

Google

Analytics

Do the right thing



Google Merchandise Store

Data Mining Principles

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Do the right thing

AGENDA

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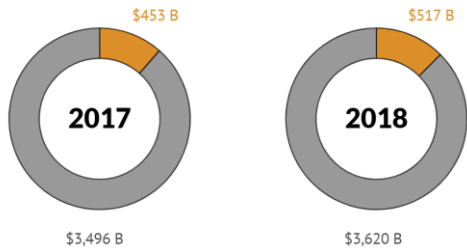
07

Sources

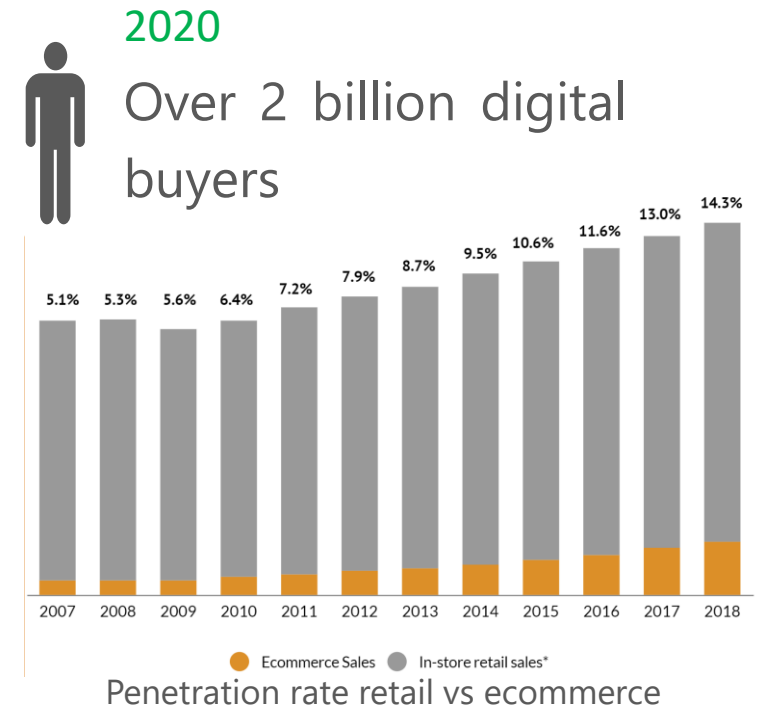
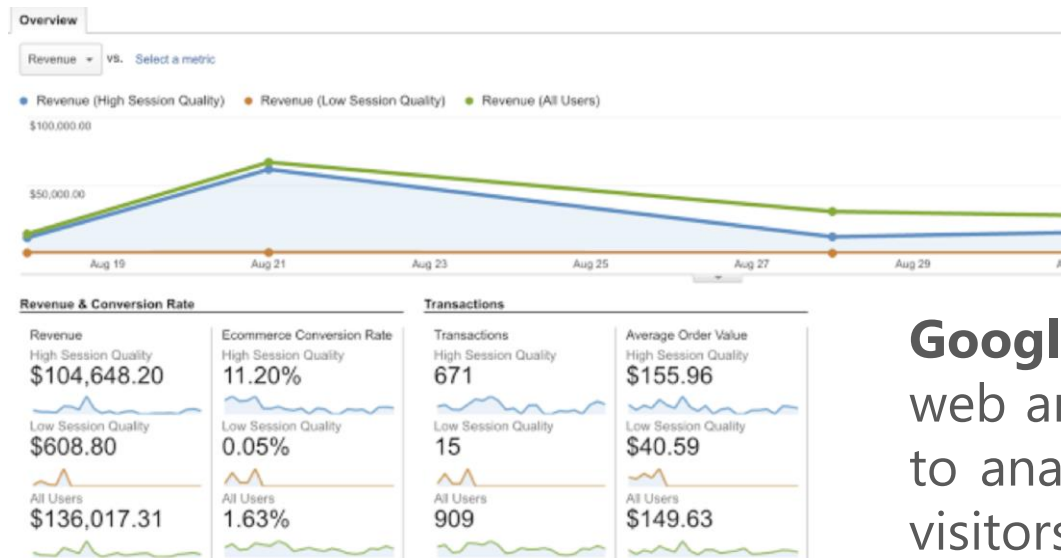
Industry Overview

2017 – 2021

Retail e-commerce sales account for \$2.3 trillion in 2017 and it is projected to grow to \$4.88 trillion by 2021



Sales in retail vs E-commerce

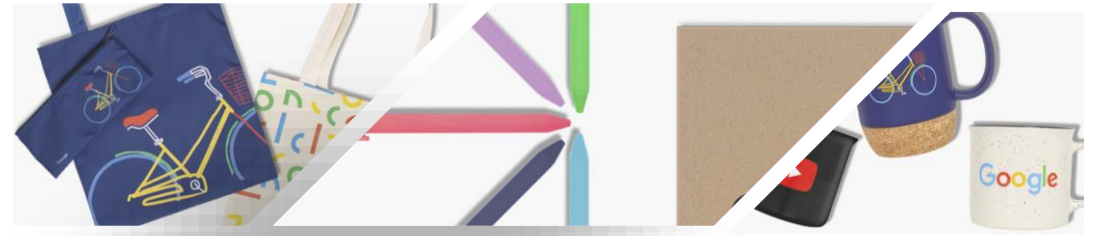


Google Analytics is a web analytics service that allows you to analyze in-depth detail about the visitors on your website.

Google Store

Do the right thing

The 80/20 rule has proven true for many businesses, only a small percentage of customers produce most of the revenue. Understanding how much each customer spends will allow companies to place actionable operations to better allocate their marketing budgets.



Problem Statement

Predict the natural log of the sum of all transactions per user

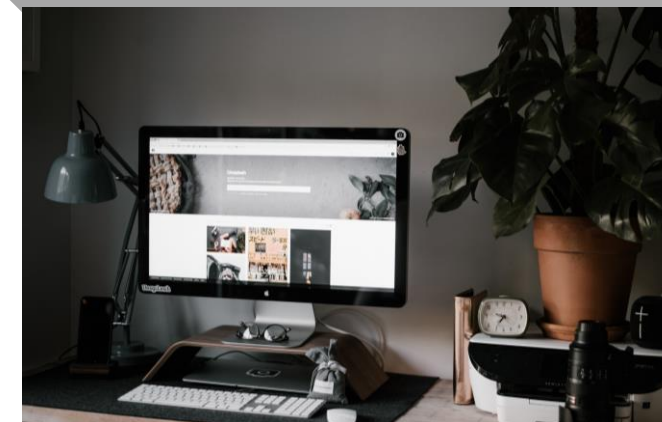
$$y_{user} = \sum_{i=0}^n transaction_{user_i}$$

$$target_{user} = \ln(y_{user} + 1)$$

Data Preparation

Dataset Columns

- ☐ fullVisitorId
- ☐ channelGrouping
- ☐ date
- ☐ visitId
- ☐ visitNumber
- ☐ visitStartTime



Jason Columns

- ☐ device
- ☐ geoNetwork
- ☐ totals
- ☐ trafficSource

fullVisitorId	channelGrouping	date	visitId	visitNumber	visitStartTime	device	geoNetwork	totals	trafficSource
3162355547410993243	Organic Search	20171016	1508198450	1	1508198450	{"browser": "Firefox", "browserVersion": "not ..."	{"continent": "Europe", "subContinent": "Weste..."}	{"visits": "1", "hits": "1", "pageviews": "1", ...	{"campaign": "(not set)", "source": "google", ...
8934116514970143966	Referral	20171016	1508176307	6	1508176307	{"browser": "Chrome", "browserVersion": "not a..."}	{"continent": "Americas", "subContinent": "Nor..."}	{"visits": "1", "hits": "2", "pageviews": "2", ...	{"referralPath": "/a/google.com/transportation..."}
7992466427990357681	Direct	20171016	1508201613	1	1508201613	{"browser": "Chrome", "browserVersion": "not a..."}	{"continent": "Americas", "subContinent": "Nor..."}	{"visits": "1", "hits": "2", "pageviews": "2", ...	{"campaign": "(not set)", "source": "(direct)" ...

