

Introduction to Discrete Choice Experiments

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Including joint work with DARTH workgroup

Discrete Choice Experiments

- Discrete choice experiments are a stated preference method where subjects must choose between two or more options that are defined through a set of *attributes*.
- Each attribute can take values from a pre-defined set of *levels*.

Your doctor tells you that your current **risk of stroke is 4%** and **risk of bleed is 2%**. Suppose you have the following options:

	No treatment	Drug A	Drug B
Risk of stroke	 (4 out of 100)	 (2.4 out of 100)	 (0 out of 100)
Risk of bleed	 (2 out of 100)	 (2 out of 100)	 (3 out of 100)
Antidote available	N/A	<u>No</u> antidote	<u>Yes</u> antidote
Monthly blood test	N/A	<u>No</u> monthly blood tests	<u>Yes</u> monthly blood tests
Dose frequency	N/A	<u>Once</u> a day dosing	<u>Twice</u> a day dosing
Interactions likely	N/A	<u>No</u> interactions likely	<u>Yes</u> interactions likely
Cost per month	\$0 per month	<u>\$0</u> per month	<u>\$10</u> per month

Which would you choose? ☐ No treatment ☐ Drug A ☒ Drug B

From the remaining two options, which would you choose? ☐ No treatment ☒ Drug A ☐ Drug B

Discrete Choice Experiments

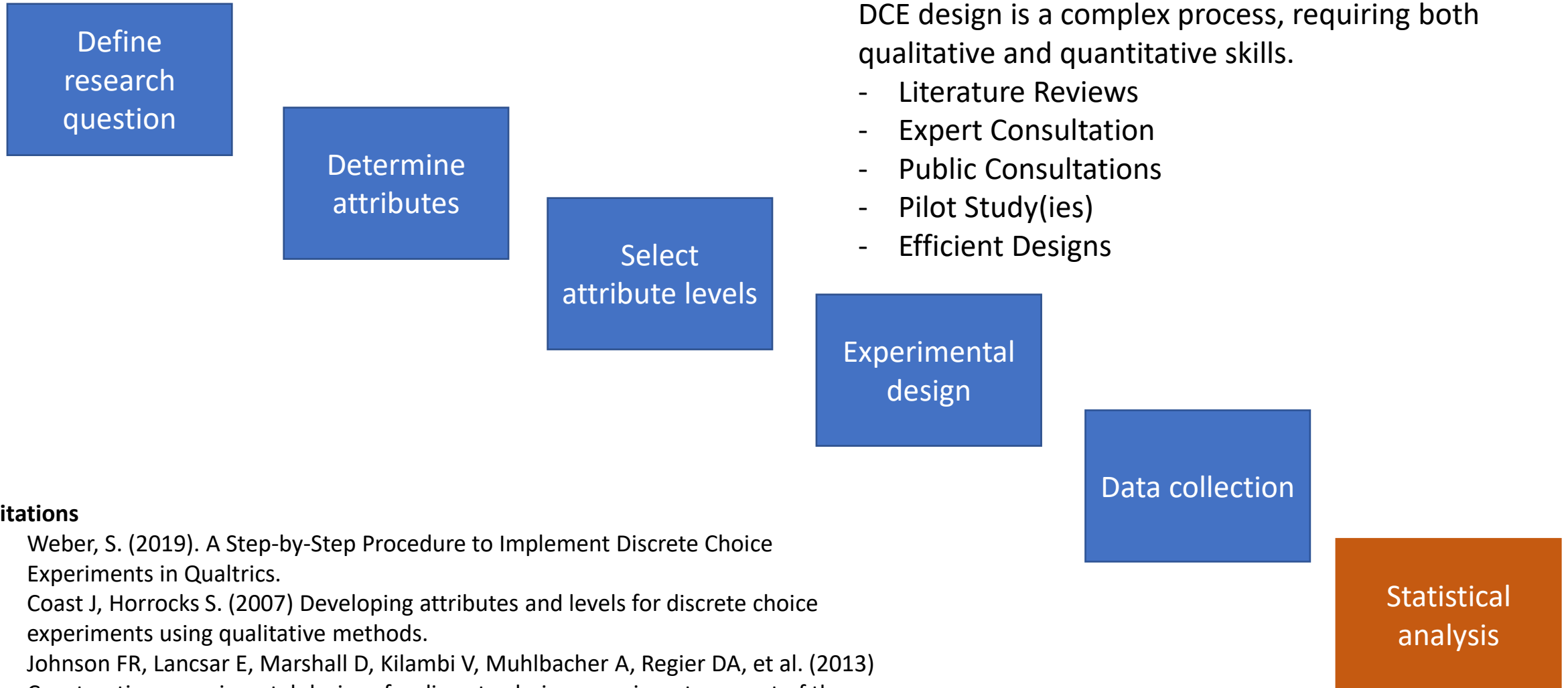
- Data from a DCE gives information about preferences.
- These data can be used to estimate utilities and willingness-to-pay

