Including Value Orientations in Choice Models to Estimate Benefits of Wildlife Management Policies

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Abstract

Value orientations towards wildlife affect the way people perceive nature and their connection with animals. In particular, the social psychological literature within the environmental field suggests that there are two main orientations of people towards wildlife: mutualism and domination. This body of literature has shown how wildlife value orientations can serve as predictors of attitudes and behaviours toward wildlife and form the foundation of human-wildlife conflicts. A common approach in the non-market valuation literature is to include information on attitudes and values in the deterministic part of the utility function, leading to problems of endogeneity bias. To avoid this, analysts have recently shifted their attention to approaches based on latent variables. This paper presents an application of a latent variable and latent class model, to understand how latent orientations influence choices, in a case study in the Italian Alps. The intuition is that different underlying individual value orientation affects preferences and the level of willingness to pay and should be therefore considered in choice models. The latent variable is used to explain class membership of respondents. Results indicate that the latent variable has a significant effect in class allocation and that the hybrid model performs better than a simple two class model. Results provide guidance on the social acceptability of management interventions and can support public decision-makers in the modulation of wildlife management policies for balancing the needs of conservation and outdoor recreation, explicitly considering existing human-wildlife conflicts.

Keywords: choice experiment, hybrid model, value orientation, endangered species, latent class model 2010 MSC: 00-01, 99-00

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1. Introduction 1

Conservation of wildlife is of primary importance worldwide, due to the alarming rate of biodiversity 2 loss affecting many natural areas. The population of several species has sharply decreased, mainly due to 3 hunting and habitat depletion. Economic valuation of biodiversity with stated preference methods, within 4 this framework, may be extremely beneficial to inform policy makers about people's attitudes and preferences 5 towards management alternatives. It is well-recognised in the literature that, when making choices in the en-6 vironmental field, people are led by several cognitive variables, such as attitudes, values and social norms. In 7 particular, value orientations (VOs) seem to play an important role in the individual choice process. Wildlife 8 VOs are defined as representing broad, cultural ideological believes, that orient and provide personal meaning 9 to basic values related to wildlife (Teel and Manfredo, 2010). In the literature, it is possible to identify two 10 main orientations towards wildlife (Manfredo, 2008): dominance and mutualism. People with a dominant 11 orientation tend to think that wildlife should be managed for the benefits of mankind. On the other hand, 12 a person with a mutualism orientation place humans and animals nearly on the same level; animals are seen 13 as creatures deserving rights and care. 14 Choice experiments (CE) are typical examples of techniques aiming at studying how people make choices. 15 Value orientations affect choices, which means that they should be therefore taken into account by the ana-16 lyst in CE surveys. A common approach in the non-market valuation literature is to use information about 17 18

attitudes and values, which are collected by means of Likert scales, in the deterministic part of the utility function. However, such indicators are likely correlated to other non-observed individual characteristics, thus 19 leading to problems of endogeneity bias (Hess and Stathopoulos, 2013). For the cited reasons, latent variable 20 methods are gaining popularity. Such methods acknowledge that what is observed is only answers to VO 21 questions and not the real orientation. In this vision, a latent variable is included in the model, in order to 22 explain simultaneously the behaviour of the respondent in the stated choice survey and in the value orienta-23 tion questions. Value orientations are no longer in the deterministic part of the utility function but treated

24 as dependent variables, estimated simultaneously to the choice model, thus eliminating problems connected 25

with endogeneity. 26

In this paper, we apply the latent variable approach in a CE hybrid estimation for valuing tourists pref-27 erences for endangered species management. The main methodological contribution is to include specific 28 value orientations as measures of individual attitudes in the set of structural equations. We expect that 29 value orientations are linked to willingness to pay (WTP) and that they can be used to explain preference 30 heterogeneity for wildlife conservation. Structural equations are modelled as ordered logit, while the choice 31 model with a latent class model. The latent variable is used as explanatory variable for the value orientation 32 and then enters the choice model in the class allocation function. We show how the inclusion of value orienta-33 tions can improve WTP estimates, with relevant policy implications. The method is applied to a case study 34 about wildlife management in the Italian Alps, the Province of Trento (Trentino). We consider three wildlife 35 species, i.e. wolf, lynx and salamander, and we ask to a sample of local tourists whether they are willing to 36 pay for an increase in their population. Wolves and lynx were naturally abundant in Trentino until the end 37 of the 19th century. Later on, due to hunting and bad habitat conditions, their population decreased rapidly, 38 bringing to their extinction. In recent years, the increased habitat quality provoked a natural return of some 30 specimen from close areas, but their number is not enough to assure reproduction. At the same time, the 40 case of salamander is interesting as well. In fact, a particular sub-species of Salamander, called salamander of 41 Aurora, lives only in a limited area of Trentino and in a valley of a neighbouring region. Establishing a viable 42 population for these species is a primary challenge for local decision-makers, in order to assure a long-lasting 43 conservation, in this context the investigation of tourist preferences might help in designing more effective 44 policies.

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2. Value Orientation Theory 46

Human-widlife relationships and interactions derive from the cognitive basis that forms human thought 47 and behaviour toward wildlife (Teel and Manfredo, 2010). A cognitive hierarchy model has been developed to 48 study the cognitive foundation of these relationships (Fulton et al., 1996; Whittaker et al., 2006; Manfredo, 49 2008; Teel and Manfredo, 2010). This theory is based on the value-attitude-behavior model (Homer and 50 Kahle, 1988), focusing on the fact that cognition exists on different linked levels of abstraction. The cognitive 51

⁵² hierarchy model includes values at the base, then going higher in the hierarchy VOs, attitudes and norms,

⁵³ behavioural intentions and behaviours. Values are the most abstract cognitions in the human mind, they are

⁵⁴ few in numbers, slow to change, central to beliefs and transcend to situations. The values of a person are ⁵⁵ shaped in the early years of life and are strongly influenced by the sociological context. Going up through the

⁵⁶ cognitive hierarchy, cognitions become more numerous, quick to change, peripheral and specific to situations.

⁵⁷ Value orientations are networks of basic and core beliefs that serve as intermediary between values and

attitude (Manfredo, 2008). They are reflective of the cultural ideology (Manfredo et al., 2009) and provide a
 contextual meaning for values within a given domain of interest such as wildlife (Teel and Manfredo, 2010).

⁶⁰ Since they are less abstract than values, VOs can better explain specific thoughts and behaviours.

A first articulated classification of attitudes towards wildlife was proposed by Kellert (1980). In more recent 61 years, the literature has suggested that people tend to show mainly two different and opposing VOs toward 62 wildlife: mutualism and domination (Manfredo, 2008; Manfredo et al., 2009; Teel and Manfredo, 2010; Teel 63 et al., 2010). A domination orientation stems from a utilitarian view of the relationship between humans 64 and wildlife; it follows that wildlife should be managed for human benefit. Domination is one of the oldest 65 VO showed by the humankind. A person with this VO tends to believe in the human mastery over the 66 animals and is more prone to accept control measures resulting in death or harm to wildlife and more 67 likely to engage in behaviours such as hunting and fishing. On the other hand, a person with a mutualism 68 orientation tends to place humans and animals nearly on the same level; animals are seen as creatures with 69 their own personalities and emotions. Such people recognize also that animals need care and have rights. A 70 strong mutualism orientation render people less likely to accept control measures towards wildlife, as well 71 as management options involving killing or hurting specimens but more likely to exhibit behaviors such 72 as wildlife viewing and feeding. Mutualism is strongly related to the modernization, to the importance of 73 wildlifes non-consumptive value and seems more consistent with a biocentric philosophy (Manfredo et al., 74 2016). The dualism mutualism-domination can be viewed in economic terms as utilitarian versus intrinsic 75 views (Rolston III, 1983; Rolston, 1994). The dichotomy is usually seen as a continuum and the two different 76 wildlife VOs often occurs in various levels. According to Teel and Manfredo (2010), gradients between these 77 two main orientations can be found; in particular these authors suggest two other subclasses: distance and 78 pluralism. The distant orientation includes people who do not care or who do not care very much about 79 wildlife. On the other hand, pluralists do not show a particular orientation and their opinions on wildlife can 80

⁸¹ be influenced by the contingent situation.

