# Shopping Trip Classifier

What does your cart say about your shopping trip?

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## Objective

Classify the type of shopping trip
based on
Products bought

## **Data Description**

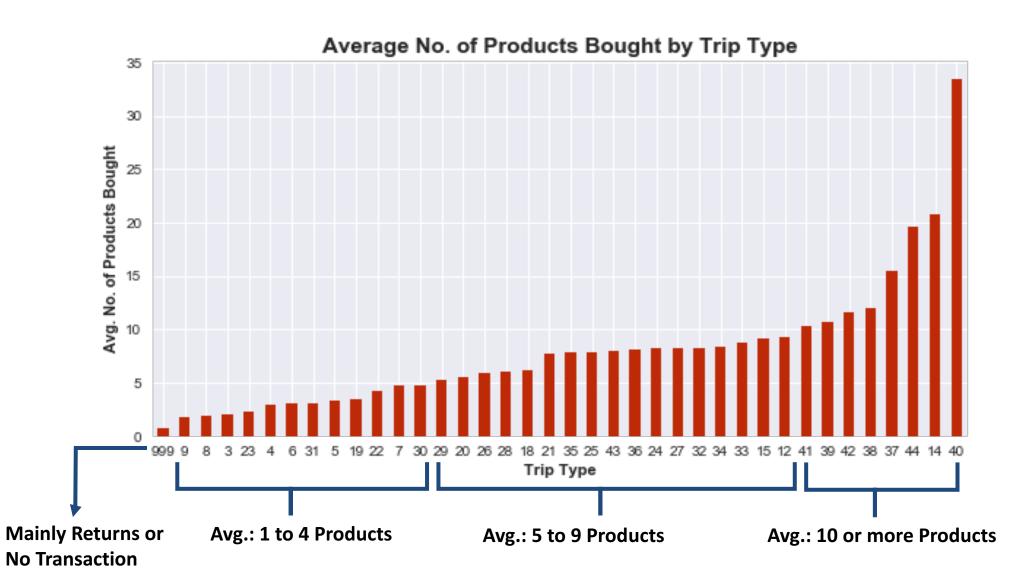
- Data acquired from Walmart Kaggle Competition
- Label : Trip Type
  - An id code to describe the trip type
  - 38 different types based on Walmart's proprietary method
- Features:
  - Visit Number: Id for an unique visit (over 95K visits)
  - Department Description 68 Product Departments/ Product Categories
  - Scan Count No. of products bought

# Data Exploration

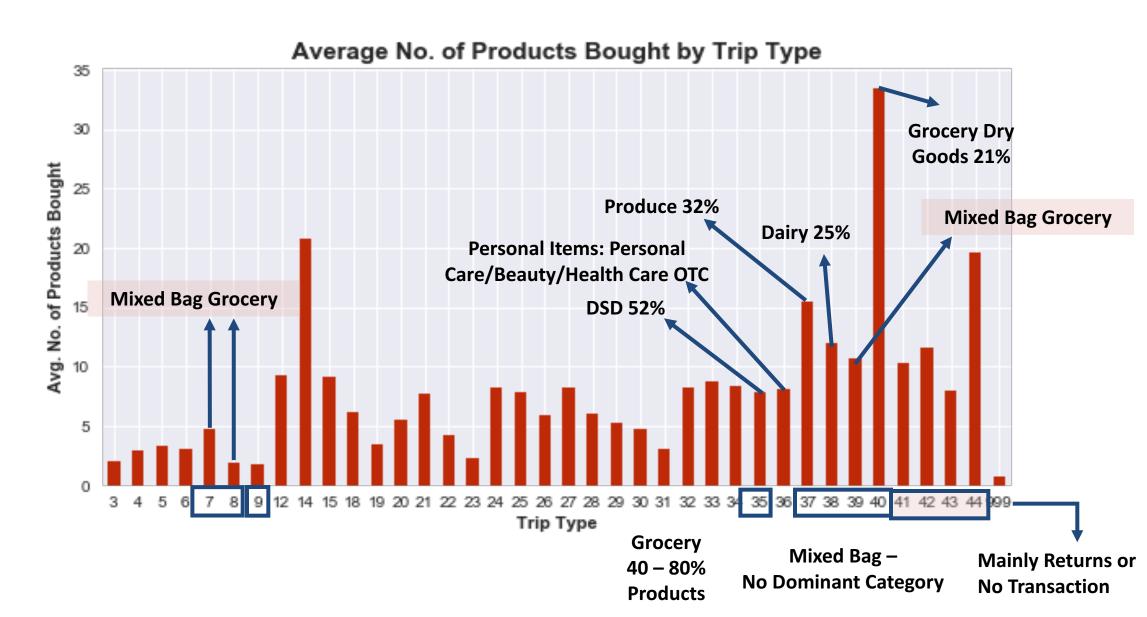
Deciphering Walmart's Trip Type Coding Based on:

- No. of Products Bought
- Product Categories Bought

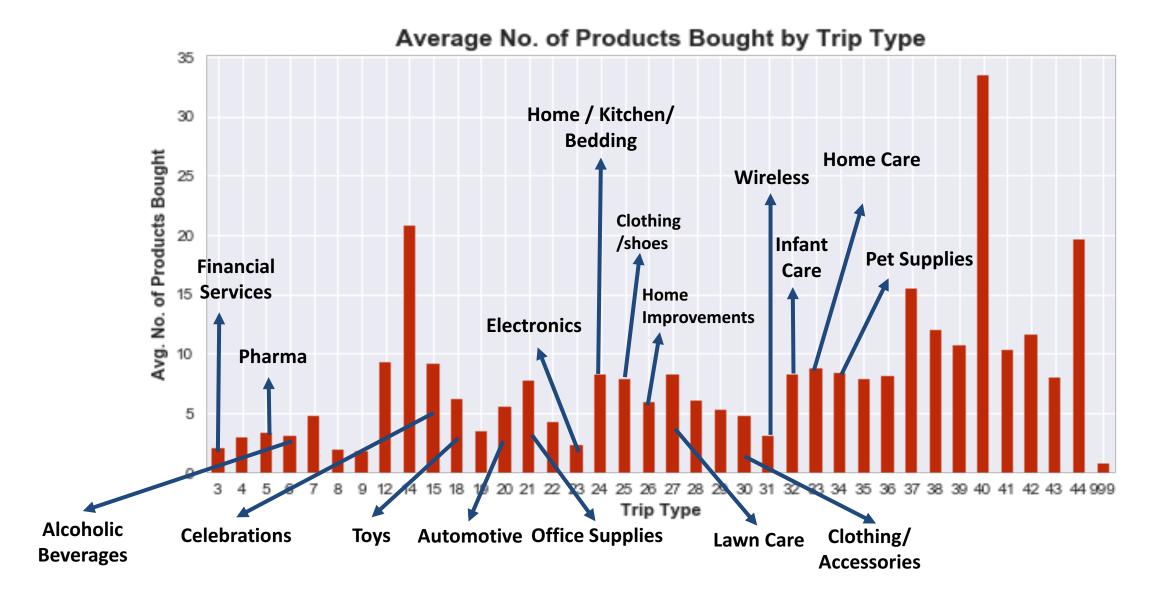
## Trip Types



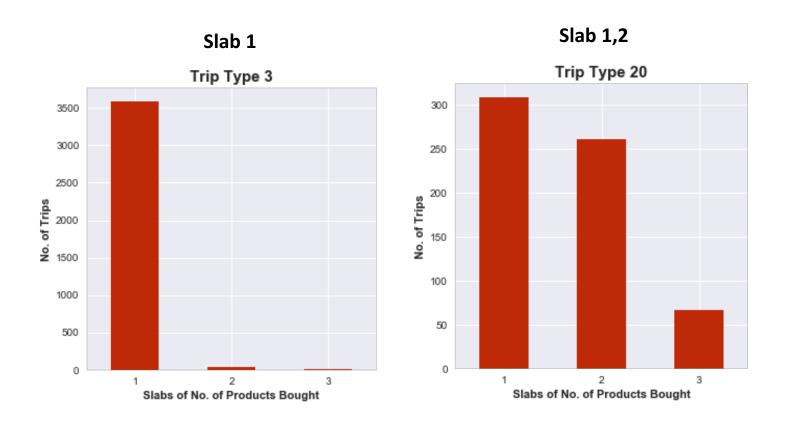
## Splitting Trip Type by Categories Bought



