

PROFILING & ANALYSING CUSTOMER BEHAVIOUR — USING CLUSTER & ASSOCIATION ANALYSIS

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SPONSOR BACKGROUND

- A local startup launched in May 2015
- Bridge the gap between customers & beauty professionals
- Customers can book beauty services at their convenience
- Beauty professionals grow their brand & customer base



PROJECT OBJECTIVES

Customers

- 1. To determine the customer segmentation (different groups of customers) from the current booking patterns
- 2. To understand customers' behaviour. When was the last time a customer used the app? How frequent does a customer use the app? How much does a customer spend on average and in total?
- 3. To understand how customers select the type of services within each booking

PROJECT OBJECTIVES

Beauty Professionals

- To determine the type of services that are commonly published
- To determine the different price points for the different types of services

METHODOLOGY





DATA COLLECTION

- Access to their MongoDB database
- Date range used from Jan 2015 to Dec 2016
- Narrowed down from 59 to 7 relevant data tables



DATA CLEANING

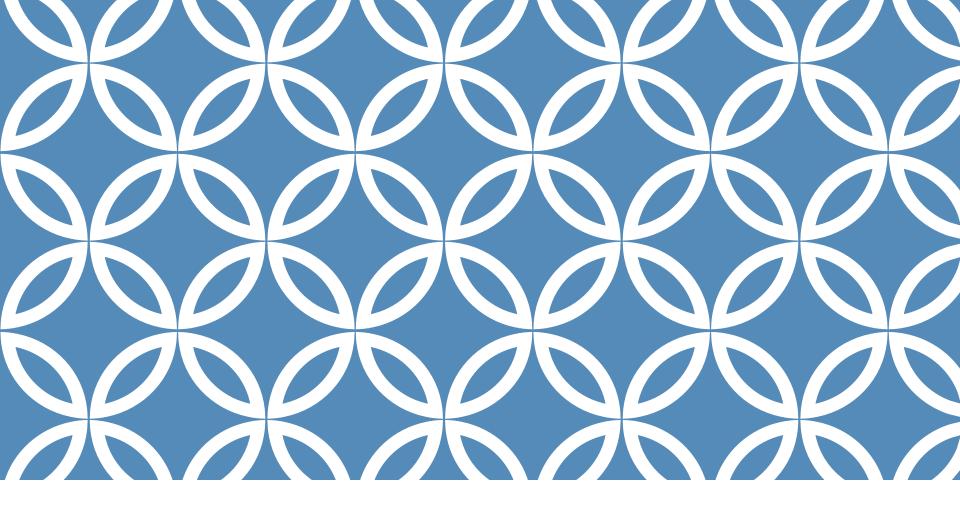
Filtering out test data

- Removed by specific columns (test_at, is_test, deleted_at)
- Removed test data across data tables
- o e.g. remove admin users AND bookings made by them

