



A LOOK AT CUSTOMER LIFETIME VALUE

Rachelle Perez

December 2019

WHO IS OLIST?

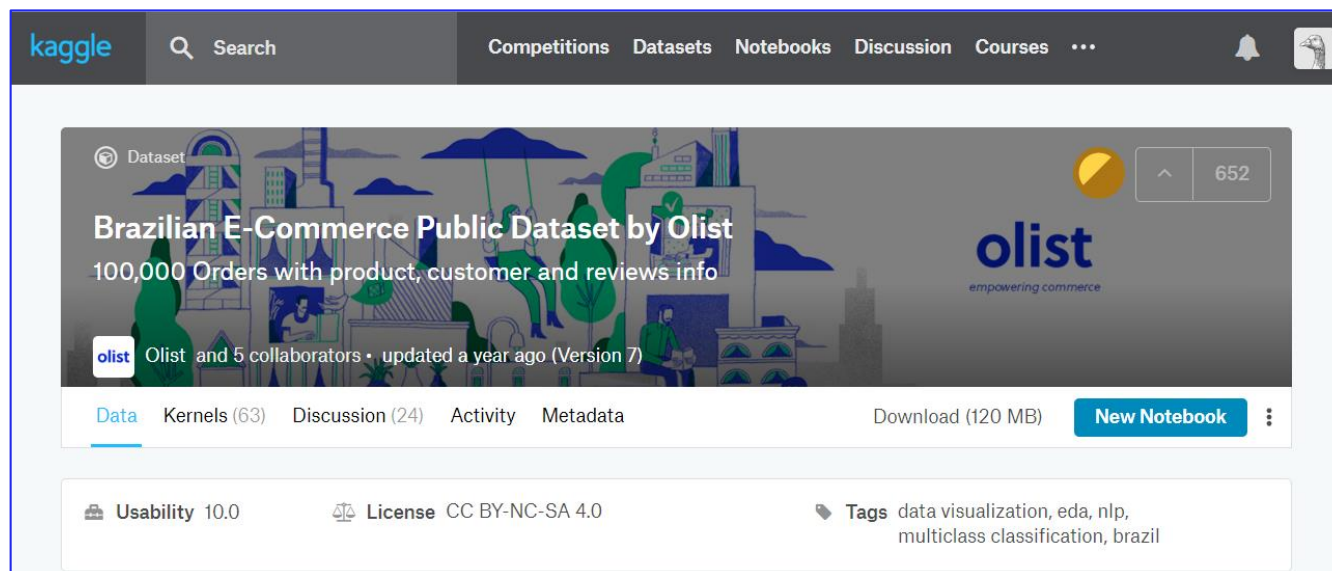
- **E-Commerce** business
- Small business merchants ("sellers") **sell their products** to customers through Olist and **ship them directly** to customer **using Olist logistics partners** ("carrier")
- olist.com

The Olist logo is displayed in a large, bold, blue sans-serif font. The word "olist" is written in lowercase letters, with the 'o' being slightly larger than the other letters.

PROBLEM: What **factors** affect 6-month Customer Lifetime Value (LTV)?