- ☐ Nr of rows: 928,860
- ☐ Nr of unique users: 716,705



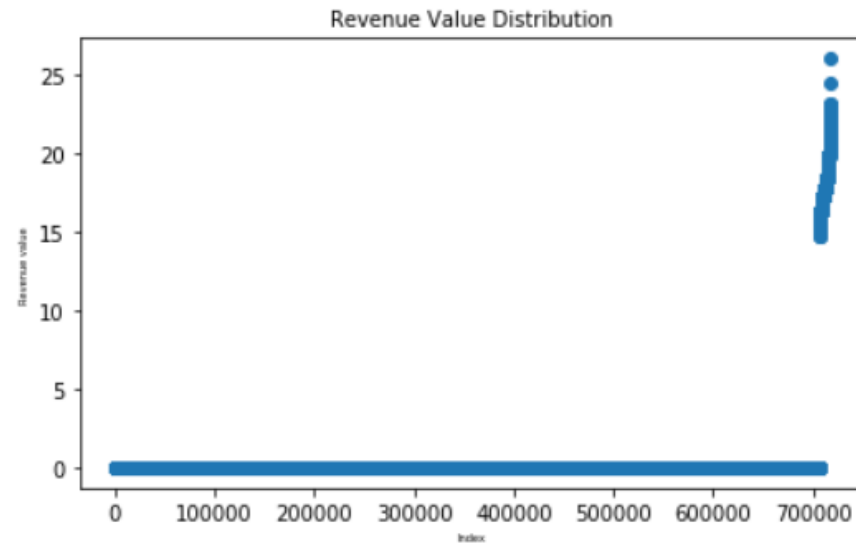
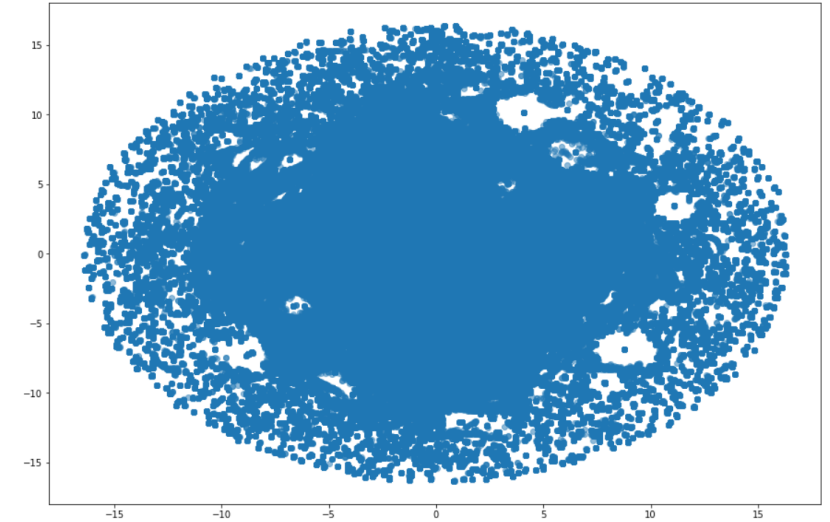
Data Exploration & Feature Engineering

Tsne

- ❑ Majority of transactions share similar attributes

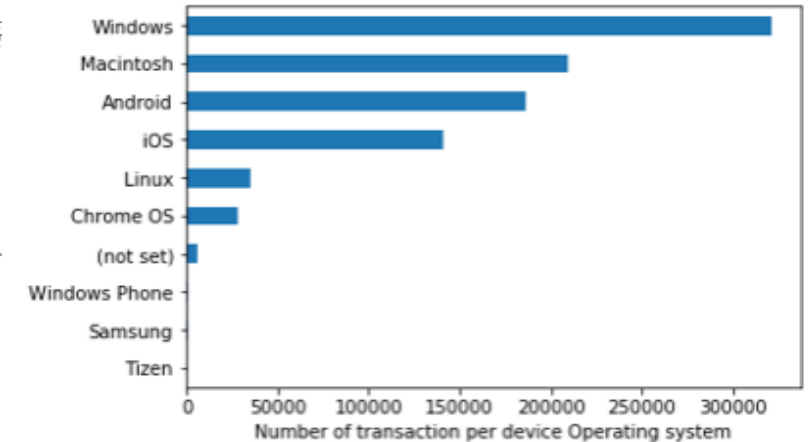
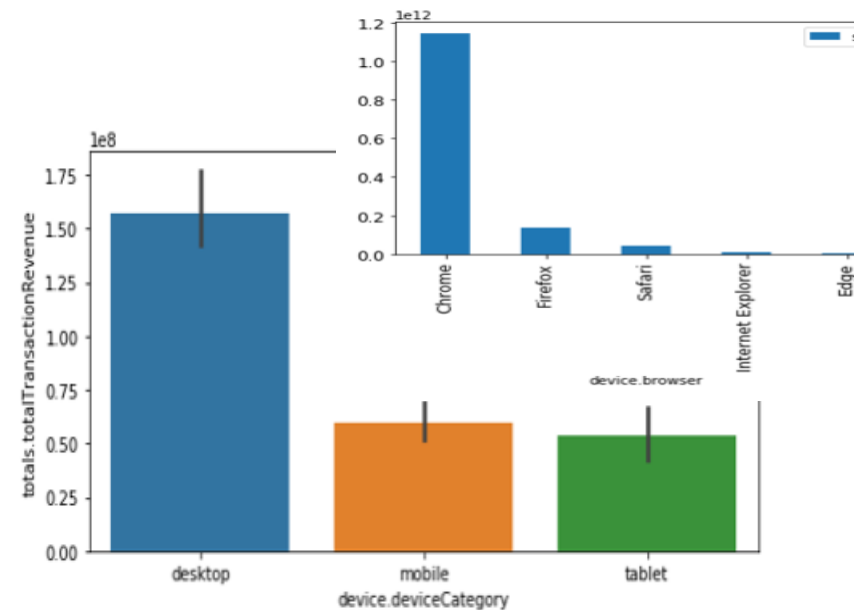
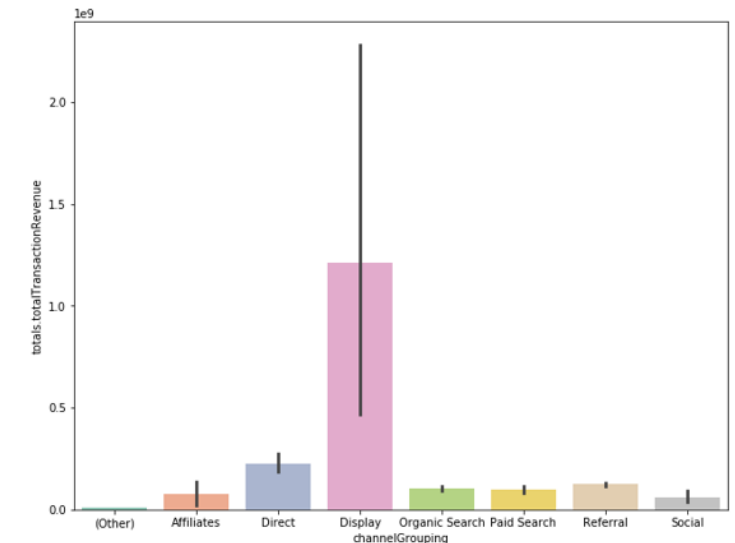
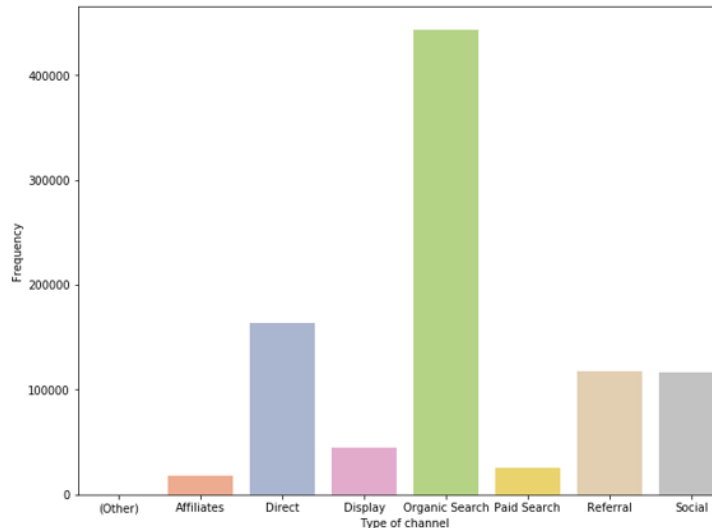
Overview of total transaction revenue:

- ❑ For Log of revenue is slightly **skewed to right**
- ❑ Only **1.2% of the transactions** contribute to total revenue



Data Exploration & Feature Engineering

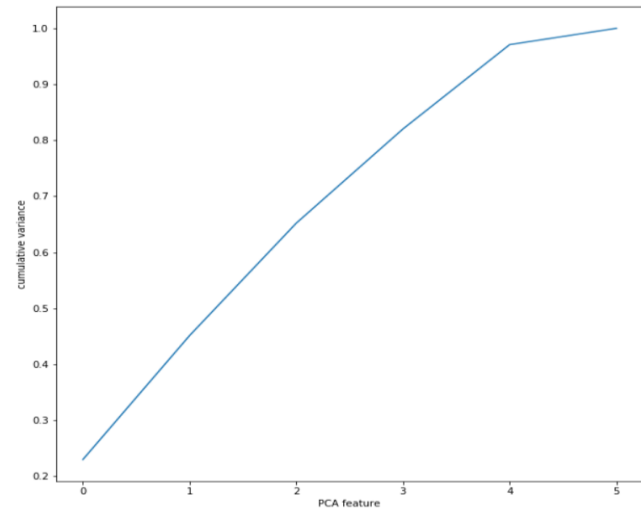
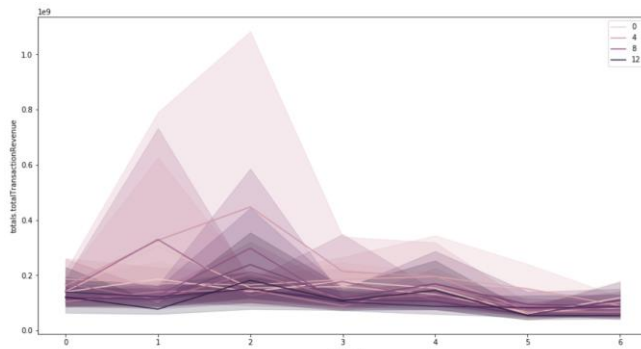
- ❑ **Group channel:** Most common channel to access GStore: Organic Search; Direct, Referral and Social Media
- ❑ **Display:** the channel with the highest contribution in terms of revenue
- ❑ **Operating System:** the first 4 options represent more than 90% of the revenue generated
- ❑ **Device:** Desktop is the most used device
- ❑ **Browser:** Chrome is the most used



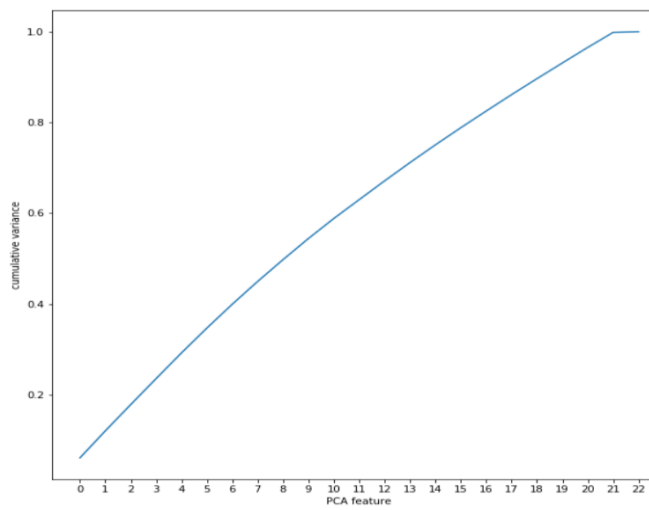
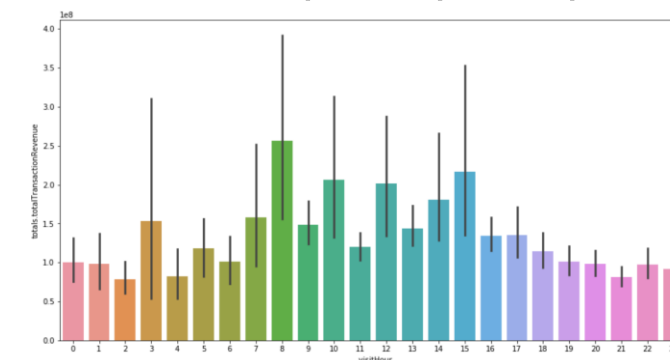
Data Exploration & Feature Engineering

Visitor time analysis:

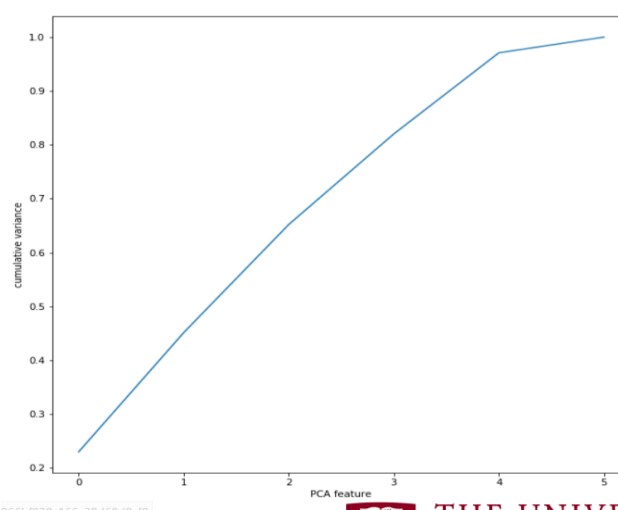
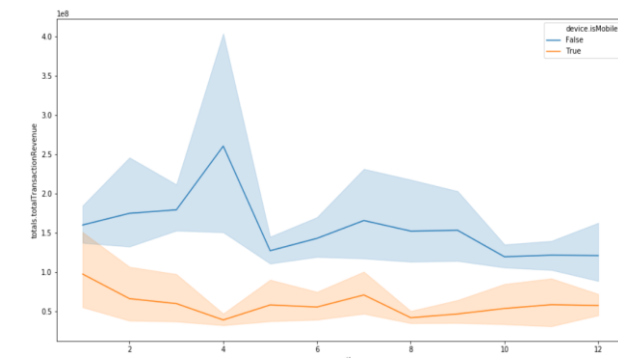
□ Tuesday is the pick day



□ Highest revenue: 3am, 8am, 10am, 12pm, 4pm, 5pm



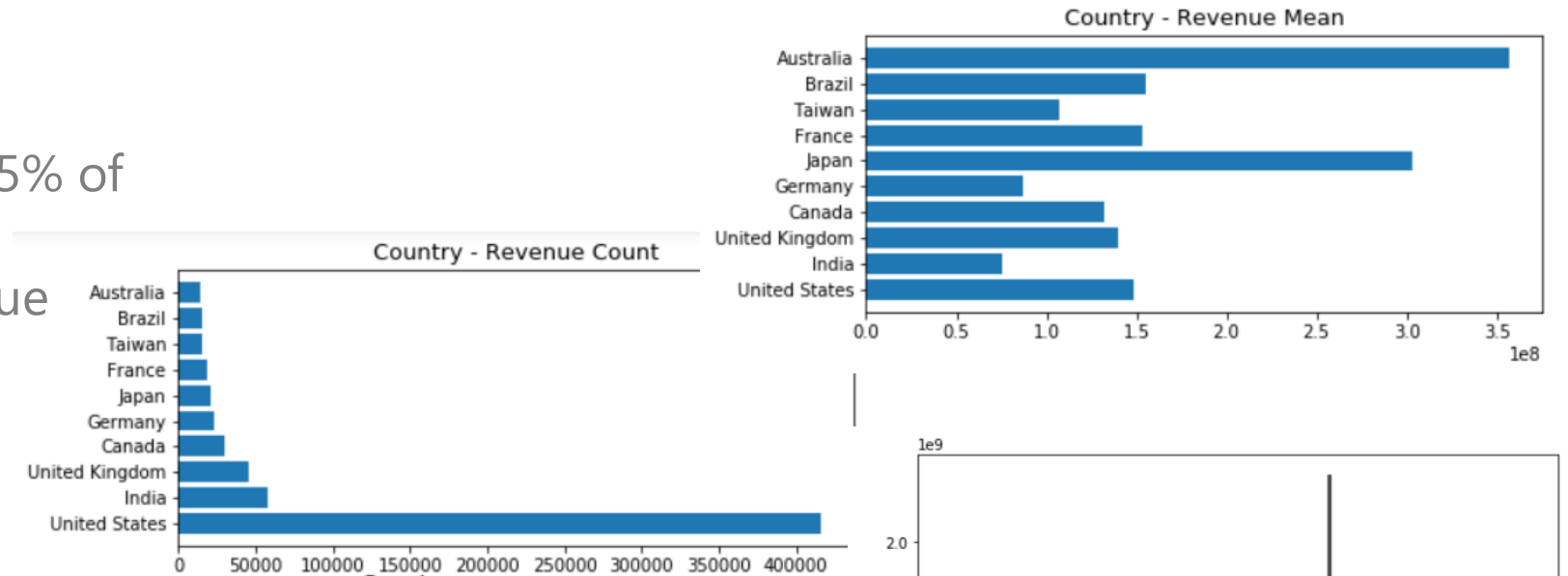
□ April is the pick month



Data Exploration & Feature Engineering

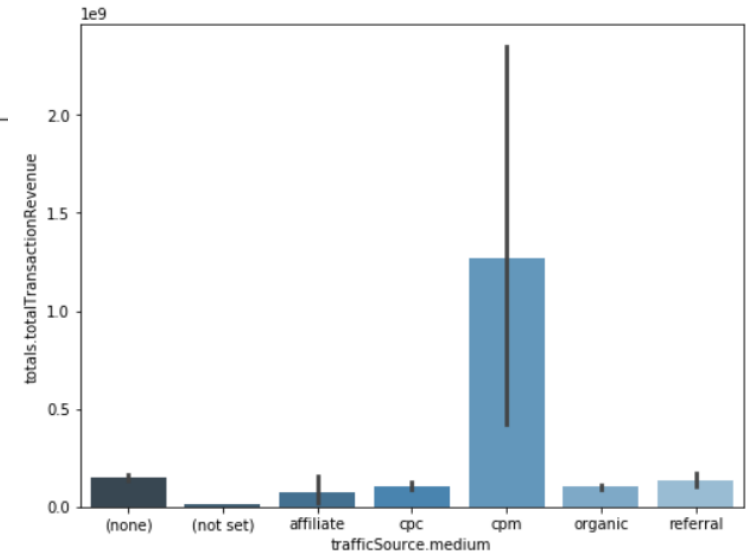
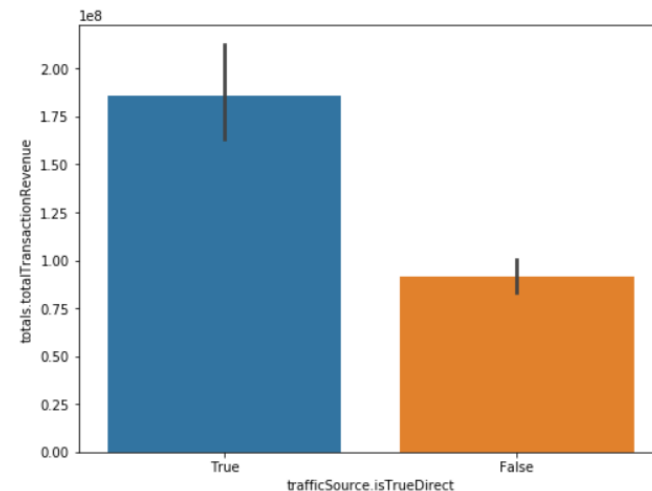
Country & Region:

- ☐ United States contributed 95% of total revenue
- ☐ Australia has highest Revenue mean
- ☐ Create new feature called isUnitedStates, isAustralia



Traffic Source:

- ☐ CPM contributes main transaction revenues for medium of traffic source
- ☐ There is difference between True direct and False direct

