

# **B.** Choice Experiments

- DCE is an attribute-based survey method for exploring preferences, measuring benefits (utility) and assessing economic values.
- Respondents are presented with hypothetical scenarios (<u>choice sets</u>). The choice sets comprise two or more alternatives, which vary along several characteristics or attributes of interest and individuals are asked to choose one alternative.
- Most commonly, each respondent faces several choice questions within a single survey, drawn from statistical design principles.
- DCE originates in the market research and transport literature and more recently has been applied to environment and risk valuation.





		Option A	Option B	Current situation	
	Flood frequency	Once every 25 years	Once every 50 years	Once every 5 years	
Example of Choice Scenario (Brouwer et al., 2016)	Water quality	Good	Very good	Moderate	
	Increase water bill	€3 / year	€10/year	€0	
	I prefer:	Option A	D Option B	D Neither	
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# Comparing formats: Choice Experiment vs Contingent Valuation

Question X: Do you support the proposal to protect the environment that will ensure:

- an increase in the number of endangered species present from 5 to 10
- an increase in the area of healthy native vegetation from 1500 ha to 1800 ha
- an increase in the number of visitors from 2000 pa to 3000 pa

to be funded by a one-off levy of \$20 on your income tax, or do you oppose it?

Please circle the option that most closely represents your view:

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I support the proposal with a $20 levy ...... 1
I oppose the proposal and the $20 levy ...... 2
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Figure 3.1 Typical contingent valuation method question

Question Y: Consider carefully each of the following three options. Suppose these options were the only ones available, which one would you choose?

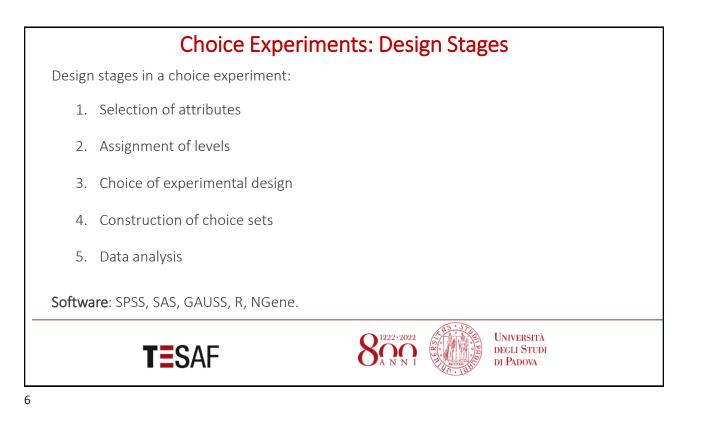
Alternative Attribute	'Status Quo' alternative	Proposed alternative 1	Proposed alternative 2
Number of endangered species	5	15	15
Hectares of healthy native vegetation	1500	1800	2100
Visitor days per annum	2000	3000	2000
Cost to you (S)	0	20	10
Please circle you	r preferred optio	n.	
I would choose the	ne status quo at ne	o cost to me	1
I would choose a	Iternative 1 at a \$	20 cost to me	2
I would choose a	Iternative 2 at a \$	10 cost to me	3

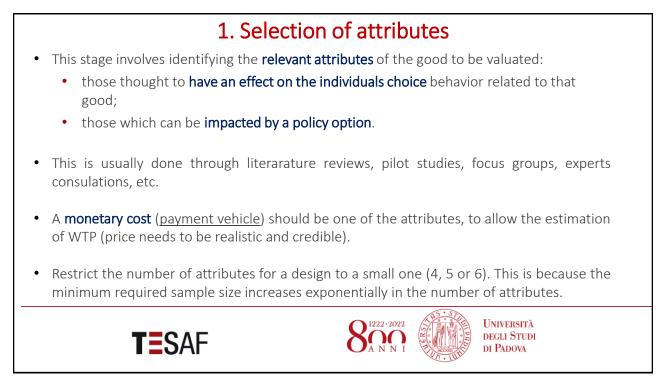
Figure 3.2 Choice modelling choice set





# Choice Experiments DCE can provide us with useful insights in risk valuation research: 1. which attributes are significant determinants of the values people place on risk mitigation options; 2. the implied ranking of these attributes among population; 3. the total economic value of each attribute and of the a good as a whole TEERF





# 2. Assignment of levels

- It involves the assignment of **realistic levels** for the attributes:
  - "bracket" the existing level of an attribute with higher and lower values;
  - maximum and minimum levels can be identified by scientists and policymakers;
  - the "do nothing" or status quo level should be included.
- Relevant price ranges could be estimated from pilot studies, literature and focus groups. They have to be commensurate with the levels of the attributes.
- Too low prices will always be accepted (small/zero price coefficient); too high prices will always be rejected (as above).