DATA AVAILABLE

SOURCE



kaggle.com/olistbr/brazilian-ecommerce

towardsdatascience.com/

DATE RANGE

9/4/2016 - 8/29/2018

MAIN TABLES

99,441 Orders

96,096 Customers

SUPPORTING TABLES

Order Items

Products

Sellers

Payments

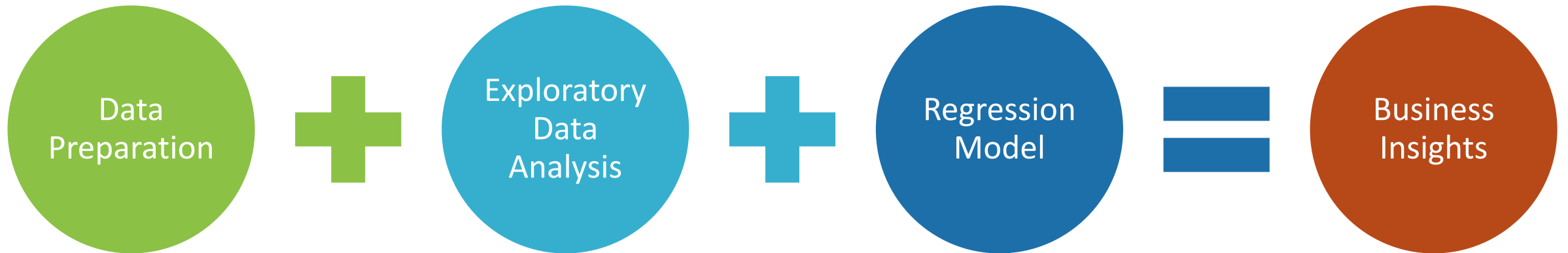
Reviews

REFERENCE TABLES

Geolocation

Product Category Translations

PROJECT STEPS

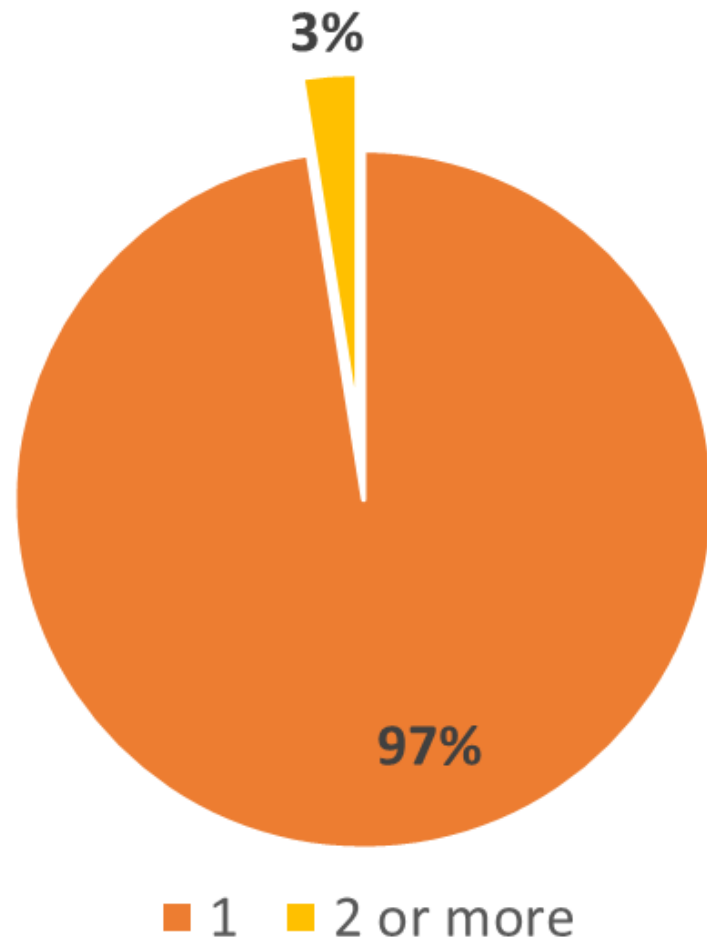


BUSINESS INSIGHTS

What is the overall state of the business?

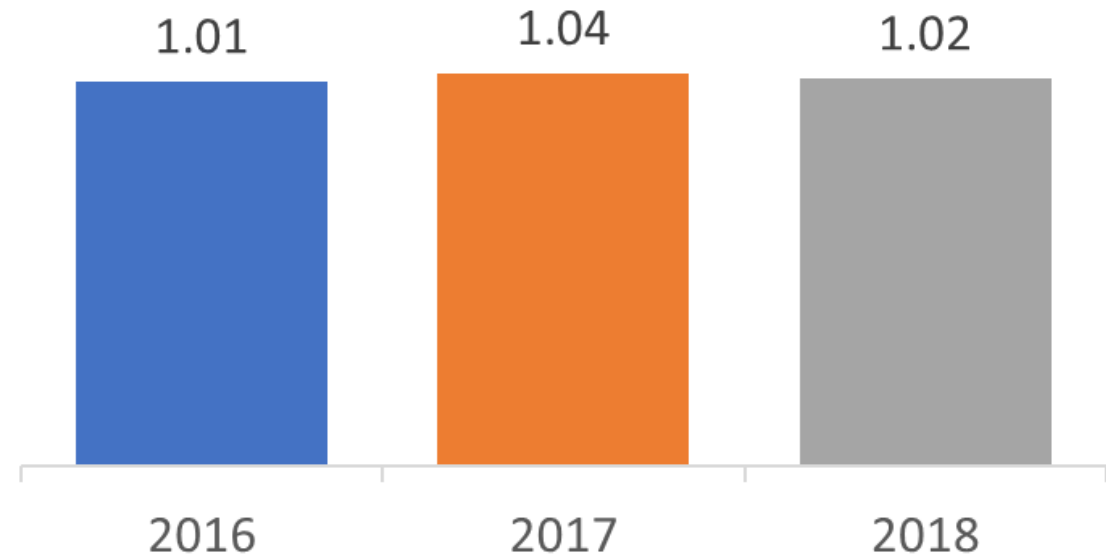
HOW OFTEN CUSTOMERS CHURN?

Customers split by 1 order vs. 2+ orders



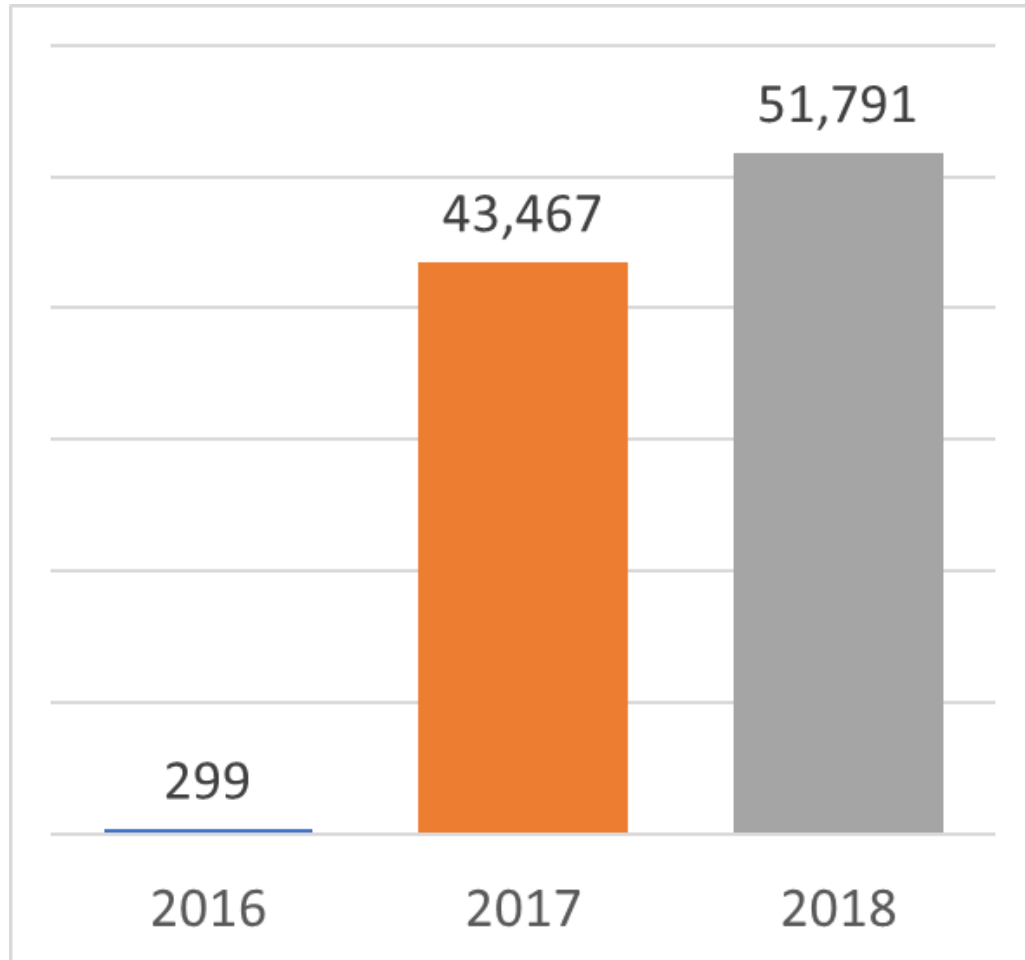
- **97%** new customers churn (they make only 1 order)
- **1 order per customer** has been a consistent pattern across the years.