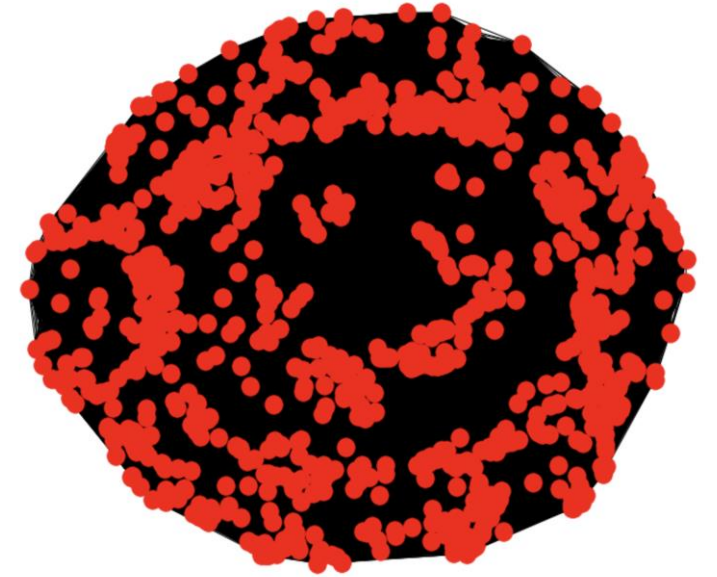


Graph Analysis



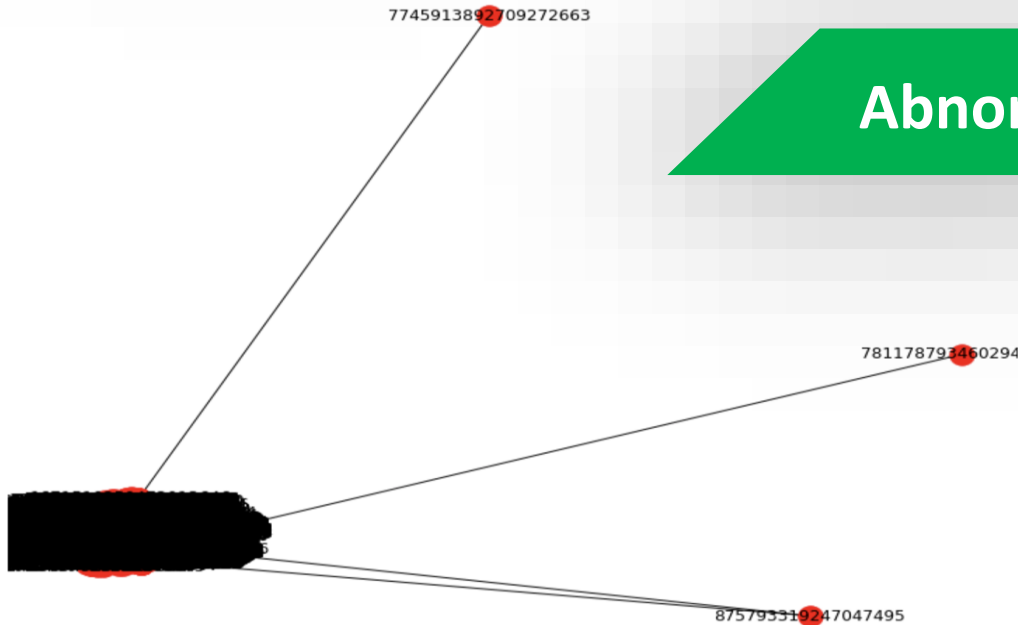
Revenue

- ❑ Node is User
- ❑ Edge revenue group for each transaction
 - Log revenue rounded to tens place
 - Remove zeros
- ❑ 1000 transactions



Abnormal User

- ❑ Subset of Users
 - Had transactions from more than one country
- ❑ Node and Edge similar to Revenue group
 - Edge not transformed
- ❑ (Not Set) country code



Two different approaches

Customer level

```
(total 46 columns):
TS_adwordsClickInfo.page_max 716705 non-null int32
BS_Firefox_max                716705 non-null float64
BS_Chrome_max                 716705 non-null int32
BS_Safari_max                 716705 non-null float64
BS_IE_max                     716705 non-null float64
BS_Android_max                716705 non-null float64
OS_Windows_max                716705 non-null float64
OS_Macintosh_max              716705 non-null float64
OS_Android_max                716705 non-null float64
OS_iOS_max                    716705 non-null float64
subCont_NorthernAmerica_max   716705 non-null float64
subCont_Western Africa_max    716705 non-null float64
country_USA_max               716705 non-null float64
country_Australia_max         716705 non-null float64
medium_cpm_max                716705 non-null float64
CG_Affiliates_max              716705 non-null float64
CG_Direct_max                  716705 non-null float64
CG_Display_max                 716705 non-null float64
CG_organicSearch_max          716705 non-null float64
CG_paidSearch_max             716705 non-null float64
CG_Referral_max                716705 non-null float64
CG_Social_max                  716705 non-null float64
device_Mobile_max              716705 non-null float64
device_Desktop_max            716705 non-null float64
cont_Africa_max                716705 non-null float64
cont_Americas_max              716705 non-null float64
cont_Asia_max                  716705 non-null float64
cont_Europe_max                716705 non-null float64
cont_Oceania_max               716705 non-null float64
TS_isTrueDirect_max            716705 non-null float64
TS_sessionQuality.100_max      716705 non-null float64
TS_Slot.RHS_max                716705 non-null float64
TS_Slot.Top_max                716705 non-null float64
TS_Network.Content_max         716705 non-null float64
TS_Network.GSearch_max         716705 non-null float64
TS_Network.PSearch_max         716705 non-null float64
dtypes: float64(8), int32(2), int64(14), object(1), uint8(21)
memory usage: 145.6+ MB
```

Transaction level

```
RangeIndex: 928860 entries, 0 to 928859
Data columns (total 57 columns):
date                928860 non-null object
id                  928860 non-null int64
totals.hits          928860 non-null int64
totals.pageviews     928860 non-null int64
totals.bounces       928860 non-null int64
totals.newVisits     928860 non-null int64
totals.timeOnSite    928860 non-null float64
revenue             928860 non-null float64
TS_adwordsClickInfo.page_max 928860 non-null int32
month_4              928860 non-null int64
isTuesdays          928860 non-null int64
BS_Firefox           928860 non-null int64
BS_Chrome            928860 non-null int64
BS_Safari            928860 non-null int64
BS_IE                928860 non-null int64
BS_Android           928860 non-null int64
OS_Windows           928860 non-null int64
OS_Macintosh         928860 non-null int64
OS_Android           928860 non-null int64
OS_iOS               928860 non-null int64
subCont_NorthernAmerica_max 928860 non-null int64
subCont_Western Africa_max   928860 non-null int64
country_USA          928860 non-null int64
country_Australia    928860 non-null int64
medium_cpm           928860 non-null int64
CG_Affiliates         928860 non-null uint8
CG_Direct             928860 non-null uint8
CG_Display            928860 non-null uint8
CG_organicSearch      928860 non-null uint8
CG_paidSearch         928860 non-null uint8
CG_Referral           928860 non-null uint8
CG_Social             928860 non-null uint8
Hour_0                928860 non-null uint8
Hour_2                928860 non-null uint8
Hour_3                928860 non-null uint8
Hour_8                928860 non-null uint8
Hour_9                928860 non-null uint8
Hour_10               928860 non-null uint8
Hour_11               928860 non-null uint8
Hour_12               928860 non-null uint8
Hour_13               928860 non-null uint8
Hour_14               928860 non-null uint8
Hour_15               928860 non-null uint8
device_Mobile         928860 non-null uint8
device_Desktop        928860 non-null uint8
cont_Africa           928860 non-null uint8
cont_Americas         928860 non-null uint8
cont_Asia             928860 non-null uint8
cont_Europe           928860 non-null uint8
cont_Oceania          928860 non-null uint8
TS_isTrueDirect        928860 non-null uint8
TS_sessionQuality.100  928860 non-null uint8
TS_Slot.RHS           928860 non-null uint8
TS_Slot.Top           928860 non-null uint8
TS_Network.Content     928860 non-null uint8
TS_Network.GSearch     928860 non-null uint8
TS_Network.PSearch     928860 non-null uint8
dtypes: datetime64[ns](1), float64(5), int32(2), int64(16), object(1), uint8(32)
memory usage: 198.4+ MB
```



