

Course: Customer Analytics

# CHOICE ANALYSIS

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# AGENDA

- I. Discuss the objective and design of choice analysis
- II. Run the application in R and interpret
- III. Hand-on application with new data

# OBJECTIVE

- To assess consumer preferences
- To understand **how features** (e.g., package size, brand, flavor, price, privacy, ...) affect which product a customer will choose
- To understand how they trade off desirable features against each other
- The outcome (DV) is not a number, but **a product choice** among several options

# SCENARIO FRAMING

- **Setting:** A company designs a new line of minivans with different characteristics
- **Research problem:** How large the minivan should be and what type of engine it should have?
- **Objective:** The experiment/survey aims to identify the factors most customers desire. For this experiment, the company selects five main characteristics (factors or numeric):
  - Carpool
  - Seat
  - Cargo
  - Engine
  - Price

# FULL FACTORIAL DESIGN

## Factors:

1. Carpool
2. Seat
3. Cargo
4. Engine
5. Price

## Levels:

yes, no  
6,7,8  
2ft; 3ft  
gas, hyb, elec  
30, 35, 40

## Max product profiles

$$2*3*2*3*3 = 108$$

# SURVEY QUESTION

Each respondent answers several questions, where options have varying attributes.

**Which of the following minivans would you buy?**

Assume all three minivans are identical other than the features listed below.

	Option 1	Option 2	Option 3
	6 passengers 2 ft. cargo area gas engine \$35,000	8 passengers 3 ft. cargo area hybrid engine \$30,000	6 passengers 3 ft. cargo area gas engine \$30,000
I prefer (check one):	<input type="checkbox"/>	<input type="checkbox"/>	<input checked="" type="checkbox"/>

**Fig. 13.1** An example choice-based conjoint survey question

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**ALTERNATIVES** (points to the header row)

**ATTRIBUTES** (points to the feature descriptions in Option 3)

**LEVEL OF ATTRIBUTE** (points to the price in Option 3)

**Fig. 13.1** An example choice-based conjoint survey question

# SEVERAL QUESTIONS PER CUSTOMER

- Each question has the same structure, but varies the levels of the attributes for the alternatives
- In this study, each respondent sees 15 questions. ( $15 \times 3 = 45$  product profiles out of 108 possible)
- A typical study might have 5–10 attributes and include 10–20 questions for each respondent
- The number of product profiles to choose from can also vary



# FULL VS. PARTIAL FACTORIAL DESIGN

- **FULL** = Tests every possible combination of factors and levels
  - Comprehensive: Captures all possible interactions between factors.
  - Clarity: Provides clear and detailed information about the effects of each factor and their interactions.
  - Resource-Intensive: Requires a large number of experiments as the number of factors increases.
  - Time-consuming: More experimental runs mean more time needed for completion.
- **PARTIAL / FRACTIONAL** = Tests only a subset of the combinations of factors and levels
  - Efficient: Requires fewer experiments, saving time and resources
  - Practical: Ideal for preliminary studies or when resources are limited
  - May miss some interactions between factors.
  - The choice of which combinations to test must be strategic to ensure relevant data is collected.

# METHODS OF ANALYSIS

- Choice-based conjoint analysis
  - Individuals choose among products => DV is categorical
  - Multinomial logit model (focus in this lecture)
- Rating-based (metric) conjoint analysis
  - Individuals give ratings to single products => DV is metric
  - Linear model

# DATA STRUCTURE

- Depending on the software
- In R, long format

	resp.id	ques	alt	carpool	seat	cargo	eng	price	choice
1	1	1	1	yes	6	2ft	gas	35	0
2	1	1	1	yes	8	3ft	hyb	30	0
3	1	1	1	yes	6	3ft	gas	30	1
4	1	2	1	yes	6	2ft	gas	30	0
5	1	2	2	yes	7	3ft	gas	35	1
6	1	2	3	yes	6	2ft	elec	35	0
7	1	3	1	yes	8	3ft	gas	35	1
8	1	3	2	yes	7	3ft	elec	30	0
9	1	3	3	yes	8	2ft	elec	40	0
10	1	4	1	yes	7	3ft	elec	40	1
11	1	4	2	yes	8	3ft	gas	40	0
12	1	4	3	yes	6	2ft	hyb	30	0
13	1	5	1	yes	6	3ft	elec	30	1
14	1	5	2	yes	6	3ft	elec	40	0
15	1	5	3	yes	6	3ft	hyb	40	0
16	1	6	1	yes	6	2ft	elec	40	0
17	1	6	2	yes	7	2ft	gas	35	1
18	1	6	3	yes	7	2ft	hyb	35	0
19	1	7	1	yes	6	3ft	hyb	35	1
20	1	7	2	yes	7	3ft	gas	40	0
21	1	7	3	yes	6	2ft	elec	30	0
22	1	8	1	yes	6	3ft	elec	35	0
23	1	8	2	yes	8	3ft	hyb	40	0
24	1	8	3	yes	7	3ft	hyb	30	1

# STEPS OF ANALYSIS IN R

1. Ensure correct data structure
2. Perform some descriptive statistics, e.g.
  - *Summary()* - How many times each level appear in the questions
  - *Xtabs()* - the number of times respondents chose an alternative at each feature level.
3. Perform multinomial logit model with “mlogit” package
  - It requires transforming data in a special format with *logit.data()* function
  - The syntax for setting the model is standard

# RESULTS

- The estimated parameter (part worth coefficient) for each level, relative to the base levels of each attribute
- Customers disliked electric and hybrid engines (relative to the base level, which is gas) and disliked the \$40K and \$30K price (relative to the base level price of \$30)
- All parameter estimates are on the logit scale and typically range between -2 and 2

```
Call:
mlogit(formula = choice ~ 0 + seat + cargo + eng + price, data = cbc.mlogit,
        method = "nr")
```

```
Frequencies of alternatives:choice
```

```
      1      2      3
0.32700 0.33467 0.33833
```

```
nr method
```

```
5 iterations, 0h:0m:0s
```

```
g'(-H)^-1g = 7.84E-05
```

```
successive function values within tolerance limits
```

## Coefficients

	Estimate	Std. Error	z-value	Pr(> z )	
seat7	-0.535280	0.062360	-8.5837	< 2.2e-16	***
seat8	-0.305840	0.061129	-5.0032	5.638e-07	***
cargo3ft	0.477449	0.050888	9.3824	< 2.2e-16	***
enghyb	-0.811282	0.060130	-13.4921	< 2.2e-16	***
engelec	-1.530762	0.067456	-22.6926	< 2.2e-16	***
price35	-0.913656	0.060601	-15.0765	< 2.2e-16	***
price40	-1.725851	0.069631	-24.7856	< 2.2e-16	***

```
---
```

```
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
Log-Likelihood: -2581.6
```

# MAIN REFERENCES

Chapman C. & Feit E. M. (2015). *R for Marketing Research and Analytics*, Springer. Chapter 13.  
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Research Project:

[https://docs.google.com/forms/d/e/1FAIpQLSdBgLnkUTZliFAEx0FYHlcLT4zcz0vKf2gyvRJCmi2fL2R7qg/viewform?usp=sf\\_link](https://docs.google.com/forms/d/e/1FAIpQLSdBgLnkUTZliFAEx0FYHlcLT4zcz0vKf2gyvRJCmi2fL2R7qg/viewform?usp=sf_link)