## Splitting Trip Type by Predominant Categories Bought



## Binning Trip Types based on No. of Products bought



#### **Slab Definitions**

Slab0: No Transactions Slab1: 1 to 4 Products Slab2: 5 to 9 Products

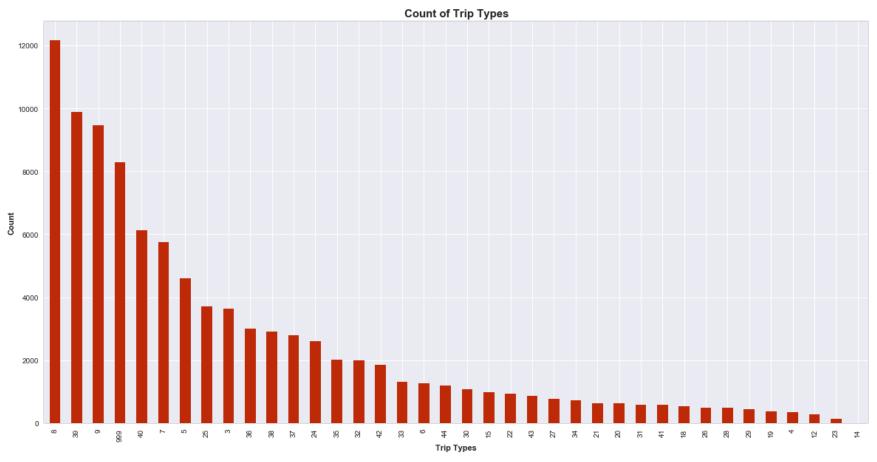
Slab3: 10 or more Products

### Features Added/Transformed

- Recreated each trip with the category & count of products bought
- Dummy Variables For Weekdays
- Total No. of Categories Bought
- Total No. of Products Bought
- Returns Only

# Modeling

## **Unbalanced Classes**



#### **Balanced Classes**

Up-sampled Minority Classes and Down sampled Majority Classes to 1800 samples

Dropped 7 Classes with very low samples

## Model Performance

Model	Cross Validated Accuracy Score (%)
Random Forest	74.54
SVM Linear Kernel	70.64
SVM RBF Kernel	70.01
Logistic Regression (Multinomial)	67.02
KNN (k=1)	65

