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O1
Context & Problem
Statement

# Context

#### **Demand Forecasting**

Critical for business success
Optimize operations + Maximize profits
Over-stocking => extra business cost
Under-stocking => loss of revenues

# **Importance**



Operations Optimization

Resources allocation



Inventory Management

Storage & Logistics



**Profitability** 

Cost efficiency & Cashflow



**Customer Satisfaction** 

Growth & Sustainability



# Challenges

#### **Data**

Availability & Quality

## Human Errors & Bias

Personal judgement & stakeholders conflicting interests

## Market Uncertainty

Business dynamics & Economic conditions

#### **Lack Expertise**

Data Analysis, Statistics & Domain Knowledge

## **Customer Behavior**

Trends & Seasonality changes rapidly

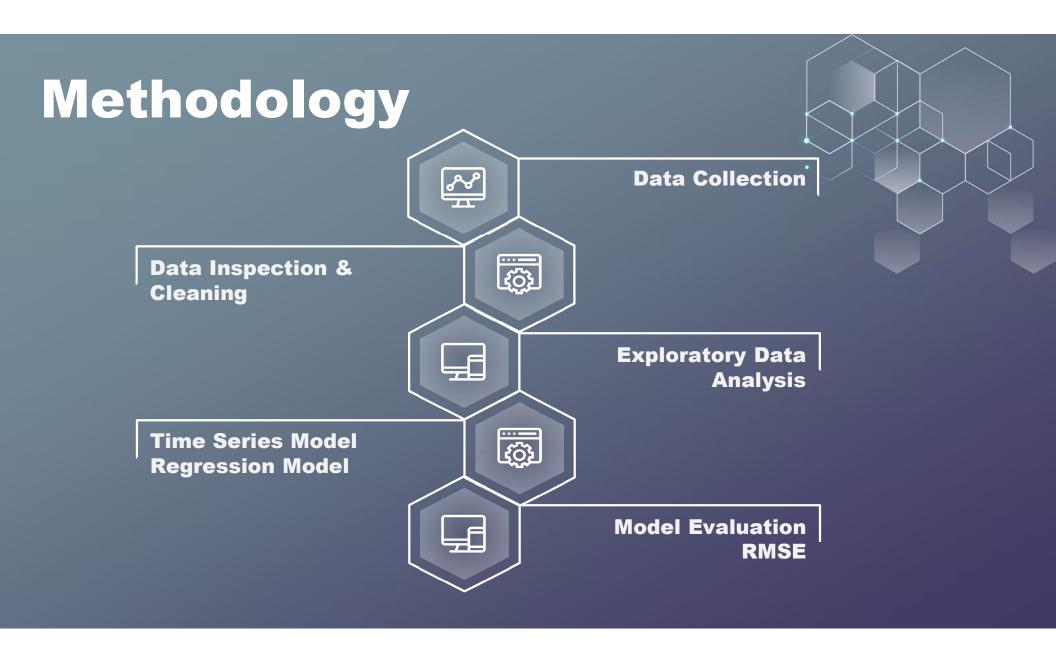
### Complex Supply Chain

Global sourcing & interdependency



# Problem Statement

Develop a short term forecasting model to help business effectively plan for demand surge in next 3 months in conjunction with festive season and special events, as current rule-based practice not able to predict seasonality





# O2 Data Collection & Cleaning

# **Data Collection & Cleaning**

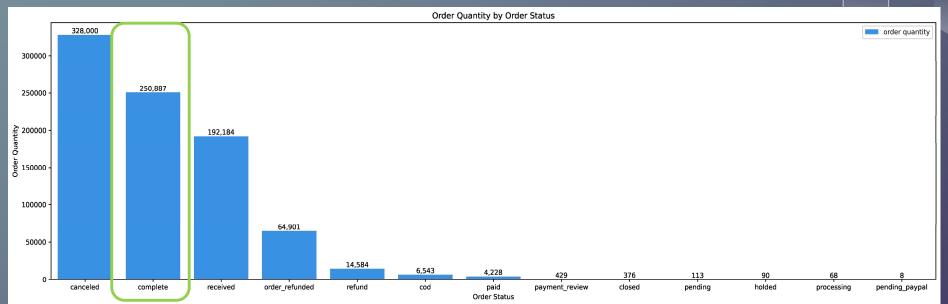
- 1. Data source: online ecommerce sales datasets 286,000 rows, 35 features
- 2. Period: October 2020 to September 2021 (12 months of daily sales)
- 3. Inspection for nulls, duplication, data types
- 4. Standardization of columns
- 5. Convert date columns to datetime index
- 6. Further inspection on mathematical accuracy, nulls, duplication after cleaning



