Transferring landscape values using discrete choice experiments: Is meta-analysis an option?

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SUMMARY: Discrete Choice Experiments (DCEs) are an increasingly used valuation method to guide policy-makers in their landscape management decision-making. However, due to time and budget constraints, conducting an original DCE for every single policy decision may not be possible in practice. In this context, meta-analyses are becoming a popular tool for supporting benefit transfer in environmental valuation. This paper raises the question whether meta-analysis is an option for transferring landscape values using DCEs within Europe. We argue that the use of specific and different attributes, definitions and units of measurement makes it hard to compare willingness-to-pay values across current DCE studies.

KEYWORDS: Discrete choice experiment, landscape valuation, meta-analysis, review, willingness-to-pay.

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1. Introduction

Landscapes are continuously changing and evolving through natural and human-induced processes and activities. Landscape changes are often characterised by the loss of diversity and identity, which brings an urgent need to preserve and develop landscapes sustainably. In this context, measuring and quantifying the economic value of changes in landscapes according to social preferences would guide policy-makers in landscape management decision-making. This practice is known as landscape valuation, aiming at aiding welfare optimization by incorporating public preferences that are not appropriately reflected by market mechanisms.

Since landscapes are complex environmental goods involving several attributes, the Discrete Choice Experiment (DCE) method is growing rapidly to enrich landscape management decision-making (Adamowicz et al., 1998; Bennett and Blamey, 2001; Bateman et al., 2002). However, carrying out an original valuation study, such as a DCE, for every single landscape management policy decision may not be feasible in practice. It would not match the time and budget constraints of most policy-evaluation exercises. There is often interest from policy-makers in transferring the results from previous and similar studies rather than conducting new ones, this being faster and cheaper. This practice is known as benefit transfer (BT), although a more general term would be value transfer (Rolfe, 2006; Navrud and Ready, 2007).

BT is the application, with the necessary adjustments, of valuation information from an original study (i.e. the study site), or multiple studies, to a different context, where such information is required to evaluate a new policy (i.e. the policy site). Over the last few years there has been substantial research regarding the accuracy and validity of BT, in particular investigating in which situations environmental value transfer is valid and reliable (Muthke and Holm-Mueller, 2004; Brouwer and Bateman, 2005; Rosenberger and Stanley, 2006; Rolfe, 2006; Colombo et al., 2007; Santos, 2007; Scarpa et al., 2007; Colombo and Hanley, 2008; Johnston and Rosenberger, 2010; Bateman et al., 2011; Martin-Ortega et al., 2012). Two main BT approaches have been followed in the environmental valuation literature (Navrud and Ready, 2007; Pattanayak et al., 2007; Colombo and Hanley, 2008): (i) unit value transfer (simple unit transfer and unit transfer with adjustments) and (ii) function transfer (benefit function from one study and meta-analysis).
The focus of this paper is on the meta-analysis, which refers to the study of studies (Barrio and Loureiro, 2010). More formally, it is the statistical analysis of a large collection of analysis results from individual studies for the purpose of integrating the findings (Glass, 1976). Meta-analysis has been mainly developed in medicine, psychology, marketing or education and it is gaining popularity in the field of environmental economics. According to Shrestha et al. (2007), BT using meta-analysis has several advantages. First, information from multiple studies can be added, thus providing more rigorous measures of a central tendency than can usually be derived from any single study. Second, methodological differences in the original studies can be controlled for when estimating a value from the meta-regression equation. And third, by setting the explanatory variables specific to the policy site, the analyst can potentially account for the heterogeneity in the policy site features. Given this, and compared with the other BT methods, the benefit estimate based on meta-analysis might be a better approximation of the value of the resource at the policy site (Rosenberger and Loomis, 2000; Shrestha et al., 2007).

Most of the meta-analyses concerning environmental valuation have been based on Contingent Valuation (CV) studies so far, despite the increasing application of DCEs in this field. This raises the question whether meta-analysis is an option for transferring landscape values using estimates from DCEs. Thus, the objective of this paper is to analyse and discuss the opportunities to use the results from DCE studies in meta-analysis regarding landscape values in Europe. The paper is structured as follows. Section 2 presents meta-analysis methodology and reviews meta-analysis studies in environmental valuation. Section 3 discusses the suitability of European DCEs for landscape valuation for meta-analysis and Section 4 provides discussion and the main conclusions of the paper.