Several studies have demonstrated that wildlife VOs can serve as a predictor of attitudes toward wildlife 82 (Hartel et al., 2015), wildlife management options (Kansky et al., 2016; Sponarski et al., 2015; Hermann 83 et al., 2013; Jacobs et al., 2014), wildlife viewing (Manfredo et al., 2016; Teel et al., 2010) and toward 84 hunting (Hrubes et al., 2001; Gamborg and Jensen, 2017) and fishing (Riepe and Arlinghaus, 2014). This 85 body of research has shown that a specific behaviour towards wildlife can be explained by different VOs. 86 87 and that VOs are at the basis of the conflicting attitudes for wildlife management actions, but none of these studies have used an economic approach. VOs can be linked to the economic theory of value as 88 ideals affecting choices and actions. Steinhoff (1980) and subsequently Brown (1984) offered an interesting 89 preference-related theory of value. They distinguish between held values and assigned values. A held value 90 is the basis for preference about things, a 'conception of the preferable' (Brown 1984, p.232), a first order 91 preference affecting second order preferences (i.e. choices and actions). Examples of held values provided by 92 Brown are model of behaviour (e.g., bravery), end-states and qualities. In contrast, assigned values is the 93 economic value of an object. Held values are values of the subjects and assigned values are of the objects. 94 Within this framework, we can think about VOs as held values affecting WTP, i.e. economic values assigned 95 to wildlife. 96

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97 3. Brief Overview of biodiversity in CE

Non-market valuation techniques, and CE in particular, have been extensively used in valuing biodiversity. Most of the available studies do not focus on the economic valuation of biodiversity but rather on a single species (Pearce, 2001). For example, Han et al. (2010) implemented a CE survey, for assessing tourists' perceived best management alternative for the conservation of the goral, in Woraksan National Park (South Korea). Similarly, Hanley et al. (2003) evaluate the benefits provided by wild geese. Delibes-Mateos et al.

(2014) considered the quantity of partridges likely to be shot in game activities as an attribute, while another 103 attribute was the possibility to have additional (not specified) species. CE applications, in which several 104 species are included in the study as different attributes, are less common in the literature. Hanley et al. 105 (2010) evaluate simultaneously the worth of two Scottish species, namely hen harries and golden eagle, esti-106 mating people's WTP for an increase in their populations. Di Minin et al. (2013), investigate people's WTP 107 for conserving several endangered species in South Africa, including lions, leopards, rhinos and buffaloes. 108 Similarly, Wallmo and Lew (2011) evaluates the perceived benefits of conserving some marine species, cur-109 rently under the stewardship of the National Marine Fisheries Service (NMFS) in the United States. Finally, 110 Campbell et al. (2012) and Campbell et al. (2015) use a CE to estimate the existence value of a number of 111 rare and endangered fish species in Ireland. 112

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3.1. The inclusion of attitudes and values in biodiversity evaluation studies

While environmental attitudes are sometimes considered in CE surveys (see, among others, Choi and 115 Fielding, 2013; Milon and Scrogin, 2006; Birol et al., 2006; Morey et al., 2008), and value orientations in 116 contingent valuation under the Environmental Concern framework (Ojea and Loureiro, 2007; Spash, 2006), 117 to be best of our knowledge VOs have never been included in choice experiments applied to biodiversity eval-118 uation. Specifically Ojea and Loureiro (Ojea and Loureiro, 2007), in their Contingent Valuation study have 119 tried to value the importance of three generic environmental VOs of the attitude towards the environment-120 the altruistic, egoistic, and the biospheric value—in WTPs for the recovery of a marine bird, the common 121 murre, in Galicia (Spain). Spash (Spash, 2006) have analysed the role of the same generic value orientations 122 in estimating WTPs for wetlands recreation. The novelty of our study is that we have modeled specific latent 123 value orientation to explain taste heterogeneity of respondents towards wildlife conservation in a CE. 124

As already highlighted, including attitudes in the utility function may not represent an optimal solution. 125 In particular, measurement errors may occur. In addition, attitudinal questions may be correlated with 126 other unobserved characteristics of the respondent, thus causing correlation between the deterministic and 127 stochastic components of the utility. In an attempt to address this important issue, hybrid models have 128 been recently developed (Ben-Akiva et al., 1999, 2002; Bolduc et al., 2005). The innovative approach is to 129 use latent variables to explain attitudes and psychological constructs. The latent variable is a function of 130 socio-demographic characteristics and enters the model both in the choice model and as explanatory variable 131 for attitudinal questions (Bolduc and Alvarez-Daziano, 2010). In this way, attitudinal questions are treated 132 as a function of the latent variable, eliminating endogenity problems. Hybrid choice models have been ap-133 plied extensively in transportation research (see, among others, Daly et al., 2012; Abou-Zeid et al., 2010; Hess 134 et al., 2013), to lesser extent in other fields. In environmental valuation, an example of application is given by 135 the paper of Hess and Beharry-Borg (2012), which analyzes values of improving the quality of coastal water. 136 Another interesting paper is provided by Hoyos et al. (2015), in which the authors use a psychometric scale, 137 called awareness of consequences (AC), in order to understand choices of respondents while valuing land-use 138 policies for Natura 2000 network sites. Finally, a hybrid approach to include environmental attitudes for 139 valuing forest ecosystem services has been recently proposed (Agimass et al., 2017). In this paper, we make 140 use of two latent variables, one describing the mutualism orientation and one for the dominant orientation. 141 Following the idea of Hoyos et al. (2015), latent variables enter the model in the class allocation function of 142 the latent class model. 143

¹⁴⁴ 4. Methods

145 4.1. Study Area

Data for this case study originated from a questionnaire survey, administrated face-to-face in Trentino, a mountainous province in the north- east of the Italian Alps. Trentino is an important tourist destination, with around three million tourists per year and a good balance between winter and summer tourists. This area is important for nature conservation, because of the presence of several rare and endangered species. The province includes one national park (Parco Nazionale dello Stelvio), two regional parks (Adamello-Brenta and Paneveggio Pale di San Martino) and several other Natura 2000 sites, occupying 34 percent of the total area. Among several interesting species, this study focuses on tourists' preferences for conserving the wolf (Canis lupus Linnaeus), the lynx (Linx linx) and the salamander of Aurora (Salamandra atra aurorae), a rare subspecies of the alpine salamander. Wolf and lynx, extinct in Trentino around the end of the 19th century, naturally came back from neighbouring areas, in particular the wolf from the Italian Appennine and the lynx from Switzerland. At present, there are seven wolves and only one lynx in the regional area, therefore the population size is not enough to assure the survival of the species. The salamander of Aurora, conversely, is a rare amphibious living only in a limited area of the Province with a population of about ten specimen.

