



Retail Customer Segmentation using SAS

April 2014

Calgary SAS Users Group meeting

Jenny Chen

Data Science, LoyaltyOne



a LoyaltyOne
program

Agenda

- Overview
- Applications
- Objectives
- Types of Segmentation
- A Real Example with SAS Code
- Further Reading



a LoyaltyOne
program

Overview

- **Customer Segmentation is the practice of classifying your customers into distinct groups based on the similarities they share with respect to any characteristics you deem relevant to your business**
- **Key components in developing proper, actionable segmentation**
 - Understand business needs and objectives
 - Customer satisfaction
 - Share of wallet
 - Market share
 - Available customer information
 - Demographic
 - Geographic
 - Behavioral
 - Attitudinal



a LoyaltyOne
program

Applications

| Applications | Examples |
|----------------------------------|---|
| Customer Investment Allocation | <ul style="list-style-type: none">• Determine segment's current and potential profitability• Define differentiated value proposition by segment |
| Customer Relationship Management | <ul style="list-style-type: none">• Better meet customer needs, improve customer satisfaction• Develop customer loyalty to your brand• Improve retention strategy |
| Tailor Marketing Program | <ul style="list-style-type: none">• Build relevant campaign, communication• Optimize marketing channel mix |
| Guide New Product Development | <ul style="list-style-type: none">• Product preference• Potential targeting audience |
| Guide Product, Service Pricing | <ul style="list-style-type: none">• Price sensitivity• Stimulate customer demand |

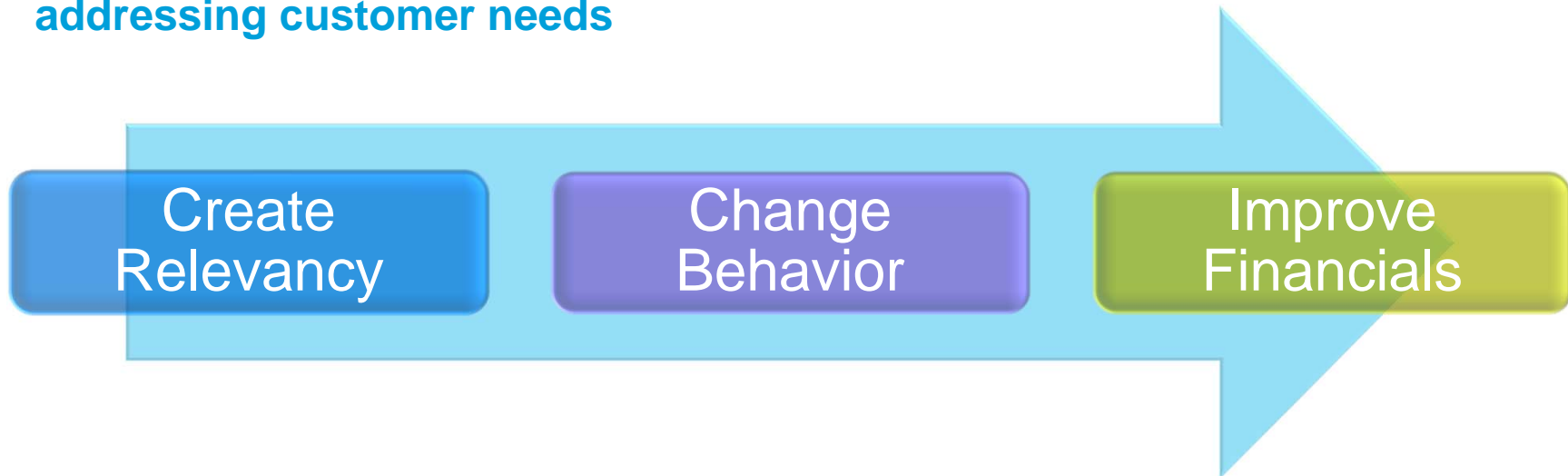
...and many more



a LoyaltyOne
program

Objectives

Segmentation enables business to increase profitability while positively addressing customer needs



- Better defined target universe
- Right offer, communication, channel, time for right customers
- Personalized experience between clients and their customers

- Higher response rates
- Combat attrition of high profit customers
- Grooming lower current value high potential customers
- Identify cross-sell & up-sell opportunities by customer

- Measureable improvement in sales, traffic, revenue
- More cost effective campaign deployment and ROI
- Opportunity identification by category, department, store, etc.



