## Do numerical probabilities promote informed stated preference responses under inherent uncertainty?

### Insight from a coastal adaptation choice experiment

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## Stated preference methods

- Provide estimates of economic value of non-market goods (e.g., clean air)
- Help determine the value of a good to society
  - Estimates of benefits for benefit-cost analyses
  - The value of losses from environmental damages (e.g., loss of recreation opportunities after oil spill)
- Wide range of applications: transportation, health, environment, culture, etc.
- Value estimates derived from preferences stated in surveys
  - Typically large survey studies on representative samples of respondents
  - Preferences are often elicited through discrete choice experiments





(1) Uncertainty in scientific models and predictions

(2) Uncertainty in the effectiveness of policy interventions

(3) Inherent ur	5		
(CHOOSE ONLY ONE) I vote for	I vote for NO NEW ACTION	I vote for PROTECTION OPTION A	I vote for PROTECTION OPTION B

### Inherent outcome uncertainty (tied to ecological systems)

- Uncertainty that is invariant across policy scenarios
- Example: The effect of installing new coastal flood defenses depends on a probability of severe storms that is fixed in the study area
- Very little attention in the stated preference literature
- Most surveys provide no formal communication of inherent uncertainty
- Often (unstated) assumptions that scenario outcomes are certain, that presented attribute levels reflect expected values, etc.
- These assumptions can have important implications for the interpretation and validity of value estimates
- <u>To our knowledge, there have been no tests of **inherent** uncertainty communication formats for stated preference studies</u>

Our research question:

Do numerical probabilities help respondents make more informed choices in stated preference surveys under inherent uncertainty?

## Data – discrete choice experiment

- Policy scenario: coastal flood adaptation to protect homes and natural systems such as beaches and wetlands from flooding and erosion
- In Old Saybrook, Connecticut, USA
- The survey distributed via mail
- May July 2014
- 282 complete surveys returned

**PROTECTION OPTION A** and **PROTECTION OPTION B** are possible protection options for . Old Saybrook. **NO NEW ACTION** shows what is expected to occur with no additional protection.

Methods and Effects of Protection	Result in 2020s with NO NEW ACTION	Result in 2020s with PROTECTION OPTION A	Result in 2020s with PROTECTION OPTION B	
	No Change in Existing Defenses	More Emphasis on HARD Defenses	More Emphasis on HARD Defenses	
Homes Flooded in Category 2 Storm	<b>28%</b> 1,411 of 5,034 homes expected to flood in a Category 2 storm	20% 1,007 of 5,034 homes expected to flood in a Category 2 storm	20% 1,007 of 5,034 homes expected to flood in a Category 2 storm	
Homes Flooded Only in Category 3+ Storm	<b>23%</b> 1,174 of 5,034 homes expected to flood only in a Category 3+ storm	<b>27%</b> 1,359 of 5,034 homes expected to flood only in a Category 3+ storm	<b>19%</b> 956 of 5,034 homes expected to flood only in a Category 3+ storm	
Wetlands Lost	5% 25 of 497 wetland acres expected to be lost	2% 10 of 497 wetland acres expected to be lost	2% 10 of 497 wetland acres expected to be lost	
Beaches and Dunes Lost	10% 3 of 30 beach acres expected to be lost	<b>16%</b> 5 of 30 beach acres expected to be lost	<b>10%</b> 3 of 30 beach acres expected to be lost	
Seawalls and Coastal Armoring	<b>24%</b> 12 of 50 miles of coast armored	<b>24%</b> 12 of 50 miles of coast armored	<b>35%</b> 18 of 50 miles of coast armored	
S Cost to Your Household per Year	<b>\$0</b> Increase in annual taxes or fees	\$35 Increase in annual taxes or fees	\$155 Increase in annual taxes or fees	
HOW WOULD YOU VOTE? (CHOOSE ONLY ONE) I vote for	l vote for NO NEW	I vote for PROTECTION	I vote for PROTECTION OPTION B	

- Three choice tasks per respondent
- We focus on the inherent uncertainty related to the protection of homes vulnerable to flooding during storms of different intensities (the Saffir-Simpson Hurricane Wind Scale)
- These storms have some inherent probabilities of occurrence
- The storm probabilities do not vary across the protection scenarios (not included as an attribute)
- The effect of flood adaptation measures depends on the inherent storm probabilities