2. Duplicate & missing values

o e.g. missing monetary values within bookings data



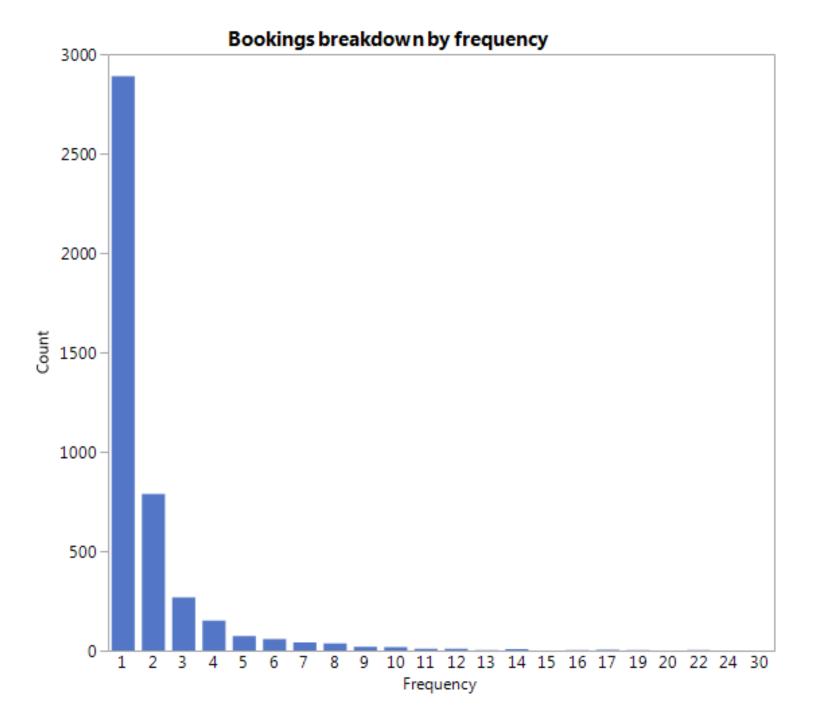


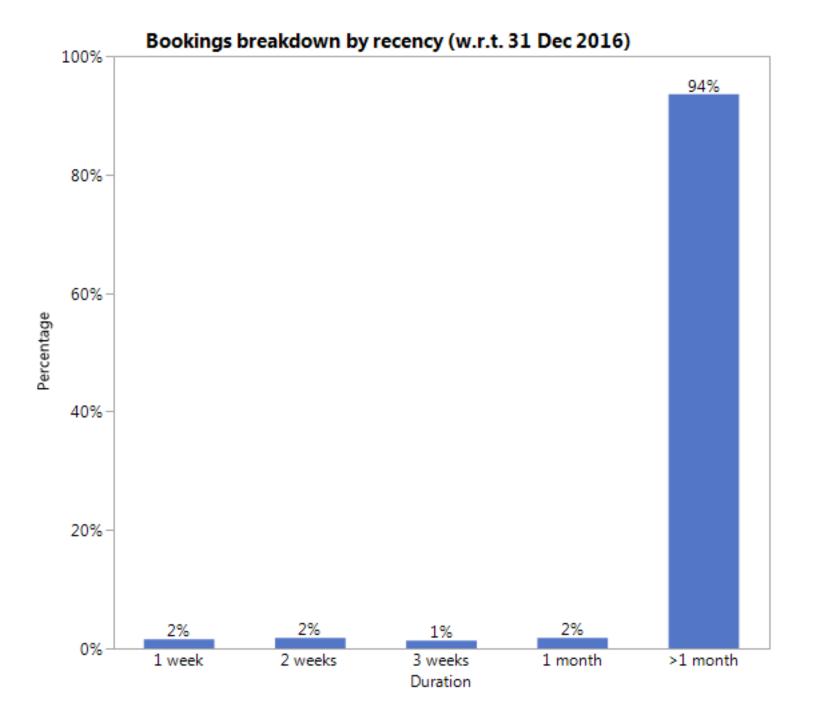
EXPLORATORY DATA ANALYSIS

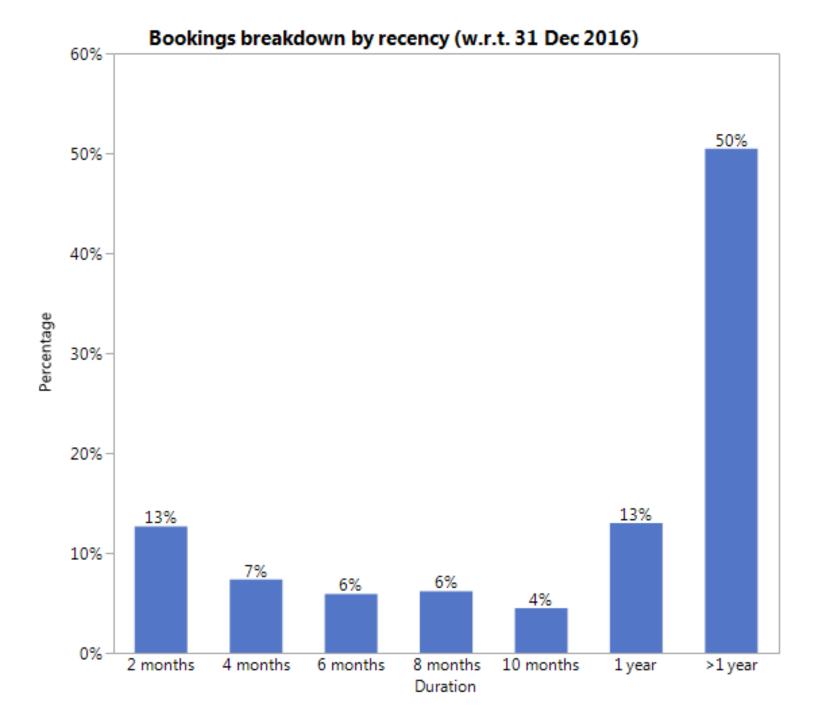
Recency Frequency Monetary

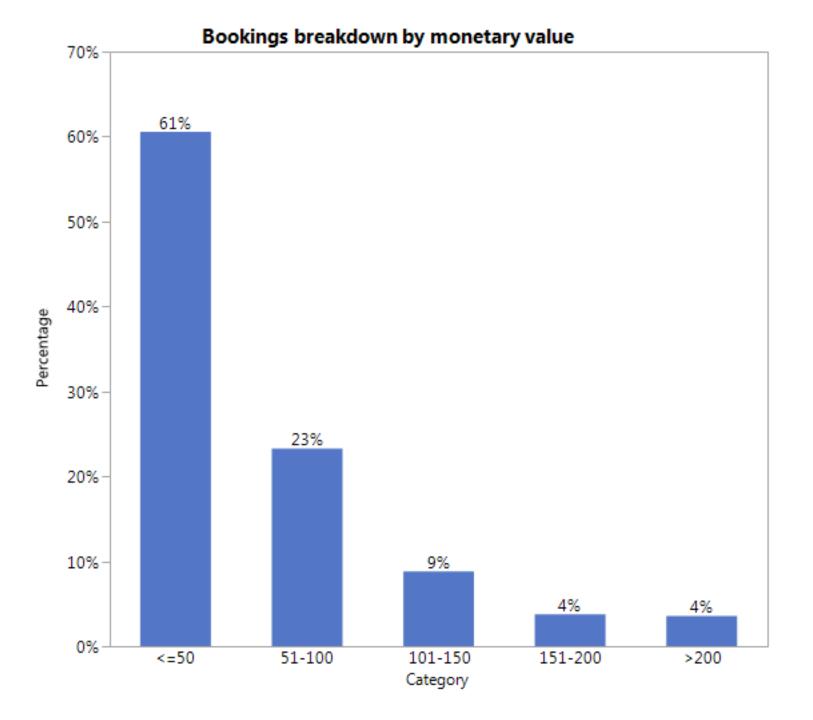
EXPLORATORY DATA ANALYSIS (EDA)

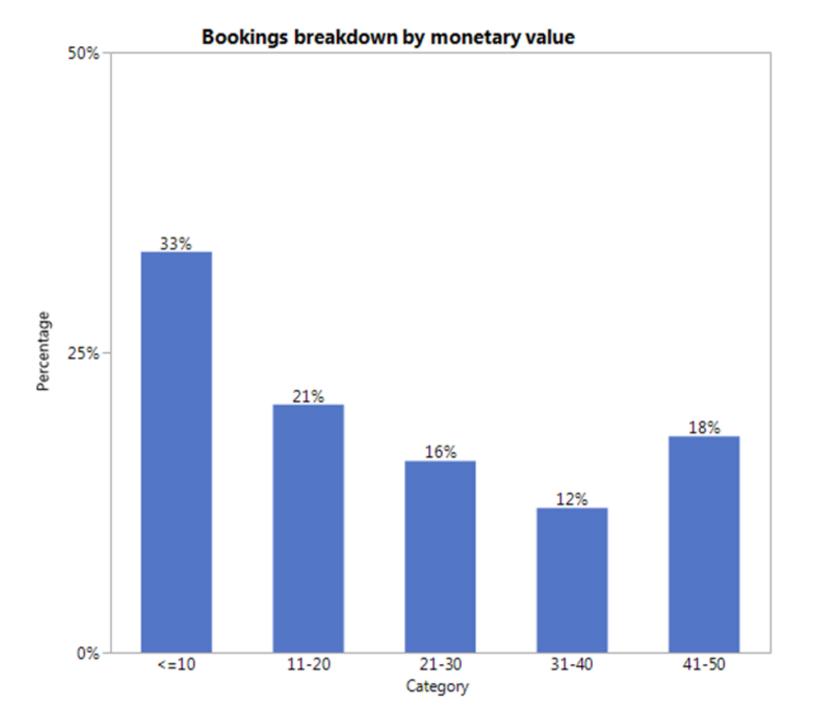
- Users breakdown by type
- Customers breakdown by age, gender
- Professionals breakdown by age, gender
- Bookings breakdown by type
- Bookings breakdown by status
- Bookings breakdown by year, month, day
- Bookings breakdown by category
- Bookings breakdown by frequency, recency, monetary amount
- Bookings breakdown by duration from sign up to 1st online booking.
- Bookings breakdown by service count
- Services breakdown by price, category