a LoyaltyOne
program

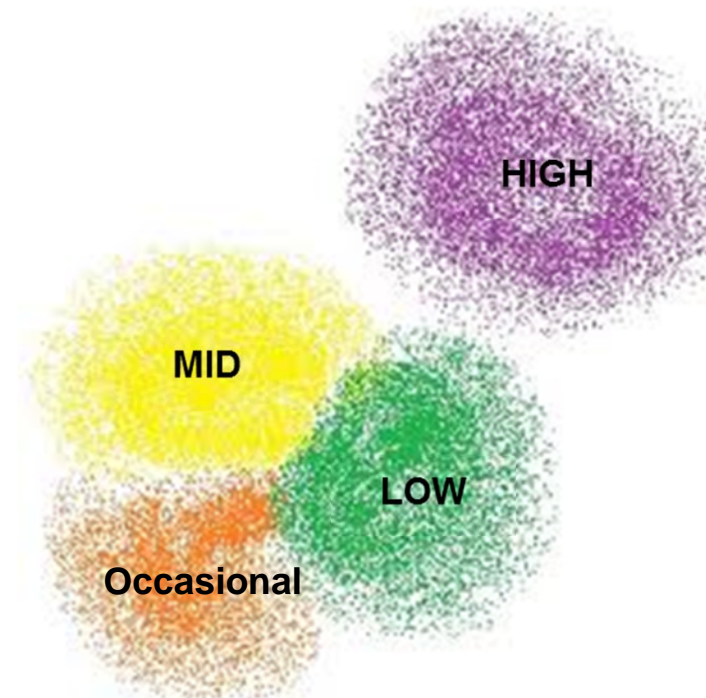
Segmentation Type

Data source

- Demographic, geographic
- Behavior, attitudinal
- Life stage, life style

Approach

- Business rules
 - Profit ranking
 - RFM (Recency, Frequency, Monetary)
- Supervised clustering
 - Decision tree
- Unsupervised clustering
 - K-means clustering



a LoyaltyOne
program

A Behavior Based Segmentation Example

Using unsupervised clustering segmentation for a grocery chain which would like better product assortment for its high profitable customers

Potential Inputs

Value

- Basket Size
- Visit Frequency

Basket

- Spend by category
- Type of category
- Brand spend (i.e. private label)

Promotions

- % bought on targeted promotion
- % bought from flyer

Time

- Time of day
- Day of week

Location

- Store format
- Area population density

Clustering
approach



Deal Seeking Mom

Key Differentiators



- Full store shop
- High avg. basket size / # trips



- High % purchased on promotion
- Rewards seeker



- High spend categories
 - Fresh produce
 - Organic food
 - Multipack juice, snack



a LoyaltyOne
program

Segmentation Building Process I

Define business objective

- Identify high profitable customer groups
- Improve product assortment for them
- Design relevant, effective marketing campaigns for them
- Quantify campaign measurements

Gather information

- Customer transactional data
- Promotional data
- Client store, department, category merchandising hierarchy

Align with client on project scope

- Business Objectives
- Data Sources
- Methodology
- Measurement plan



a LoyaltyOne
program

Segmentation Building Process II

Data preparation

- Data merge purge, data cleaning, dealing with missing
- Convert all attributes to numeric metrics, cap outliers
- Standardize all metrics/variables
- Standardization enables each metric have similar contribution to the future cluster building

```
PROC STDIZE DATA=SAS_DATA.ALL_INFO  
OUT=SAS_DATA.STANDARD_STD  
OUTSTAT=SAS_DATA.SUMMARY  
(WHERE=( _TYPE_ IN ( 'LOCATION' , 'SCALE' ) ) );  
VAR &ALL_VAR;  
RUN;
```



a LoyaltyOne
program

Segmentation Building Process III

Data exploration, variable reduction

- Check descriptive statistics, correlation among all potential variables
- Use principle component, factor analysis (**PROC PRINCOMP**, **PROC FACTOR**)
- Use the Variable Clustering node in SAS Enterprise Miner to create variable cluster constellation plot and variable cluster tree diagram

Measure similarity among customers

- Using Euclidean distance - this measures the distance between 2 points in multi-dimensional space
- A very common method of unsupervised clustering with large data using Euclidean distance is K-means clustering



a LoyaltyOne
program

Segmentation Building Process IV

K-means clustering

- Using the distances to group customers into K clusters where each customer is with the nearest centroid
- The centroid is calculated as the multi-dimensional set of the means of the variables used for the particular cluster
- Pre-determine a range of number of clusters, use bottom-up approach
- Test the FASTCLUS procedure repeatedly using different starting points and different number of clusters until we obtain stabilized centroids and desired distance between clusters

```
PROC FASTCLUS DATA=SAS_DATA.STANDARD_STD MAXCLUSTERS=29
OUT=SAS_DATA.RESULT OUTSEED=SAS_DATA.CENTRES
OUTSTAT=SAS_DATA.CLUSTER_STAT REPLACE=RANDOM
RANDOM=187413849 CONVERGE=0.001 MAXITER=350;
VAR &VAR_LIST;
RUN;
```