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161 4.2. Data

The CE survey was carried out by means of personal interviews, conducted by three trained interviewers, 162 to a sample of tourists of the Province of Trento from July and September 2015. We focused on tourists 163 as they comprise a large share of the direct users of the parks and natural areas in Trentino. Wildlife is 164 an important regional resource and attracts many visitors annually (Tattoni et al., 2017). We also acknowl-165 edge that local communities represent a relevant actor for wildlife management and should be considered in 166 decision-making, but this important issue was out of the scope of the present research. Respondents were 167 randomly selected in tourist areas and invited to take part to the interview. Personal interviews were be-168 lieved to facilitate the "yes-saying", however recent contributions suggest that administration method do not 169 influence SP surveys (Bell et al., 2011; Lindhjem and Navrud, 2011), even though results are mixed. The 170 NOAA panel on contingent valuation suggested personal interview as administration method (Arrow et al., 171 1993), while the most recent SP guidelines (Johnston et al., 2017) do not provide specific recommendations. 172 We therefore chose to administrate interviews personally. Moreover, some authors argue that stated CE 173 have the advantage to reduce the yeah-saying behaviour compared to contingent valuation surveys (among 174 others, Boxall et al., 1996; Hanley et al., 1998; Winslott Hiselius, 2005). A pilot study with 63 tourists was 175 conducted in June 2015 to fine-tune the questionnaire, check attributes and levels and collect priors to be 176 used for generating a Bayesian D-efficient design (Bliemer et al., 2008). The questionnaire was composed by 177 34 questions, organized in four thematic sections. The questionnaire was a tool used for a broader research 178 and not all questions were considered in this study. In section number one, environmental indicators for the 179 individual VO were collected with fourteen propositions, to which respondents had to express their degree 180 of agreement on a 7-point-Likert scale. These questions replace the typical attitudinal questions that are 181 included in stated preferences applications. Statements to capture orientations were taken from the ques-182 tionnaire used by Sponarski et al. (2015) and previously by Teel and Manfredo (2010), which were exactly 183 replicated to better capture orientations. Out of these, seven propositions were related to the mutualism 184 orientation and seven to the domination orientation. Attitudinal questions in the form of Likert scales might 185 be subject to acquiescence, i.e. yeah-saying bias (Ray, 1990), because people might tend to show a more 186 'environmentally-friendly' attitude compared to their real values. For this reason, we tried to reduce the 187 occurrence of yeah saying bias by training interviewers to be as neutral as possible while administrating 188 the questionnaire and allow people self-filling in the answers to VOs questions. Although questions were 189 taken from the literature and already tested, we also checked wordings and consistency during the pre-test 190 phase. The inclusion of fourteen proposition to examine VOs was necessary to respect prescription from 191 the environmental psychology literature (Manfredo, 2008; Teel and Manfredo, 2010; Sponarski et al., 2014), 192 however, using all these pieces of information into the choice model would have complicated too much the 193 estimation, in terms of number of parameters. For this reason, only four were selected as indicators for the 194 final hybrid model (Table 1). This simplification is convenient because it reduces the number of parameters. 195 which was already large, but it might approximate the assessment of individual orientation. The inclusion 196 of four indicators is in line with several hybrid CE studies (for example, Hess and Stathopoulos, 2013; Hoyos 197 et al., 2015). In the final model, we included the four statements that interviewers reported to be the most 198 clear for respondents, however we also conducted a sensitivity analysis using different set of statements and 199 results did not change significantly. Section number two contained questions on emotions provided by respon-200 dents in their interactions with wildlife. These questions were not related to VOs and were not considered 201 in this study. Section three contained choice cards which were preceded by an explanatory text that was 202 read by interviewers before showing the cards. Respondents were informed that wolves, lynx and salaman-203 ders of Aurora are protected and managed in Trentino from the Province of Trento and from the natural 204

parks within specific European projects. Then they were informed about the actual situation (7 wolfes, 1 205 lynx and 10 salamander) and we explained that the actual situation can improve, but also deteriorate if no 206 management actions are going to be taken. We gave no description of any management actions. The null 207 alternative was presented to respondents as the 'do-nothing situation' in terms of management, what will 208 happen if management actions are not implemented. We then added that an entrance fee for the park might 209 be a solution to increase revenues and address conservation issues. Finally, we included some cheap talks to 210 inform respondents that results will be used by the Province of Trento to improve management policies and to 211 encourage accurate responses. Cheap talks were remainders that they had to consider their budget and that 212 if they agreed to pay a fee they would have less money for other purchasings. The preparation of the CE was 213 carried out following guidelines available in the literature (Hoyos, 2010; Riera et al., 2012). The attributes 214 selected for the survey, as shown in table 2, are the number of animals for wolves, lynx and salamanders; the 215 cost attribute was an entrance fee for parks and natural areas in Trentino. At present, there are no entrance 216 fees and visitors can access all natural areas for free, therefore our payment veichle is hypothetical. However 217 a ticket could be a solution to increase parks' self-funding. In fact, in Trentino, parks have to co-finance 218 activities related to biodiversity conservation, thus an entrance fee was included as a way for tourists to 219 contribute to these actions. Attributes and attribute levels were determined by experts and scientists. In 220 one-on-one interviews wildlife managers of the Province of Trento and zoologists of the Science Museum of 221 Trento stated that a viable population for wolves and lynx was of about 45-50 individuals and a maximum 222 of 90-100. Within this range the carrying capacity of the territory is respected and wolves and lynx are ex-223 pected not to compete for habitat and food. Salamanders could potentially have a bigger population, but the 224 pre-test highlighted that larger levels lead people to think the animal was not in danger and non-attendance 225 of this attribute was high. Therefore, we decided to maintain the same attribute levels also for salamanders. 226 Despite experts proposed levels for animal populations that are respectful of the local carrying capacity, there 227 could be correlation across attributes if some management actions are proposed. For example, if we stated 228 that larger populations had to be achieved by an increase in the habitat quality, this would imply a better 229 ecosystem for all the animals and their population would increase simultaneously. For this reason, we were 230 careful in avoiding management suggestions, so that we were able to capture uncorrelated preferences for 231 each of the species¹. During the pre-test phase we also asked some questions that can be used to understand 232 whether respondents perceived attributes to be correlated. For this purpose, we asked respondents their opin-233 ion on the current size of the populations of wolves, lynx and salamanders. Since most of the respondents 234 reported high numbers of animals and thought that the populations were increasing, this brings additional 235 evidence that respondents, on average, did not perceive particular problems in the coexistence of the animals. 236 237

After the pre-test we noticed that there was a quite large share of preferences for the null alternative, 238 which was chosen 20.5% of the time as first best and 7.7% of the time as second best. This result, together 230 240 with a previous experience on wildlife study in the same study area, which also detected a large share of SQ choices (Agnolin, 2012), led us to worry for the SQ bias. The SQ coefficient was positive, suggesting 241 positive utility for the current situation. We suspected that using the real SQ led people to be more likely to 242 choose it, because they could have a certain number of animals without paying for them. In this way policy 243 improvements provided by larger population sizes were not properly highlighted, therefore we opted for a 244 null alternative (Olsen et al., 2012; Whittington et al., 2017; Scarpa et al., 2011), which is often used in the 245 literature and contributed to improve the model². We believe that this does not impact on final estimates, as 246 our objective was to estimate marginal WTP and not consumer surplus, for which a hypothetical SQ could 247 have caused problems when applying the *log-sum* formula (Hanemann, 1984), because the comparison of the 248 policy improvement with the baseline SQ would be complicated. 249

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 $^{^{1}}$ A small effect of correlation between attribute levels might still be present and this is a limitation to consider when interpreting the results.

²As one reviewer suggested, in our experiment part of the SQ bias could be caused by the payment vehicle we chose, as an entrance ticket could be opposed by tourists. However, a different mandatory instrument was difficult to retrieve and voluntary payments are often inadequate as they tend to overestimate the WTP (Wiser, 2007). We believe that this effect is likely to be small, because paying a fee for a natural park is expected by the new law on natural areas (art. 18 L. 394/91).

	Question	Orientation	
	We should strive for a world where		
WVO1	theres an abundance of wildlife for	Domination	
	hunting		
WVO2	The needs of humans should take prior-	Domination	
W V 02	ity over fish and wildlife protection	Domination	
	We should strive for a world where hu-		
WVO3	mans and fish and wildlife can live side	Mutualism	
	by side without fear		
WVO4	I value the sense of companionship I re-	Mutuoliam	
	ceive from animals	Mutualishi	

Table 1: Questions on value orientations

Table 2: Attributes and levels in the choice tasks					
Attribute	Description	Levels			
Wolves	Number of wolves	0, 15, 30, 45, 60, 75, 90			
Linx	Number of lynx	0, 15, 30, 45, 60, 75, 90			
Salamanders	Number of salamanders	0, 15, 30, 45, 60, 75, 90			
Cost	Entrance fee for parks (in ${\ensuremath{\in}})$	0, 3, 6, 9, 12, 15, 18			

We hypothesized a non-linear relationship between population sizes and people's WTP. People might 251 be willing to pay for a viable population but, at the same time, they might want not too big populations 252 because they may generate human-wildlife conflicts (Dickman, 2010; Kubo and Shoji, 2014). For this reason, 253 we tested non-linearities in the utility with two models, one with a linear specification of the utility function 254 with a dummy for each attribute level (with twenty parameters to be estimated) and a quadratic specification 255 (with eight parameters), with a linear and quadratic coding for each attribute, in a multinomial logit model. 256 The log-likelihood were very similar (-5300 in the linear model and -5335 in the quadratic model) but the 257 BIC was lower for the quadratic model. For this reason, we decided to use the quadratic expression of the 258 utility function also in the more complicated subsequent models, which has also the advantage of a smaller 259 number of parameters to be estimated. The monetary attribute was linearly coded. 260