Average Number of Orders per Customer



HOW IS CUSTOMER ACQUISITION (I)?

Number of NEW Customers Per Year

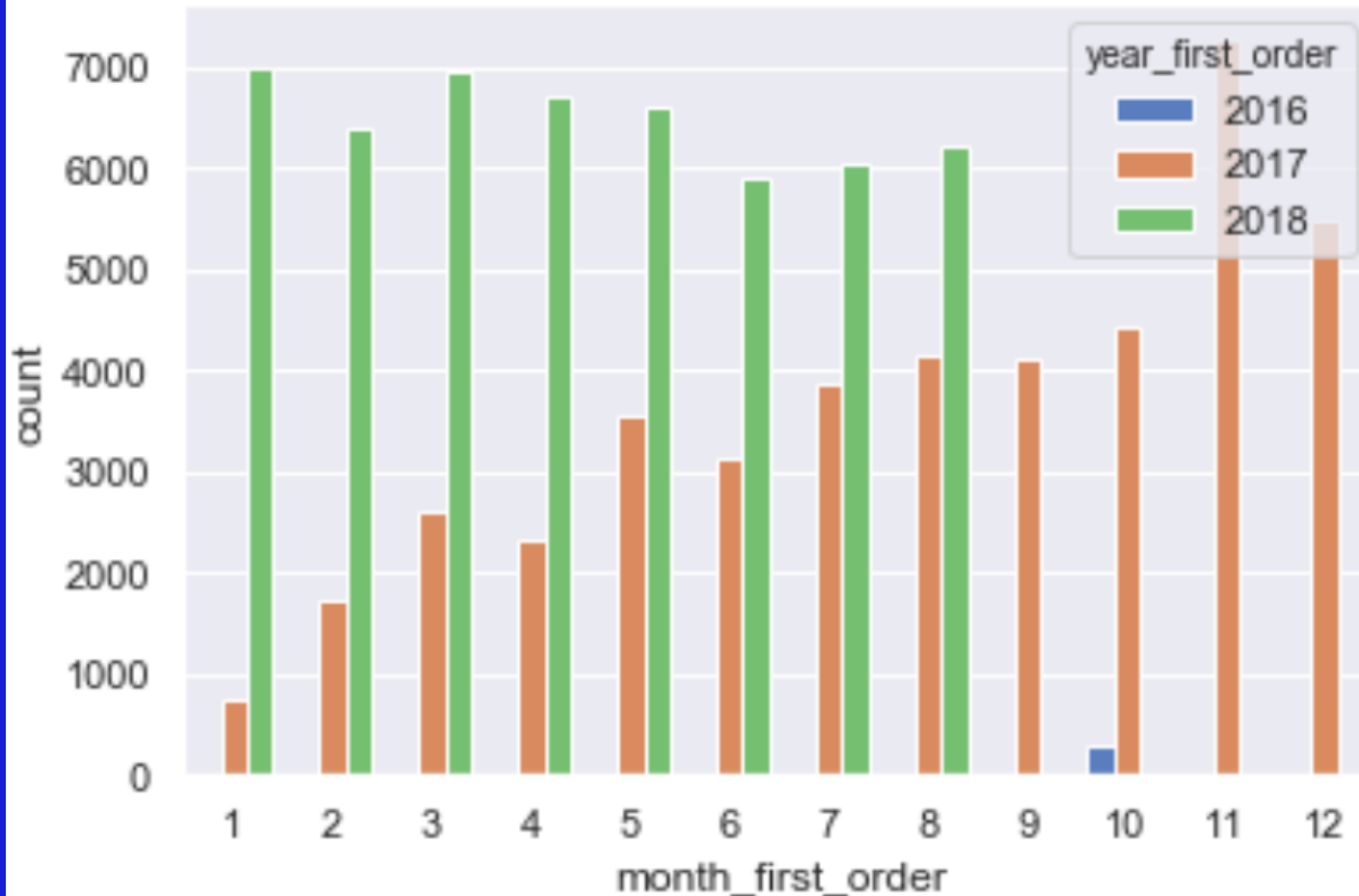


- **Positive trend** in customer acquisition
- 2018 is already **19% up** (despite only 8 months of data)

Does this paint the whole picture?

HOW IS CUSTOMER ACQUISITION (II)?