# 3. Choice of Experimental Design

- <u>Complete factorial design</u>: a factorial design combines every level of each attribute with every level of all other attributes. Each combination of attribute levels is an alternative.
- By considering all the possible combinations it would be possible to *generate n. levels to the power of n. attributes* alternatives.
- A problem of the full factorial design is that a **large number of alternatives are generated** as the numbers of attributes and levels are increased.
- Take as an example a choice experiment with 4 attributes and 3 levels.
- The possible number of alternative is **3^4 = 81**. It is difficult to ask respondents to consider 81 alternatives! Excessive cognitive burden is required.

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# Statistical experimental designs provide the means to select subsets of the total set of possible alternatives (e.g. 81 ⇒ 9) for use in an experiment (or questionnaire) in a statistically efficient manner: Fractional factorial design.

- Experimental designs were originally developed in the field of experimental science and agricultural research.
- The design has the property of **orthogonality**: each of the variables has zero correlation with any of the others. It means that the influence of changes in any of these attributes on the respondents' choice can be identified and measured.
- Possible to estimate **efficient designs** (according to different criteria, e.g. minimizing variance of estimated parameters)





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#### 4. Construction of choice sets

- You need to **package the alternatives** selected by the experiment design in **choice sets** in order to present them to respondents.
- Usually, the n. of choice sets produced by experimental design can still be too large for respondents to be able to cope with ⇒ split them into "blocks" and to offer for each respondent only one block.
- How many choice tasks the respondent can be asked to perform? It depends partially on the complexity of the issue itself.

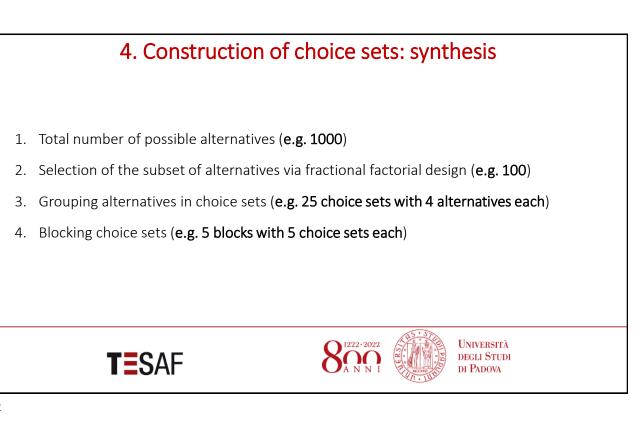
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• The fewer the n. of attributes and levels, the higher the number of choice sets for each person.







#### 5. Data Analysis

- CE has two "foot-holds" in economic theory:
  - 1. Lancaster's characteristics theory of value (Lancaster, 1966).
  - 2. Random Utility Maximization (RUM) (McFadden, 1973).
- Within RUM, an individual's preferences can be represented by the following utility function:

$$U = U(X_1, \dots, X_m)$$

• which depends on the levels of X = 1, ..., n attributes.





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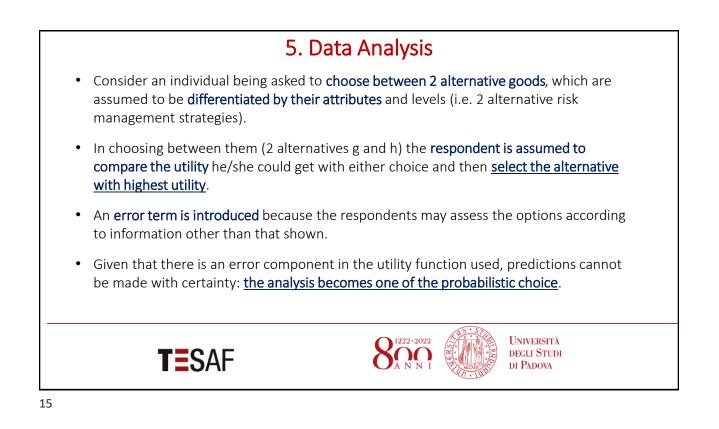
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# 5. Data Analysis