## Two survey versions

- Storm probabilities may be characterized by:
  - historical frequencies (common in media)
  - numerical percentage probabilities (common in stated preference surveys)
- Two versions of the survey that differ <u>only</u> in the uncertainty communication

#### • (1) Without numerical probabilities

 describes only historical frequencies of Category 2 and 3 storms (and asks about respondents' subjective assessments of the probabilities)

### • (2) With numerical probabilities

 provides identical information on historical frequencies but also translates these frequencies into numerical percentage probabilities

### Two survey versions

Over the last 75 years, Old Saybrook has been struck by **Category 2 storms in 1960, 1985 and 1991**, and by **Category 3 storms in 1938 and 1954**. There have been no Category 4 or 5 storms. Although hurricane Sandy was a Category 2 storm off the New Jersey coast, it weakened to below hurricane intensity before it reached Connecticut.

#### Without numerical probabilities

Based on past storm events, scientists estimate that there is approximately a 55% (or about one in two) chance that a Category 2 storm will strike Old Saybrook at least once by the mid 2020s (0% would mean there is no chance and 100% would mean it is absolutely certain).

In contrast, scientists estimate that there is approximately a **20% (or one in five) chance that a Category 3 or higher storm will strike Old Saybrook at least once by the mid-2020s** (0% would mean there is no chance and 100% would mean it is absolutely certain).

## Econometric approach

- Each model is pooled—estimated on samples from the two survey versions
- In willingness-to-pay (WTP) space: Parameters represent willingness-to-pay values in dollars per year
- Random parameters logit heterogeneous preferences described by continuous distributions of WTP parameters (all normal, except for the log-normal cost)

$$U_{ph}(\cdot) = \lambda_h (\boldsymbol{\omega}'_h \boldsymbol{X}_{ph} - \boldsymbol{C}_{ph}) + \varepsilon_{ph}$$

- An additional variable to capture systematic variation in preferences associated with the survey version ( $Num_h = 1$  for numerical probabilities);  $\omega_h = \omega_h^* + \rho Num_h$
- Latent class heterogeneous preferences described by discrete distributions – Three classes
  - Variable *Num<sub>h</sub>* used to explain class membership probabilities

# Random parameters logit

in willingness-to-pay (WTP) space

Choice	Mean WTP	Standard	Means interacted with
attributes	estimates	deviations	"numerical probabilities"
Status aug	-4.83***	10.34***	0.04
<i>Status</i> 400	(1.24)	(3.01)	(0.48)
Homesa	-1.38**	4.18***	0.35
	(0.63)	(1.18)	(0.68)
Homeso	-1.23*	4.47***	-0.44
	(0.64)	(1.23)	(0.73)
Motlands	-1.32*	3.64***	-0.17
wellunus	(0.74)	(0.99)	(0.88)
Reaches	-0.24	3.07***	-0.95
Deuches	(0.42)	(0.83)	(0.61)
Sogwalls	-0.59	1.17***	0.50
Jeuwalls	(0.38)	(0.33)	(0.39)
Hard	-1.47**	2.16***	0.66
<i>пи</i> и	(0.66)	(0.59)	(0.61)
Soft	-0.56	3.00***	0.47
	(0.52)	(0.87)	(0.56)
Cost	0.46	1.99***	0.33
- COSt	(0.53)	(0.43)	(0.47)

LL at convergence	-678.50
LL at constant(s) only	-883.88
AIC/n	1.8422
BIC/n	2.2094
Number of observations	805
Number of Sobol draws	6,000

- Mean WTP estimates with expected signs
- Substantial preference heterogeneity and not strongly significant means for parameters
- No effect of presenting numerical probabilities
- Can a latent class model better capture this heterogeneity?