	No treatment	Drug A	Drug B
Underlying Utility	$U_{\text{No Treatment}}$	$U_{\text{Drug A}}$	$U_{\text{Drug B}}$
Revealed Ranking	3	1	2

$$U_{\text{Drug A}} \geq U_{\text{Drug B}} \geq U_{\text{No Treatment}}$$

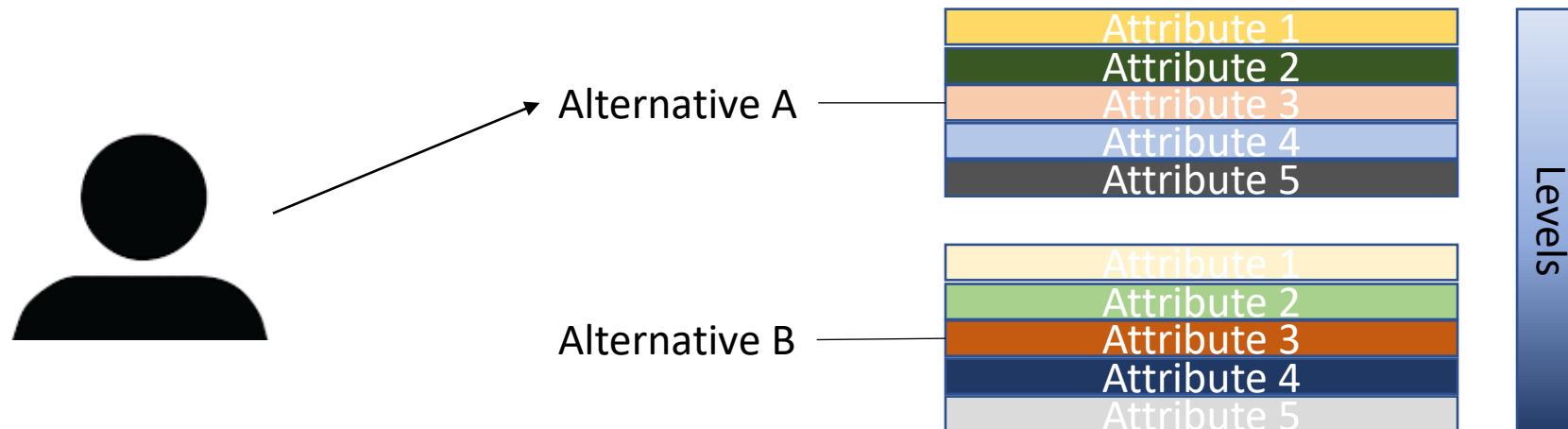
- Analysing the DCE data can estimate the utility conditional the given attributes.

Designing a Discrete Choice Experiment



Discrete Choice Experiments

- A DCE consists of asking I individuals to choose between J alternatives across S scenarios.
- A scenario is a single choice that the individual is required to make, where each of the J alternatives are defined in terms of a given set of attribute levels $A_{i,j,s}$, for $i = 1, \dots, I, j = 1, \dots, J, s = 1, \dots, S$.



Estimating Utilities

- The utility U for individual i , alternative j and scenario s is $U_{i,j,s}$
- A DCE gives data $y_{i,s}$, the choice for individual i in scenario s .
- An individual will choose option j if it has the highest utility

$$P_{i,j,s} = \text{Prob}(U_{i,j,s} > U_{i,l,s}) = \text{Prob}(y_{i,s} = j) \\ \text{for all } l \neq j$$

- As we observe choices, the scale of the utility is unidentifiable.

$$\text{Prob}(U_{i,j,s} > U_{i,l,s}) = \text{Prob}(aU_{i,j,s} + b > aU_{i,l,s} + b)$$

Multinomial Logistic Regression

- The impact of the attributes on the utility can be estimated:

$$U_{i,j,s} = \alpha_j + \beta A_{i,j,s} + \gamma_j X_i + \epsilon_{i,j,s}$$

where X_i are individual-specific covariates and $\epsilon_{i,j,s}$ is an error term.

- If $\epsilon_{i,j,s}$ is assumed to be i.i.d Gumbel distribution with shape parameter 1, then α_j , β and γ_j are estimated using *logistic regression*.
- Individual-specific slopes $\beta_{i,j}$ can be specified.
- Other distributions for the error functions require different models, i.e., normally distributed errors require probit regression.

Interpreting the Results

- The β coefficients estimate the utility weights for each attribute
- The willingness to pay is the ratio of two coefficients:

For two attributes *price* and *wait time*, $\frac{\beta_{wait}}{\beta_{price}}$ is the willingness to pay to be seen a minute sooner.

- The “true” utility is unidentifiable but external information can be used to *anchor* U to a given scale:

For EQ5D, the utility of state 11111 is set to 1 with another state estimated by Time Trade Off or Standard Gamble.

- A DCE that include price or time naturally gives an external anchor.

Data: Wide Form

ID	Choice	Attributes of A					Attributes of B				
1	A										
1	B										
1	A										
2	A										
⋮	⋮	⋮					⋮				

Data: Long Form

ID	Choice	Alternative	Attributes				
1	TRUE	A					
1	FALSE	B					
1	FALSE	A					
1	TRUE	B					
1	TRUE	A					
1	FALSE	B					
2	TRUE	A					
2	FALSE	B					
⋮	⋮	⋮	⋮				

Discrete Choice Experiment in R

- There are several packages to analyse DCEs in R.
- We will use the mlogit package which requires *long form* data
- The apollo package is more flexible but has its own data storage method - <http://www.apollochoicemodelling.com/>
- The Rchoice package is another alternative.
- Packages for generalised linear models are also suitable, e.g., glm, glmmTMB.

Citations

- Emily Lancsar, Denzil G. Fiebig, Arne Risa Hole (2017) Discrete Choice Experiments: A Guide to Model Specification, Estimation and Software, *PharmacoEconomics*, 35:697–716
- Ian Waudby-Smith, A. Simon Pickard, Feng Xie, Eleanor M. Pullenayegum (2020) Using Both Time Tradeoff and Discrete Choice Experiments in Valuing the EQ-5D: Impact of Model Misspecification on Value Sets, *Medical Decision Making*, 40(4):483-497