a LoyaltyOne
program

Output of K-means Clustering

- Centroids info

| Cluster Mean | | | | | | | | | |
|--------------|--------|--------|--------|--------|--------|--------|--------|--------|--------|
| Cluster | Var1 | Var2 | Var3 | Var4 | Var5 | Var6 | Var7 | Var8 | Var9 |
| 1 | (0.68) | (0.28) | (0.37) | (0.20) | (0.77) | 0.46 | (0.29) | 1.11 | 1.70 |
| 2 | 0.81 | 0.91 | 2.44 | (0.19) | 0.93 | 0.95 | 0.41 | (0.62) | (0.56) |
| 3 | (0.68) | (0.62) | (0.73) | (0.44) | (0.77) | (0.67) | (0.53) | 1.11 | 1.66 |
| 4 | (0.49) | (0.54) | (0.45) | (0.45) | (0.51) | (0.33) | (0.48) | (0.46) | (0.56) |
| 5 | (0.17) | 0.04 | 0.10 | (0.02) | (0.13) | 0.73 | 0.70 | (0.43) | (0.55) |
| ... | | | | | | | | | |
| 29 | (0.68) | (0.68) | (0.84) | (0.47) | (0.77) | (1.64) | (0.58) | 2.27 | 2.40 |

- Statistics for variable comparison and further variable reduction
- $RSQ/(1-RSQ)$ is the ratio of between cluster variance to within cluster variance
- Get client's buy-in for the final list of variables for cluster building

| Statistics for Variables | |
|--------------------------|---------------|
| Variable | $RSQ/(1-RSQ)$ |
| Var1 | 13.82 |
| Var2 | 16.51 |
| Var3 | 6.69 |
| ... | |
| Var 9 | 56.76 |



a LoyaltyOne program

Segmentation Building Process V

Bottom-up approach

- Flexible with how many clusters to choose initially
- Aggregate clusters up afterwards using hierarchy clustering **PROC CLUSTER**
- Create a tree diagram for clusters using **PROC TREE**
- Choose proper distance to fit into client's request of final number of segments and distribution of segments

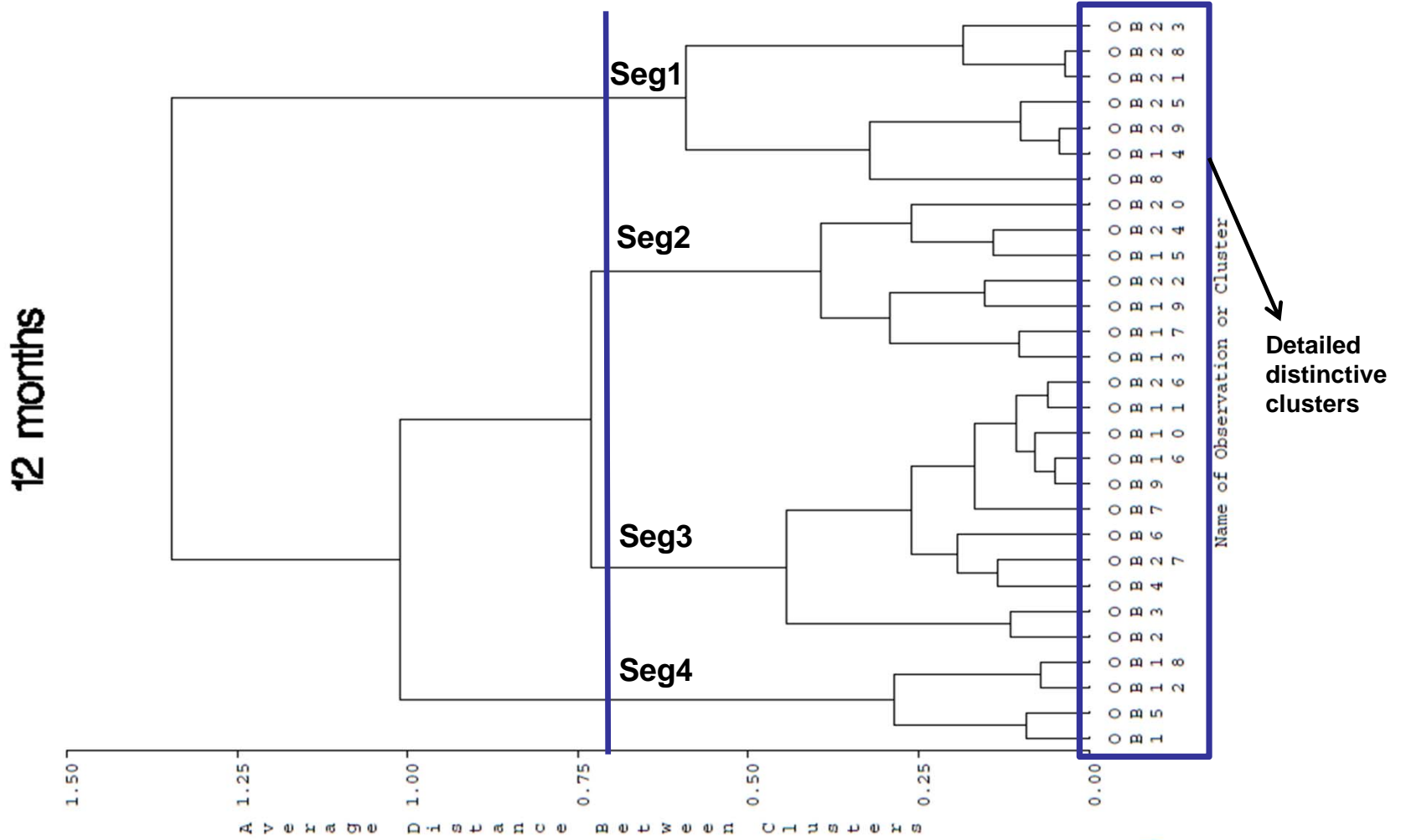
```
PROC CLUSTER DATA=SAS_DATA.CENTRES METHOD=AVE OUTTREE=TREE ;  
VAR &VAR_LIST ;  
RUN ;  
  
PROC TREE DATA=TREE ;  
RUN ;
```