Each respondent was asked to complete twelve choice tasks, composed by a null alternative and two other 261 alternatives. The chosen answer format was the best-worst scaling (BWS) (Flynn et al., 2007), through 262 which each interviewed person had to express his/her best and worst alternative, among the three available. 263 The BWS assures accurate estimates and provides a larger number of observation compared to the common 264 pick one solution, with only a small effort increase for respondents. An example of choice card is provided 265 in figure 1. The final section of the questionnaire included the usual socio-demographic characteristics (e.g., 266 gender, age, education and income). For the present analysis, the relevant population was assessed to be 267 composed by tourists in the province, aged 18 or more, which were reached and surveyed on site. 268

269 4.3. Econometric Analysis

We implement a hybrid latent class and latent variable model, following the method proposed by Hoyos 270 et al. (2015). Value orientations towards wildlife are considered to be a source of taste heterogeneity among 271 respondents but they are, at the same time, latent and unobservable. Some of the latent orientations may be 272 captured by socio-demographic characteristics. Such underlying individual VOs influence choices, through 273 class allocation probabilities, and answers to VOs questions value at the same time. Within this hybrid 274 framework, two set of structural equations have to be specified, one related to the choice model and one 275 to the latent variable model. In addition, a set of measurement equations has to be included as well. 276 Measurement equations include answers to VOs questions as dependent variables, explaining their outcome 277 with the latent variable as covariate. 278

²⁷⁹ Concerning the choice model, the theoretical foundation lies in the theory of random utility (RUM), which

		Scenario A	Scenario B	Scenario C
	Number of free wolves	90	30	0
	Number of free lynx	0	45	0
	Number of free salamander aurora	45	0	0
€	Entrance ticket	6€	12€	0€
Choose your most preferred option (mark $)$				
Choose your least preferred option (mark $$				

Figure 1: Example of choice card

allows defining the observed part of the utility through a statistical model (Manski, 1977). According to RUM, the utility that respondent n derives from alternative i in the choice situation t is given by:

$$U_{int} = V_{int} + \epsilon_{int} \tag{1}$$

where V_{int} is the observed component of the utility, while ϵ_{int} represents the non-observed random distur-282 bance. The observed component of the utility V_{int} is described by a set of attributes x_{it} and by a vector of 283 parameters β , indicating the marginal effect of the attributes on utility. Modelling utility needs assumptions 284 on the distribution of the random term ϵ . A common approach is to assume an i.i.d. type I extreme value 285 distribution for the random term, thus leading to a multinomial logit model (MNL). The MNL model assumes 286 preferences are constant across respondents and does not take into account preference heterogeneity. In order 287 to account for preference heterogeneity, we make use of an extension of the simple MNL, the latent class logit 288 model (LC), which hypothesise that individuals can be sorted into C classes (Greene and Hensher, 2003), 289 each with class-specific β_c . Preferences are now constant in each class but vary across classes, thus relaxing 290 the assumption of preference homogeneity in the entire sample. 291

In the LC framework, given class membership c, the probability of a sequence of T_n choices made by individual n is given by (Shen, 2009):

$$\Pr\left(y_n|\beta_c, x_n, c\right) = \prod_{t=1}^{T_n} \frac{\exp\left(\beta_c' x_{int}\right)}{\sum\limits_{j=1}^{J} \exp\left(\beta_c' x_{jnt}\right)},\tag{2}$$

where y_n is the choice made over the total number on choice situations T_n and J is the number of alternatives in each choice situation ³.

²⁹⁶ The second structural set of equations concerns the latent variables. We include two latent variables, one

capturing the domination behaviour, while the second capturing the mutualism behaviour. Reference categories are distant and pluralist, because indicators for these categories are not available Teel and Manfredo
(2010). The *q*th latent variable is defined as:

$$LV_{qn} = h(Z_n, \gamma_q) + \omega_{qn} \tag{3}$$

 $^{^{3}}$ A latent class model is also advantageous since it inherently allows for possible correlations between preferences for different attributes as well as how attributes were perceived by respondents

Table 3:	Explanatory	variables	used in	the	latent	variable	equations
T	4 T 7 • 1	1 • 1	1 1	<u> </u>	• .•		

Latent Variable variables	Description
Female	1 = female
	0 = otherwise
AGE_category1	1 = 18 < AGE < 30
	0 = otherwise
AGE_category2	1 = 31 < AGE < 45
	0 = otherwise
AGE_category3	1 = 46 < AGE < 60
	0 = otherwise

where $h(Z_n, \gamma_q)$ is the deterministic part of the equation; Z_n is a vector of sociodemographic characteristics and γ_{qn} a vector of parameters to be estimated; h(.) is linear. Conversely, ω_{qn} is the error component of the equation, which is assumed to follow a normal distribution with zero mean and standard deviation $\sigma_{q\omega}$. As visible in table 3, covariates in the latent variable equations were gender and age classes, using the highest age class (i.e. people aged 60 years or above) as reference level.

Together with the described two structural equation sets, a measurement equation set of variables has to be determined as well. The measurement equation set is composed by the answers to VOs questions as dependent variable, explained with the latent variable as covariate. The *l*-th indicator for respondent n is

$$I_{qnl} = f(LV_{qn}, \xi_q) + \nu_{qn} \tag{4}$$

Where I_{qnl} , i.e. the answer to VO questions, is a function of the latent variable LV_{qn} and the vector 309 of parameters to be estimated ξ_q . Value orientation questions were collected on a seven point scale (from 310 strongly disagree to strongly agree), subsequently recoded in a three level scale. Categories "strongly disagree" 311 and "disagree" were merged together and recoded as one, median classes ("mildly disagree", "neutral" and 312 "mildly agree") were coded as two and "agree" and "strongly agree" were coded as three. This should not 313 modify the result as the order of preference is maintained. This question format has an intrinsic ordering of 314 the answers, thus an ordered logit model has been used for modelling the measurement equations. Ordered 315 logit includes threshold parameters for the latent variable (Greene, 2003), which need to be estimated: 316

$$f(x) = \begin{cases} i_1 & if - \infty < LV_{qn} < \tau_{ql1} \\ i_2 & if \tau_{ql1} < LV_{qn} < \tau_{ql1} + \delta ql1 \\ \dots \\ i_k & if \tau_{ql(k-1)} < LV_{qn} < \infty \end{cases}$$
(5)

where τ_{ql1} is the first threshold parameter and δ_{ql} the width of each ordered class. The latent variables 317 $LV_{1n},...,LV_{Qn}$ are linked to the choice model as well and enter the class allocation probabilities π_{n,c_s} , that 318 are now respondent specific because of the influence of the latent variables. During the modelling phase we 319 also explored the possibility to include only one latent variable, entering the model as explanatory variable for 320 two indicators. We tried this for both the mutualism and dominant latent variable. Signs were all maintained, 321 in the choice part of the model coefficients changes only after the fourth decimal place, suggesting that using 322 one or two latent variables does not impact the computation of WTP significantly. We then decided to include 323 in the final model both the mutualist and dominant LVs to show the effect of the two types of orientations at 324 the same time. The described equations are estimated simultaneously, by jointly maximizing the likelihood 325 function of the sequence of choice and the answer to attitudinal questions, conditional on the realization of 326 the latent variable LV_{qn} . The log-likelihood function of the model is given by the integration over ω_{qn} : 327

328

305

$$LL(\beta,\mu,\gamma,\xi,\tau) = \sum_{n=1}^{N} ln \int_{\omega} (P_n \prod_{l=1}^{L_q} \prod_{q=1}^{Q} L_{Iqln}) g(\omega) d$$

Orientation	Disagree	Median	Agree
WVO1	42.62~%	44.76~%	12.62~%
WVO2	35.48~%	41.66~%	22.86~%
WVO3	3.58~%	23.57~%	72.85~%
WVO4	2.38~%	22.62~%	75.00~%

Table 4: Level of agreement with orientation questions of the respondents

329

(6)

The global model is estimated through maximum simulated likelihood in R (R Core Team, 2013), using 1,000 Sobol sequences for the simulation of the error term of latent variable.