CS
Exploratory
Data Analysis
(EDA)

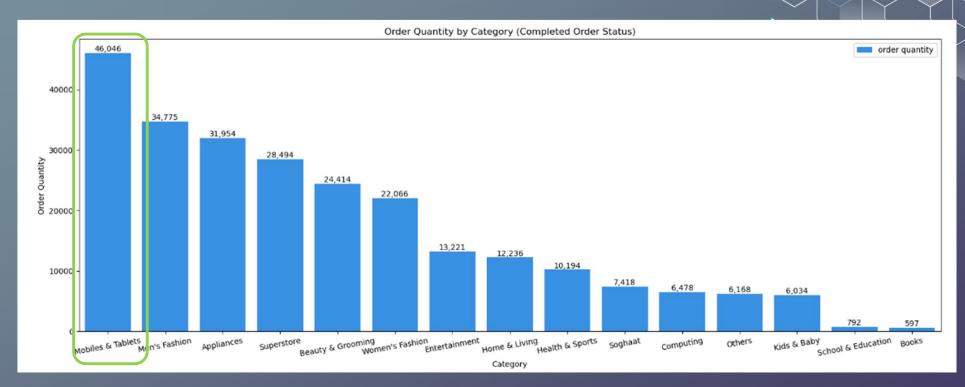
#### **Overall Order Status**





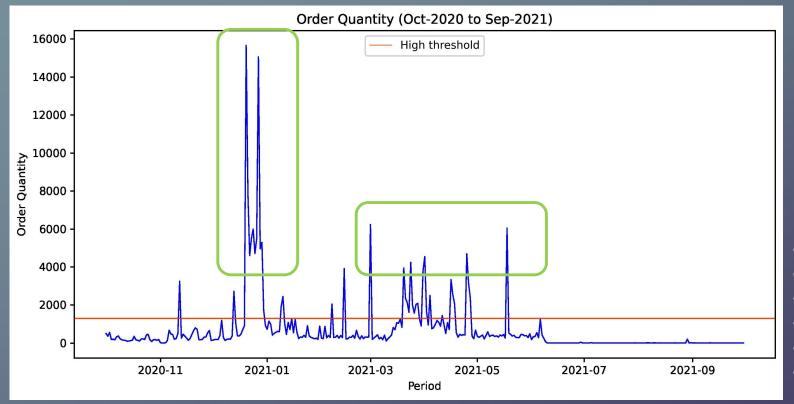
- Order status indicated as 'completed' means product is delivered and payment is received
- Cancelled order status is significant hence business should gather data on reasons of cancellation

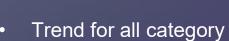
## **Order Quantity by Category**



- Mobiles & Tablets is top sellers and has more daily sales, hence it is selected for modelling
- Top 3 categories made up 45% of total completed orders

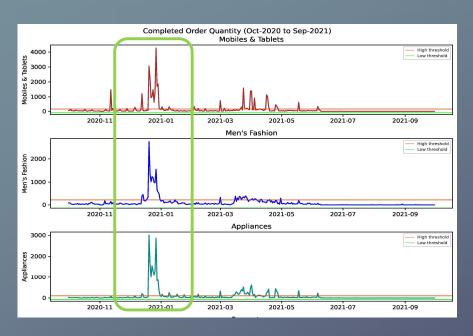
# Trends of Order Quantity

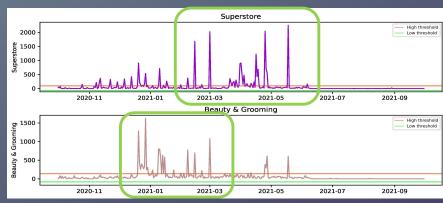




- Outliers = peaks
- Dec-2020 = Christmas
- Threshold = 1299 units
- Low demand
- Cyclical business