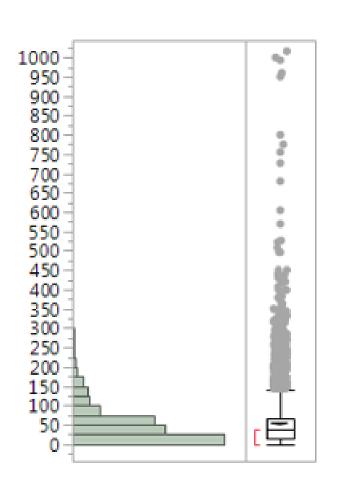






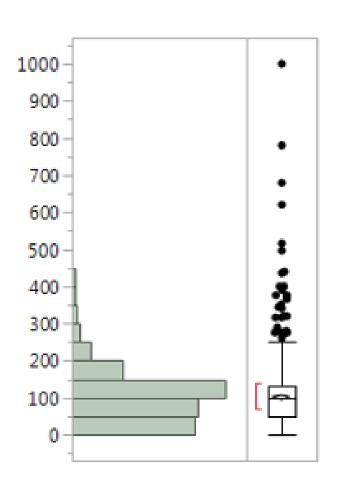


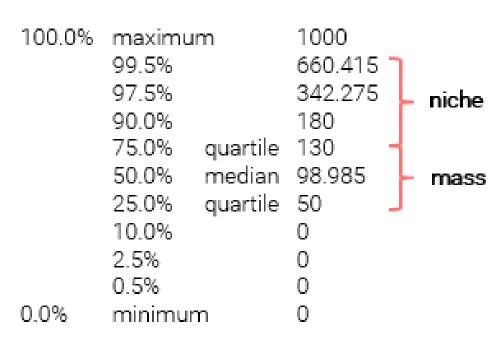
BOOKINGS (NAILS)



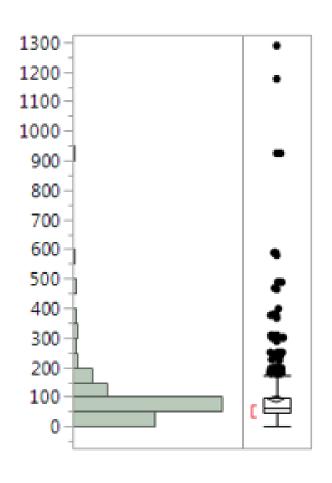
100.0%	maximu	m	1016		
99.5%		392.749	75	٦	
	97.5%		210	L	_
	90.0%		127.15		niche
	75.0%	quartile	66.485	4	
	50.0%	median	37.84	_	mass
	25.0%	quartile	15		
	10.0%	•	0		
	2.5%		0		
	0.5%		0		
0.0%	minimur	n	0		

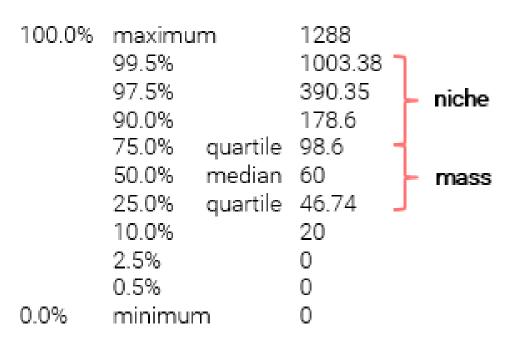
BOOKINGS (MAKEUP)



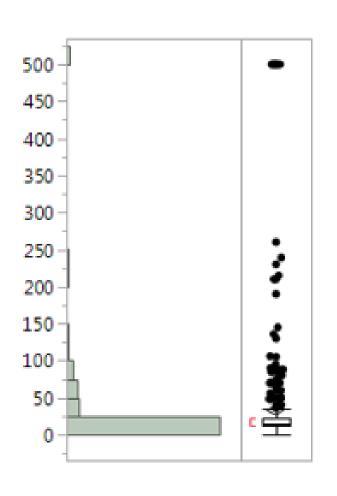


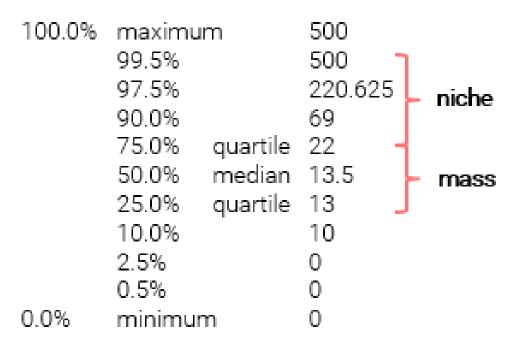
BOOKINGS (BROWS)



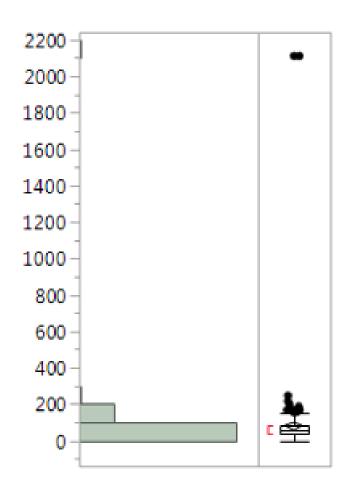


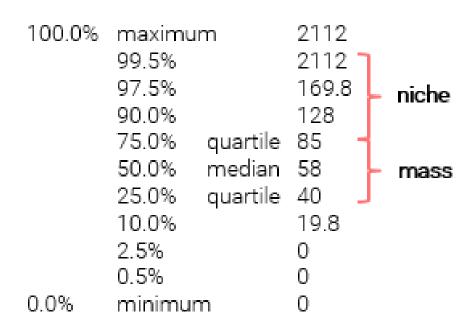
BOOKINGS (HAIR STYLING)

