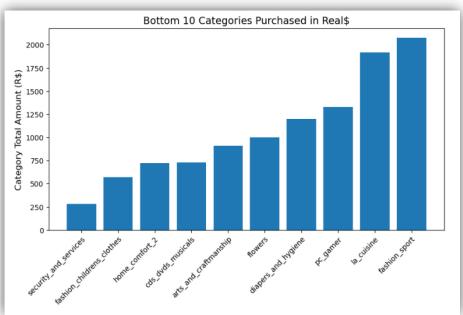
Categories

- Most items that were purchased were
 from the Bed, Bath & Table category but
 the top category in the amount purchased
 was Health & Beauty.
- Most customers purchased items in the bedroom/bathroom area or that were used in the bedroom/bathroom area.







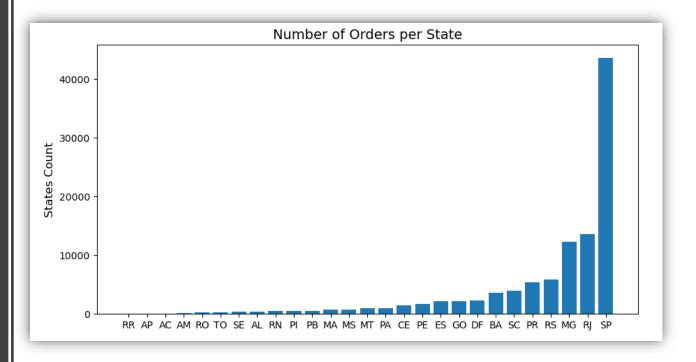


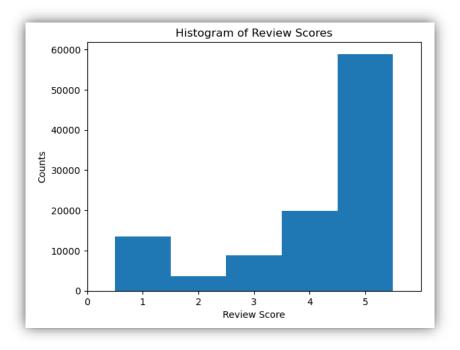
Categories (con't)

Looking at both the number of items purchased, and the amount spent, the Security and Services and Children's clothes categories are at the bottom. It could be that there are not many sellers of these items or more marketing needs to be allotted for these bottom categories.

States

The top State was São Paulo by far compared to the other States in Brazil. Marketing may investigate advertising more in States other than São Paulo.







Review Score

- The review score of 5 is by far the most used by Olist customers which is a good sign
- Looking at the review score based on the order status does show purchases that were delivered by far getting a review score of 5 and the other statuses, not surprisingly, were lower.

	customer_id	review_score	order_status	price	freight_value	product_category_name	cust_delivery_diff	est_delivery_diff
4201	d6646ea91d8cd9fc7e6882a7068779d4	5	delivered	81.99	14.51	computers_accessories	7	20
4350	679f84ceb2ee4ca5bca0c3ea34647746	5	delivered	59.90	17.67	garden_tools	20	4
13293	b4afeb58ac51bc903c5362286c6a5cfe	5	delivered	19.30	11.73	drinks	11	5
49414	10de381f8a8d23fff822753305f71cae	5	delivered	65.49	16.22	furniture_decor	19	5
54499	b7770073b02ed1d626a027ce86a4ff82	5	delivered	66.90	31.65	sports_leisure	10	44
67152	0d93f21f3e8543a9d0d8ece01561f5b2	5	delivered	20.70	16.11	housewares	8	8
67939	1ff773612ab8934db89fd5afa8afe506	5	delivered	284.99	16.87	drinks	14	18
77657	20c93357daf05d1c3a092be59aea2c2b	5	delivered	20.50	16.91	drinks	10	14
90355	0e772d9e02b17408e716f35cd1dcc222	5	delivered	36.99	11.85	bed_bath_table	10	13
96279	adb32467ecc74b53576d9d13a5a55891	5	delivered	51.00	1.20	garden_tools	14	20

	customer_id	review_score	order_status	price	freight_value	product_category_name	${\sf cust_delivery_diff}$	est_delivery_diff
6639	91f92cfee46b79581b05aa974dd57ce5	1	delivered	108.00	15.52	watches_gifts	11	13
19725	d5f2b3f597c7ccafbb5cac0bcc3d6024	1	delivered	59.00	13.43	garden_tools	14	10
24699	4a60b2ce1ee8c7b828e4bbcca5b86b41	1	delivered	137.90	38.81	computers_accessories	14	1
25887	be1c4e52bb71e0c54b11a26b8e8d59f2	1	delivered	49.99	7.10	bed_bath_table	5	11
37089	78fc46047c4a639e81ff65f0396e02fe	1	delivered	109.97	34.04	furniture_living_room	5	13
46421	be1b70680b9f9694d8c70f41fa3dc92b	1	delivered	100.00	10.12	computers_accessories	10	2
68674	cb87122c4871e202777cf243fbea2d12	1	delivered	149.91	0.14	computers_accessories	11	23
77621	a7693fba2ff9583c78751f2b66ecab9d	1	delivered	29.99	7.78	telephony	8	5
102470	fc3d1daec319d62d49bfb5e1f83123e9	1	delivered	1.20	7.89	health_beauty	14	-4
102724	9eb3d566e87289dcb0acf28e1407c839	1	delivered	5.31	15.23	housewares	10	9

Customers

Looking at the two charts, you can't distinguish between those that gave a high review score to those that gave a low review score.