## A closer look at Random Forest

**Using Confusion Matrix** 

&

Classification Report

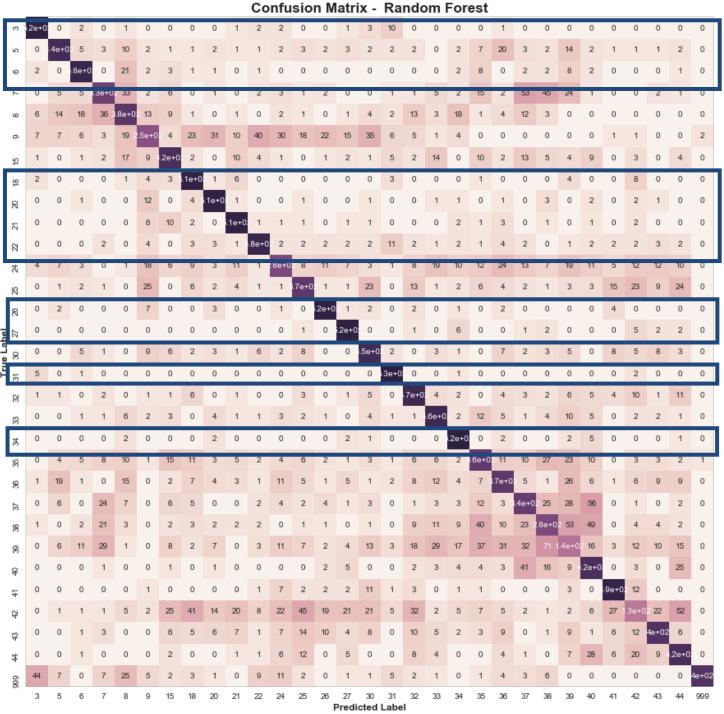
## Random Forest Successes

Classification Report: Success: F1 Score > 0.85 Failure: F1 Score < .65

**Slab Definitions** 

Slab1: 1 to 4 Products Slab2: 5 to 9 Products

Slab3: 10 or more Products



TripType:3 (Slab1)
Financial Services: 80%

Trip Type: 5 (Slab1,2)
Pharma OTC & RX: 58%

TripType:6 (Slab1) LIQUOR,WINE,BEER- 30% CANDY, TOBACCO, COOKIES- 20 %

TripType:18 (Slab1 & 2) Toys - 40.4% Impulse Merchandise -12.4%

**TripType:20** (Slab1 & 2) Automotive- 54.44% DSD Grocery -6.85%

TripType:21 (Slab1 & 2) Office Supplies- 34.42% Fabrics & Crafts -31.35%

TripType:22 (Slab1)
Electronics- 20%
Media & Gaming-12%

**Trip Type:26** (Slab1,2) Home Improvement -56%

**Trip Type:27** (Slab1,2,3) Lawn Care/ Horticulture-60%

**Trip Type:31** (Slab1) Wireless –54%

**Trip Type:34** (Slab1,2,3)

Pet Supplies – 61%

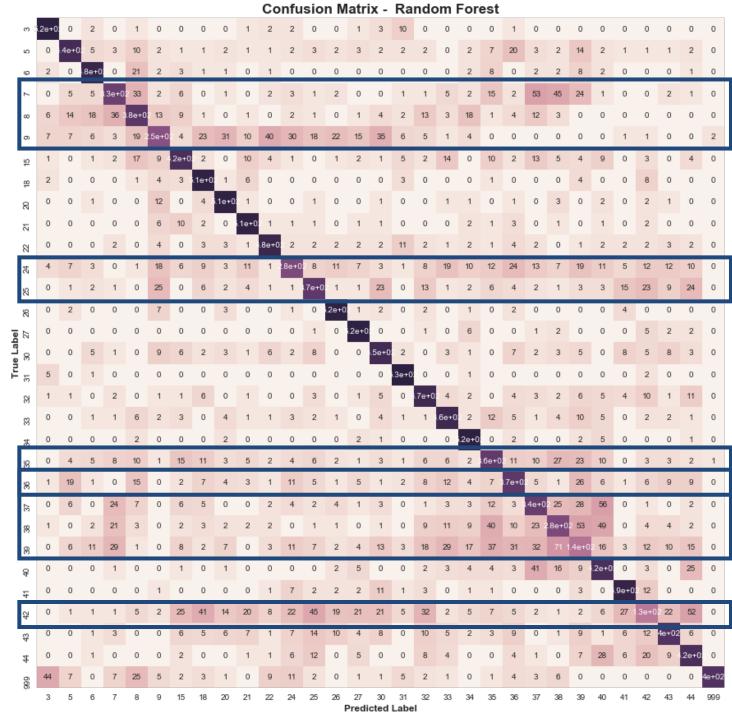
## Random Forest Failures

Classification Report: Success: F1 Score > 0.85 Failure: F1 Score < .65

#### **Slab Definitions**

Slab1: 1 to 4 Products Slab2: 5 to 9 Products

Slab3: 10 or more Products



#### Trip 24:

- Each category has low contributions ~12%
- Lots of products bought along with grocery

#### Trip 25:

- None of the individual categories are high contributors
- Trip 30(shoe/accessories/ clothing trip) is more accurate as it has lesser no. of products are bought
   Trip 36:
- It might be confusing the model as these categories appear frequently in grocery classes

