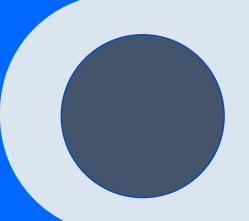
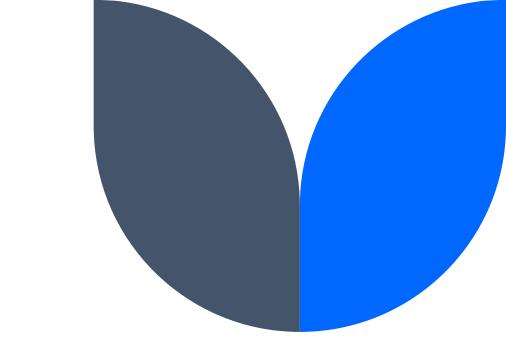
Best Buy Presentation







Introduction

Slow-selling SKUs need forecasts



Make up a large quantity of SKUs



Forecasts need to be accurate and efficient







Timeline



Exploratory Data Analysis (EDA)

Preliminary analysis to understand data better and assist in feature engineering



Approach Exploration

Compared baseline results of 4 different models



Model Results

Analyzed and interpreted model results

Based on EDA, created features to assist with model performance

Feature Engineering



Approach and considerations to improve chosen models

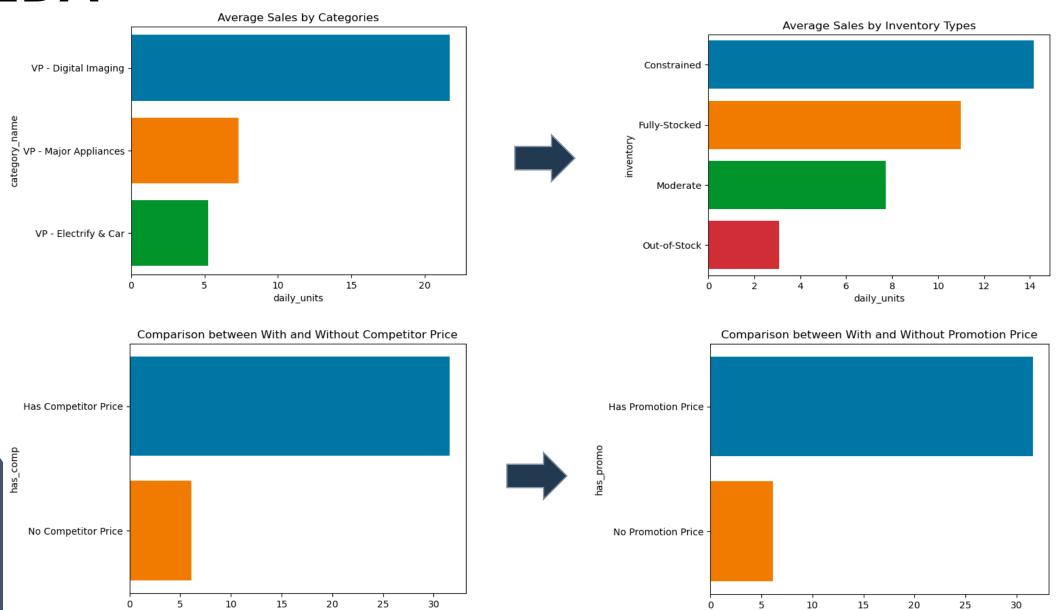
Model Refinement



Exploratory DataAnalysis



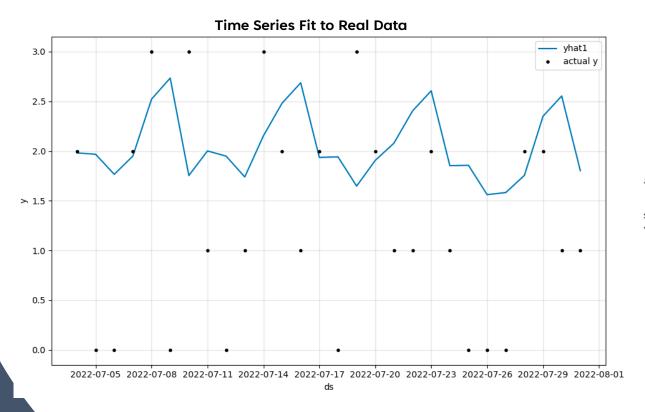
EDA



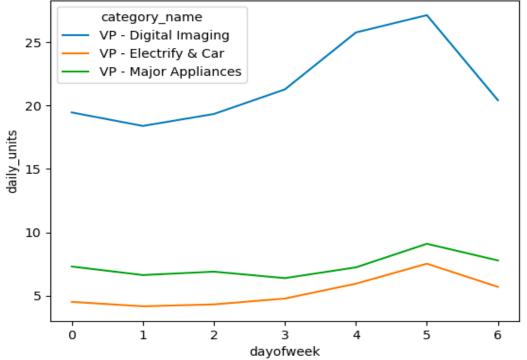
DAILY_UNITS

DAILY_UNITS

EDA continued

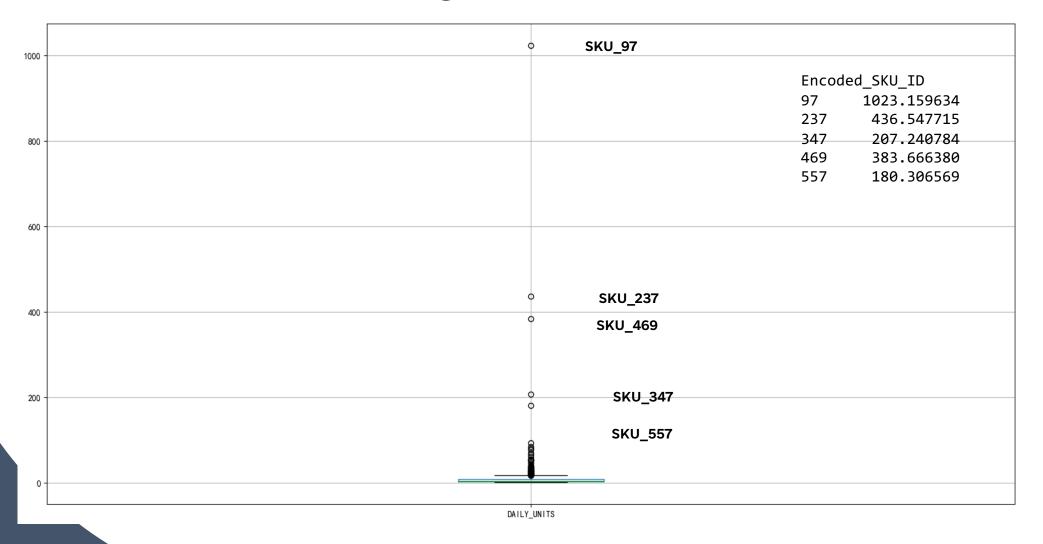


Average Sales of Day of Week Grouped by Category Name





EDA Outlier Analysis





Feature Engineering



Feature Engineering

Category Features

- Sales vary by category
- Ex: Flash
 memory has
 much higher
 sales

Price Features

- Promo prices increase sales
- Having lower price than competitors = higher sales

Inventory Features

- Sales slow when items are out-ofstock
- Sales increase when supply is limited

Time Features

- Sales increase during holidays
- Sales increase during weekends









Approaches Exploration



Approaches

Time Series Approach:

Time series is traditional, but analysis shows that the sales data is too sparse and might not be appropriate.

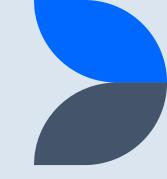
Regression Approach:

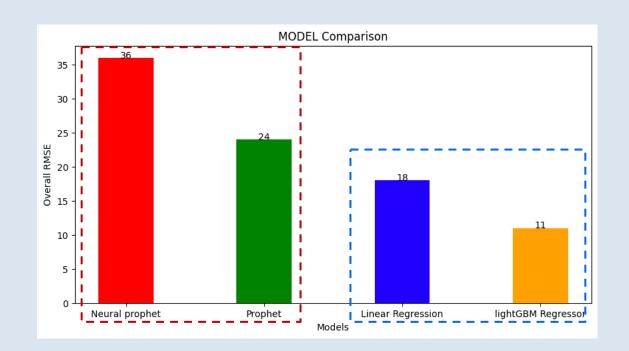
Formulate 7 regression problems for 7 days of weeks due to the clear weekly trend.

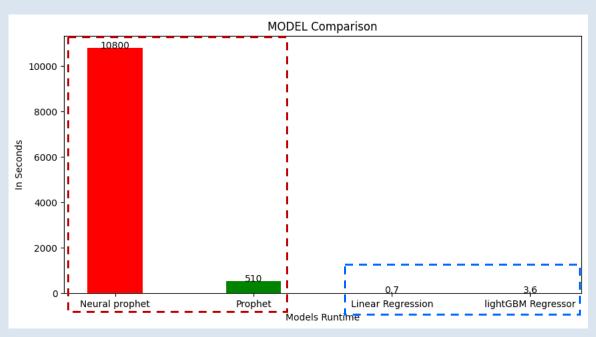
We separated data into train (before 07-04-2022) and validation (07-04-2022 to 07-31-2022).



Approach Comparison







RMSE Comparison

Runtime Comparison (seconds)

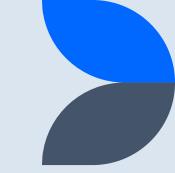
Time-series Regression approaches



Model Refinement







1

New Features

Added additional features such as: holidays, median sales same category, etc.

2

Outlier Removal

Analyzed models' errors and iteratively improved performance with new features and tuning 3

Feature Selection

Used LASSO for feature selection

4

Model Tuning

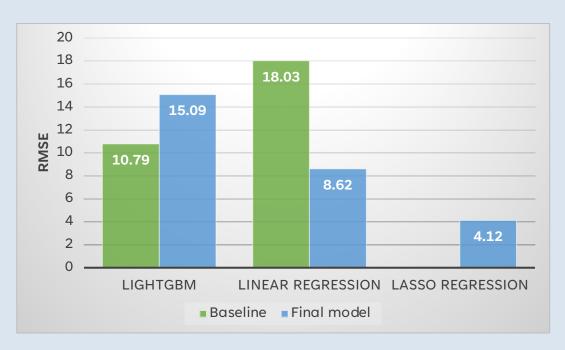
Tuned models' hyper parameters



Model Results



Model Results



RMSE by SKU distribution of lasso model

200 95%
175 150
125 100
75 50
25 0 3.73 10 20 30 40 50

RMSE baseline vs Final model

Distribution of RMSE by SKU for Lasso Regression



Summary

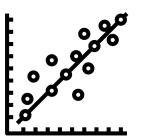


Summary

Regression approach outperformed time series approach

Improvement to 4.12 RMSE & 0.7 second run time

Can run forecast as frequently as needed









Questions?

The goal of forecasting is not to predict the future but to tell you what you need to know to take meaningful action in the present.

-Paul Saffo