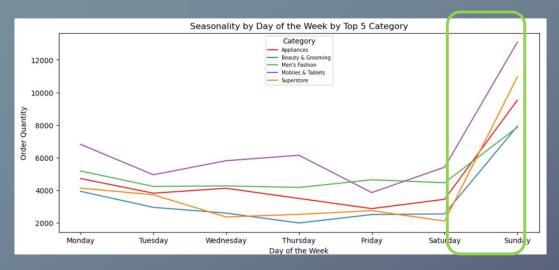
# **Trends by Top 5 Category**





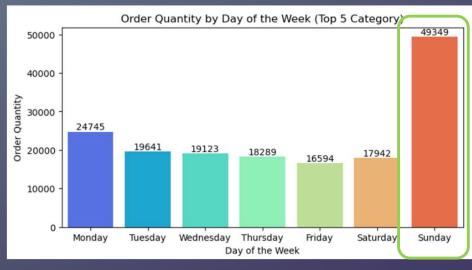
- Trend for top 3 category are similar and follows closely with all category
- Top 3 = Mobiles & Tablets, Men's Fashion and Appliances
- Next 2 category follows a different trend, hence one product category is chosen for modelling.
- Flat demand from Jun to Sep 2021

## **Day of Week Analysis**

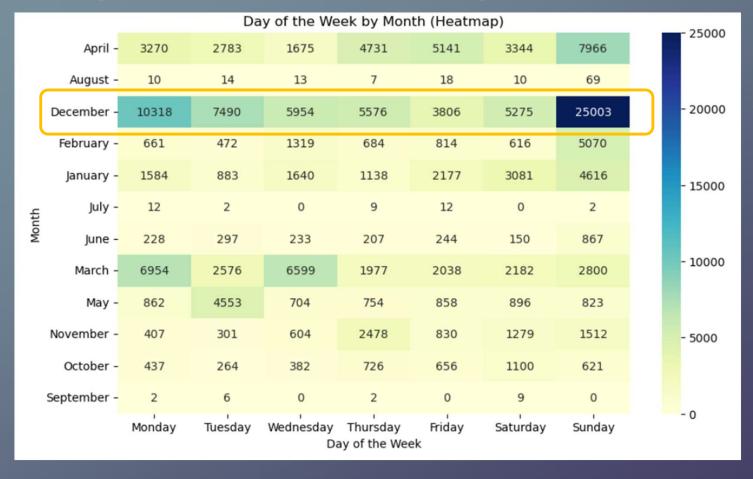


- Top 5 category shows similar trend for weekly trend
- Sundays are the popular day to do online shopping
- Fridays are the lowest due to socialising





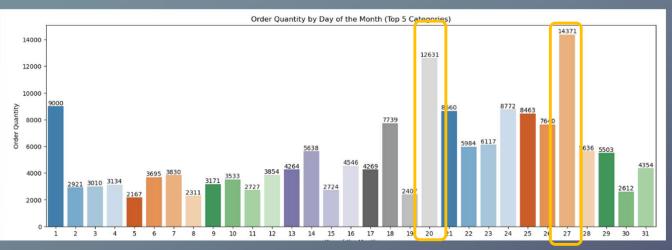
## Day of Week Analysis





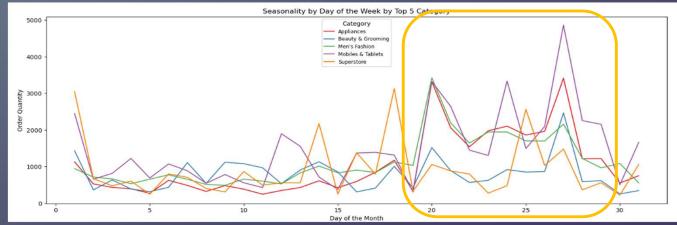
- Correlation between day of week and month
- Top 5 categories
- Sundays in December exhibit strongest correlation with orders