Number of NEW Customers by Month & Year

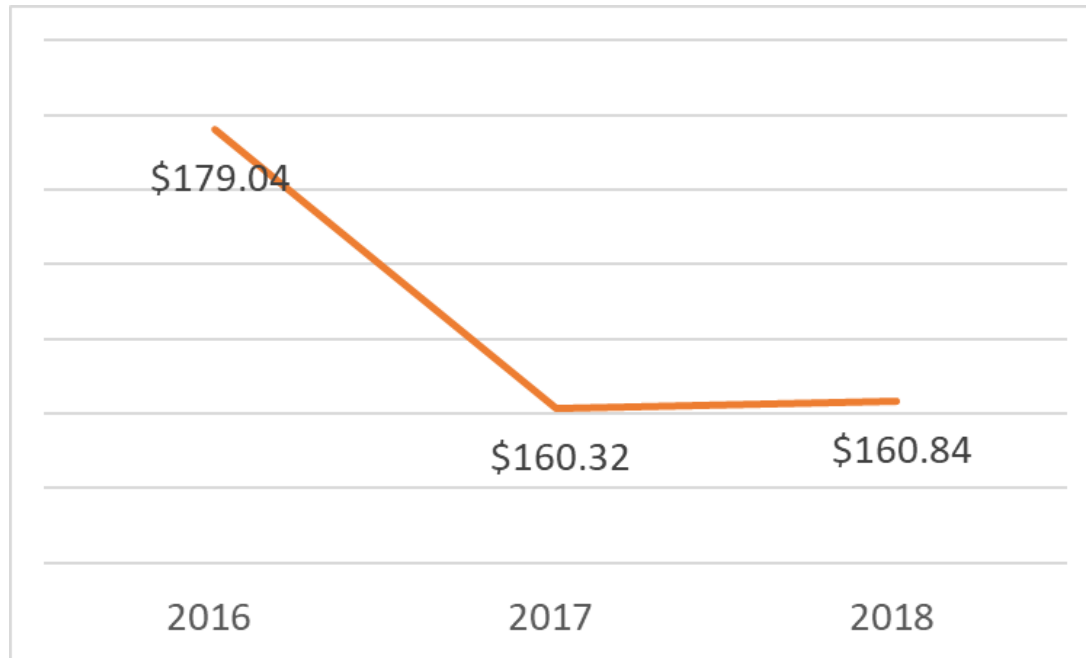


- 2017 (Orange) had a positive monthly trend in customer acquisition
- Despite total new customers up year on year, 2018 (Green) shows a **flat monthly trend** so far

WHAT IS IN AN ORDER?

PAYMENT PER ORDER

Average paid per order, by year



ORDER ITEMS PER ORDER

Average number of items per order

1.14

- Total payments per order is trending mostly **flat**.
- Number of items in one order **stagnant** at around 1 items per order.

WHAT IS GOING ON?

New customers only make 1 order and churn

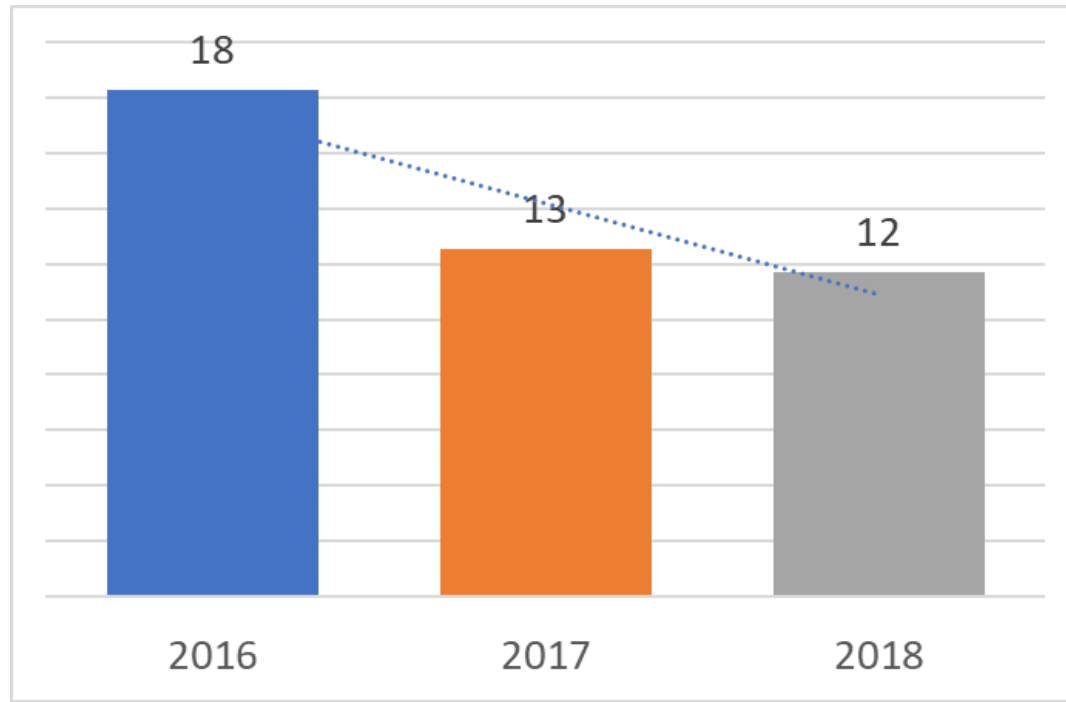
Despite this, company will grow if customer base is growing... but it is stalling

Revenue per order and the number of items in per order are also stagnant.