Review Score = 1

Final Dataset for Modeling

	customer id	order_purchase_timestamp	order id	price		
0	9ef432eb6251297304e76186b10a928d	2017-10-02 10:56:33	e481f51cbdc54678b7cc49136f2d6af7	29.99		
1	a20e8105f23924cd00833fd87daa0831	2017-08-15 18:29:31	128e10d95713541c87cd1a2e48201934	29.99		
2	26c7ac168e1433912a51b924fbd34d34	2017-08-02 18:24:47	0e7e841ddf8f8f2de2bad69267ecfbcf	29.99		
3	53904ddbea91e1e92b2b3f1d09a7af86	2017-10-23 23:26:46	bfc39df4f36c3693ff3b63fcbea9e90a	29.99		
4	b0830fb4747a6c6d20dea0b8c802d7ef	2018-07-24 20:41:37	53cdb2fc8bc7dce0b6741e2150273451	118.70		

104777	609b9fb8cad4fe0c7b376f77c8ab76ad	2017-08-10 21:21:07	e8fd20068b9f7e6ec07068bb7537f781	356.00		
104778	609b9fb8cad4fe0c7b376f77c8ab76ad	2017-08-10 21:21:07	e8fd20068b9f7e6ec07068bb7537f781	356.00		
104779	a2f7428f0cafbc8e59f20e1444b67315	2017-12-20 09:52:41	cfa78b997e329a5295b4ee6972c02979	55.90		
104780	39bd1228ee8140590ac3aca26f2dfe00	2017-03-09 09:54:05	9c5dedf39a927c1b2549525ed64a053c	72.00		
104781	edb027a75a1449115f6b43211ae02a24	2018-03-08 20:57:30	66dea50a8b16d9b4dee7af250b4be1a5	68.50		
104782 rows × 4 columns						

Modeling

RFM Analysis + K-Means Clustering

Recency, Frequency, and Monetary Analysis (RFM)

Looks at historical customer behavior to predict how might a new customer act in the future

Recency:

How recently has a customer purchased?

Frequently:

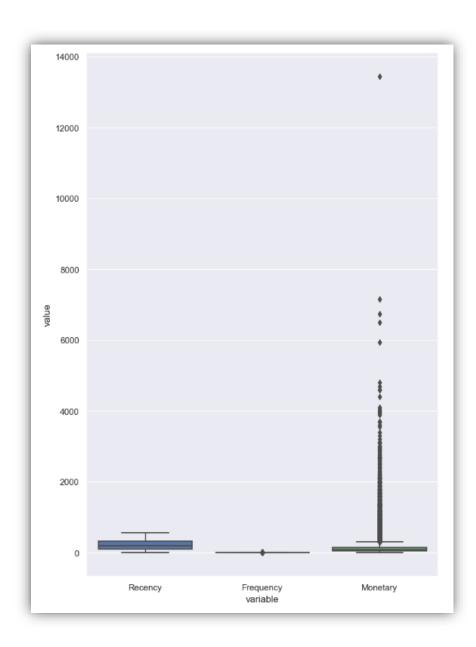
How many orders did a customer purchase?

Monetary:

How much money has a customer spent?

K-Means Clustering

Grouping customers based on historical data according to how recently the purchased, how frequently they purchased and how much money they spent on purchases.



RFM Analysis

Recency

Mean: 256 days

Min: 0 days

Max: 572 days

Frequency

Mean: 1.12 times

Min: 1 time

Max: 22 times

Monetary Value

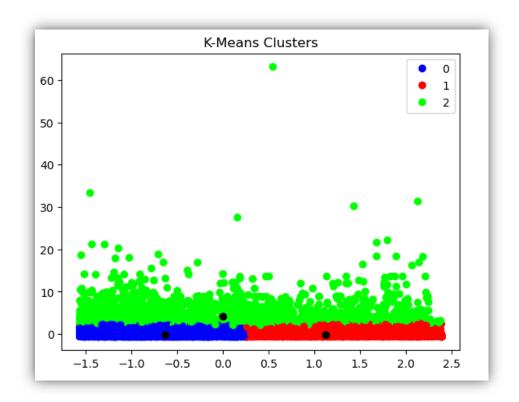
Mean: R\$138

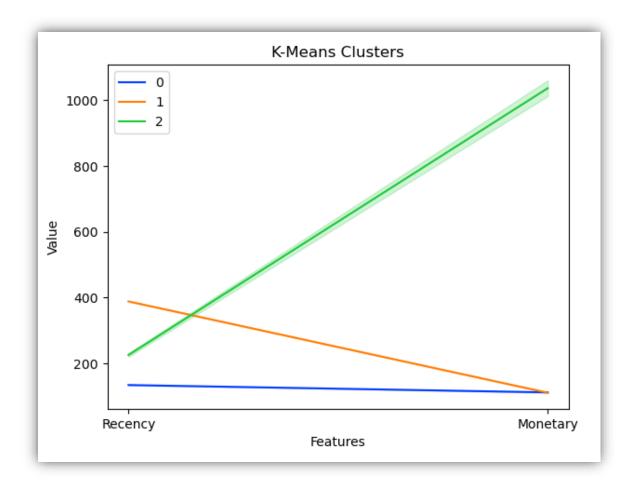
Min: R\$0.85

Max: R\$13,440

K-Means Analysis

Number of clusters: 3





Looking at the two graphs, we can see the differences between the different clusters.

- Cluster 0: Customers that haven't purchased items in a long time and their purchases amount to a very low dollar amount.
- Cluster 1: Customers that have purchased recently but their purchases amount to a very low dollar amount
- Cluster 2: These are customers that have purchased items somewhat recently but these customers' purchases amount to a very high dollar amount.