## **Day of Month Analysis**

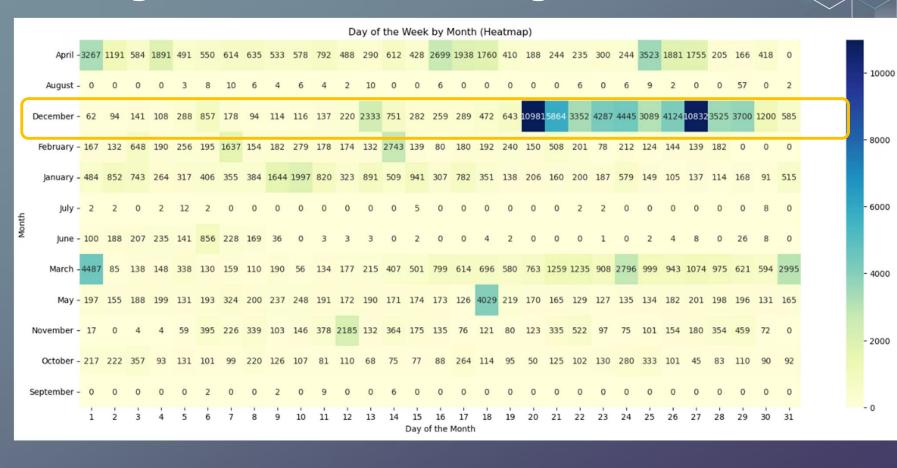




- 27<sup>th</sup> and 20<sup>th</sup> = Christmas, weekends
- Top 5 category shows similar uptrend towards end of month



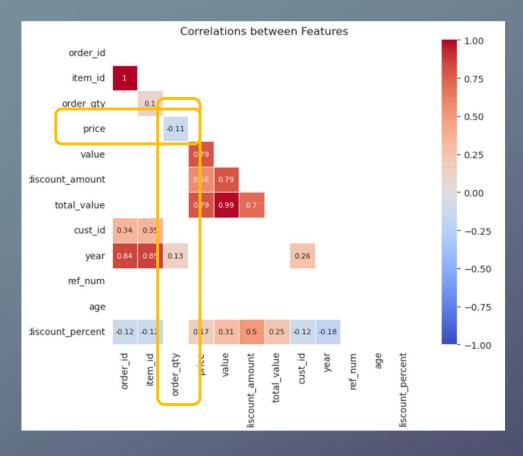
## **Day of Month Analysis**

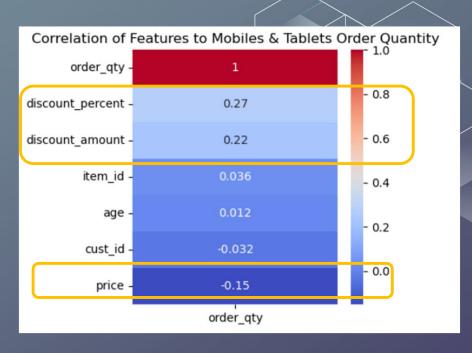


Correlation between day of month and month

20<sup>th</sup>-29<sup>th</sup> Dec 1<sup>st</sup> Mar 18<sup>th</sup> May

#### Correlation





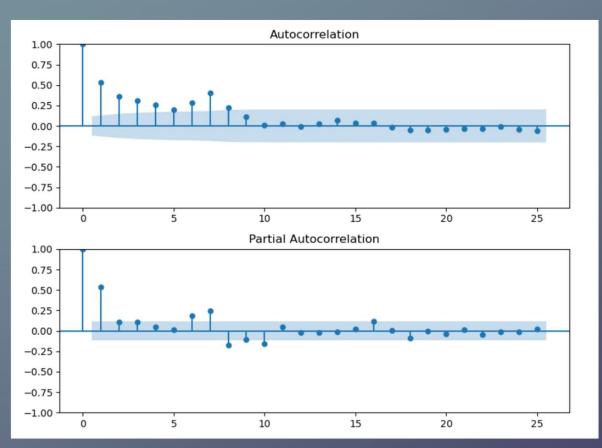
- All category: order quantity vs price = weak negative correlation
- Mobiles & Tablets: order quantity vs discount weak positive correlation
- Mobiles & Tablets: order quantity vs price weak negative correlation