- Many of these elements X are **unobservable** to the researcher or are observable with an error.
- The conventional utility **function U(.)** can be broken down into two parts:
  - 1. one <u>deterministic</u> and <u>observable</u> V(.)
  - 2. an error term  $\varepsilon(.)$ :
- $U = U(X_1, ..., X_m) = V(X) + \varepsilon(X)$
- Advantage: it is a more realistic representation of preferences
- **Disadvantage**: some assumption have to be made about the nature of error component to make any predictions from this theory, since the error is not observable.







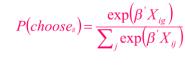
#### 5. Data Analysis

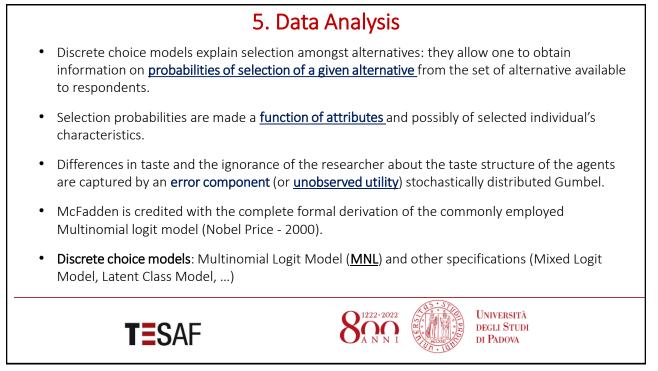
• The probability that a respondent *i* prefers option *g* in the choice set to any alternative option *h* can be expressed as the **probability that the utility associated** with option *g* exceeds that associated with all other options.

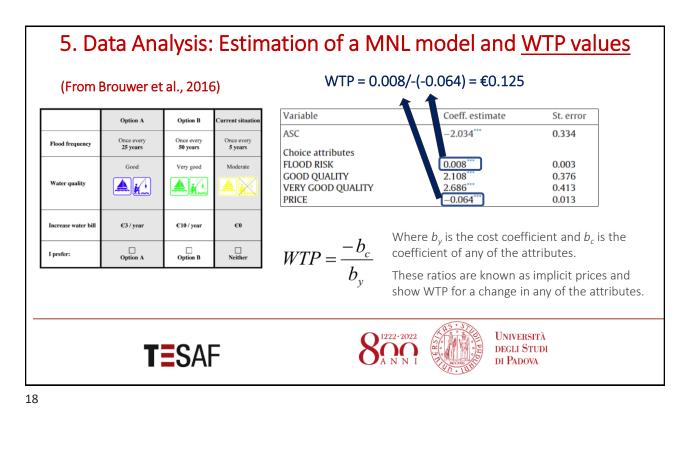
$$P(U_{ig} > U_{ih}) = \frac{\exp(\mu V_{ig})}{\sum_{j} \exp(\mu V_{ij})}$$

- **X** = vector of attributes
- $\beta$  = vector of estimated coefficient
- $\mu$  = scale parameter (usually is set to 1)
- If the dependent variable takes only two possible values (ex. A or B) then a **binary logit model** is required, if it takes 3 or more values (ex. A, B, neither) then a **multinomial logit** model is required.









# Advantages of Choice Experiments

- CE can be seen as a natural generalization of CV. A CV study (change or no change) cannot value the attributes of the change (or the attributes of the policy leading to the change). CE, because they look at more than two alternatives, provide a natural way to do it.
- CE do a better job than CV in measuring the value of changes in the characteristics of goods. This is a more useful focus from a management/policy perspective than focusing on either the gain or loss of the good.
- CE may avoid some of the response issues in CV: i.e. yea-saying.

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# **Disadvantages of Choice Experiments**

- In order to estimate the value of a good, as distinct from a change in one of its attributes, it is necessary to assume that the value of the whole is equal to the sum of the parts.
- But:
  - there may be additional attributes not included in the design which generate utility
  - is the value of the whole indeed linear additive?
- Value estimates of CE are sensitive to study design.
- Choice complexity can be a problem for the respondents.

