Know Thy Customer: Predicting How Preference Translates into Choice

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A Quick Survey (In the spirit of knowing my customer)

• How many of you have done the following:

- Conducted an Economic Value (to the Customer) Analysis?
- Conducted a conjoint or tradeoff analysis?
- Estimated attribute utilities and/or demand based on purchase data?
- None of the above?

• How many of you:

- Have used utility estimates to build a choice or demand model?
- Have used preprogrammed demand or choice models, such as Sawtooth?



- Improve your customer choice models so that they correspond more closely to the underlying choice process and real market data
- Enable improved estimates of customer response to price changes thus improving pricing decisions



Outline

- Brief review of preference measurement and traditional demand modeling
- Challenges in fitting predicted preferences to real world data
 - Accounting for the impact of marketing-related activities in choices
 - Accounting for individual customer choice decision processes
 - Accounting for group customer choice decision processes



 If you consider customer value when you set price, you will use a model of demand

Types of demand models

- *Implicit*: Using managerial judgment
- Empirical: Generic Approaches

Aggregation of individual choices

- Estimate preferences of individuals
- Predict individual's choices and how they vary with price and other product design variations
- Aggregate across individuals
- Estimate from aggregate choices
 - Estimate how aggregate sales vary over time or across markets as a function of price, marketing activities and product features



Approaches to Preference Measurement

Economic Value to the Customer Analysis

- Reference/Replacement Value: The present value of the incremental cost of the option to be replaced
- Differentiation Value: The present value of differences in cost or revenue implications of the alternative to be valued
 - Cost of use (e.g. efficiency, operating speed)
 - Cost of maintenance
 - Set-up costs (installation and training)
 - Risk (likelihood of failure and resulting cost)
 - Performance quality of output
 - Adding value to customer's customer



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Approaches to Preference Measurement

Tradeoff analysis (e.g. Conjoint)

- A survey driven approach
- Determine relevant attributes and attribute values
- Elicit preferences based on hypothetical product profiles





CONJOINT TASK ILLUSTRATION



Approaches to Preference Measurement

Tradeoff analysis (e.g. Conjoint)

Sample Output: Part-Worth Utility Estimates CONJOINT OUTPUT ILLUSTRATION Part-Worth Utilities Part Worths 4 Part Worths 4 (3.5)3 3 (2.7)2 2 (1.5)1 1 (.5)Goodyear Michelin Generic 50.000 60.000 80.000 TREAD MILEAGE BRAND Part Worths Part Worths 4 4 3 3 (2.2)2 2 (1.7)(1.5)1 1 120 100 140 Yes No PUNCTURE SEALING PRICE

The utility of an alternative Is given by the sum of the part-worth utilities associated with its attributes



Basic Choice Modeling

Strict Utility Maximization

- Customers choose the most preferred alternative with certainty

Expected Utility Maximization

- Customers make random errors in estimating the utility of alternatives



- Where β is an estimated parameter reflecting choice or model accuracy (ideally estimated based on past choices)



Challenges

• The reality check:





What could have gone wrong?



Challenges

- What could have gone wrong?
 - Preference model misspecification



- A failure to account for marketing related variables
- A failure to account for decision processes



- Problem: Customer perceptions may not correspond to what you think is reality
- Why?
 - May not believe that certain brands will have certain characteristics
 - Past experiences may not be representative
 - May not have used the product before so don't know about features or benefits
 - May not know how to use a feature so don't appreciate benefits
 - May have had bad (or good) luck in the past



- May not have been exposed to information about a product's features or benefits or that information may not have been processed
 - Some benefits may be taken for granted or ignored
 - Common in economic value to the customer analyses
- You may be mistaken



- Problem: Customer perceptions may not correspond to what you think is reality
- Pricing implications
 - People make choices based on perceptions rather than reality
 - This may work for or against you
 - Price should be set to capture the desired share of customer's perceived value rather than your perceived value
 - Perceived value may be adjusted by marketing communications
 - Be careful about charging a price to take advantage of mistaken positive beliefs if there is a good chance that those beliefs will be corrected
 - Customers should be asked about their perceptions of different product alternatives and brand choice predictions should be based on perceptions as opposed to an "objective" measure



- Problem: Customers evaluate product offerings in a context
- The attractiveness of a price will depend on
 - How it compares to a reference price based on
 - Past experience
 - The prices of other alternatives
 - A list price
 - Prices that are higher than the reference price are treated less favorably than the same price would be in the absence of a reference price
 - How the price is framed
 - A discounted price versus a lower regular price Some product categories have "reference discounts"
 - Fixed price versus a price based on the level of use, etc.
 - A statistical analysis of historical choice data is helpful to measure these effects



- Problem: The product or brand may not be in the choice set
- Why?
 - Customers may not be aware of the product
 - Their chosen vendor may not carry the product or display it effectively
 - It may be incompatible with technology they currently use
- Pricing implications
 - Price changes won't influence sales among these customers unless doing so brings the product or brand into the choice set
 - Non-price marketing efforts may need to be required to get the product considered



- Traditional choice models assume that decision-makers process all available data and they will tend to choose the alternative that yields the highest utility (value)
 - This may imply a lot of data processing on the part of the customer when making decisions, especially if there are many alternatives with many features
- Problem: Decision-makers will often take steps to simplify the decision-making process



- Example:
 - Consider a restaurant that servers hamburgers and hot dogs
 - Assume price is set so that the customer is indifferent between them
 - What is the probability of the customer's choosing each alternative?
 - What happens if a cheeseburger is added to the menu?
 - Assume the price of the cheeseburger is just high enough to leave the consumer indifferent between all three alternatives
 - What is the probability of the customer's choosing each alternative?