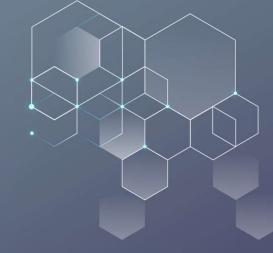


O4 Model & Evaluation



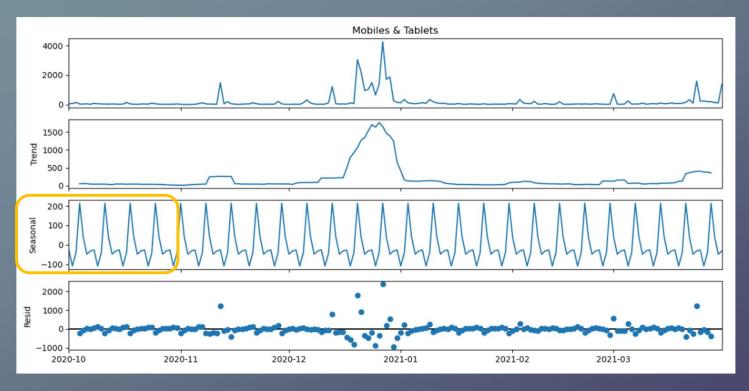
# Stationarity, AR and MA





- Augmented Dickey–Fuller test (ADF Test)
- Kwiatkowski-Phillips-Schmidt-Shin test (KPSS Test)
- p-value for ADF test < 0.05 = reject null => stationary
- p-value for KPSS test > 0.05 = => stationary
- Autocorrelation (ACF) lags not in negative zone– no differencing

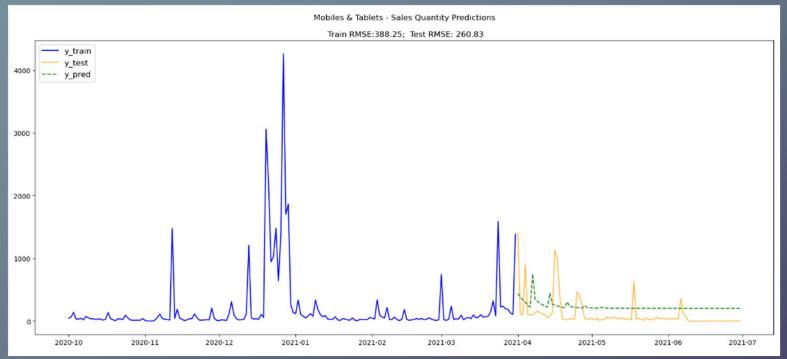
# Seasonality





- > Trend is not linear
- Seasonality within a month, each cycle lasts a week
- Residuals are high in Dec-2020

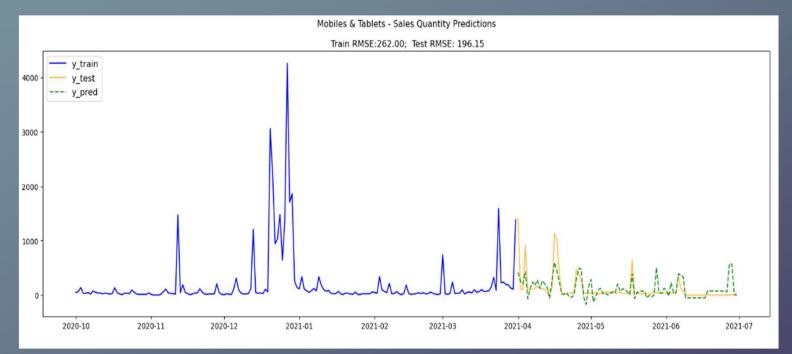
# SARIMA





- SARIMA overfitting
- Train RMSE: 388
- Test RMSE: 261
- SARIMA not managed to predict well

## **SARIMAX**

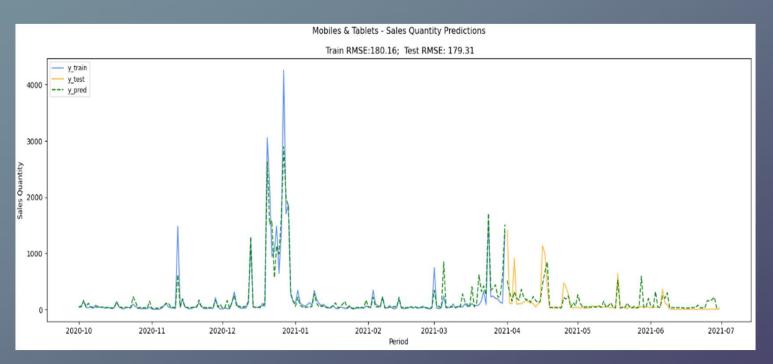




- Less overfitting with exogenous variables
  - Train RMSE: 262
- Test RMSE: 196
- SARIMAX able to predict better than SARIMA