a LoyaltyOne
program

Bottom-up Approach

Tree diagram to illustrate the arrangement of the clusters



a LoyaltyOne program

Segmentation Building Process VI

Segments profiling

- Using cluster building variables to profile the segments
- Using additional data source to profile them as well
- Paint a clear picture of your segments

Validate segmentation effectiveness

- Test campaigns with segments strategy
- Measure campaign results by segments
- Validation segments with market research survey

Review, iterate, deploy again

- Update segments regularly
- Monitor the migration of segments
- Gather feedback, campaign response results
- Maintain and improve the process

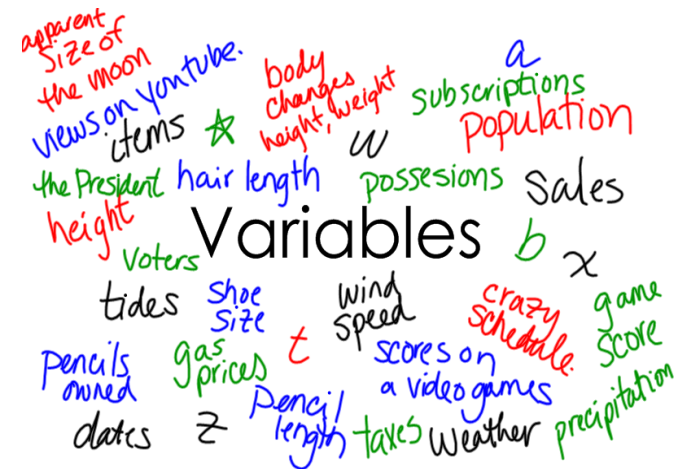


a LoyaltyOne
program

Pros and Cons of K-means Clustering

Pros

- Flexible with fewer assumptions about your customer population
- Can incorporate many relevant metrics
- Provide more comprehensive, diverse segments
- Simple algorithm fast to run with large dataset
- Prove to work well for behavioral targeting



Cons

- Purely data driven
- Pre-specify number of clusters or the minimum distance between clusters
- Time consuming to maintain: for example, certain metrics used to build the clusters were changing overtime dramatically
- May have difficulty to explain the concept to your marketing clients



a LoyaltyOne
program

Question or Comment



a LoyaltyOne
program

Reference

SAS website

<http://support.sas.com/documentation/>

<http://support.sas.com/resources/papers/>

<http://support.sas.com/events/sasglobalforum/previous/online.html>

- <http://support.sas.com/resources/papers/proceedings12/103-2012.pdf>
- <http://support.sas.com/resources/papers/proceedings12/200-2012.pdf>
- <http://support.sas.com/resources/papers/proceedings12/329-2012.pdf>

A video about LoyaltyOne: Enriching Relationships

<http://www.youtube.com/watch?v=NSed-m3Z8Ik>



a LoyaltyOne
program



AllianceData.

LoyaltyOne

© 2014 LoyaltyOne, Inc. All rights reserved.

No part of this material may be produced in any form without the written permission of the owner. No representation or warranty of any kind is made with respect to the content hereof and no liability is assumed for errors contained herein or for incidental or consequential damages in connection with the furnishing, performance or use of this publication or its contents.