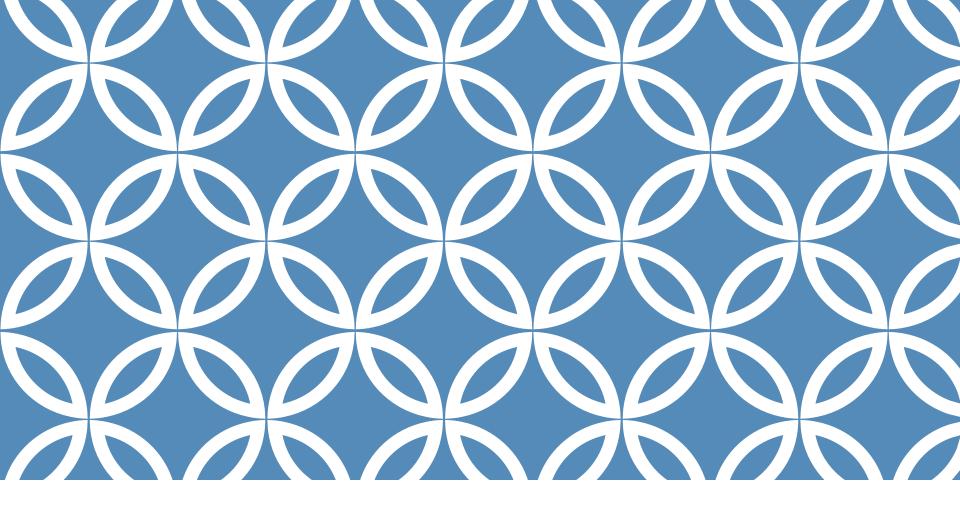




BOOKINGS (FACIAL)



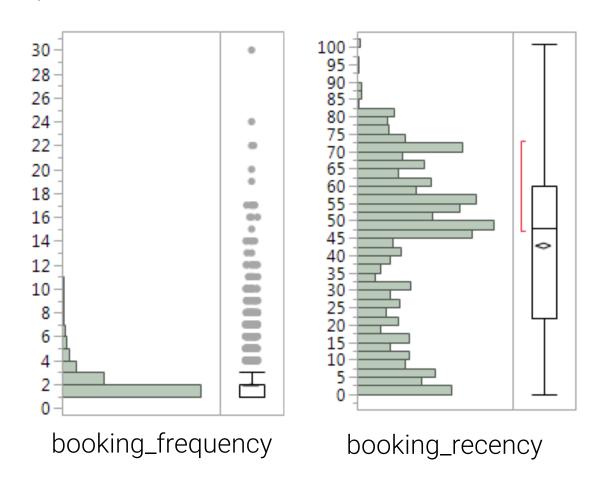


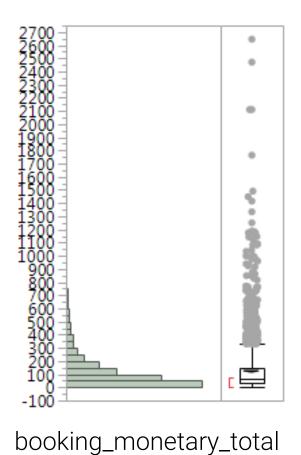


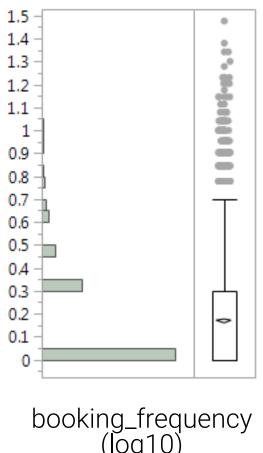
Customer Segmentation

- Aims to discover any potential customer segmentation
- K-means clustering
- Looked at customers who have made at least one booking
- Clustering variables
 - Booking recency
 - Booking frequency
 - Total monetary value

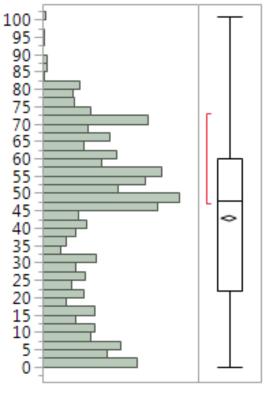




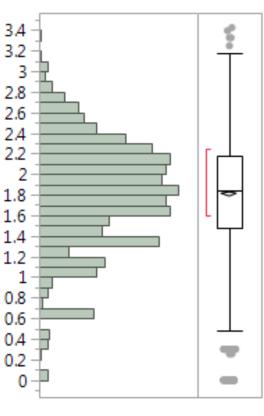




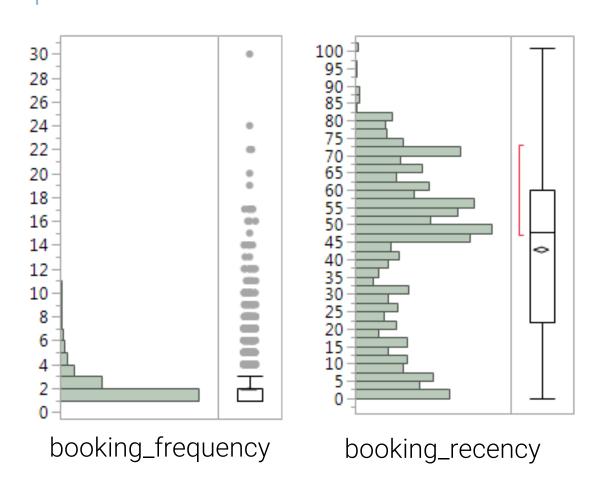
booking_frequency (log10)

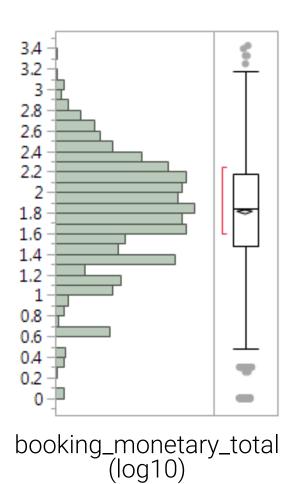


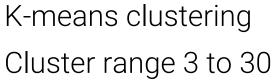
booking_recency

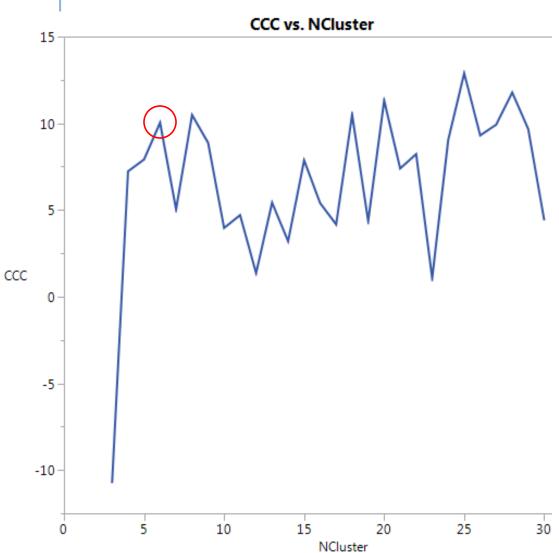


booking_monetary_total (log10)