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Costumer Level Modeling

Goal

- ✓ Predict the log of the revenue per user

Steps

- I. Aggregate data per user ID
- II. Sum of the original revenue and log of the sum per user
- III. Build Model

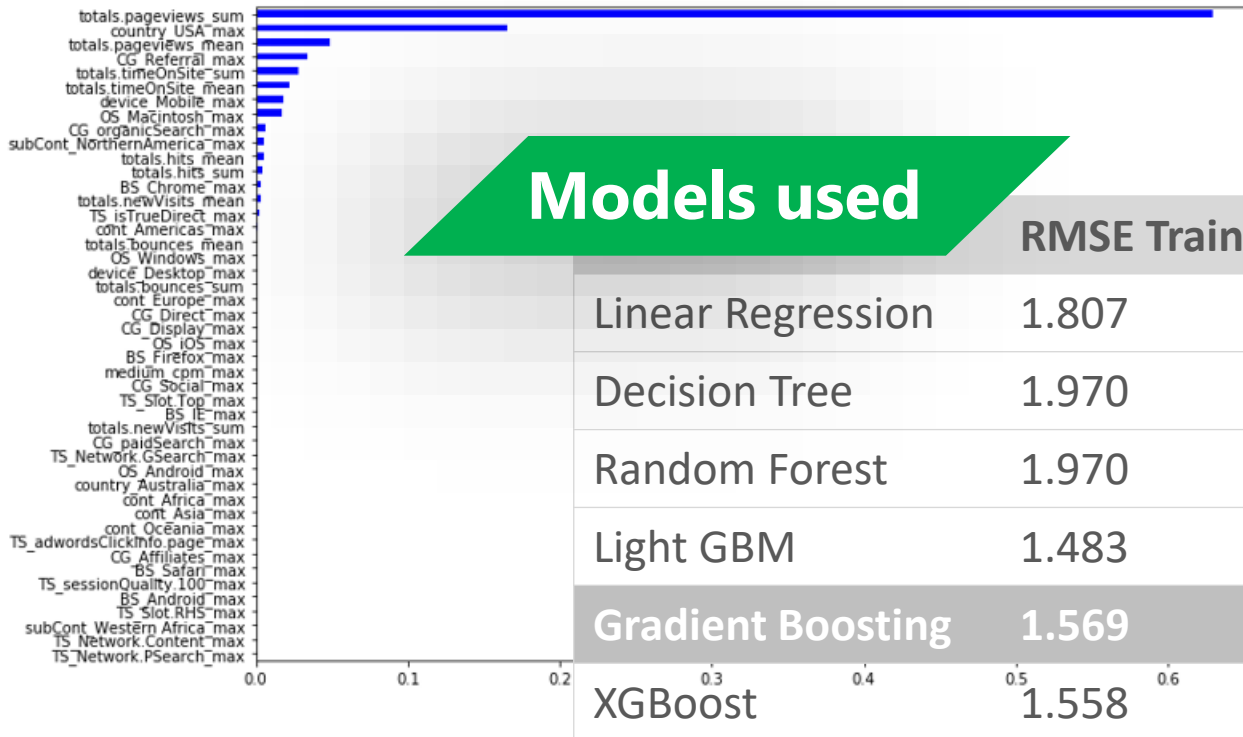
Y^t = The revenue at transaction level

Y^a = The revenue at user level

$$\sum Y^t_{user} = \xrightarrow{\text{Aggregate}} Y^a \xrightarrow{\text{Transform}} Y^a_{log} \xrightarrow{\text{Model}} \hat{Y}^a_{log}$$



Modeling: Customer level



Transaction Level Modeling

Goal

✓ Want to predict the transaction level revenue as well as compare our results to the aggregated data

Steps

- I. Log Revenue
- II. Build Model
- III. Exponentiate Predicted values
- IV. Aggregate Sum over User ID
- V. Log Predicted
- VI. Compare values to Aggregate Log Revenue

Y^t = The revenue at transaction level

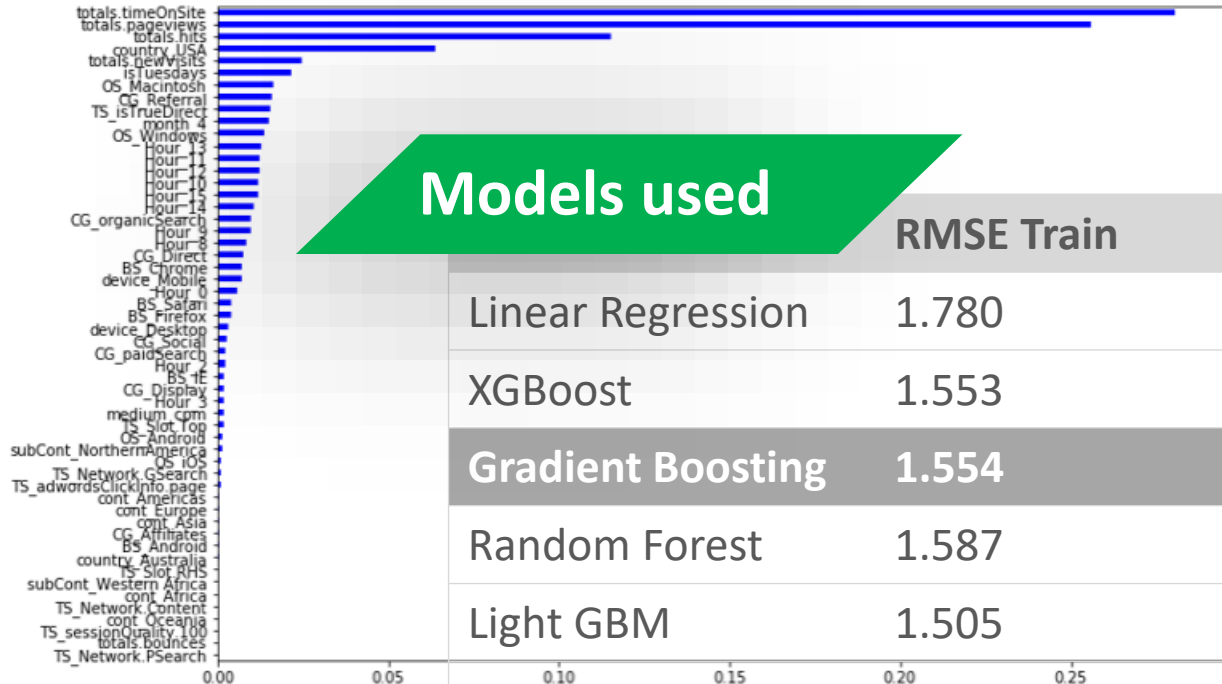
Y^a = The revenue at user level

$$Y^t \xrightarrow{\text{Transform}} Y_{log}^t \xrightarrow{\text{Model}} \hat{Y}_{log}^t \xrightarrow{\text{Transform}} e^{\hat{Y}_{log}^t} = \hat{Y}^t \xrightarrow{\text{Aggregate}} \sum \hat{Y}_{user}^t = \hat{Y}^a \xrightarrow{\text{Transform}} \hat{Y}_{log}^a$$

$$\text{Error} = Y_{log}^a - \hat{Y}_{log}^a$$



Modeling: Transaction Level



Models used

	RMSE Train	RMSE Test	Cross Validation
Linear Regression	1.780	1.724	1.781
XGBoost	1.553	1.556	1.573
Gradient Boosting	1.554	1.557	1.575
Random Forest	1.587	1.575	1.596
Light GBM	1.505	1.552	1.560
Decision Tree	1.488	2.188	2.226