332 5. Results and Discussions

Interviewers were able to collect a sample of 420 respondents (65 percent response rate); however, only 333 341 were useful for the present analysis, for a total number of 8184 observations. Respondents were on aver-334 age 43 years old and females accounted for the 53.3 percent of the sample (males constituted the remaining 335 46.7 percent). Most of respondents had a high school degree (41 percent) but the presence of people with a 336 university diploma was high as well (around 37 percent). Sample characteristics are in line with the average 337 tourists visiting Trentino, although we oversampled a bit national tourists compared to foreigners. In fact, 338 according to local statistics on tourism 4 summer visitors from abroad represent 26% of the total, while in 339 our sample they accounted for 10.3%. (35 respondents) 340

Answers to questions on VOs are available in table 4, in which WVO1 and WVO2 are questions related to 341 the domination orientation, while WVO3 and WVO4 are connected with the mutualism orientation. It is 342 possible to see that most of respondents tend to disagree with the propositions related to the domination 343 orientation and to agree more with the mutualism propositions, thus suggesting that most of respondents 344 show a mutualism orientation. A paper by Vaske et al. (2011), which reports a study in the Netherlands, 345 found that 44% of respondents showed mutualism, 21% a dominant orientation and 35% none of the two. 346 They also pointed out that their sample was mainly composed by people living close to natural areas and 347 that there is a dichotomy between urban and rural residents, with people living in rural areas to be more in 348 favour of an anthropocentric mentality, involving hunting and the use of wildlife for people's benefit Teel and 349 Manfredo (2010). Looking at the frequency of responses in our sample, the share of mutualist people seems 350 to be larger and the place of origin of local tourists might explain the result. In fact, only 11% of respondents 351 lived in rural territories while the vast majority lived in urban areas (34% in small town and 45% in medium 352 of large cities). Part of this result might be caused by acquiescence but, more importantly, results might 353 be driven by the different reference population between the cited study and this one, which is focused on 354 tourists. Tourists usually visit natural areas for short periods, so their psychological distance with wildlife is 355 larger. Serenari et al. (2015) explored the differences in wildlife value orientations among visitors, local and 356 tourists, to Chiles Tamango National Reserve. They found that local residents were more likely to belong 357 to the mixed protection group (57.6% of residents), while tourists to the strong protections group (73.4% of)358 tourists). Tourists were more supportive of huemul conservation policies and were also more likely to pay the 359 reserve entry fee. 360

³⁶¹ The large share of people with high scores in the mutualist orientation statements, as well as small shares

of agreement with dominant orientation statements confirms a societal shift from dominance to mutualism,

which is a result of modernization (Manfredo, 2008; Manfredo et al., 2009, 2016). We conducted a χ^2 test

to check the association between the four orientation propositions. All tests were significant at one percent

 $_{265}$ confidence level (p-value = 0.002 and p-value = 0.000, respectively), thus the null hypothesis of independence

³⁶⁶ between answers was rejected. This result strengthens the idea that answers to domination and mutualism

⁴http://www.statistica.provincia.tn.it/binary/pat_statistica_new/turismo/ITuristiNellaStagioneEstiva2015. 1447238555.pdf

Parameter	Estimate	St. err.	sign.
Constant	7.57	1.09	****
Female	$-1.97 \cdot 10^{-01}$	$7.91 \cdot 10^{-01}$	
AGE_category1	-8.12	1.01	****
AGE_category2	-6.87	1.26	***
AGE_category3	-8.56	$6.29 \cdot 10^{-01}$	****
ω_{dom}	7.07	$7.90 \cdot 10^{-01}$	****
Mutualism LV			
Parameter	Estimate	St. err.	sign.
Constant	1.53	$4.61 \cdot 10^{-01}$	****
Constant Female	$\frac{1.53}{2.24 \cdot 10^{-01}}$	$\frac{4.61 \cdot 10^{-01}}{2.23 \cdot 10^{-01}}$	****
Constant Female AGE_category1	$ \begin{array}{r} 1.53 \\ 2.24 \cdot 10^{-01} \\ -4.75 \cdot 10^{-01} \end{array} $	$ \begin{array}{r} 4.61 \cdot 10^{-01} \\ 2.23 \cdot 10^{-01} \\ 5.82 \cdot 10^{-01} \end{array} $	****
Constant Female AGE_category1 AGE_category2	$\begin{array}{c} 1.53 \\ 2.24 \cdot 10^{-01} \\ -4.75 \cdot 10^{-01} \\ 6.81 \cdot 10^{-01} \end{array}$	$\begin{array}{r} 4.61 \cdot 10^{-01} \\ 2.23 \cdot 10^{-01} \\ 5.82 \cdot 10^{-01} \\ 5.92 \cdot 10^{-01} \end{array}$	*
Constant Female AGE_category1 AGE_category2 AGE_category3	$\begin{array}{c} 1.53\\ 2.24\cdot 10^{-01}\\ -4.75\cdot 10^{-01}\\ 6.81\cdot 10^{-01}\\ 3.36\cdot 10^{-01}\end{array}$	$\begin{array}{r} 4.61 \cdot 10^{-01} \\ 2.23 \cdot 10^{-01} \\ 5.82 \cdot 10^{-01} \\ 5.92 \cdot 10^{-01} \\ 5.66 \cdot 10^{-02} \end{array}$	****
Constant Female AGE_category1 AGE_category2 AGE_category3 ω_{mut}	$\begin{array}{c} 1.53 \\ 2.24 \cdot 10^{-01} \\ -4.75 \cdot 10^{-01} \\ 6.81 \cdot 10^{-01} \\ 3.36 \cdot 10^{-01} \\ -1.75 \end{array}$	$\begin{array}{r} 4.61 \cdot 10^{-01} \\ 2.23 \cdot 10^{-01} \\ 5.82 \cdot 10^{-01} \\ 5.92 \cdot 10^{-01} \\ 5.66 \cdot 10^{-02} \\ 7.59 \cdot 10^{-01} \end{array}$	**** * *** **

 Table 5: Coefficients for the specification of latent variables

 Domination LV

371

Table 5 displays coefficients for the latent variables, which are assumed to depend on gender and age. 372 Differences in coefficients between the two latent variables indicate that they both have a role and explain 373 dissimilarities between mutualism-dominance and other orientations, such as distance and pluralism. LVs 374 enter the class allocation function, therefore signs indicate the probability of belonging to class one (for iden-375 tification reason, we set parameters for class two fixed at zero (Scarpa and Thiene, 2005)). The coefficient 376 for females is positive in the mutualism latent variable, while negative for the domination orientation. This 377 means that, all else held constant, females have a higher likelihood of having a mutualism orientation, albeit 378 neither are statistically significant. Difference in signs is also visible for the dummy variables indicating 379 age classes. The baseline age class is given by respondents who are 60 years of age or over. Thus, in the 380 domination latent variable, the negative and statistically significant coefficients indicate that people younger 381 than the baseline are less likely to show a domination orientation. In contrast, age categories 2 and 3 are 382 significant in the mutualism latent variable, but not class 1. Thus, the general indication from our model is 383 that younger tourists are more likely to be mutualist, rather than older ones, all else held equal. A negative 384 relationship between age and interest towards wildlife is a quite common result and found, for example, by 385 Bjerke et al. (1998). In the broader context of environmental protection, two main reasons have been detected 386 for such a result: first, older people have less time to benefit from long-term preservation of the resources 387 (Carlsson and Johansson-Stenman, 2000). Secondly, younger people are often found to be more concerned 388 about environmental problems (Howell and Laska, 1992). 389

390

Table 7 reports the results of a latent class model and the hybrid latent class and latent variable model. 391 both of which are specified having two latent classes. The first model is considered the baseline for our 392 analyses. The share of SQ as first choice was roughly 4% and only seven people out of 420 systematically 393 chose the SQ alternative over the 12 choice tasks, therefore protest responses do not seem to be a problem 394 for the survey. In both classes the cost coefficient is negative, which is expected and in line with the 395 economic theory, because it is an indication that, all else being equal, as the cost increases utility decreases. 396 Nevertheless, we note that it is statistically significant only in the second class. In the first class, coefficients 397 for the population of animals is always positive, thus meaning that people are willing to contribute to actions 398

are statistically associated. Measurement equations, linking latent variables to VO questions, are displayed in table 6. Even in this system of equations, it is possible to see that coefficients associated with the latent variable (ξ) are always statistically significant, meaning that latent variables have explanatory power for modelling VOs.