## Latent class model

#### in willingness-to-pay (WTP) space

Attributes	ttributes Class 1		Class 3	
Status quo	0.44**	-2.76***	1.55	
	(0.22)	(0.44)	(1.16)	
Homes 2	1.3 <sup>8***</sup>	-0.50***	-0.34	
	(0.35)	(0.17)	(0.50)	
Homes 3	0.75*	-0.57***	-0.42	
	(0.40)	(0.19)	(0.67)	
Wetlands	0.14	-0.60**	-0.83	
	(0.43)	(0.27)	(0.68)	
Beaches	1.07***	-0.28**	-0.02	
	(0.31)	(0.14)	(0.38)	
Seawalls	0.51***	-0.28*	0.14	
	(0.16)	(0.16)	(0.28)	
Hard	-0.51***	-0.54**	-0.23	
	(0.09)	(0.24)	(0.71)	
Soft	0.43***	-0.19	0.63	
	(0.09)	(0.19)	(0.58)	
-Cost -31.54		1.07***	1.51*	
(51.54)		(0.21)	(0.81)	

LL at converg	ence	-681.09
LL at constan	-883.88	
AIC/n		1.7692
BIC/n		1.9498
Number of ob	805	
Class 1	Class 3	
embership probabilit	y function	

Class membersh			
Constant	-1.44***	0.68***	
	(0.43)	(0.19)	
"Numerical	1.03**	-0.14	
probabilities"	(0.52)	(0.29)	
Average class pr	robabilities		
	13%	57%	30%

• Standard neoclassical tradeoffs, in line with expectations

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• "Numerical probabilities" do not influence the probability of being in this class

## Latent class model

### in willingness-to-pay (WTP) space

Attributes	Class 1	Class 2	Class 3
Status quo	0.44**	-2.76***	1.55
	(0.22)	(0.44)	(1.16)
Homes 2	1.38***	-0.50***	-0.34
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	(0.09)	(0.24)	(0.71)
Soft	0.43***	-0.19	0.63
	(0.09)	(0.19)	(0.58)
– Cost	-31.54	1.07***	1.51*
	(51.54)	(0.21)	(0.81)

LL at convergence	-681.09
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	Class 1	Class 2	Class 3
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probabilities"	(0.52)	(0.29)	
Average class pr	robabilities		
	13%	57%	30%

- Pay attention only to cost
- A common pattern that some do not care about climate change adaptation measures and their environmental effects

## Latent class model

#### in willingness-to-pay (WTP) space

Attributes	Class 1	Class 2	Class 3		L	L at converge	ence	-681.09
Status quo	0.44** (0.22)	-2.76*** (0.44)	1.55 (1.16)	AIC/n BIC/n			-003.00 1.7692	
Homes 2	1.38*** (0.35)	-0.50*** (0.17)	-0.34 (0.50)	-	Number of observations			805
Homes 3	0.75*	-0.57***	-0.42	-	Class manhar	Class 1	Class 2	Class 3
	(0.40)	(0.19)	(0.0/)	_ <u>Class membership probability function</u>				
Wetlands	(0.43)	(0.27)	(0.68)		Constant	-1.44^^^ (0.43)	(0.19)	
Pagchas	1.07***	-0.28**	-0.02	-	"Numerical	1.03**	-0.14	
Deuches	(0.31)	(0.14)	(0.38)	_	probabilities"	(0.52)	(0.29)	
Segwalls	0.51***	-0.28*	0.14		Average class p	robabilities		
Jeuwalls	(0.16)	(0.16)	(0.28)	_		13%	57%	30%
Hard	-0.51*** (0.09)	-0.54** (0.24)	Signs for many Highly random	y paramete	ers are opposit	te to expect	tations;	
Soft	0.43***	-0.19	ringing randon	i choices				
<i>50j1</i>	(0.09)	(0.19)	Choices incons	sistent with	h standard neo	oclassical as	ssumption	s; These
-Cost	-31.54	1.07***	could be peop	le who wer	re confused, re	ejected scer	narios, pro	tested, etc
	(51.54)	(0.21)	• "Numerical pr	obabilities'	" increase the	probability	of being ir	n this class

#### Conclusions

## Do numerical probabilities promote informed stated preference responses under inherent uncertainty?

• Not necessarily

- The use of numerical probabilities to communicate inherent uncertainty leads to more "randomness" in stated preferences
- This may suggest increased symptoms of scenario rejection, protest responses, confusion, among others
- Our findings contradict a common (perhaps naïve) expectation that the use of numerical probabilities necessarily enhances the validity of stated preferences
- Numerical probabilities may not always be an effective way to communicate inherent uncertainty in environmental stated preference questionnaires

# THANKYOU!

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