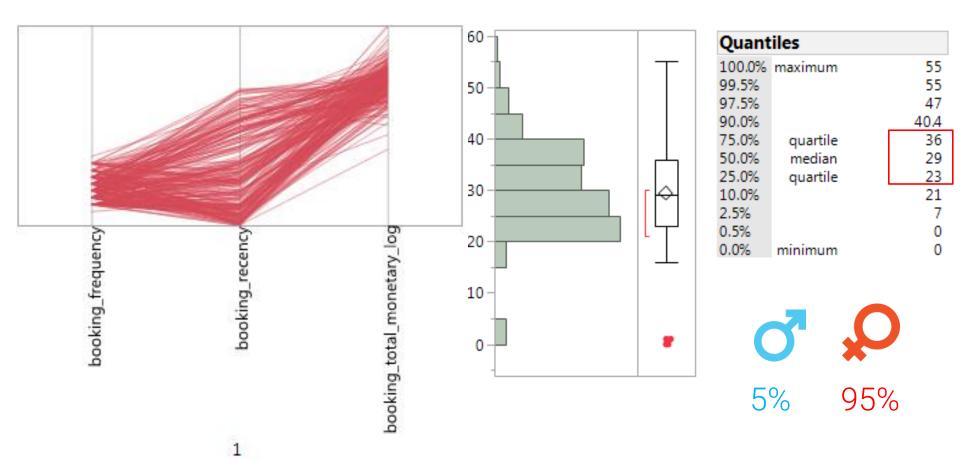


Cluster Compar	ison		
Method	NCluster	CCC	Best
K-Means Clustering	3	-10.75	
K-Means Clustering	4	7.23595	
K-Means Clustering	5	7.92032	_
K-Means Clustering	6	10.0372	
K-Means Clustering	7	5.04086	_
K-Means Clustering	8	10.4839	Optimal CCC
K-Means Clustering	9	8.88005	
K-Means Clustering	10	3.96878	
K-Means Clustering	11	4.70582	
K-Means Clustering	12	1.36393	
K-Means Clustering	13	5.42802	
K-Means Clustering	14	3.20073	
K-Means Clustering	15	7.87067	
K-Means Clustering	16		
K-Means Clustering	17		
K-Means Clustering	18	10.5065	
K-Means Clustering	19	4.36813	
K-Means Clustering	20	11.326	
K-Means Clustering	21		
K-Means Clustering	22		
K-Means Clustering	23	1.07648	
K-Means Clustering	24	9.03791	
K-Means Clustering	25	12.8891	
K-Means Clustering	26		
K-Means Clustering		9.92589	
K-Means Clustering	28	11.7841	
K-Means Clustering	29		
K-Means Clustering	30	4.40883	

No. of clusters = 6

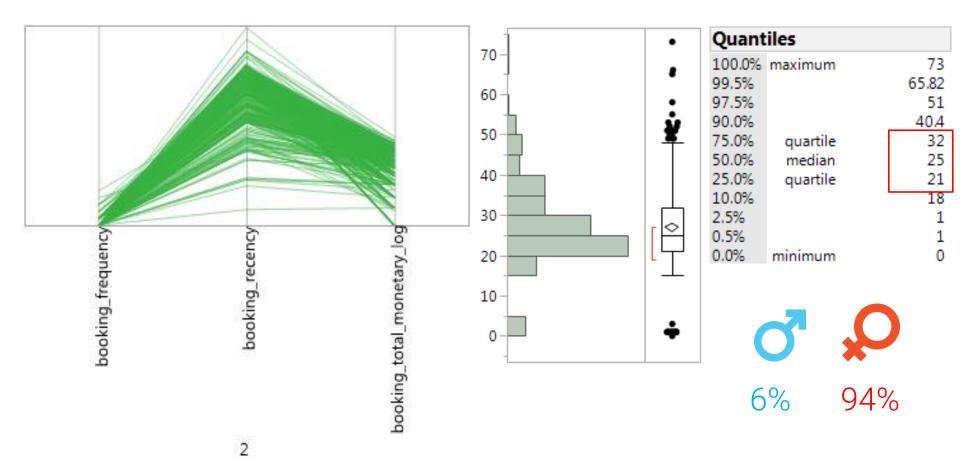
Cluster Summary					
Cluster	Count		Step	Criterion	
1	258		55	0	
2	620				
3	42				
4	1121				
5	793				
6	1020				

Cluster Means				
Cluster	booking_frequency	booking_recency	booking_total_monetary_log	
1	6.29844961	27.0232558	2.47834188	
2	1.1516129	68.4483871	0.97168195	
3	14.6190476	16.1904762	2.74301569	
4	1.29884032	56.8412132	1.73241173	
5	1.85119798	47.0063052	2.30626783	
6	1.46666667	12.7460784	1.81295839	



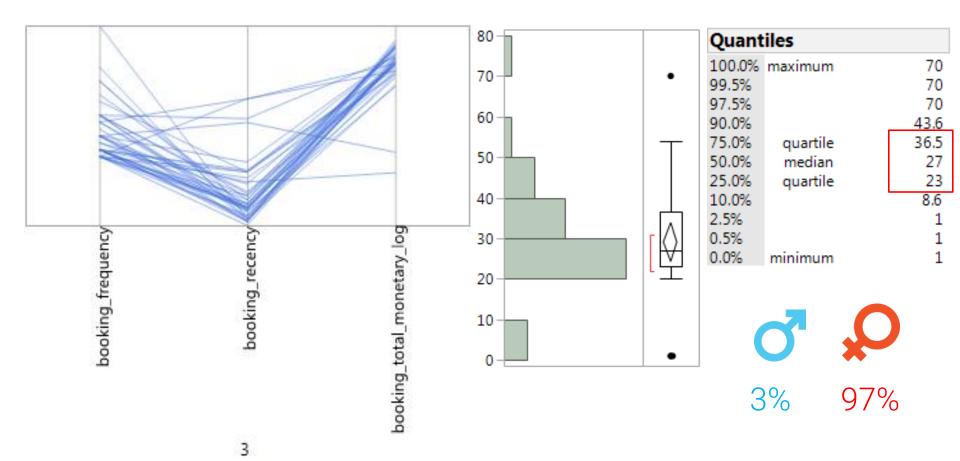
- Moderate frequency
- Widespread recency
- High monetary value