## **Random Forest Model**

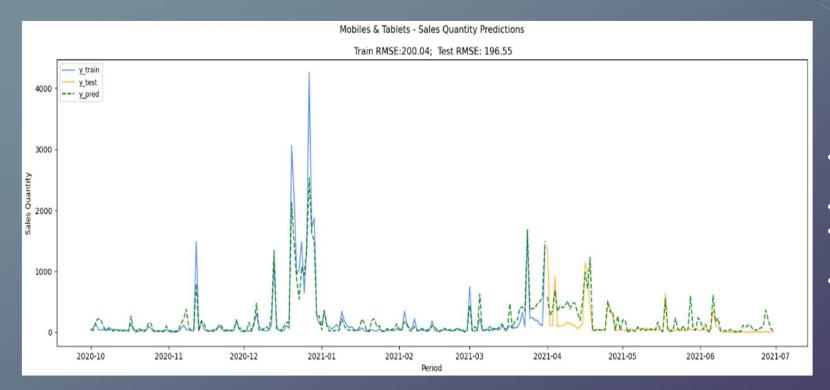




	feature	importance
12	$ohe\_payment\_method\_easypay\_voucher$	0.605682
40	ss_discount_amount	0.203963
41	ss_age	0.089821
2	oheday_of_week_Sunday	0.049896
7	ohe_week_in_month_3	0.024299

- Good fit
- Predicted trends very well
- Lowest RMSE
- Train RMSE: 180
- Test RMSE: 179
  - Feature importance:
    payment method, discount,
    age, Sundays, week 3 of
    the month

## **XGBoost Model**



Good fit

Train RMSE: 200

Test RMSE: 197

Predicted trends better than time series models

# **Model Comparison**





#### **SARIMAX**

Train RMSE: 262 Test RMSE: 196 2



#### **Random Forest**

Train RMSE: 180 Test RMSE: 179 3

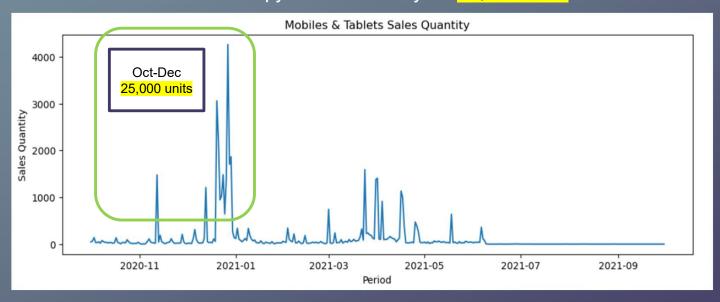


#### **XGBoost**

Train RMSE: 200 Test RMSE: 197

# **Cost-Benefit Analysis**

- Mobiles & Tablets average price = \$712
- Mobiles & Tablets margin = 43%
- Mobiles & Tablets average cost = \$306 [\$712 \* 43%]
- Current state historical mean = 169 units
- Oct-Dec forecasted qty = 169 \* 90 days = 15,210 units





- Loss of business = 10,139 units
- Product margin loss = \$3,136,587 !!!
- Future business loss



# O4 Conclusion & Recommendation

## Conclusion

#### **Best Model**

#### **Random Forest**

Good fit
Hyperparameter
tuning
Feature importance

# Importance of Forecasting

Loss of current orders

Loss of FUTURE business

#### Data

Quality & Quantity Prediction Accuracy Features Selection

# Recommendations

#### **Data Collection**

More Historical Data Product Attributes

#### **Feature Engineering**

Interactive terms
Market Basket Analysis
Holidays
Special Events

#### **Model Selection**

Different Models
Seasonality
Trends

