

# Managing Customer Heterogeneity II Factor Analysis for Perceptual Maps

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## Agenda



- Discuss perceptual mapping methods for measuring and portraying how consumers perceive brands
- We will discuss:
  - Positioning
  - Perceptual mapping
  - Factor analysis







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## Consumer perceptions







Safety

Performance

Luxury

Fuel efficient









## Positioning vs. differentiation

- Differentiation = What you do to an offering
- Positioning = What you do to the minds of consumers

## Perception is reality

- Know your customer
- Know your brand's perception
- Find touchpoints that connect
- Manage your perceptions









### Positioning statements

- For [target segment], the [product/service] is [most important claim] because [single most important support]
- For [females of age-group18-49 yrs who possess dry damaged hair and believe they cannot achieve truly healthy/shiny hair], Pantene is a [hair care system (shampoo/conditioner/styling aids)] that offers ["hair so healthy it shines"] because it ["penetrates from root to tip" through its patented Pro-Vitamin B5 formula].









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## Positioning base













## Perceptual mapping



- Positioning can be influenced by consumers perceptions
- Perceptual mapping is a technique used to visually display consumer's views
- Consumers are asked questions about their perceptions of the product
  - Select a set of brands of interest to the target group of customers
  - Identify <u>a set of key attributes</u> on which these brands need to be evaluated by the target group
  - Have customers evaluate the products on the chosen set of attributes
- Results are employed either improving existing product or developing a new one





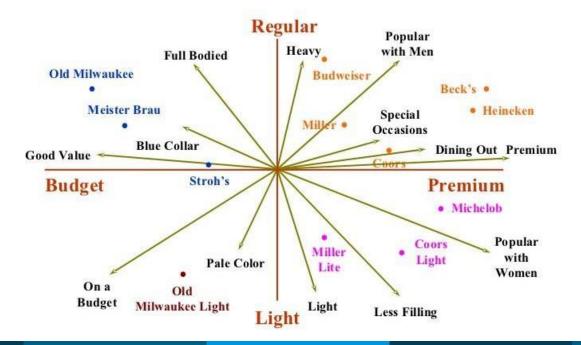




A perceptual map provides a picture of how consumers perceive different competitors

### **Beer Market**

Perceptual Mapping



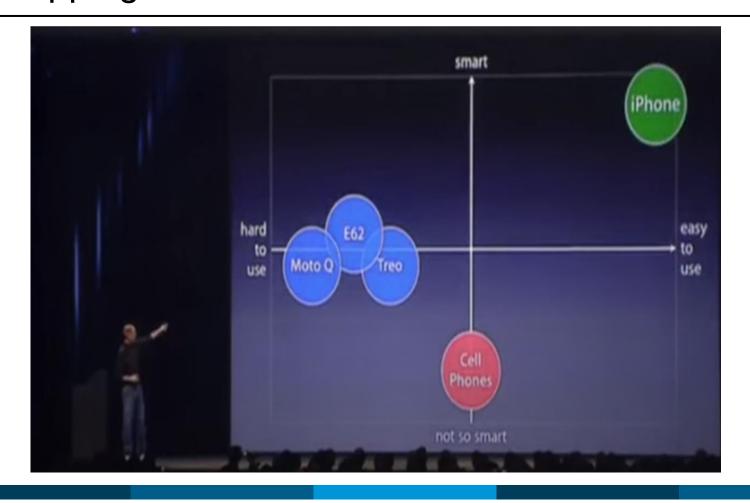






## Perceptual mapping











## Perceptual mapping



#### Procedure:

- Generate a matrix of rating data consisting of each customer's (C1, C2,...)
   ratings of each brand (B1, B2, ...) on each of the attributes (A1, A2, A3,....)
- Compute average ratings of each brand on each attribute
- Use factor analysis for dimension reduction
- Interpret the underlying key dimensions of the map using the directions of the individual attributes
- Articulate the implications of how customers' view the competing products and concepts







## **Factor Analysis**



- Factor analysis is a data reduction technique that can be used to identify a small number of latent "factors" that explain the variation in a large number of observed variables.
- It is employed to construct perceptual maps for product positioning, by condensing a large pool of potential customer needs, wants, and preferences into a short set of similar characteristics.