**Trip Type:42** (Slab1,2,3)

Mixed Bag

**Trip Type:24** (Slab1,2,3)

Home , Kitchenware & Bedding Items ~12% each

**Trip Type:25** (Slab1 2,3) Clothing /shoes – 56%

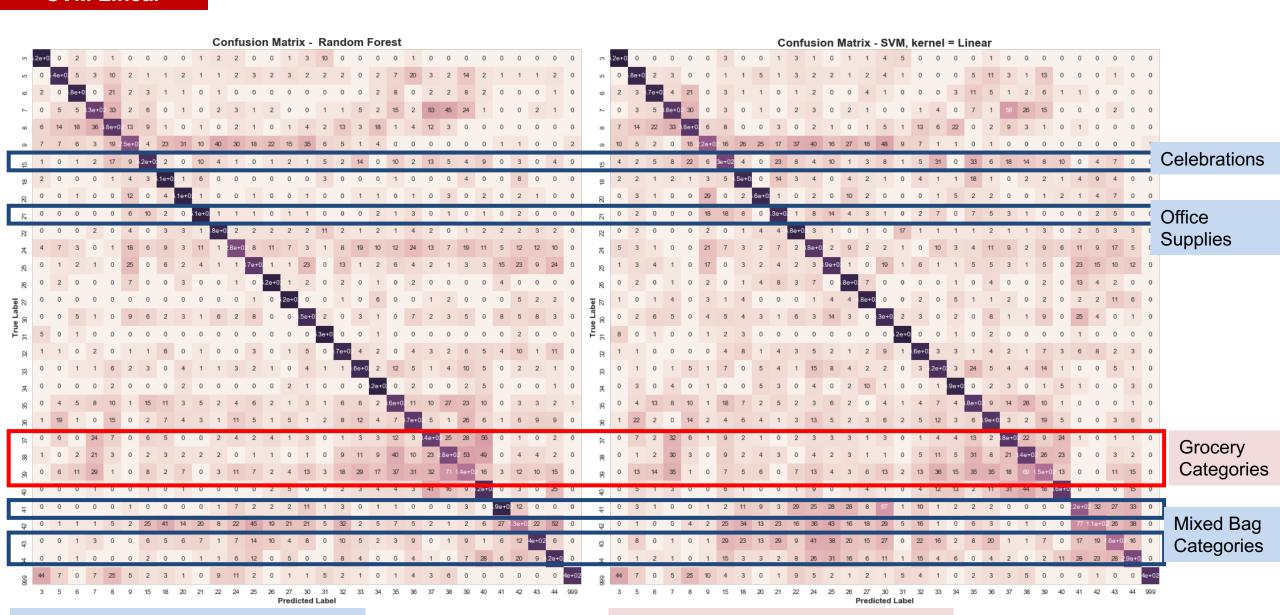
**Trip Type:36** (Slab1,2,3) Personal Items: – 70%