WVO1 indicator							
Parameter	Estimate	Std. Error	sign.				
au	$-9.71 \cdot 10^{-02}$	$1.10 \cdot 10^{-01}$					
δ_1	$-2.21 \cdot 10^{+00}$	$1.20 \cdot 10^{-01}$	****				
δ_2	$-5.62 \cdot 10^{-06}$	$1.76 \cdot 10^{+00}$					
ξ	$3.33 \cdot 10^{-02}$	$1.82 \cdot 10^{-02}$	*				
WVO2 indicator							
Parameter	Estimate	Std. Error	sign.				
au	$-2.90 \cdot 10^{+00}$	$1.34 \cdot 10^{+00}$	**				
δ_1	$-1.53 \cdot 10^{+01}$	$1.18 \cdot 10^{+00}$	****				
δ_2	$-5.56 \cdot 10^{-06}$	$1.28 \cdot 10^{+00}$					
ξ	$1.93 \cdot 10^{+00}$	$2.65 \cdot 10^{-01}$	****				
WVO3 indicator							
Parameter	Estimate	Std. Error	sign.				
au	$-8.32 \cdot 10^{+00}$	$6.05 \cdot 10^{-01}$	****				
δ_1	$-1.06 \cdot 10^{+01}$	$1.01 \cdot 10^{+00}$	****				
δ_2	$1.77 \cdot 10^{-05}$	$1.58 \cdot 10^{+00}$					
ξ	$5.25 \cdot 10^{+00}$	$2.27 \cdot 10^{+00}$	**				
WVO4 indicator							
Parameter	Estimate	Std. Error	sign.				
au	$-3.66 \cdot 10^{+00}$	$6.10 \cdot 10^{-01}$	****				
δ_1	$-3.25 \cdot 10^{+00}$	$4.89 \cdot 10^{-01}$	****				
δ_2	$-2.02 \cdot 10^{-06}$	$7.96 \cdot 10^{-01}$					
ξ	$7.73 \cdot 10^{-01}$	$1.57 \cdot 10^{-01}$	****				
LL	-1379						
$^{****} p < 0.001, ^{***} p$	$\frac{1}{2} + \frac{1}{2} + \frac{1}$						

Table 6: Results of the orientation questions WVO1 indicator

for their conservation. Conversely, the coefficients associated with quadratic number of animals is negative. This result indicates a concave relationship between individuals' utility and number of specimen in the animal populations.

In class number one, significance levels of the estimated parameters is rather poor. In fact, only the ASC 402 coefficient is statistically significant at 5% level. In this class, respondents seem not interested to increase 403 the population of wolves, lynx and salamanders. Probably, the insignificance of parameters is due to the fact 404 that in this class members do not care much about wildlife. In the second class, the trend is different in terms 405 of significance, the estimated parameters are in fact almost all statistically different from zero. The mag-406 nitude of the coefficients is slightly different, indicating different marginal utilities (and perhaps differences 407 in scale) compared to the first class. In general, this model does not appear to be fully satisfactory. Class 408 allocation is only given by a constant term and it does not appear to be appropriate to model heterogeneity 409 in answers. Probably, a simple two class model that does not recognize latent attributes of value orientation 410 is not adequate for explaining preference heterogeneity across respondents. Class probabilities were assessed 411 to be 34 percent for class number one and 66 % for class number two. 412

We now move to the hybrid model. The log-likelihood function of the choice component of the model is 413 higher, suggesting a better fit of the choice data. Similarly, AIC and BIC are smaller than those of the simple 414 latent class. Cost coefficients are negative in both classes, as in the baseline model, but, importantly, are now 415 found to be statistically significant in both classes. In general, it can be observed that the sign of coefficients 416 are the same as in the baseline model, only their magnitude changes. The number of statistically significant 417 parameters is higher. The latent variables for the mutualism orientation was found to have a significant 418 effect in class allocation. This result suggests that the underlying VO has a good explanatory power for 419 determining class membership and therefore preference heterogeneity for the populations of wolves, lynx and 420 salamanders. The coefficient expressing the effect of the mutualism latent variable is negative, suggesting that 421