## Factor Analysis: Intuition

#### Example

The manager of an online retailor collected customer rating data (on a 100 scale) from a survey of 1,000 customers on 6 aspects of the company's focal product and other competitors' products (10 brands in total)

Customer	Brand	Diversity	Speciality	Price	Delivery	Customer Care	Communicat ion
1	1	Х	Х	х	Х	Х	Х
1	2	X	Х	X	x	Х	X
1	3	X	X	X	X	Х	x
1	4	X	Х	X	X	Х	x
1	5	X	X	X	X	Х	x
1	6	X	x	X	X	X	x
1	7	X	Х	X	X	Х	x
1	8	X	Х	X	X	Х	x
1	9	X	x	X	X	X	x
1	10	X	Х	X	X	X	x
2	1	Х	Х	х	x	Х	Х
2	2	X	Х	X	X	Х	x
2	3	Х	x	X	X	X	x
2	4	Х	x	X	X	X	x
2	5	Х	Х	X	Х	X	x
2	6	Х	Х	X	Х	X	x
2	7	х	X	x	Х	X	Х
2	8	X	X	x	Х	X	X
2	9	X	X	x	Х	X	X
2	10	X	Х	X	Х	Х	Х
3	1	X	Х	x	Х	Х	Х
3	2	X	X	x	X	X	X
	•••						

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## Factor Analysis: Intuition



How would you summarize such rating data and draw a perceptual map? The average of all the customers' ratings on the 6 aspects of the 10 brands are calculated.

Brand	Diversity	Speciality	Price	Delivery	Customer Care	Communication
1	90	85	88	56	60	49
2	23	33	31	90	78	79
3	65	59	49	60	58	55
4	88	92	94	90	93	98
5	22	19	33	45	34	12
6	82	78	79	85	79	90
7	96	95	94	45	67	53
8	43	36	45	89	93	87
9	66	76	45	34	44	40
10	70	80	75	60	65	70



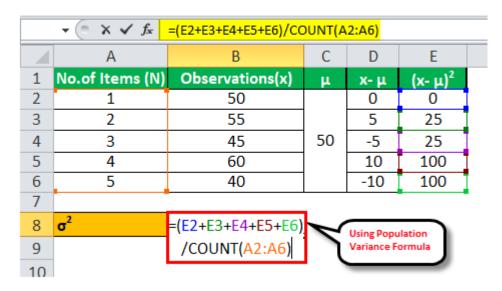




## **Variance Calculation**



$$\sigma^2 = \frac{\sum_{i=1}^n (xi - \mu)^2}{N}$$



$$\sigma^2 = 250/5$$







## **Factor Analysis**

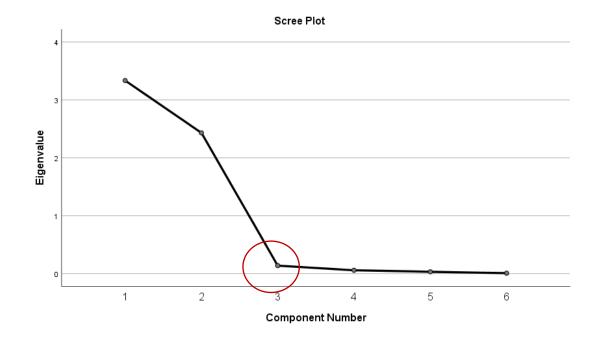


## How many factors/components?

Total Variance Explained							
Component	Eigenvalues	% of Variance	Cumulative %				
1	3.335	55.581	55.581				
2	2.431	40.521	96.102				
3	0.137	2.288	98.390				
4	0.057	0.956	99.346				
5	0.032	0.531	99.877				
6	0.007	0.123	100.000				

#### Criteria:

- Eigenvalue > 1
- Elbow from the scree plot
- Evaluate the percent of variance explained









## Factor Analysis: Intuition



- A factor is a group of variables ...
  - measuring something in common
  - highly correlated
- Two "latent" factors seems to be able to capture majority of the variance of the six attributes:
  - Factor 1 is highly associated with product diversity, specialty, and price; it can be interpreted as the "product" factor.
  - Factor 2 is highly associated with delivery service and customer service; and is labeled the "service" factor.