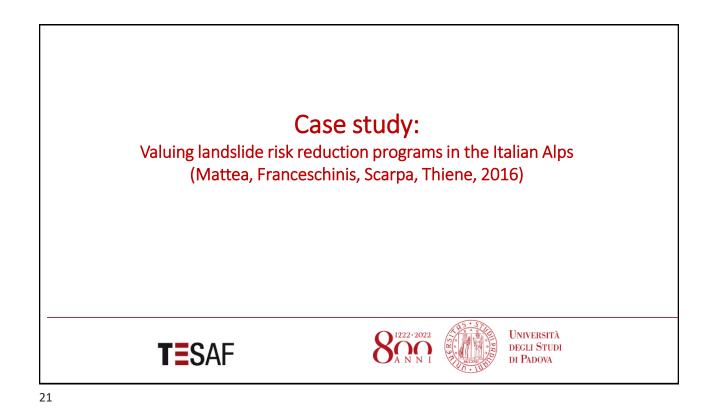


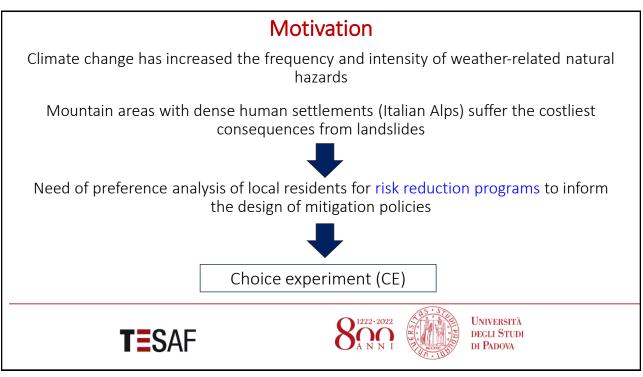


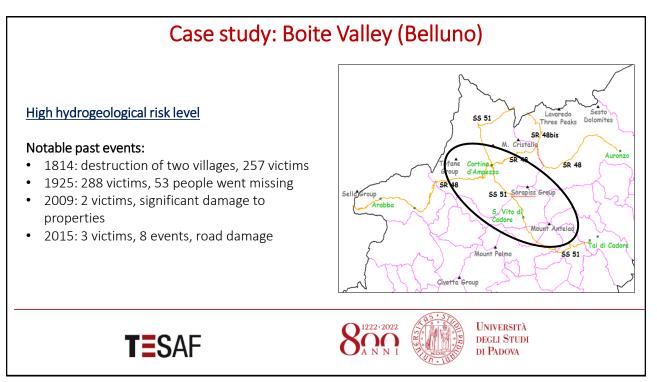
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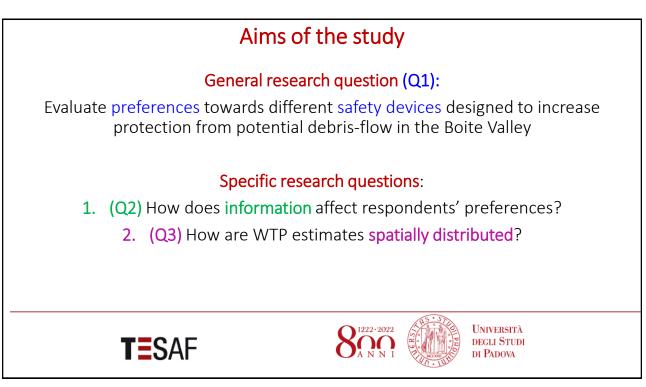