ParameterEstimateStd. ErrorEstimateStd. Error $Wolf.c1$ $5.51\cdot10^{-3}$ $2.38\cdot10^{-2}$ $2.82\cdot10^{-2}$ $1.28\cdot10^{-2} **$ $Wolf^2.c1$ $-5.90\cdot10^{-5}$ $1.97\cdot10^{-4}$ $-3.28\cdot10^{-4}$ $1.11\cdot10^{-4} ***$ $Linx.c1$ $2.53\cdot10^{-2}$ $2.00\cdot10^{-2}$ $4.10\cdot10^{-2}$ $1.54\cdot10^{-2} ***$ $Linx^2.c1$ $-2.51\cdot10^{-4}$ $1.99\cdot10^{-4}$ $-4.19\cdot10^{-4}$ $1.44\cdot10^{-4} ***$ $Salam.c1$ $1.78\cdot10^{-3}$ $8.70\cdot10^{-5}$ $-7.91\cdot10^{-5}$ $1.08\cdot10^{-2}$ $Salam^2.c1$ $-4.70\cdot10^{-5}$ $7.48\cdot10^{-5}$ $-7.91\cdot10^{-5}$ $9.85\cdot10^{-5}$ Cost.c1 $-2.99\cdot10^{-2}$ $1.89\cdot10^{-2}$ $-8.80\cdot10^{-2}$ $2.03\cdot10^{-2} ****$ Asc.c1 1.09 $5.06\cdot10^{-1} **$ $6.23\cdot10^{-1}$ $4.04\cdot10^{-1}$ $Wolf.c2$ $8.79\cdot10^{-2}$ $1.74\cdot10^{-2} ***$ $5.32\cdot10^{-2}$ $5.30\cdot10^{-3} ****$ $Kolf^2.c2$ $-9.55\cdot10^{-4}$ $2.96\cdot10^{-4} ***$ $-5.43\cdot10^{-4}$ $4.66\cdot10^{-5} ****$ $Linx.c2$ $5.27\cdot10^{-2}$ $2.14\cdot10^{-2} *$ $4.95\cdot10^{-4}$ $4.59\cdot10^{-5} ****$ $Salam.c2$ $1.87\cdot10^{-2}$ $9.83\cdot10^{-3} *$ $-4.85\cdot10^{-4}$ $4.59\cdot10^{-5} ****$ $Salam.c2$ $-1.77\cdot10^{-4}$ $9.87\cdot10^{-5}$ $-1.49\cdot10^{-4}$ $4.7\cdot10^{-5} ****$ $Salam.c2$ $-1.75\cdot10^{-4}$ $9.87\cdot10^{-5}$ $-1.49\cdot10^{-4}$ $4.7\cdot10^{-5} ****$ $Salam.c2$ $-1.7\cdot10^{-4}$ $9.87\cdot10^{-5}$ $-1.49\cdot10^{-4}$ $4.7\cdot10^{-5} ****$ $Salam.c2$ $-1.5\cdot10^{-1}$ $7.7\cdot10^{-2} **$ </th <th></th> <th>Latent class</th> <th></th> <th>Hybrid model</th> <th></th>		Latent class		Hybrid model	
$ \begin{array}{c ccccccccccccccccccccccccccccccccccc$	Parameter	Estimate	Std. Error	Estimate	Std. Error
$\begin{array}{cccccccccccccccccccccccccccccccccccc$	Wolf.c1	$5.51 \cdot 10^{-3}$	$2.38 \cdot 10^{-2}$	$2.82 \cdot 10^{-2}$	$1.28 \cdot 10^{-2} **$
$\begin{array}{cccccccccccccccccccccccccccccccccccc$	$Wolf^2.c1$	$-5.90 \cdot 10^{-5}$	$1.97 \cdot 10^{-4}$	$-3.28 \cdot 10^{-4}$	$1.11 \cdot 10^{-4} ***$
$\begin{array}{cccccccccccccccccccccccccccccccccccc$	Linx.c1	$2.53 \cdot 10^{-2}$	$2.00 \cdot 10^{-2}$	$4.10 \cdot 10^{-2}$	$1.54 \cdot 10^{-2} ***$
$\begin{array}{cccccccccccccccccccccccccccccccccccc$	$Linx^2.c1$	$-2.51 \cdot 10^{-4}$	$1.99 \cdot 10^{-4}$	$-4.19 \cdot 10^{-4}$	$1.44 \cdot 10^{-4} ***$
$\begin{array}{cccccccccccccccccccccccccccccccccccc$	Salam.c1	$1.78 \cdot 10^{-3}$	$8.70 \cdot 10^{-3}$	$6.04 \cdot 10^{-3}$	$1.08 \cdot 10^{-2}$
$\begin{array}{cccccccccccccccccccccccccccccccccccc$	$Salam^2.c1$	$-4.70 \cdot 10^{-5}$	$7.48 \cdot 10^{-5}$	$-7.91 \cdot 10^{-5}$	$9.85 \cdot 10^{-5}$
$\begin{array}{cccccccccccccccccccccccccccccccccccc$	Cost.c1	$-2.99 \cdot 10^{-2}$	$1.89 \cdot 10^{-2}$	$-8.80 \cdot 10^{-2}$	$2.03 \cdot 10^{-2} ****$
$\begin{array}{cccccccccccccccccccccccccccccccccccc$	Asc.c1	-1.24	$4.25 \cdot 10^{-1} ***$	$-5.64 \cdot 10^{-2}$	$6.83 \cdot 10^{-1}$
$\begin{array}{cccccccccccccccccccccccccccccccccccc$	Scale.c1	1.09	$5.06 \cdot 10^{-1} *$	$6.23 \cdot 10^{-1}$	$4.04 \cdot 10^{-1}$
$ \begin{array}{cccccccccccccccccccccccccccccccccccc$	Wolf.c2	$8.79 \cdot 10^{-2}$	$1.74 \cdot 10^{-2} ***$	$5.32 \cdot 10^{-2}$	$5.30 \cdot 10^{-3} ****$
$\begin{array}{cccccccccccccccccccccccccccccccccccc$	$Wolf^2.c2$	$-9.55 \cdot 10^{-4}$	$2.96 \cdot 10^{-4} **$	$-5.43 \cdot 10^{-4}$	$4.66 \cdot 10^{-5} ****$
$\begin{array}{cccccccccccccccccccccccccccccccccccc$	Linx.c2	$5.27 \cdot 10^{-2}$	$2.14{\cdot}10^{-2}$ *	$4.95 \cdot 10^{-2}$	$4.94{\cdot}10^{-3}$ ****
$\begin{array}{cccccccccccccccccccccccccccccccccccc$	$Linx^2.c2$	$-5.49 \cdot 10^{-4}$	$1.85 \cdot 10^{-4} **$	$-4.85 \cdot 10^{-4}$	$4.59 \cdot 10^{-5} ****$
$\begin{array}{cccccccccccccccccccccccccccccccccccc$	Salam.c2	$1.87 \cdot 10^{-2}$	$9.83 \cdot 10^{-3} *$	$1.65 \cdot 10^{-2}$	$4.39 \cdot 10^{-3} ****$
$\begin{array}{cccccccccccccccccccccccccccccccccccc$	$Salam^2.c2$	$-1.17 \cdot 10^{-4}$	$9.87 \cdot 10^{-5}$	$-1.49 \cdot 10^{-4}$	$4.17 \cdot 10^{-5} ****$
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	Cost.c2	$-1.58 \cdot 10^{-1}$	$7.73 \cdot 10^{-2} **$	$-9.00 \cdot 10^{-2}$	$8.87 \cdot 10^{-3} ****$
$\begin{array}{cccccccccccccccccccccccccccccccccccc$	Asc.c2	-5.50	$7.70 \cdot 10^{-1} ****$	-5.27	$9.62 \cdot 10^{-1} ****$
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	Scale.c2	$5.33 \cdot 10^{-1}$	$3.02 \cdot 10^{-1} *$	$9.07 \cdot 10^{-1}$	$1.33 \cdot 10^{-1} ****$
$\begin{array}{ccccccc} \text{IV_dom.cl1} & & -2.92 & 1.83 \\ \text{IV_mut.cl1} & & -4.33\cdot10^1 & 2.58\cdot10^{-1} **** \\ \hline \text{IL} (\text{choice model}) & -4147 & -4013 \\ \text{IL} (\text{global}) & -4147 & -4333 \\ \text{AIC} & 8332 & 8024 \\ \text{BIC} & 8465 & 8017 \\ \text{Observations} & 8184 & 8184 \\ \hline \text{Respondents} & 341 & 341 \\ \end{array}$	$class_constant$	$6.469 \cdot 10^{-1}$	$3.958 \cdot 10^{-1}$	7.48	2.36 ***
$\begin{array}{c c c c c c c c c c c c c c c c c c c $	LV_dom.cl1			-2.92	1.83
LL (choice model) -4147 -4013 LL (global) -4147 -4333 AIC 8332 8024 BIC 8465 8017 Observations 8184 8184 Respondents 341 341	LV_mut.cl1			$-4.33 \cdot 10^{1}$	$2.58{\cdot}10^{-1}$ ****
LL (global) -4147 -4333 AIC 8332 8024 BIC 8465 8017 Observations 8184 8184 Respondents 341 341	LL (choice model)		-4147		-4013
AIC 8332 8024 BIC 8465 8017 Observations 8184 8184 Respondents 341 341	LL (global)		-4147		-4333
BIC 8465 8017 Observations 8184 8184 Respondents 341 341	AIC		8332		8024
Observations81848184Respondents341341	BIC		8465		8017
Respondents 341 341	Observations		8184		8184
	Respondents		341		341

Table 7: Results of the 2 class model and the 2 class hybrid model

****p < 0.001, *** p < 0.01, ** p < 0.05, * p < 0.1

respondents with a mutualism orientation are more likely to belong to the second latent class. Coefficients in this class are larger than class number one, indicating that increasing the number of animals provides higher utility for people in this class. This is also reflected in larger WTPs for the three animals, which will be discussed later. This result was anticipated, because people showing mutualism are expected to be willing to pay more than dominants for wildlife conservation. Prior research suggested that VOs may be predictors of attitudes towards wildlife and wildlife management options; our study suggests that VOs may also influence the economic behaviour of the individual.

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Concerning welfare measures, the quadratic coding for the number of animals of each species allowed a
 non-linear representation of WTP levels for each population size, expressed per person per visit to one of the
 regional park. The calculation of WTPs was computed by means of the following formula:

$$WTP_n = \frac{-(\beta * n + \beta * n^2)}{\beta_{cost}} \tag{7}$$

where *n* is the population size of the species of interest, while β and β_{cost} are, respectively, the estimated coefficients from the choice model for the species of interest and the cost attribute. Table 8 shows estimated WTP with confidence intervals for some representative sizes of the populations. Results of the WTP computation per each animal is shown in figures 2, 3 and 4, together with the confidence intervals calculated through the Krinsky-Robb method, using 5,000 draws (Hole, 2007). WTP is presented separately for the two classes. The continuous lines represents the mean class-specific marginal WTP, while dashed lines the lower

		Class one					
Animal	Ν	Lower	Mean	Upper	Lower	Mean	Upper
	20	0.21	5.52	6.86	6.86	9.56	6.86
Wolf	40	-0.61	7.77	9.99	9.99	14.22	9.99
	60	-2.53	6.73	9.37	9.37	13.98	9.37
	80	-5.91	2.43	4.83	4.83	8.84	4.83
	20	1.56	8.17	19.69	6.44	8.99	12.23
Linx	40	2.00	12.17	29.89	9.65	13.61	18.61
	60	1.35	12.00	30.60	9.53	13.83	19.14
	80	-0.71	7.66	22.30	6.13	9.68	13.86
	20	-2.43	1.42	7.82	1.29	3.08	5.14
Salamander	40	-4.07	1.94	11.81	1.96	4.81	8.11
	60	-5.03	1.55	12.12	1.99	5.19	8.97
	80	-5.49	0.25	8.85	1.22	4.21	7.85

Table 8: WTP and confidence intervals for wolves, lynx and salam anders (in $\textcircled{\mbox{\ }}$



Figure 2: Krinsky-Robb confidence intervals for WTP for conserving wolves, in class one and two (in \in)

Figure 3: Krinsky-Robb confidence intervals for WTP for conserving lynx, in class one and two (in \in)



and upper borders of the confidence interval, estimated at a confidence level of 95 percent. Individual WTPs
in each class follow a similar concave trend, given by the marginal utility, which is increasing until a certain
size of the population and then decreasing. People are willing to contribute with higher amounts to increase
the number of animals until a certain threshold. This might be explained by the fact that people are aware
of the importance of biological conservation but, at the same time, they do not want too many specimen for
public safety or public security reasons. Beyond the threshold, people show a still positive WTP but with a

Figure 4: Krinsky-Robb confidence intervals for WTP for conserving salamanders, in class one and two (in \in)



decreasing trend, until reaching zero WTP at a certain population level, which differ by animal.