			Factor 1			Factor 2	
						λ	
		Diversity	Speciality	Price	Delivery	Customer Care	Communication
•	Diversity	1.00					
ions	Speciality	0.97	1.00				
Correlations	Price	0.92	0.89	1.00			
Corr	Delivery	-0.14	-0.16	0.04	1.00		
	Customer care	0.21	0.19	0.34	0.88	1.00	
	Communication	0.25	0.27	0.34	0.87	0.95	1.00







## **Pearson Correlation Coefficient**



#### Formula

$$r = rac{\sum \left(x_i - ar{x}
ight)\left(y_i - ar{y}
ight)}{\sqrt{\sum \left(x_i - ar{x}
ight)^2 \sum \left(y_i - ar{y}
ight)^2}}$$

r = correlation coefficient

 $oldsymbol{x_i}$  = values of the x-variable in a sample

 $ar{oldsymbol{x}}$  = mean of the values of the x-variable

 $y_i$  = values of the y-variable in a sample

 $ar{m{y}}$  = mean of the values of the y-variable







## Factor analysis



<u>Factor loading</u> is the correlation between a variable (an attribute) and a factor.

Factor Loading Matrix

	Product performance	Service performance
Variable	Factor 1	Factor 2
Diversity	0.99	0.03
Speciality	0.98	0.02
Price	0.95	0.18
Delivery	-0.17	0.97
Customer care	0.18	0.97
Communication	0.22	0.96

High factor loading Low cross-factor loading

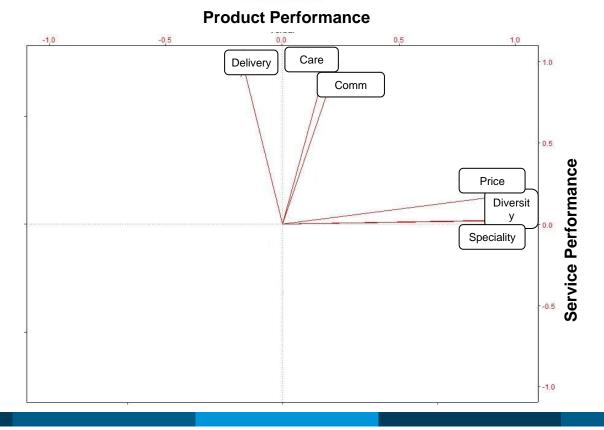
Note: this is the matrix after rotation







A summary map of the variables on the two factors (using factor loading)









## Factor analysis



Factor score is a weighted sum of the variables. The weights are regression-like coefficients which are output by the computer program. **Product** Service

							Score	Score
Brand	Diversity	Speciality	Price	Delivery	Care	Comm.	Factor 1	Factor 2
1	90	85	88	56	60	49	87.67	55.00
2	23	33	31	90	78	79	29.00	82.33
3	65	59	49	60	58	55	57.67	57.67
4	88	92	94	90	93	98	91.33	93.67
5	22	19	33	45	34	12	24.67	30.33
6	82	78	79	85	79	90	79.67	84.67
7	96	95	94	45	67	53	95.00	55.00
8	43	36	45	89	93	87	41.33	89.67
9	66	76	45	34	44	40	62.33	39.33
10	70	80	75	60	65	70	75.00	65.00

Factor1 = 1/3 Div +1/3 Spec + 1/3 Price + 0 Deli + 0 Care + 0 Comm

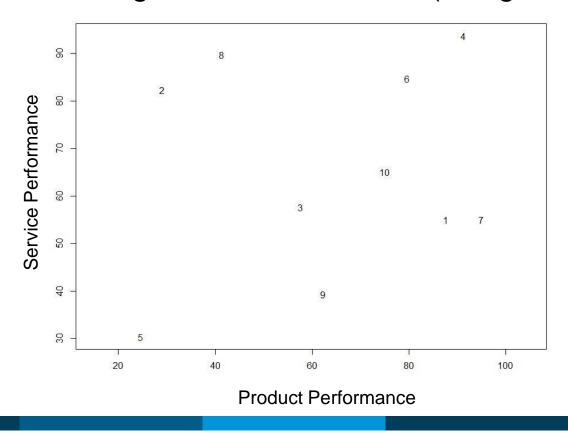
Factor2 = 0 Div + 0 Spec + 0 Price + 1/3 Deli + 1/3 Care + 1/3 Comm







A summary map of the ratings on the two factors (using factor score)



