

Customers Behavior Analysis of Taobao & TMall



天猫 TMALL.COM



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Agenda

I. Introduction and Overview

II. Interesting Findings Through Analysis

III. Returned Customer Analysis

Introduction & Overview

Dataset: traffic data of Taobao & TMall from 05/2018 to 11/2018,
found it from Database Lab of Xiamen University

Variables:

user_id | item_id | cat_id | merchant_id | brand_id

month | day

action | age_range | gender | province

Tools: Spark

Objective: better understand the e-commerce market &
customer behavior

Finding About Shopping Festival



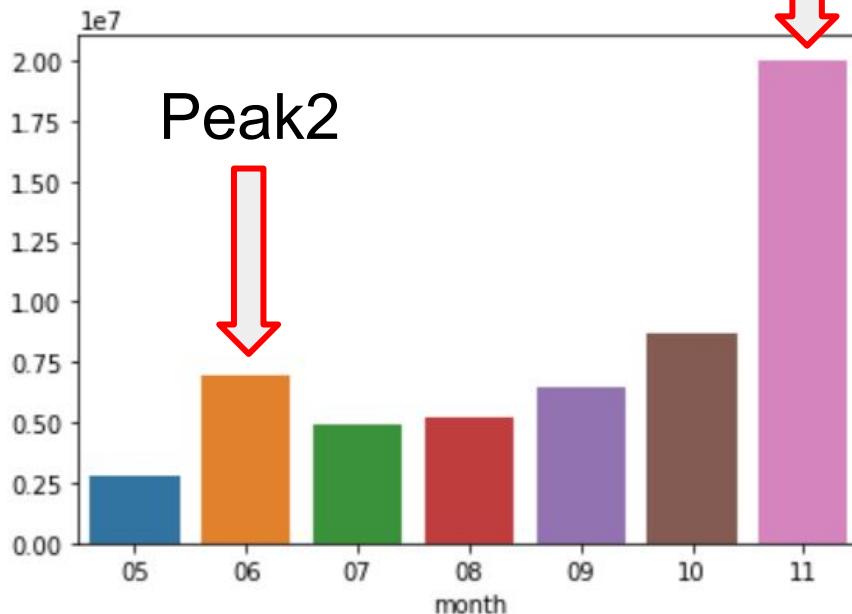
Pattern Through Time



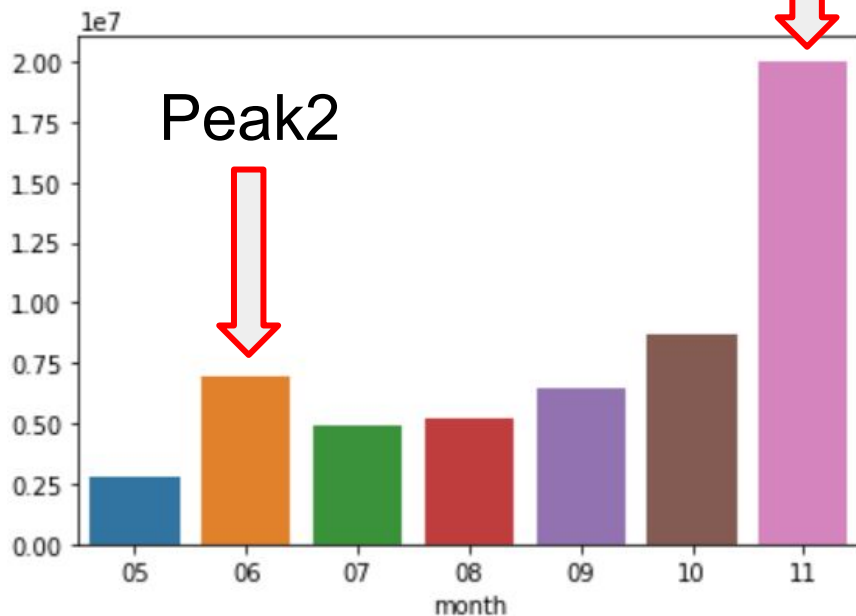
Peak1



Where does the Peak 2 come from?



Peak1



618 Festival

month day
06 18

1395895

month day
11 11

10582633

10 times larger!



618 Festival

Why 618 Festival is far less influential than double 11 Festival?

More About Shopping Festival



Maybe we can find some hint from the data to verify the difference.

Follow Up Validation

6.18

Cat.ID	Count
662	144344
1505	64673
602	57396
737	56980
389	44413
1577	42659
1271	33382
407	31548
302	27642
821	27067

Blowing Up !

Overlap

11.11

Cat.ID	Count
1349	640718
656	484089
389	429461
1142	410082
662	371260
737	276413
1208	252022
602	235353
1553	215133
177	202225

Returned Customer Analysis

Data Manipulation & Variables

How to define repeat customer?

1. User_ID + Merchant_ID
2. Different purchased date

If one user purchased in three different shops --> Three unique observations

If one user purchased in the same shop for more than one time --> Returned customer

Variables Used:

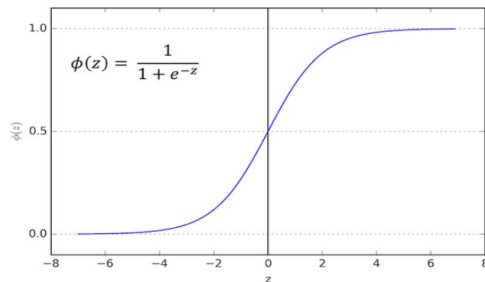
Age_Group:

Gender: Male / Female

- 1 [0 ,18]
- 2 [18,24],
- 3 [25,29],
- 4 [30,34],
- 5 [35,39],
- 6 [40,49]

Model & Result Analysis

- **Logistics Regression:**



- **Result Formular:**

$$y = \text{sigmoid} (-3.19 + 0.302 \times \text{Age_Range} + 0.044 \times \text{Is_female})$$

- **Accuracy** = 96.27% (based on 15% of test set)
- **AUC** = 0.5413 (slightly better than random guess)
- **Why** → **Super imbalanced data**
(3.71% Positive observations)

What can we do next?

1. Adding more Features: Item category, Merchant, Brand, City etc. (has to be done with one-hot encoding because of high cardinality)
2. Features engineering - date variables, extract extra features (is holiday, is weekend)
3. Try more models (SVM, RF, GBT) / Ensemble
4. Data augmentation - upsampling / downsampling
5. Clustering customers and build model to identify future customer segments