MARKET & RETAIL ANALYSIS

BY: BHAVENDRA SAHU



AGENDA

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- Data methodology
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To manage the inventory cost of E-commerce company Olist.

OBJECTIVE



To help OList to identify the product categories which they can get rid of without significantly impacting business.



To identify top products that contribute to the revenue.



Use market basket analysis to analyze the purchase behavior of individual customers to estimate with relative certainty, what items are more likely to be purchased individually or in combination with some other products.



OList is one such e-commerce company that has faced some losses recently and they want to manage their inventory very well so as to reduce any unnecessary costs that they might be bearing.

BACKGROUND



They need to identify top products that contribute to the revenue.



They want to analyze the purchase behavior of individual customers to estimate with relative certainty, what items are more likely to be purchased individually or in combination with some other products.

KEY FINDINGS

- The graph shows the top 20 products ordered by quantity.
- The most ordered product has been purchased 467 times.
- Most of the top products ordered are from 'toys' category.



TOP 20 PRODUCTS BY REVENUE

- The highest revenue generated by the top product is **63,885**.
- The product that generated highest revenue belongs to the 'Toys' category.
- Most of the products belongs to the 'toys' category contributing to higher revenue.



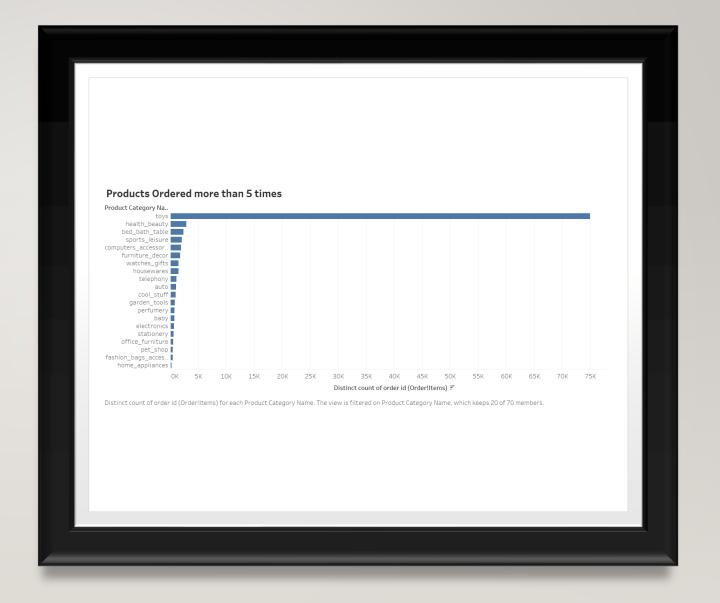
REVENUE PARETO

- This Revenue Pareto shows the number of orders, percentage of total running revenue and revenue generated for each product id.
- About 25% of the top products having higher revenues contributing to around 80% of total revenue.
- It helps to identify the contribution of the products towards total revenue.

Revenue Analysis Product Id bb50f2e236e5eea0.. 63,885 6cdd53843498f928.. 0.91% 54,730 d6160fb7873f1840.. d1c427060a0f73f6b.. 47,215 99a4788cb2485696.. 43,026 25c38557cf793876c.. 38,907 aca2eb7d00ea1a7b.. 53b36df67ebb7c41.. 5f504b3a1c75b73d.. e0d64dcfaa3b6db5c.. d285360f29ac7fd97.. 7a10781637204d8d.. f819f0c84a64f02d3.. 28,292 422879e10f466829.. 16c4e87b98a9370a.. a62e25e09e05e6faf... 2b4609f8948be188.. a5215a7a9f46c418.. 21,740 389d119b48cf3043... 368c6c730842d780... 6.60% 20,593 53759a2ecddad2bb.. 52c80cedd4e90108. 1dec4c88c685d5a0.. 19c91ef95d509ea3.. 1a080577618e7fe4... 165f86fe8b799a708..

PRODUCT CATEGORIES ORDERED MORE THAN 5 TIMES

- The graph shows the top 20 categories ordered more than 5 times.
- Toys, health_beauty ,
 bed_bath_tables, sports_leisure &
 computer_accessories etc. are the
 top products ordered more than 5
 times.
- The 'Toys' category is the most ordered category with a total of 74,929 orders.



MARKET BASKET ANALYSIS

- Market basket analysis is performed to identify the frequently ordered category association.
- 'bed_bath_table' associated very well with 'toys'.
- 'Furniture_decor', 'computers_accessories' go along pretty well with 'toys'.
- 'health_beauty', 'housewares', 'watches_gifts' etc. are also be considered to order with 'toys'.
- Hence the combination of all of these few categories with toys & each other are frequently ordered.



RECOMMENDATIONS

Target customers who are more likely to buy 'toys' such as parents of newly born and 1-2 years child's.

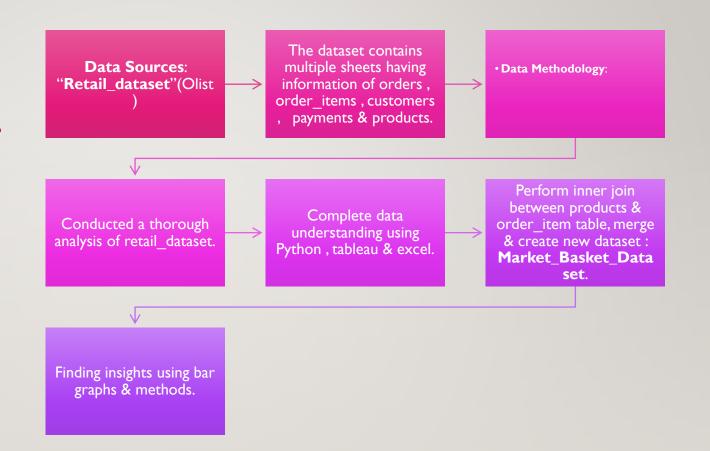
Offer some discounts or coupons on frequently ordered category associations and the most ordered products to attract more customers.

Olist can consider removing unnecessary products & categories which are not much of a major contributors neither in terms of maximum number of orders nor in terms of revenue.

Olist can optimize average price & shipping charges for some of the products which are not much of contributing to revenue.

'**Toys**' should be held in enough stocks as they are contributing to significant amount of revenue.

APPENDIX



DATA ASSUMPTIONS & EXECUTIVE SUMMARY

- Only the cases having order status as 'delivered' are to be considered.
- The columns 'order_approved_at' and 'order_delivered_timestamp' are assumed to be equivalent to/same as the columns 'order purchase timestamp' and 'order estimated delivery date' respectively.

Data Cleaning using Python:

- ORDERS: Only the orders with order status as 'delivered' are considered for this case study since 97% of the records are successfully delivered. The missing values for 'order_approved_at' are replaced with respective 'order_purchase_timestamp'. Similarly, the missing values for 'order_delivered_timestamp' are replaced with respective 'order_estimated_delivery_date'.
- Order Items: No missing values, no duplicate values, No outlier treatment done as it may lead to data loss.
- <u>Customers</u>: No missing values, Duplicate customer id's were dropped, No outliers detected.
- Payments: No missing values, Duplicate order id's were dropped, No outlier treatment done as it may lead to data loss.
- <u>Products</u>: The missing values of product category name has been imputed with the mode, product weight, height, length & width imputed with median, No duplicate values.