- 258 customers
- Age around 23 to 36
- Average total monetary spending of \$125



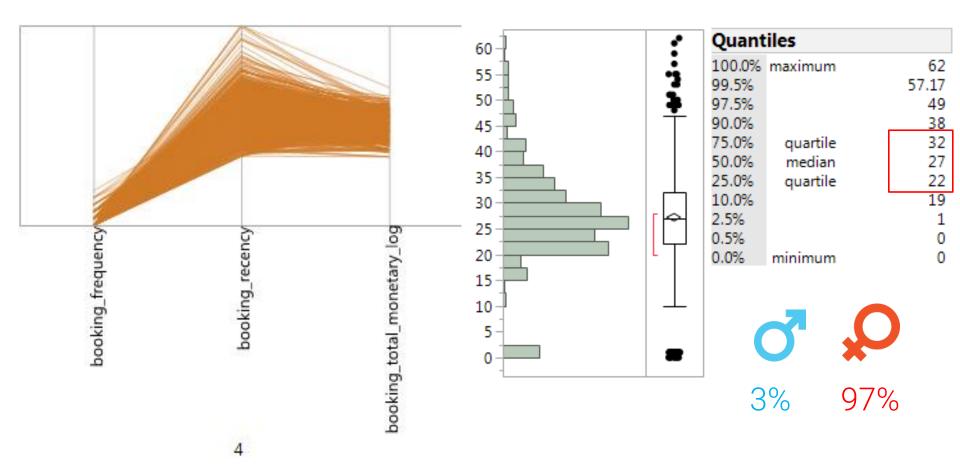
- Low frequency
- High recency (haven't booked in a year)
- Low monetary value

- 620 customers
- Age around 21 to 32
- Average total monetary spending of ~ \$12



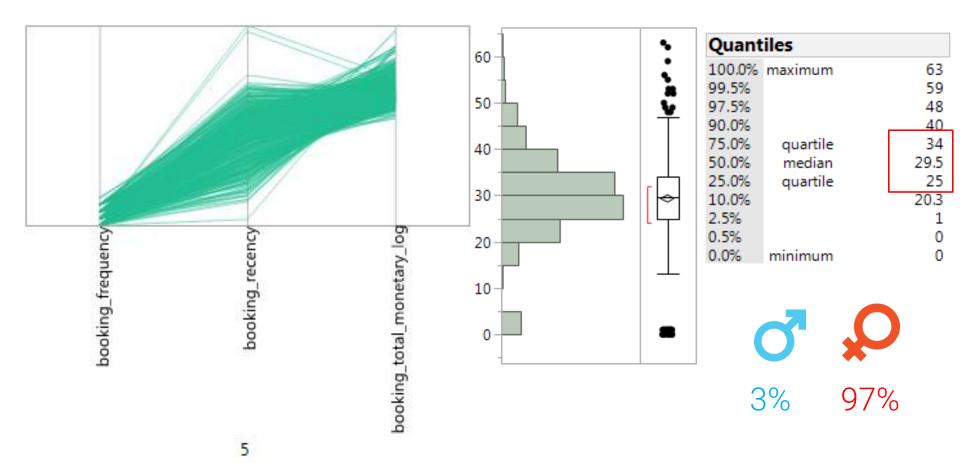
- High frequency
- Low recency (<4months)
- High monetary value

- 42 customers
- Age around 23 to 36
- Average total monetary spending of ~ \$700



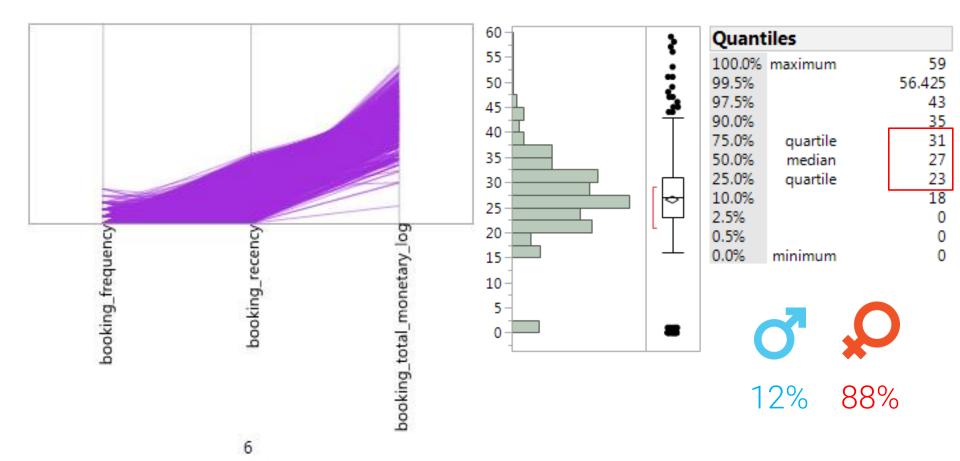
- Low frequency
- High recency (haven't booked in a year)
- Medium monetary value

- 1121 customers
- Age around 22 to 32
- Average total monetary spending of ~ \$60



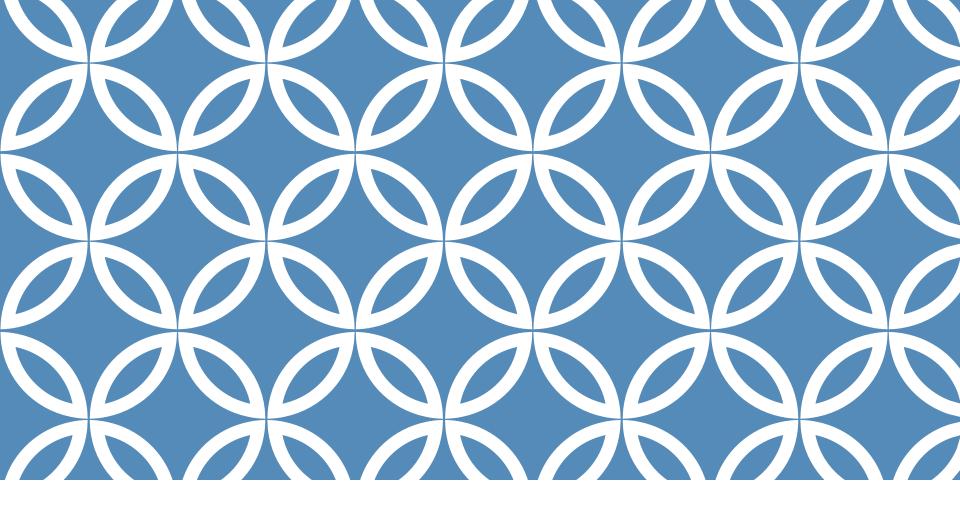
- Low frequency
- Moderate recency
- High monetary value

- 793 customers
- Age around 25 to 34
- Average total monetary spending of ~ \$238



- Low frequency
- Low recency
- Medium monetary value

- 1020 customers
- Age around 23 to 31
- Average total monetary spending of ~ \$89



Market Basket Analysis

Item – each item is a category of a service (i.e. nails, makeup etc.) that the customer includes in a booking.