Customer Segmentation & Life Time Value Prediction

Customer Segmentation

LTV Prediction

Jan

Sept

Dec, 2017

Recency



Frequency



logRevenue

RecencyCluster	count	mean
0	116766	236.708854
1	122078	167.864210
2	116446	95.801110
3	140699	29.319469

FrequencyCluster	count	mean
0	470010	1.097385
1	24609	3.980007
2	1336	14.991018
3	34	89.029412

logRevenueCluster	count	mean
0	488670	0.000000
1	7319	17.937419

=

OverallScore	Recency	Frequency	logRevenue
0	236.827423	1.082989	0.000000
1	170.698931	1.185075	0.120600
2	100.203454	1.250945	0.245042
3	33.465895	1.252588	0.227297
4	35.571747	4.055868	3.653922
5	33.389222	8.849634	12.900719
6	25.857143	23.529101	16.950539
7	14.000000	130.200000	19.778151



Segment	count	mean_logRevenue_9
Low-Value	111388	0.000000
Mid1-Value	373489	0.198640
Mid2-Value	10918	4.926860
High-Value	194	17.023416

logRevenue (last 3 months)

Segment	count	mean_logRevenue_12
Low-Value	111388	0.000311
Mid1-Value	373489	0.005806
Mid2-Value	10918	0.191567
High-Value	194	1.820208



LTV Cluster

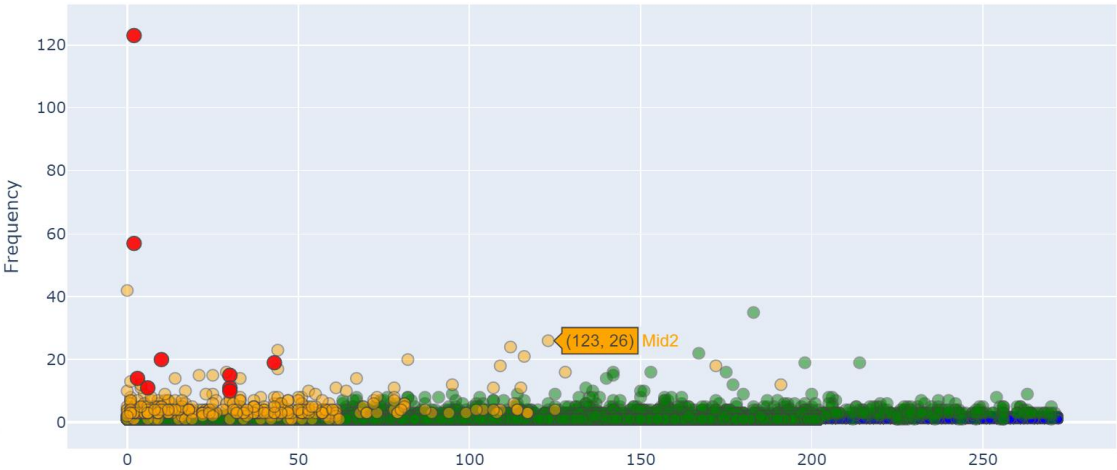
LTVCluster	count	mean_logRevenue_12
0	495733	0.000000
1	112	17.055235
2	98	18.394497
3	46	20.319989



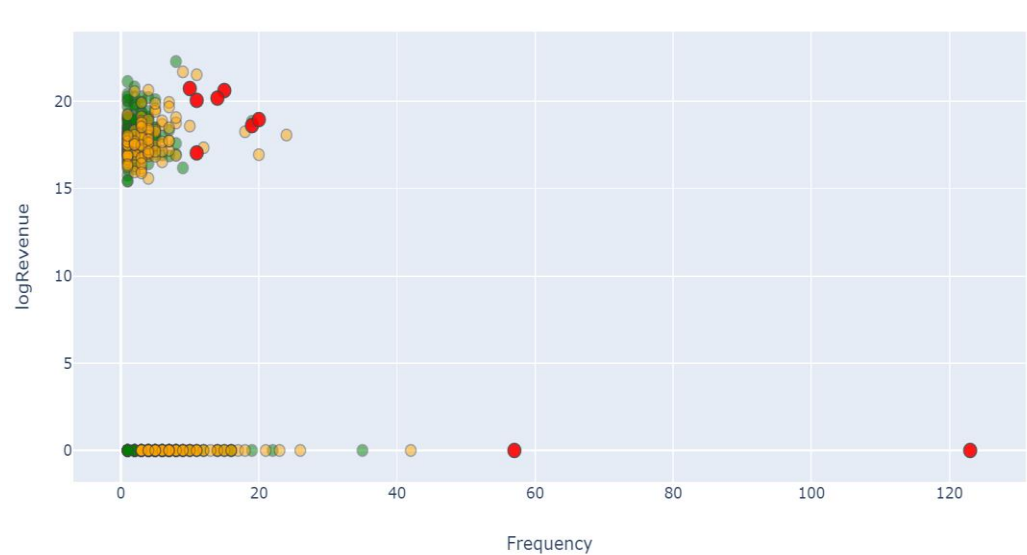
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Customer Segmentation

