

SI 671 Final Project: E-commerce User Behavior Analysis and Modeling Xinyu Zhang (xyuzhang@umich.edu)



OVERVIEW

Every business decision takes much financial and human resource costs. It is important to understand customers' interests and behavior to drive sales. This project aims to conduct user behavior analysis for e-commerce platforms to gain insights and make effective business decisions.

Research Questions:

- What are the peak activity times in purchasing?
- Are top-selling categories the ones that receive the most views?
- Can we predict when a user is likely to stop using the platform or reduce their engagement?
- What is the next interaction of a user towards one item?

DATASET

The dataset used is sampled from *User Behavior Data* from *Taobao for Recommendation*.

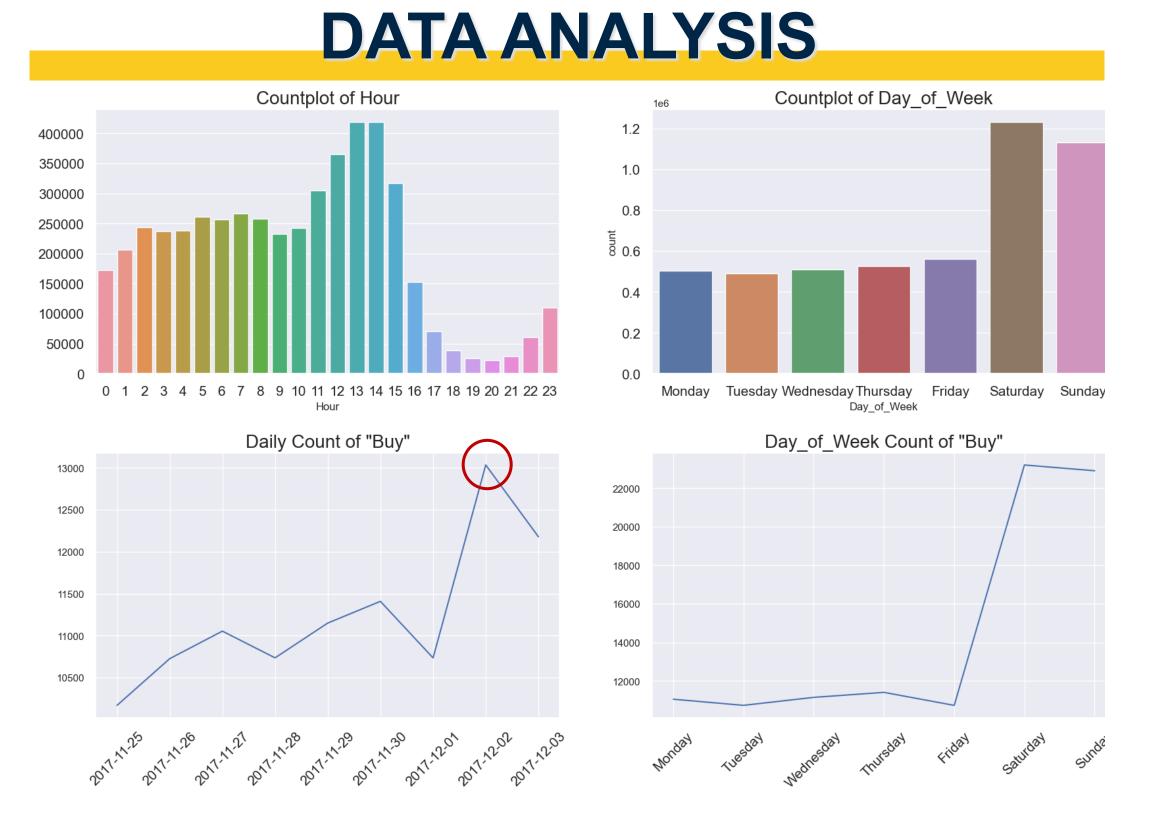
- Records of one million users between November 25, 2017 and December 3, 2017
- Four behaviors: page views, purchases, adding to the cart and favoriting products
- Data Prerocessing: data sampling, cleaning, format transformation ...