ID – the ID of the transaction that the item belongs to (i.e. booking ID)

Support – proportion of transactions in which an item set appears

Confidence – proportion of transactions that contain the consequent item set, given that the condition item set is in the transaction

Lift – ratio of an association rule's confidence to its expected confidence

service_count	service_ids.0	service_ids.1
1	570a8572d65405706d6aee56	
2	570a8573d65405706d6af2ed	570a8572d65405706d6aee75
2	570a8572d65405706d6aee4a	570a8572d65405706d6aee4b

booking_id	service_count	label	service_id	category_1
570a8716d65405706d6b7956	1	service_ids.0	570a8574d65405706d6afa92	Nails
570a8716d65405706d6b7964	2	service_ids.0	570a8574d65405706d6afb43	Nails
570a8716d65405706d6b7964	2	service_ids.1	570a8573d65405706d6af480	Nails
570a8716d65405706d6b7965	1	service_ids.0	570a8572d65405706d6aef94	Makeup
570a8716d65405706d6b7966	1	service_ids.0	570a8572d65405706d6aef94	Makeup
570a8716d65405706d6b7967	1	service_ids.0	570a8573d65405706d6af7d8	Makeup
570a8716d65405706d6b7968	1	service_ids.0	570a8573d65405706d6af7d8	Makeup
570a8716d65405706d6b796a	1	service_ids.0	570a8573d65405706d6af7d8	Makeup

Including bookings with service count = 1

Minimum Support	0.001
Minimum Confidence	0.001
Minimum Lift	0.001
Maximum Antecedents	3
Maximum Rule Size	99

Support	N Items
75%	1
8%	1
7%	1
5%	1
4%	1
2%	1
1%	2
1%	1
0%	2
0%	2
0%	2
0%	1
0%	2
0%	2
	8% 7% 5% 4% 2% 1% 0% 0% 0% 0%

Frequent Item Sets

Rule			
Condition	Consequent	Confidence	Lift
Hair Removal	Nails	23%	0.303
Brow & Lash	Nails	12%	0.156
Hair Removal	Facial	8%	2.169
Facial	Brow & Lash	6%	0.878
Hair Removal	Brow & Lash	6%	0.788
Facial	Hair Removal	5%	2.169
Brow & Lash	Facial	3%	0.878
Facial	Nails	3%	0.04
Brow & Lash	Hair Removal	2%	0.788
Nails	Brow & Lash	1%	0.156
Nails	Hair Removal	1%	0.303
Nails	Facial	0%	0.04

Rules

Excluding bookings with service count = 1

Minimum Support	0.01
Minimum Confidence	0.01
Minimum Lift	0.01
Maximum Antecedents	3
Maximum Rule Size	99

Frequent Item Sets				
Item Set	Support	N Items		
{Nails}	92%	1		
{Brow & Lash}	7%	1		
{Hair Removal}	5%	1		
(Brow & Lash, Nails)	3%	2		
{Facial}	2%	1		
{Hair Removal, Nails}	2%	2		
{Makeup}	1%	1		

Rules					
Ru	ule				
Condition	Consequent	Confidence	Lift		
Brow & Lash	Nails	49%	0.533		
Hair Removal	Nails	42%	0.453		
Nails	Brow & Lash	4%	0.533		
Nails	Hair Removal	2%	0.453		

booking_id	service_count	label	service_id	category_1
570a86fed65405706d6b757f	3	service_ids.0	570a8572d65405706d6af11f	Nails
570a86fed65405706d6b757f	3	service_ids.1	570a8572d65405706d6af120	Nails
570a86fed65405706d6b757f	3	service_ids.2	570a8572d65405706d6af124	Nails
570a86ffd65405706d6b7596	1	service_ids.0	570a8572d65405706d6af09b	Nails
570a86ffd65405706d6b759a	1	service_ids.0	570a8573d65405706d6af36d	Nails
570a86ffd65405706d6b75b1	3	service_ids.0	570a8572d65405706d6af071	Hair Styling
570a86ffd65405706d6b75b1	3	service_ids.1	570a8572d65405706d6af073	Hair Styling
570a86ffd65405706d6b75b1	3	service_ids.2	570a8572d65405706d6af06c	Hair Styling

booking_id	service_count	label	service_id	category_1	category_2
570a86fed65405706d6b757f	3	service_ids.0	570a8572d65405706d6af11f	Nails	Gel
570a86fed65405706d6b757f	3	service_ids.1	570a8572d65405706d6af120	Nails	Nail Art
570a86fed65405706d6b757f	3	service_ids.2	570a8572d65405706d6af124	Nails	Nail Art
570a86ffd65405706d6b7596	1	service_ids.0	570a8572d65405706d6af09b	Nails	Gel
570a86ffd65405706d6b759a	1	service_ids.0	570a8573d65405706d6af36d	Nails	Gel
570a86ffd65405706d6b75b1	3	service_ids.0	570a8572d65405706d6af071	Hair Styling	Colour
570a86ffd65405706d6b75b1	3	service_ids.1	570a8572d65405706d6af073	Hair Styling	Colour
570a86ffd65405706d6b75b1	3	service_ids.2	570a8572d65405706d6af06c	Hair Styling	Cut