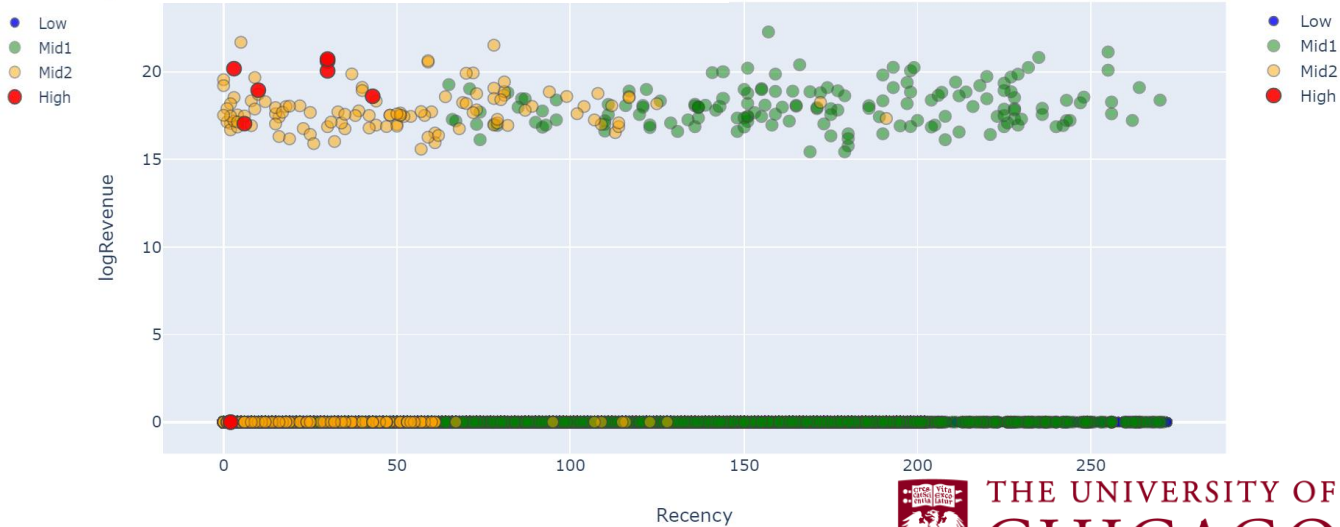
Segments



Segments



Recency



Customer LTV Prediction



9-month
Features

SMOTE
Predict

3-month
LTV Cluster

Add Features

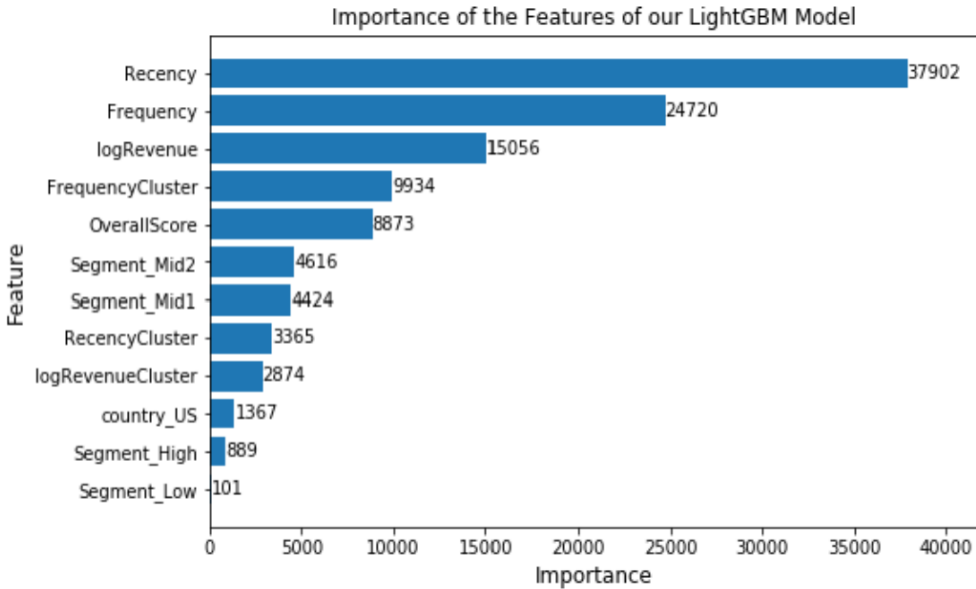
1. Continent only

2. Continent and US

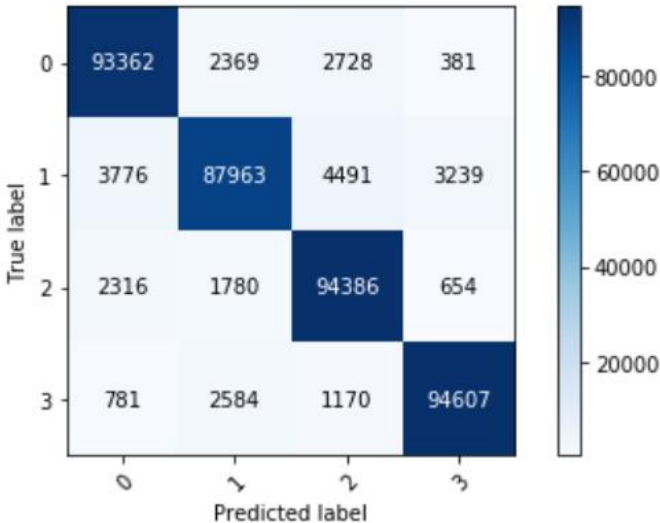
3. US only

SMOTE
Predict

3-month
LTV
Cluster

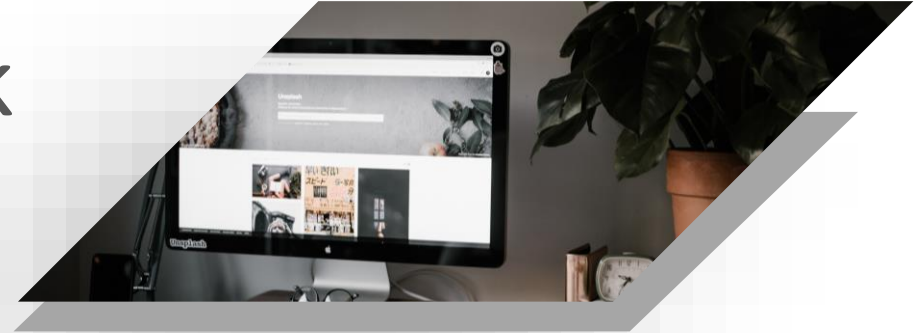


Confusion matrix, without normalization



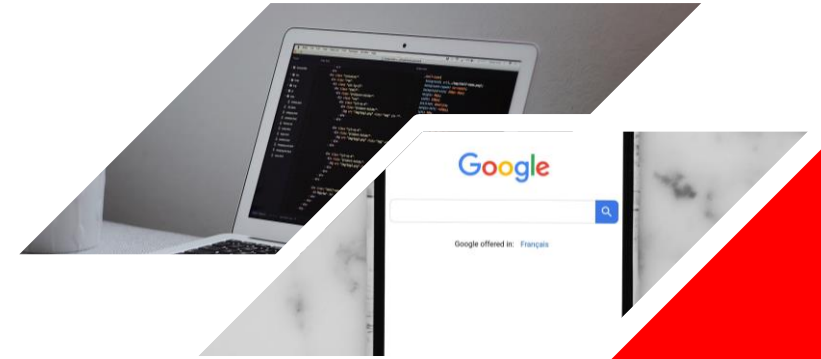
	precision	recall	f1-score	support
0	0.94	0.95	0.94	98840
1	0.93	0.89	0.91	99469
2	0.92	0.95	0.94	99136
3	0.96	0.95	0.96	99142
accuracy			0.94	396587
macro avg	0.94	0.94	0.94	396587
weighted avg	0.94	0.94	0.94	396587

Lessons Learned & Future Work



- ❑ Meaningful insights from large datasets sometimes more complicated
- ❑ Do proper research on models that better suit the specificities of our data
- ❑ Large data may limit the number of models to use
- ❑ Keep in mind the business goal throughout the project
- ❑ Use product level data for more insights
- ❑ Use several years of data – identify seasonality
- ❑ Apply more models to Customer segmentation – LTV
- ❑ Tune models more

Sources



- ❑ Data: Google Analytics Customer Revenue Prediction – [link](#)
- ❑ Statistics digital buyers – [link](#)
- ❑ Statistics on ecommerce – [link](#) & [link](#)
- ❑ Customer lifecycle prediction – [link](#)

Q&A

Google

Thank You