	User_ID	Product_ID	Category_ID	Behavior	Timestamp
1747	1000060	2178355	4962280	pv	1511642462
1748	1000060	857323	1320293	pv	1511822512
1749	1000060	4983347	4069500	pv	1511822555

Original Dataset: 100150807 records with 987994 unique users

	Datetime	Day_of_Week	Hour	Date
1747	2017-11-25 20:41:02	Saturday	20	2017-11-25
1748	2017-11-27 22:41:52	Monday	22	2017-11-27
1749	2017-11-27 22:42:35	Monday	22	2017-11-27

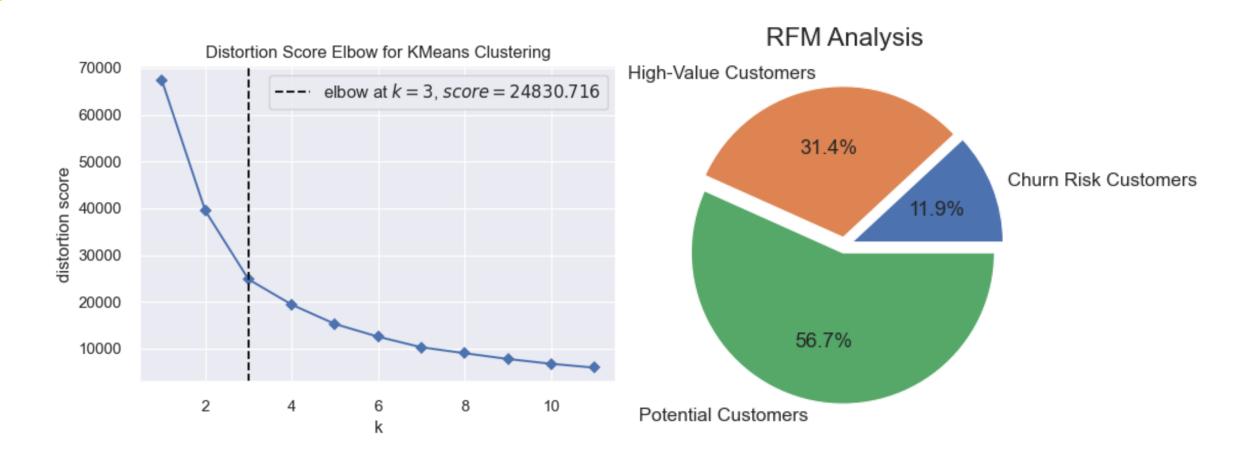
Processed Dataset: 4952632 records with 49400 unique users



CUSTOMER SEGMENTATION

RFM Analysis with K-means Clustering

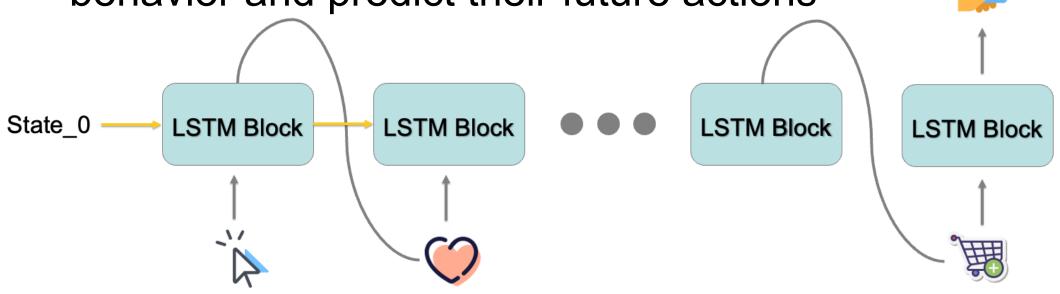
- RFM Analysis uses past purchase behavior to divide customers into groups for future marketing strategy Recency: days since last purchase Frequency: total number of purchases
- ❖ K-means Clustering with Recency and Frequency
 Use Elbow method to select the optimum k value
 ☐ Churn Risk Customers: high recency + low frequency
- □ Potential Customers: moderate recency
- ☐ <u>High-Value</u> Customers: low recency + high frequency



BEHAVIOR MODELING

Method Overview

Long Short-Term Memory (LSTM) is a variant of recurrent neural network (RNN), using memory cell structure to capture sequential dependencies in user behavior and predict their future actions



Experimental Results

		Training Loss Curve
Methods	Test Accuracy	100 - Training Loss
Random Choice	0.25	40 -
RNN	0.73	20 -
LSTM	0.95	
		0 1000 2000 3000 4000 5000 6000 7000 8000

CONCLUSION & DISCUSSION

- Users interests of purchasing greatly increase during weekends and shopping festivals
- The conversion rates are exceedingly low, mostly below 4% even among the top-selling categories
- 11.9% of users are churn-risk users but the overall customer groups contribute positively to Taobao
- LSTM model predicts the next user behavior with an accuracy of 95%, showcasing its feasibility

FUTURE

- Use frameworks like Spark or computing resources like GPU to process the whole dataset
- LSTM models can be further applied with recommendation systems to predict the user preferences of different products and categories