OPERATIONS REVIEW

LEAD

Average Lead Time per customer, YOY



Average Review Score (Scale 1-5)

4.08

- Lead Time = Time interval between order and delivery
- Lead time is trending **down**. Customer are getting their orders faster than ever.
- Average review score is **high** and continues to trend positively

REGRESSION MODEL (based on Customers)

PROCESS

- What are we looking to explain? – **6-month Customer Lifetime Value (LTV)**
- Selected 17 possible factors to test for
- Entered into Model
- Interpreted Results

RESULTS



OLS Regression Results

Dep. Variable:	total_paid_first_6_months	R-squared:	0.877				
Model:	OLS	Adj. R-squared:	0.877				
Method:	Least Squares	F-statistic:	5.664e+04				
Date:	Mon, 09 Dec 2019	Prob (F-statistic):	0.00				
Time:	02:19:32	Log-Likelihood:	-5.5396e+05				
No. Observations:	95557	AIC:	1.108e+06				
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Df Model:	12						
Covariance Type:	nonrobust						
		coef	std err	t	P> t	[0.025	0.975]
	Intercept	-92.7899	1.372	-67.611	0.000	-95.480	-90.100
	C(ordered_from_top_10_prod_category_bol)[T.1]	2.3119	0.539	4.288	0.000	1.255	3.369
	perc_orders_unavailable	179.6209	3.338	53.816	0.000	173.079	186.163
	avg_item_count_per_order	104.7614	0.558	187.655	0.000	103.667	105.856
	avg_product_count_per_order	-23.8962	1.369	-17.454	0.000	-26.580	-21.213
	average_price_per_unit	1.0418	0.002	672.494	0.000	1.039	1.045
	avg_freight_cost_per_order	1.2673	0.018	72.197	0.000	1.233	1.302
	avg_installments	0.7760	0.103	7.504	0.000	0.573	0.979
	perc_orders_boleto_voucher	1.4128	0.476	2.965	0.003	0.479	2.347
	avg_days_seller_processing_time	-1.0578	0.281	-3.766	0.000	-1.608	-0.507
	avg_days_transit_time	-1.3576	0.278	-4.879	0.000	-1.903	-0.812
	avg_days_lead_time	1.3025	0.278	4.691	0.000	0.758	1.847
	avg_days_survey_lag	1.0362	0.214	4.850	0.000	0.617	1.455
Omnibus:	284112.843	Durbin-Watson:	2.004				
Prob(Omnibus):	0.000	Jarque-Bera (JB):	82359323782.783				
Skew:	42.215	Prob(JB):	0.00				
Kurtosis:	4550.324	Cond. No.	2.97e+03				

REGRESSION MODEL (Results to note)

No Significance	Significant + Positive Correlation	Significant + Negative Correlation
Month customer acquired % orders shipped late % orders from sellers with perfect review	% orders from top 10 categories # payment installments % orders paid with boleto or voucher	Seller Processing Time Transit Time

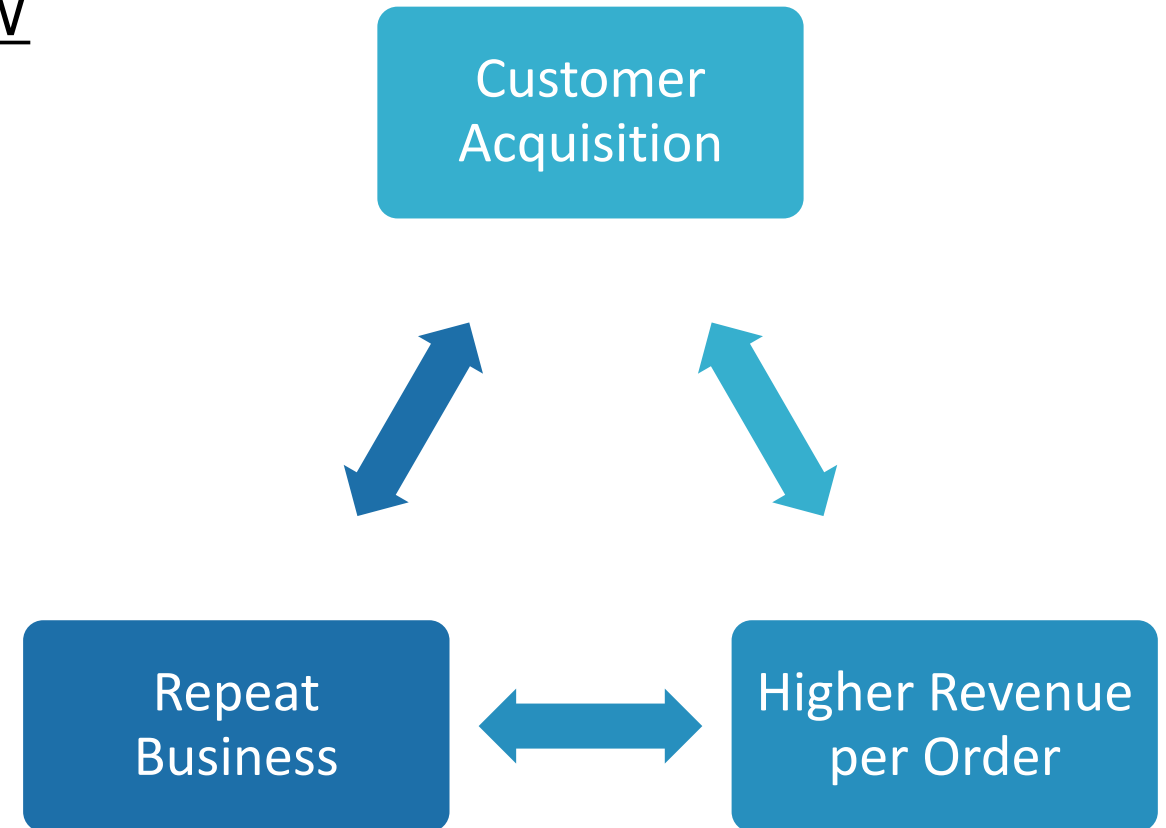
INITIAL RECOMMENDATIONS

MARKETING STRATEGY REVIEW



INVENTORY REVIEW

- 63% of customers order from your **top 10 categories**, despite having 71 available.
- Is this intentional?
- Should Olist consider focusing on those?