From table 8, it can be noticed that, on average, the highest levels of WTP for wolves is found for a popula-446 tion size between 40 and 60 individuals, although confidence intervals are quite large and exact estimates are 447 difficult to establish. Including VOs resulted in consistent differences between classes, in particular for what 448 concern the range of confidence interval, which is always much larger in the first class. In class number one 449 people are willing to pay roughly between $\in 6.70$ and $\in 7.70$ for such population sizes, while in class number 450 two WTP rises to around $\in 14$, with a much smaller interval range compared to class number one. In class 451 one, the lower confidence interval for WTP is very close to zero and turn negative at about 30 wolves, while 452 in class number two WTP becomes negative only in correspondence of about 90 individuals. The maximum 453 WTP for lynx and salamanders may be found for similar population sizes, but with different amount of 454 money. There are not big differences between classes in terms of average WTP for the lynx, in fact people 455 are willing to pay a maximum of \in 12-13.50 in both classes. Eventually, salamanders showed the lowest level 456 of preference from respondents. This result was expected, because the scientific literature on biodiversity 457 evaluation indicates people are generally willing to pay more for mammals than for reptiles and other species 458 (Martin-Lopez et al., 2008). This result is consistent with the similarity principle, stating that man prefers 459 the species most similar to himself (Tisdell et al., 2005; Metrick and Weitzman, 1996). The maximum WTP 460 is between 40 and 60 salamanders, in the first class people are willing to pay a maximum of about $\in 1.50-2$ 461 for a viable population, while in class number two WTP rises up to about $\in 5$ for the same population size. 462 463

In general, it can be observed that respondents that are more likely to be in class number two have higher 464 WTPs for the three animals. These people are also more likely to have a mutualism orientation, their be-465 haviour is therefore explained by a higher positive attitudes towards animals' right of existence. Conversely, 466 WTPs in class number one are, on average, considerably smaller but with a higher variability. Class one 467 is mainly composed by respondents with a domination orientation, whose willingness to conserve wildlife is 468 typically smaller than for people with a mutualism orientation. For this reason, the smaller average WTP is 469 understandable. The high variability of the confidence interval might be explained by the fact that, according 470 to answer to value orientation questions (table 4), few people showed a domination orientation, which may 471 have increased the variance of answers to choice tasks. At the same time, the relative small interest towards 472 wildlife might have caused a small engagement in taking the survey, thus leading to a higher variability of 473 answers (Hess and Stathopoulos, 2013). 474

Results from this study suggest that VOs are important to explain heterogeneity in preferences, bringing to a more accurate estimation of WTP for wildlife conservation. The use of a latent variable model was justified by the need of avoiding endogeneity, which is demostrated to have relevant consequencies in CE studies (Budziski and Czajkowski, 2017; Hoyos et al., 2017). Considering VOs in wildlife management has interesting policy implications. Understanding how orientations of tourists affect preferences about wildlife facilitates an *ex-ante* evaluation of the effects of new management strategies. Such information is important to communicate management decisions effectively and reduce the probability of conflicts. The human-wildlife conflict

452 is the most substantial limitation to wildlife conservation policies (Dickman, 2010; Kubo and Shoji, 2014)

and can influence a number of economic and recreational activities related to wildlife (Burns, 2004a). Un-483 derstanding the individual factors and context for human-wildlife relationships is useful for the development 484 of conservation management actions that depend on public support (Dietsch et al., 2016). The segmentation 485 approach presented in this paper can be a useful tool for understanding target audiences and to improve 486 WTP estimates. Results can provide guidance on the social acceptability of management interventions and 487 can help the identification of suitable tools aimed at increasing social consensus. They can support public 488 decision-makers in the modulation of wildlife management policies for balancing the needs of conservation 489 and outdoor recreation, explicitly considering existing human-wildlife conflicts. 490

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⁴⁹² Our study was limited to the evaluation of preferences among tourists, however another critical issue in ⁴⁹³ the human-wildlife interactions is represented by opinions of the locals. Local communities play an important ⁴⁹⁴ role in conservation and, at the same time, are also a group that could be affected by decisions on natural ⁴⁹⁵ resource management (Paletto et al., 2014; Parkins and Mitchell, 2005). VOs and preferences of the local ⁴⁹⁶ may be different, therefore understanding opinions of a wider audience would lead to better decisions.

497 6. Conclusions

This paper discussed a methodology to include tourists' VO towards wildlife in a choice model. A system 498 of equations, in which indicator equations and choice models are linked by means of two latent variables, was 499 created in order to include VOs questions in the model, thus overcoming problems of endogeneity. Latent 500 variables, representing mutualism and domination orientations, entered the indicator model as explanatory 501 variables and the choice model as a variable affecting class membership of the respondents. The procedure 502 was tested in a case study in the Italian Alps, the province of Trento, where a natural return of wolves 503 and lynx from neighbouring areas was experienced in the recent years. Moreover, the area is inhabited by 504 a unique sub-species of salamander, called salamander of Aurora. It important to consider some aspects of 505 this study when interpreting the results. In particular, a real management of wildlife might have problems 506 related to correlated population sizes for the animals we considered. In addition, the use of a hypothetical 507 null alternative as opt out could be avoided in future studies if a SQ bias is expected not to influence results. 508 We also recognize that wildlife management should be done with the investigation of preferences of all stake-509 holder groups, therefore local communities should also be surveyed for an effective policy-making. 510 511

512 Results showed that latent variables describing domination and mutualism orientations have a significant effect on latent class allocation. Differences among classes were found in the marginal utilities associated with 513 each attribute. In terms of WTPs, the two classes were proved to be different concerning the maximum level 514 of WTP, in the range of confidence intervals and in the maximum acceptable number of specimen. In class 515 one, people showed the highest average WTP for lynx and then for wolves. WTP for wolves was assessed to 516 be around $\in 12-13.50$ for a population of about 40–60 animals, while WTP for wolves was about $\in 6.70-7.70$ 517 for the same range of population size. In class two, people showed higher WTP for the wolves rather than 518 lynx, ie. \in 14, while the average maximum WTP for lynx is almost the same as in class one. In both classes, 519 WTP for salamander is considerably lower than for the two mammal species. Confirming previous literature, 520 our results demonstrated the positive attitude of the majority of tourists towards wildlife, indicated by the 521 size of the class where mutualists are more likely to belong and by the amount they are willingness to pay for 522 increasing specimens. This finding is significant because it confirms that VOs are relevant when one wishes 523 to explain heterogeneity in WTP and should be included to improve welfare analysis. In addition, our study 524 further provides evidence of a societal shift from domination to mutualism. 525

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The characterization of tourists that the model proposed is useful not only to circumvent statistical problems such as endogeneity, but also for decision-makers, to tailor effective conservation policies. Humanwildlife conflict is the most substantial limitation to wildlife conservation policies (Dickman, 2010; Kubo and Yasushi, 2014) and can influence a number of economic and recreational activities related to wildlife (Burns, 2004b).The described hybrid model is able to provide a description of preferences based on individual characteristics, so that preference heterogeneity can be effectively modelled.

What emerged from the study is that lynx and wolf are seen by tourists as symbols of the Alps and their 534 presence probably is one of the major reasons that make Trentino one of the strongest national attractions in a 535 naturalistic environment. This phenomenon may have some policy implications on economic and recreational 536 activities related to wildlife, such as wildlife tourism. One of the key points for a successful wildlife tourism 537 experience is the attitude of tourists towards wildlife (Reynolds and Braithwaite, 2001). Our psychological 538 and economic results show that wildlife tourism can be developed in Trentino. Wildlife tourism could bring 539 further benefits in terms of environmental education and awareness both for tourists and local population, 540 helping in reducing the human wildlife conflict, which in recent years has increased in Trentino. As the case of 541 the salamander aurora shows, a lack of knowledge and familiarity with a particular wild species can affect the 542 preferences of tourists and therefore availability to support any protection and conservation plans. Tourist 543 experience based on observation, familiarization with the characteristics and habits of these wild animals, 544 can produce positive effects both for sustainable tourism development and wildlife conservation. 545

⁵⁴⁶ 7. Acknowledgements

⁵⁴⁷ We are thankful to professor J.J. Vaske for sharing the questionnaire on value orientation and to Valentina ⁵⁴⁸ Manini, Serena Nichele and Alessio Tamanini for the help with data compilation.

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