Minimum Support 0.01

Minimum Confidence 0.01

Minimum Lift 0.01

Maximum Antecedents 3

Maximum Rule Size 99

Including bookings with service count = 1

Frequent Item Sets		
Item Set	Support	N Items
{Classic}	40%	1
{Express}	24%	1
{Nail Art}	14%	1
{Removal}	9%	1
(Classic, Nail Art)	7%	2
{Lash}	6%	1
{Extension}	5%	1
{Nail Spa}	4%	1
{Classic, Removal}	4%	2
(Dinner & Dance)	4%	1
{Cut}	3%	1
{Express, Nail Art}	3%	2
{Express, Removal}	2%	2
{Classic, Express}	2%	2
{Nail Art, Removal}	2%	2
{Waxing}	2%	1
{Brow}	2%	1
{Others}	1%	1
(Classic, Nail Art, Removal)	1%	3
{Extension, Nail Art}	1%	2
{Classic, Nail Spa}	1%	2

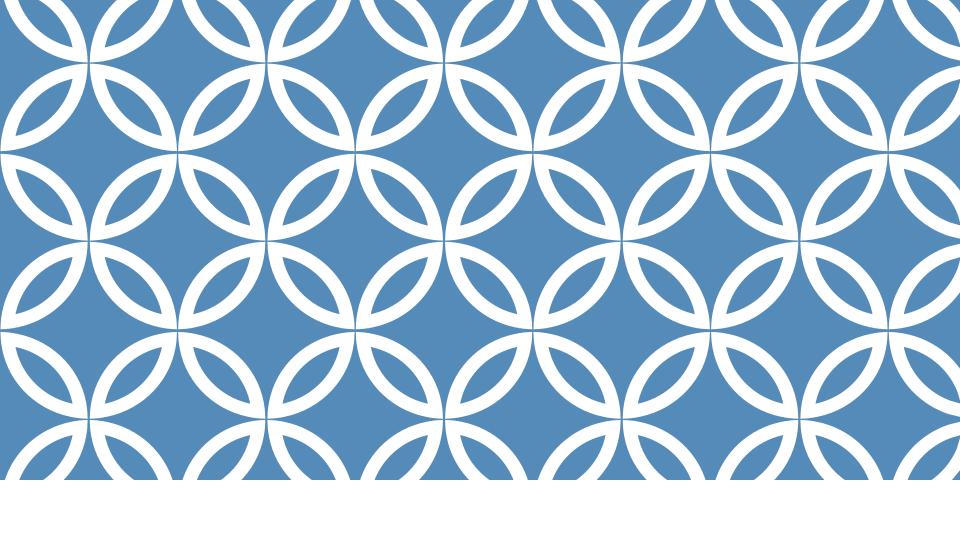
Rules			
Rule			
Condition	Consequent	Confidence	Lift
Nail Art, Removal	Classic	61%	1.524
Removal	Classic	48%	1.197
Nail Art	Classic	48%	1.195
Classic, Removal	Nail Art	31%	2.242
Nail Spa	Classic	29%	0.716
Extension	Nail Art	25%	1.841
Removal	Express	25%	1.056
Removal	Nail Art	24%	1.761
Nail Art	Express	22%	0.913
Classic, Nail Art	Removal	19%	2.246
Classic	Nail Art	17%	1.195
Nail Art	Removal	15%	1.761
Removal	Classic, Nail Art	15%	2.246
Express	Nail Art	13%	0.913
Classic	Removal	10%	1.197
Nail Art	Classic, Removal	9%	2.242
Express	Removal	9%	1.056
Nail Art	Extension	9%	1.841
Express	Classic	9%	0.224
Classic	Express	5%	0.224
Classic	Nail Art, Removal	3%	1.524
Classic	Nail Spa	3%	0.716

Excluding bookings with service count = 1

Minimum Support	0.1
Minimum Confidence	0.1
Minimum Lift	0.1
Maximum Antecedents	3
Maximum Rule Size	99

Frequent Item Sets			
Item Set	Support	N Items	
{Classic}	60%	_ 1	
{Nail Art}	42%	1	
{Express}	29%	1	
{Removal}	26%	1	
(Classic, Nail Art)	26%	2	
{Classic, Removal}	16%	2	
{Express, Nail Art}	12%	2	

Rules			
Rule			
Condition	Consequent	Confidence	Lift
Removal	Classic	61%	1.002
Nail Art	Classic	60%	0.998
Classic	Nail Art	42%	0.998
Express	Nail Art	40%	0.934
Nail Art	Express	27%	0.934
Classic	Removal	26%	1.002



RECOMMENDATIONS

RECOMMENDATIONS

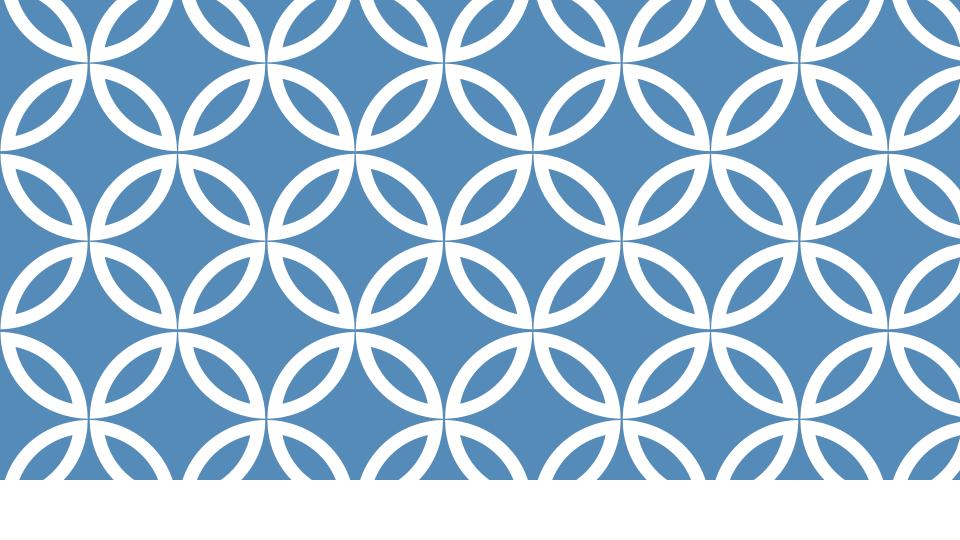
Acquire Enhance Retain

- Who are the customers
- Exploratory Data Analysis
- Cluster Analysis

- Enhance profitability of existing customers
- Cross-selling
- Association Analysis

RECOMMENDATIONS

- Targeted marketing at different groups of customers
- 2. Keeping track of RFM attributes for each customer
- Have campaigns that are tailored to the association between
 2 or more services in a booking
 - e.g. 20% off classic service if there is removal service



THANK YOU