POTENTIAL NEXT STEPS FOR PROJECT

- Churn Rate or Survival Analysis
- Customer Segmentation
- Logistics Audit
- Content Review (NLP)
- Sentiment Analysis
- Seller Patterns
- Inventory Review

TOOLS USED:

- Postgres SQL
- Python:
Pandas, Numpy, Matplotlib, Statsmodel, Seaborn, Scipy

MORE INFORMATION ON THIS PROJECT:

- Appendix
 - Data Preparation
 - Model Optimization
- github.com/rachelleaperez
- linkedin.com/in/rachelleperez/

THANK YOU



APPENDIX TO:
A LOOK AT CUSTOMER LIFETIME VALUE

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DATA PREPARATION

What data do we have to address the problem?

DATA SOURCE

The image shows the Kaggle website interface. At the top is a dark navigation bar with the Kaggle logo, a search bar, and links for Competitions, Datasets, Notebooks, Discussion, and Courses. A notification bell and user profile icon are on the right. Below the navigation bar is a large banner for the 'Brazilian E-Commerce Public Dataset by Olist'. The banner features a colorful illustration of a city scene with people and buildings. Text on the banner includes 'Dataset', 'Brazilian E-Commerce Public Dataset by Olist', '100,000 Orders with product, customer and reviews info', and the Olist logo with the tagline 'empowering commerce'. Below the banner is a tabbed interface with 'Data' selected, followed by 'Kernels (63)', 'Discussion (24)', 'Activity', and 'Metadata'. To the right of these tabs are 'Download (120 MB)' and a 'New Notebook' button. At the bottom, there is a section for 'Usability 10.0', 'License CC BY-NC-SA 4.0', and 'Tags' including 'data visualization', 'eda', 'nlp', 'multiclass classification', and 'brazil'.

kaggle Search Competitions Datasets Notebooks Discussion Courses ...

Dataset

Brazilian E-Commerce Public Dataset by Olist
100,000 Orders with product, customer and reviews info

olist and 5 collaborators • updated a year ago (Version 7)

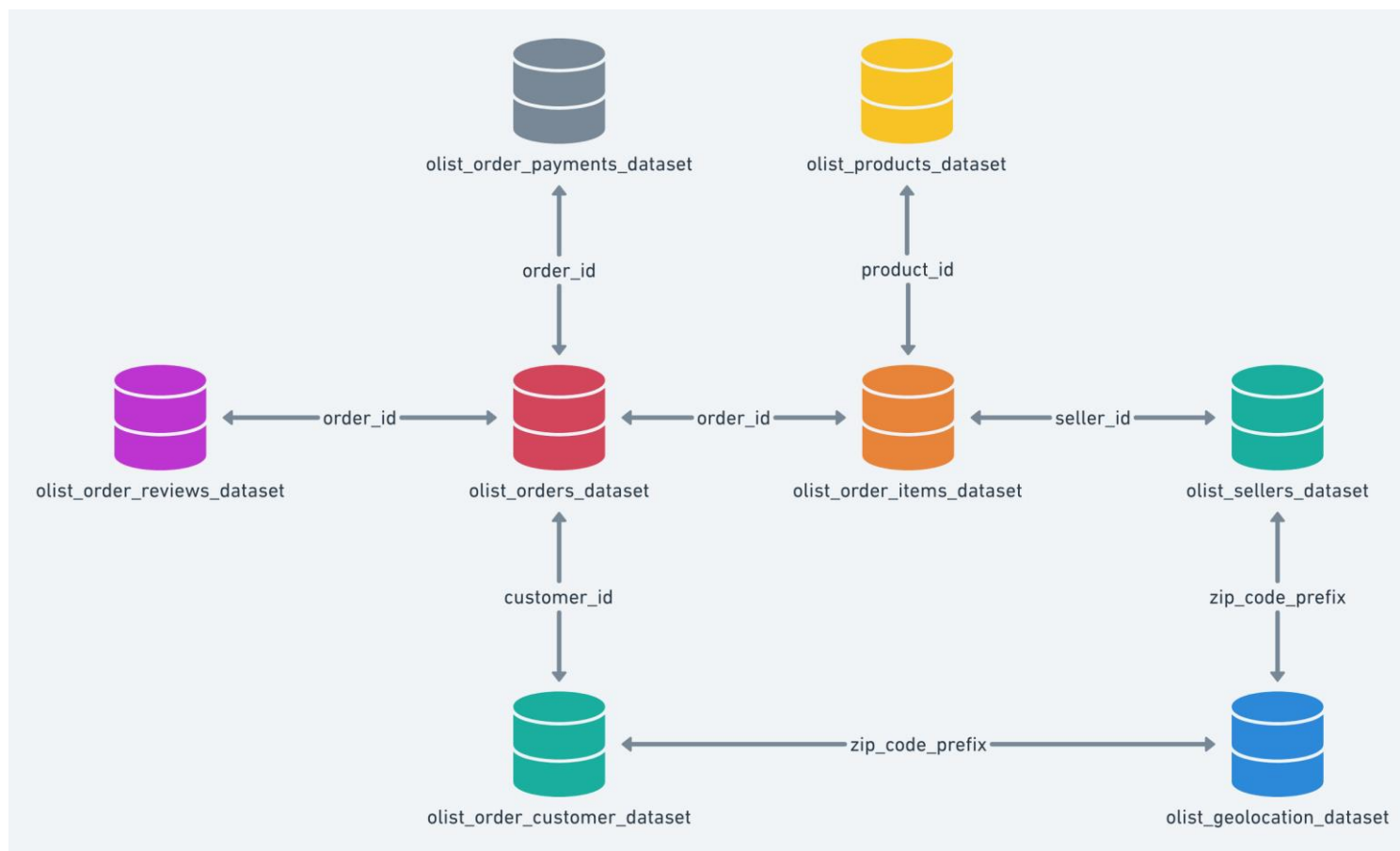
Data Kernels (63) Discussion (24) Activity Metadata Download (120 MB) New Notebook