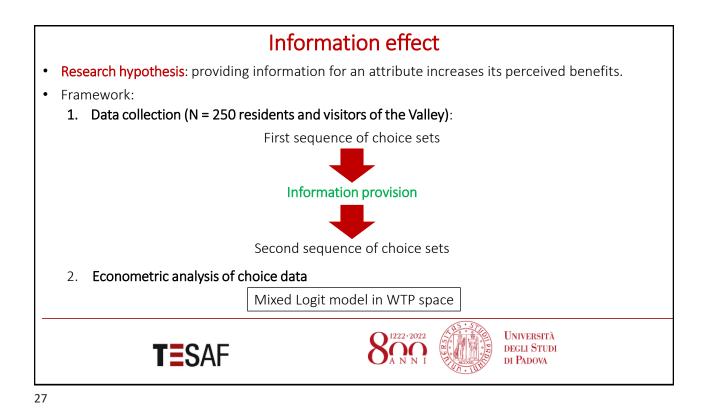


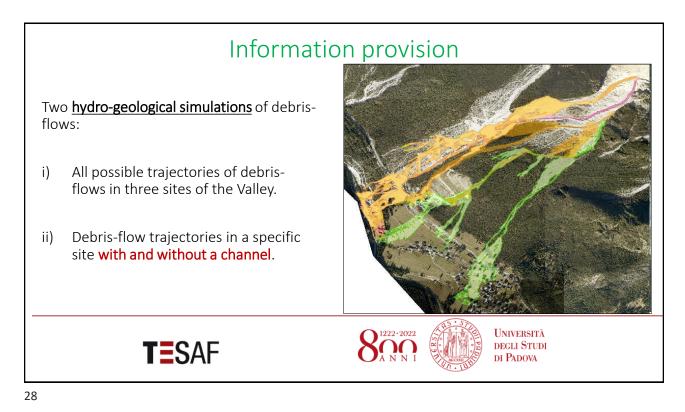


Attribute	Description	Levels
Channel	Diverging channel is a man-made channel built to redirect water.	1 if present 0 otherwise
Basin	Retaining basin is a dam where the solid and liquid mass is collected prior to damage roads and villages.	1 if present 0 otherwise
Video cameras	Video cameras monitor the debris-flow during the night and, in case of emergency, they will activate the alarm system and the traffic lights on the road.	1 if present 0 otherwise
Acoustic sensors	Acoustic sensors detect soil movement in slopes prior to landslides and provide an acoustic signal to give early warnings of landslide occurrence.	1 if present 0 otherwise
Road toll	A road toll to pay for eight months (from April to November of a specific year) daily for transit in the valley by car for residents and tourists.	€1 €2 €3 €4

Safety devices	A	В	С	D	E	F	Status quo
Channel	-	-	-	channel	channel	channel	-
Basin	-	basin	basin	basin	-	-	insuff. basin
Video cameras	video	-	video	-	-	video	_
Acoustic sensors	-	-	sensors	-	sensors	sensors	-
Road toll	€2	€1	€4	€2	€3	€3	€0
Your choice	0	$\bigcirc$	0	0	$\bigcirc$	0	0







	Results – Choice model estimates (Q1 & Q2)						
	Value	Std. Err.	<i>p</i> -value				
Mean parameters							
μ BAS	1.83	0.36	< 0.001	Perceived benefits			
μ CHAN	2.12	0.47	< 0.001	I EICEIVEU DEHEIILS			
μ SENS	1.26	0.21	< 0.001	for all devices			
μ VIDEO	1.19	0.29	< 0.001				
μ ASC_SQ	-1.98	0.97	< 0.001	(passive > active)			
$\mu \ln(\lambda)$	-2.05	1.12	< 0.001	( /			
Interaction paramete	rs						
Info $\times$ BAS	0.13	0.16	0.24	Desitive offect of			
Info $\times$ CHAN	0.42	0.20	< 0.001	Positive effect of			
Info $\times$ SENS	0.34	0.31	0.19	information on WTP			
$Info \times VIDEO$	0.08	0.14	0.56				
Info × TOLL	0.04	0.24	0.81	for channel			
Info $\times$ ASC_SQ	-0.15	0.09	0.03				
Standard deviation pa	arameters						
σ BAS	1.21	0.35	< 0.001				
σ CHAN	1.36	0.38	< 0.001	High variation of			
σ SENS	0.99	0.41	< 0.001	nrafaranaaa			
σ VIDEO	1.01	0.58	< 0.001	preferences			
σ ASC_SQ	0.87	0.63	< 0.001	towards devices			
$\sigma \ln(\lambda)$	1.81	0.95	< 0.001	LOWALUS UEVICES			
Log-likelihood	-2402.88						
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