Nested Decision-Making

- Customers will break alternatives down into categories
 - They may eliminate some categories
 - They will choose among acceptable categories
 - When a category is chosen they will choose among alternatives within the category



Nested Decision-Making

- Examples:
 - Decide on brand first and then which alternative to choose within the brand (traditional car shopping model)
 - Decide on product form first and then brand (beverages: bottles vs. cans, diet vs. non-diet, cola vs. non-cola, etc.)
 - Decide on vendor first and then choose an alternative offered by that vendor
 - Vendors may have different product offerings (implying different choice sets), different pricing, and different promotional activity
 - Vendor choice may or may not be influenced by the product category being considered



Predicting Choices with Nested Decision-Making

• Example



Procedure:

- 1. Estimate choice probabilities within each category based on product utility e.g. Pr(Hamburger|Burgers), Pr(Cheeseburger|Burgers)
- 2. Estimate the expected utility of the category

e.g. EU(Burgers) =

Pr(Hamburger|Burgers)xU(Hamburger)+ Pr(Cheeseburger|Burgers)xU(Cheeseburger)



Predicting Choices with Nested Decision-Making

Example



3. Estimate category choice probabilities based on the expected utility of the category



Predicting Choices with Nested Decision-Making

• Example



4. Estimate product choice probabilities:e.g. Pr(Hamburger) = Pr(Hamburger|Burgers)xPr(Burgers)



Nested Decision-Making

- Deciding which nested structure is appropriate
 - Ask customers about their decision-making process
 - Try different structures and see what fits the best

• Pricing implications:

- Price response estimates may be biased if they don't separately account for:
 - The impact of price on category choice
 - The impact of price on band choice within the category



Eliminating "Unacceptable" Alternatives

- Customers may have a set of "deal-breaking" criteria and will eliminate any alternatives that don't satisfy those criteria regardless on how they perform on other attributes
- This effectively eliminates them from the choice set even with awareness

• Examples:

- The customer is purchasing under a strict budget and prices above that level won't be acceptable regardless of quality
- An imaging device that doesn't have improved resolution over may not be considered



Eliminating "Unacceptable" Alternatives

- Dealing with the issue:
 - Survey customers to find out what the deal breakers are and eliminate alternatives that don't satisfy these criteria from their choice sets
 - Note: different customers may have different deal breakers so they may have different choice sets
 - To the extent possible eliminate the unacceptable from choice tasks used to estimate utility
 - It may be desirable to give different customers different conjoint tasks
 - New adaptive conjoint methodologies allow the exclusion of "deal breaking" attribute values



Simplifying the evaluation of alternatives

- Customers may not consider all attributes
 - Products often have many relevant attributes but customers often only consider a few of them
 - Economic value to the customer and conjoint analyses will often overestimate the importance of less-important attributes

Dealing with the issue:

- Omit "less important" attributes when computing estimated utility or value of the different alternatives in the choice set
- Test to see if doing so improves fit with real world data
- Note: Customers may vary in the attributes they consider to be important



- Problem: There may be multiple parties who influence decisions
- Different parties may place different values on different features
- If you are estimating the value of only one party, your choice predictions will be biased
 - Example: Medical equipment
 - Physician / Practitioner will value reliability, quality and ease of use
 - Administrators will be more interested in cost and the value of the equipment in driving business



Group Decision Processes

- **Dealing with the issue:** ۲
 - Estimate a joint utility function
 - Have the relevant parties take a conjoint (or similar) survey jointly
 - Have the relevant parties take a conjoint (or similar) survey independently - Estimated utility can be approximated by a weighted average of the respondents
 - Treat the decision as a sequential process (say with two decision makers)
 - View one party as reducing the choice set by eliminating unacceptable alternatives
 - The second party can be viewed as choosing among the alternatives in the reduced choice set.
 - Example: Medical Equipment:
 - Administrator sets maximum cost or minimum economic value
 - Practitioner chooses from alternatives that satisfy the administrator's criteria





- Traditional demand modeling methodologies may provide models that have substantial biases that may give misleading predictions about customer responses to price
- Things that should be considered when building demand models:
 - The impact of marketing-related activities in choices
 - Accounting for individual customer choice decision processes
 - Accounting for group customer choice decision processes



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