Usability 10.0 License CC BY-NC-SA 4.0 Tags data visualization, eda, nlp, multiclass classification, brazil

<https://www.kaggle.com/olistbr/brazilian-ecommerce>

DATA PREVIEW

SCHEMA



INCLUDES

- Date Range: 9/4/2016 – 8/29/2018
- 96,096 Customers
- 99,441 Orders
- 32,951 Unique Products
- 3,095 Sellers

CHALLENGES

- Dependent variable (LTV) by customer and the only customer variables given are city, state, and zip. Aggregates must be created
- Aggregates are difficult as 1) Schema not linear 2) data is split for each **customer**, each **order** per customer, each **product** per order, and each **item** per product

DATA CLEANING

POSTGRESQL

Column	Type
customer_unique_id	character varying(50)
date_first_order	timestamp without time zone
year_first_order	double precision
month_first_order	double precision
total_orders_first_6_months	bigint
total_paid_first_6_months	double precision
order_count_unavailable	bigint
avg_payment_processing_time	interval
avg_seller_processing_time	interval
avg_transit_time	interval
avg_lead_time	interval
avg_item_count_per_order	numeric
avg_product_count_per_order	numeric
orders_shipped_late	bigint
avg_quantity_by_product	numeric
average_price_per_unit	double precision
avg_freight_cost_per_order	double precision
ordered_from_top_10_prod_category_bol	integer
ordered_from_seller_perfect_avg_review_bol	integer
avg_survey_lag	interval
avg_review_lag	interval
avg_review_score	numeric
order_count_boleto_voucher	bigint
order_count_card	bigint
avg_installments	numeric
avg_days_payment_processing_time	double precision
avg_days_seller_processing_time	double precision
avg_days_lead_time	double precision
avg_days_lead_time	double precision
avg_days_survey_lag	double precision
avg_daysreview_lag	double precision

- Create dataframe with aggregates
- Remove cancelled orders & orders not from customer's first 6 months

PYTHON

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 95557 entries, 0 to 95556
Data columns (total 31 columns):
customer_unique_id
date_first_order
year_first_order
month_first_order
total_orders_first_6_months
total_paid_first_6_months
order_count_unavailable
avg_payment_processing_time
avg_seller_processing_time
avg_transit_time
avg_lead_time
avg_item_count_per_order
avg_product_count_per_order
orders_shipped_late
avg_quantity_by_product
average_price_per_unit
avg_freight_cost_per_order
ordered_from_top_10_prod_category_bol
ordered_from_seller_perfect_avg_review_bol
avg_survey_lag
avg_review_lag
avg_review_score
order_count_boleto_voucher
order_count_card
avg_installments
avg_days_payment_processing_time
avg_days_seller_processing_time
avg_days_transit_time
avg_days_lead_time
avg_days_survey_lag
avg_daysreview_lag
dtypes: float64(17), int64(6), object(8)
memory usage: 22.6+ MB
```

COUNTS

NULLS

WRONG TYPE

```
95557 non-null object
95557 non-null object
95557 non-null int64
95557 non-null int64
95557 non-null int64
95557 non-null int64
95557 non-null float64
95557 non-null int64
95538 non-null object
94217 non-null object
93306 non-null object
93327 non-null object
94978 non-null float64
94978 non-null float64
85918 non-null float64
94978 non-null float64
94978 non-null float64
94978 non-null float64
95557 non-null int64
95557 non-null int64
85569 non-null object
88879 non-null object
95557 non-null float64
22800 non-null float64
75113 non-null float64
95557 non-null float64
95538 non-null float64
94217 non-null float64
93306 non-null float64
93327 non-null float64
85569 non-null float64
88879 non-null float64
```