Personal Care - 40%

# Evaluating Random Forest vis-à-vis Other Top Models

**Using Confusion Matrix** 

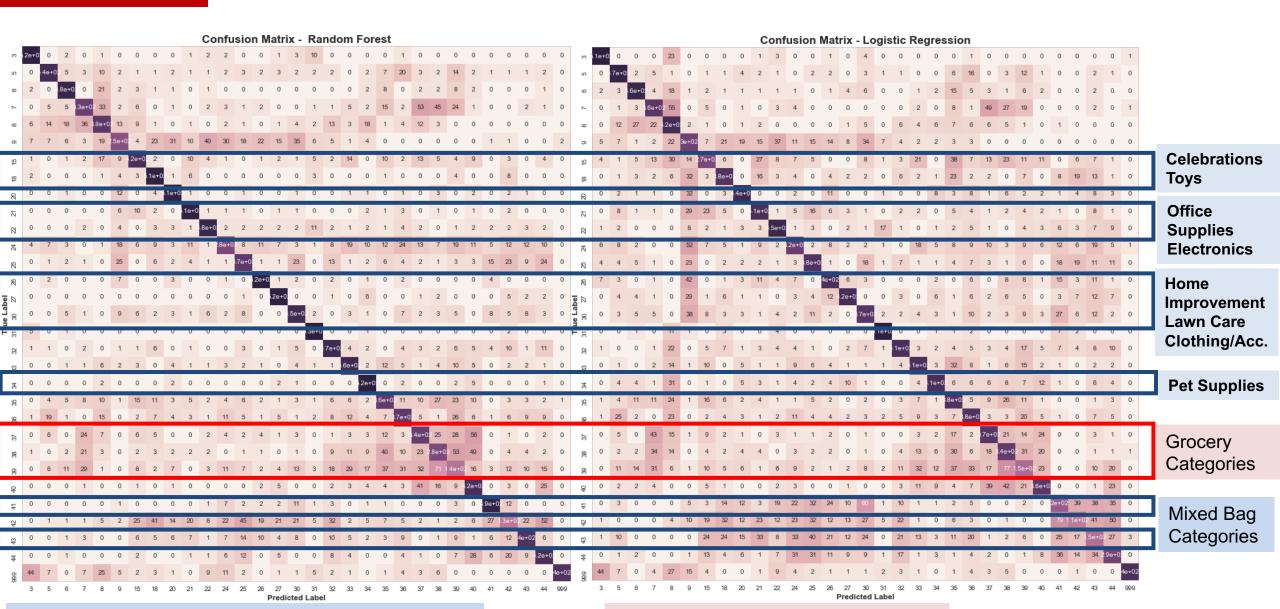
#### Random Forest Vs SVM Linear



RFC Sores Better: 15, 21, 41, 43, 44

SVM(Linear) Scores Better: 37, 38, 39

#### Random Forest Vs Logistic Regression



RFC Sores Better: 15,18, 21, 22, 26, 27,34, 41, 43

SVM(Linear) Scores Better: 37, 38, 39

## Conclusion

#### Key Findings

- At an overall level Random Forest performance is superior to all models
- Random Forest does well in classifying focused trips where:
  - Specific and less frequent categories are purchased
  - Purchases are limited to fewer categories and total products

#### Further Feature Development

• Within mix bag categories for both grocery and non-grocery find features to distinguish these classes

## Thank You!!

# Appendix

Trip Type:3 Slab1 Financial Services Impulse Merchand
Trip Type:15 Slab1,2,3

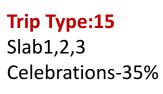
**Trip Type: 5** Slab1,2 Pharma OTC & RX: 58% dise

Trip Type:6
Slab1
LIQUOR,WINE,BEER- 30.%
CANDY, TOBACCO,
COOKIES- 20%

Trip Type:7
Slab1,2
Mix Bag Grocery-
73%

Trip Type: 8
Slab1
Mixed Bag Grocery – 50%
Personal Items: -27%

Trip Type:9
Slab1
Mixed Bag
Non- Food



**Trip Type:18** Slab1,2, Toys-40%

**Trip Type:26** 

Trip Type:34

**Trip Type:20** Slab1 & 2 Automotive- 54.44% DSD Grocery -6.85%

**Trip Type: 21** Slab1,2 Office Supplies - 34 Fabrics & crafts - 31 Camera: 35%

**Trip Type: 22** Slab1,2 Electronics/Gaming/

Trip Type:24
Slab1,2,3
Home , Kitchenware
& Bedding Items -46%

Trip Type:25
Slab1 2,3

Slab1 2,3	Slab1,2
Clothing /shoes – 56%	Home Improvement -
	56%

#### **Trip Type:27** Slab1,2,3 Lawn Care/ Horticulture-60%

irip Type:30
Slab1,2
Clothing/Accessories/
Shoes -60%

Trip Type:31	Trip Type:32
Slab1	Slab1,2,3
Wireless –54%	Infant Care –55%

Trip Type:33	
Slab1,2,3	

Tip Type.55	
Slab1,2,3	Slab1,2,3
Home Care –56%	Pet Supplies – 61%

#### **Trip Type:35** Slab1 Grocery – 86% DSD Grocery 21%

#### Trip Type:36 Slab1,2,3 Personal Items: – 70% Grocery – 82% Personal Care - 40% Beauty – 17%

#### **Trip Type:37 Trip Type:38** Slab1 Slab1 Grocery – 80% Produce – 30% Dairy - 25%

#### **Trip Type:39** Slab 2,3

Grocery – 60% Mix Bag of Grocery

#### Trip Type: 40 Slab 3 Grocery: – 70% Grocery Dry Goods -20%

## Weekend Vs Weekday