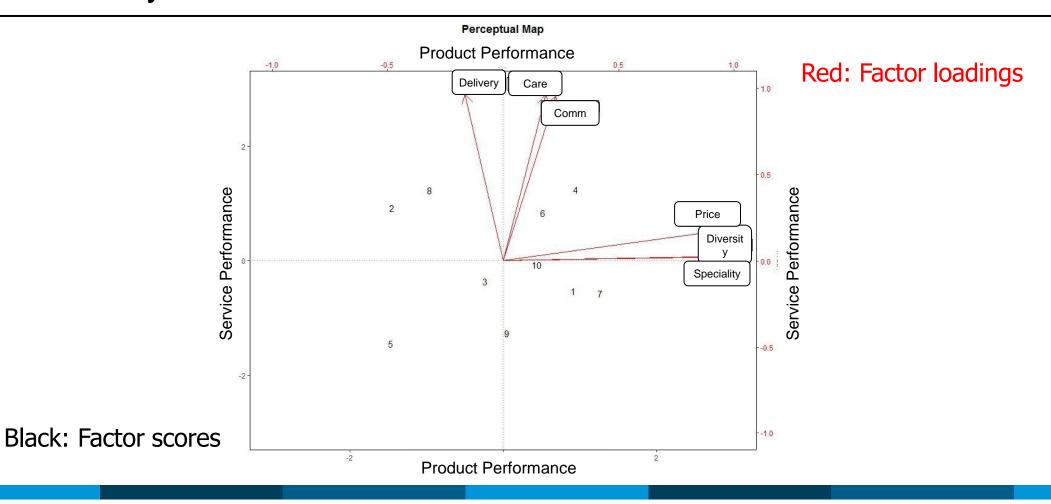






## Factor analysis











## Factor analysis



#### Summary of the steps

- We begin with a large number of measured variables of customer survey measures.
- Factor analysis synthesizes a large number of measured variables into smaller sets (e.g., 3–4) of latent "factors" that capture the essence of the meaning in the larger number of variables.
- To choose the total number of factors to retain, we observe how many factors have an <u>Eigenvalue</u> greater than 1.
- The strength of the association between a measured variable and its factor is called "factor loading." We
  categorize the measured variable with a factor with the highest loading (e.g., if a measured variable has factor
  loadings of 0.01 and 0.8 with Factors 1 and 2, we would associate the measured variable with Factor 2).
- Finally, we interpret each latent factor by surmising the conceptual commonality underlying the measured variables' loading on the factor.







## Other issues: standardisation



#### Standardized data

Brand	Diversity	Speciality	Price	Delivery	Care	Comm.	Factor 1	Factor 2
1	0.95	0.73	0.98	-0.44	-0.37	-0.54	0.89	-0.45
2	-1.54	-1.20	-1.28	1.15	0.56	0.59	-1.34	0.77
3	0.02	-0.23	-0.57	-0.25	-0.47	-0.31	-0.26	-0.35
4	0.87	0.99	1.22	1.15	1.34	1.31	1.03	1.27
5	-1.58	-1.71	-1.20	-0.95	-1.71	-1.i94	-1.50	-1.53
6	0.65	0.47	0.62	0.92	0.62	1.01	0.58	0.85
7	1.17	1.10	1.22	-0.95	-0.01	-0.39	1.16	-0.45
8	-0.80	-1.09	0.73	1.10	1.34	0.89	-0.87	1.11
9	0.06	0.40	-0.73	-1.47	-1.19	-0.88	-0.09	-1.18
10	0.20	0.54	0.46	-0.25	-0.11	0.25	0.40	-0.04









- Factor Analysis is useful for data reduction (reducing the number of variables)
- Decide on the number of factors to be retained
  - Use eigenvalue criterion and own judgment
  - Evaluate the percent of variance explained
- Use the rotated factor loadings to interpret the factor structure
  - Loadings should be greater than 0.5
- Save factor scores for subsequent analyses







## You should be able to ...



- Understand that perception influences positioning and thus perceptual maps are useful for positioning.
- Explain the purpose of Factor Analysis and how it works
- Draw conclusions regarding the positioning of brands based on perceptual mappings







## Workshop





 The manager of Intelligentsia Coffee collected consumers' ratings of 10 coffee brands ("a" to "j") on 9 attributes from N= 100 respondents. They are interested in any insights about how consumers see these brands.

Data file: Data\_Factor\_Analysis.csv

Handout: Workshop3\_Factor\_Analysis.pdf

Perceptual attributes	Survey question 1 = completely disagree, 10 = completely agree
Perform	Brand has strong performance
Leader	Brand is a leader in the field
Latest	Brand has the latest products
Fun	Brand is fun
Serious	Brand is serious
Bargain	Brand products are a bargain
Value	Brand products are a good value
Trendy	Brand is trendy
Rebuy	I would buy from Brand again