```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 95557 entries, 0 to 95556
Data columns (total 37 columns):
customer_unique_id
date_first_order
year_first_order
month_first_order
total_orders_first_6_months
total_paid_first_6_months
order_count_unavailable
avg_payment_processing_time
avg_seller_processing_time
avg_transit_time
avg_lead_time
avg_item_count_per_order
avg_product_count_per_order
orders_shipped_late
avg_quantity_by_product
average_price_per_unit
avg_freight_cost_per_order
ordered_from_top_10_prod_category_bol
ordered_from_seller_perfect_avg_review_bol
avg_survey_lag
avg_review_lag
avg_review_score
order_count_boleto_voucher
order_count_card
avg_installments
avg_days_payment_processing_time
avg_days_seller_processing_time
avg_days_transit_time
avg_days_lead_time
avg_days_survey_lag
avg_daysreview_lag
Active?
perc_orders_unavailable
perc_orders_shipped_late
perc_orders_boleto_voucher
perc_orders_credit_debit
day_first_order
dtypes: datetime64[ns](1), float64(20), int32(1), int64(6), object(3), timedelta64[ns](6)
memory usage: 26.6+ MB
```



- Replace or drop nulls & update data types
- Columns with counts turned to proportion of total orders to fairly compare customers

VARIABLES AVAILABLE (36)

CUSTOMER BEHAVIOR

customer_unique_id
date_first_order
year_first_order
month_first_order
total_orders_first_6_months
avg_review_lag
total_paid_first_6_months
avg_quantity_by_product
avg_item_count_per_order
avg_product_count_per_order
average_price_per_unit
ordered_from_top_10_prod_category_bol
ordered_from_seller_perfect_avg_review_bol
avg_review_score
order_count_boleto_voucher
order_count_card
avg_installments
perc_orders_boleto/voucher
perc_orders_credit/debit
Active?

LOGISTICS

order_count_unavailable
avg_payment_processing_time
avg_seller_processing_time
avg_transit_time
avg_lead_time
orders_shipped_late
avg_freight_cost_per_order
avg_survey_lag
avg_days_payment_processing_time
avg_days_seller_processing_time
avg_days_transit_time
avg_days_lead_time
avg_days_survey_lag
perc_orders_unavailable
perc_orders_shipped_late

18 Highlighted = Predictors for Model

REGRESSION MODEL

Iterations and Results

REGRESSION MODEL - LTV

MODEL USED

- Statsmodel **Ordinary Least Squares (OLS)** Regression
- Model to **explain** correlation between variables and 6-month customer's lifetime value

RESPONSE VARIABLE

total_paid_first_6_months

PREDICTOR VARIABLES (17)

month_first_order (C)	avg_freight_cost_per_order
ordered_from_top_10_category_bol (C)	avg_installments
ordered_from_seller_perfect_avg_reviews bol (C)	perc_orders_shipped_late
perc_orders_unavailable	perc_orders_boleto_voucher
avg_review_score	avg_days_payment_processing_time
avg_item_count_per_order	avg_days_seller_processing_time
avg_product_count_per_order	avg_days_transit_time
average_price_per_unit	avg_days_survey_lag
	avg_daysreview_lag

MODEL #1

- **22** Variables (including "dummy variables" from categorical data)
- R-squared = **0.877**
- **Missing** timedelta variables
- **4** insignificant variables (P-Score < 0.05)

Dep. Variable:	total_paid_first_6_months	R-squared:	0.877
Model:	OLS	Adj. R-squared:	0.877
Method:	Least Squares	F-statistic:	3.089e+04
Date:	Sun, 08 Dec 2019	Prob (F-statistic):	0.00
Time:	20:17:39	Log-Likelihood:	-5.5397e+05
No. Observations:	95557	AIC:	1.108e+06
Df Residuals:	95534	BIC:	1.108e+06
Df Model:	22		
Covariance Type:	nonrobust		

	coef	std err	t	P> t	[0.025	0.975]
Intercept	-89.2818	1.945	-45.903	0.000	-93.094	-85.470
C(month_first_order)[T.2]	-0.0388	1.267	-0.031	0.976	-2.522	2.444
C(month_first_order)[T.3]	0.1672	1.220	0.137	0.891	-2.225	2.559
C(month_first_order)[T.4]	-0.4776	1.235	-0.387	0.699	-2.899	1.944
C(month_first_order)[T.5]	1.0265	1.203	0.853	0.394	-1.332	3.385
C(month_first_order)[T.6]	-2.0731	1.236	-1.677	0.094	-4.496	0.350
C(month_first_order)[T.7]	-0.4061	1.211	-0.335	0.737	-2.780	1.967
C(month_first_order)[T.8]	-1.5415	1.198	-1.286	0.198	-3.890	0.807
C(month_first_order)[T.9]	5.8527	1.539	3.803	0.000	2.836	8.869
C(month_first_order)[T.10]	-1.8868	1.470	-1.283	0.199	-4.768	0.995
C(month_first_order)[T.11]	0.2644	1.302	0.203	0.839	-2.288	2.817
C(month_first_order)[T.12]	-0.1119	1.407	-0.080	0.937	-2.870	2.646
C(ordered_from_top_10_prod_category_bol)[T.1]	2.2730	0.540	4.213	0.000	1.216	3.330
C(ordered_from_seller_perfect_avg_review_bol)[T.1]	-1.2899	2.627	-0.491	0.623	-6.440	3.860
perc_orders_unavailable	177.5627	3.456	51.376	0.000	170.789	184.337
avg_review_score	-0.2671	0.200	-1.337	0.181	-0.659	0.124
avg_item_count_per_order	104.7951	0.560	187.250	0.000	103.698	105.892
avg_product_count_per_order	-24.0295	1.368	-17.566	0.000	-26.711	-21.348
average_price_per_unit	1.0420	0.002	673.569	0.000	1.039	1.045
avg_freight_cost_per_order	1.2655	0.017	73.331	0.000	1.232	1.299
avg_installments	0.7602	0.103	7.348	0.000	0.557	0.963
perc_orders_shipped_late	-1.3082	0.891	-1.468	0.142	-3.054	0.438
perc_orders_boleto_voucher	1.5403	0.475	3.245	0.001	0.610	2.471

Omnibus:	283960.659	Durbin-Watson:	2.004
Prob(Omnibus):	0.000	Jarque-Bera (JB):	81936180578.570
Skew:	42.153	Prob(JB):	0.00
Kurtosis:	4538.626	Cond. No.	3.10e+03

MODEL OPTIMIZATION

Model #	Change Description	N. Of Variables (including dummy)	R-Squared	N. of Variables with p-square > 0.05
1		22	0.877	4
2	Adds timedelta variables as n. of days	28	0.877	6
3	Removes month_first_order (C)	17	0.877	5
4	Removes perc_orders_shipped_late	16	0.877	4
5	Removes ordered_from_seller_perfect_review (C)	15	0.877	3
6	Removes avg_days_payment_processing_time	14	0.877	2
7	Removes avg_daysreview_lag	13	0.877	1
8	Removes avg_review_score	12	0.877	0
9	Normalizes data (z-score)	12	0.877	0

FINAL MODEL

- **Positive** Correlation to LTV
 - **ordered from top 10 categories**
 - % orders unavailable
 - Item count per order
 - price per unit
 - freight cost per order
 - **# payment installments**
 - **% orders paid boleto or voucher**
 - Survey Lag (from Olist to customer)
- **Negative** Correlation to LTV
 - product count per order
 - **seller processing time**
 - **transit time**

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Omnibus:	284112.843	Durbin-Watson:	2.004
Prob(Omnibus):	0.000	Jarque-Bera (JB):	82359323782.783
Skew:	42.215	Prob(JB):	0.00
Kurtosis:	4550.324	Cond. No.	2.97e+03