2. Meta-analysis methodology

An explicit specification of the meta-regression model can be described generally as follows:

\[ y_{in} = \alpha + \beta_1 x_{1,in} + \beta_2 x_{2,in} + \cdots + \beta_K x_{K,in} + \varepsilon_i + u_n, \]  

where \( i \) subscript denotes which candidate study the value estimate comes from \((i=1, 2, \ldots, I)\), \( n \) denotes the value estimate reported in the study \((n=1, 2, \ldots, N_i)\) and \( x_{1,in}, x_{2,in}, \ldots, x_{K,in} \) refer to the explanatory variable \( x_{k,in} \) \((k=1, 2, \ldots, K)\). In the case
where each study $i$ provides a single estimate $n_i$, then $N_i=1$ and $\varepsilon_i$ collapses into $u_n$. However, in the case where a study gives more than one value estimate, then it is necessary to account for the common error across estimates ($u_n$) and the group-specific or panel error within a study ($\varepsilon_i$). The total number of estimates is $N=\sum_{i=1}^{I} N_i$. Thus, the dependent variable $y_{in}$, is the estimate of the $n$-th value for the environmental change found in study $i$ (often a willingness-to-pay, WTP, per unit change for the resource change specified), $\alpha$ represents an intercept term and $\beta_1, \beta_2, \ldots, \beta_K$ are the coefficients to be estimated associated with the explanatory variable $x_{k,in}$ (e.g. amenity characteristics, socio-economic characteristics, methodological characteristics). There are several approaches to estimating this model depending on assumptions regarding the error variance-covariance matrix (Lindhjem, 2007; Nelson and Kennedy, 2009). The most common specifications of the functional form of the regression equation are linear, double-log, semi- and trans-log (Johnston et al., 2005).

After a wide bibliographic review, meta-analysis for BT should only rely on a subset of studies which fulfil the necessary conditions (see Boyle et al., 2009). Colombo and Hanley (2008) found that BT errors depend on the selection of the study sites and that additional data collection does not always reduce transfer errors. There are some specific challenges in performing a successful meta-analysis, including the difficulties of ensuring commensurability across data sets, the variation in methods and approaches because the experiments are not controlled, limited data sets and inadequate methods of analysis (Florax et al., 2002; Smith and Pattanayak, 2002; Nelson and Kennedy, 2009; Johnston and Rosenberger, 2010). An essential requirement is to define clearly and accurately the environmental issue of interest. To be comparable across studies analysed, it is very important to ensure that the dependent variable measures from the primary studies are all measuring the same thing or concept and in a common unit.

### 2.1. Meta-analysis in environmental valuation

Recent meta-analyses have been conducted in the field of environmental valuation. Table 1 classifies meta-analysis studies conducted recently according to the area of application, valuation methods, the number of value observations, as well as the definition of the dependent variable of the meta-regression model. When facing the challenge of defining a common dependent variable for the policy site, analysts should
reconcile commodity definitions across the studies selected. For this, analysts often add explanatory variables to the right hand side of the equation [1]. For example, in Ghermandi et al. (2008), the wetland value estimates per hectare and year are for different wetland types, sizes or goods and services provided. Similarly, in Lindhjem (2007), WTP estimates per household and year for protection of forests are for different dimensions of geography, country, time of year or scope (forest area).

Another challenge related to the definition of the dependent variable is its measurability. Even some studies of seemingly similar non-market goods or services may estimate values for differing underlying quantities or qualities. For example, Johnston et al. (2005) reconciled measurements of water quality change by mapping all water quality measures to the Resources for the Future (RFF) water quality ladder. In those studies in which such information was not provided, they used descriptive information to approximate the baseline level of water quality and the magnitude of the change. Likewise, Van Houtven et al. (2007) converted the specified water quality changes into a common metric by constructing a 10-point water quality index (WQI10). Alternatively, Liu and Stern (2008) addressed the marginal changes in the valuation of ecosystem services by adding dummies as explanatory variables regarding the type and degree of the change. It is noteworthy that in all cases the process of converting comparable commodity definitions and measurements across studies required subjective judgments and involved uncertainty. Furthermore, in many cases assumptions required to reconcile commodity definitions and measurements are not clearly specified in the studies reviewed.

Analysts should also specify whether Marshallian consumer surplus or Hicksian compensating surplus is used to define the dependent variable, which depends on the valuation method(s) considered. As can be seen from Table 1, six meta-analyses (out of 17) incorporated results from different Stated Preference (SP) and Revealed preference (RP) valuation methods, among others (e.g. Ghermandi et al., 2008; Brander et al., 2007; Brander et al., 2006). However, this type of analysis is limited and leads to criticism that values may not be commensurable (Smith and Pattanayak, 2002; Bergstrom and Taylor, 2006; Johnston and Rosenberger, 2010). Selecting estimates derived from similar valuation methods might be advantageous to ensure a common concept for the dependent variable (Van Houtven et al., 2007). The remaining eleven meta-analysis studies are based on the same valuation approaches or just on one
valuation method. According to Table 1, seven meta-analyses draw value estimates from only one type of valuation method, where the majority (6 studies) are based on studies using the CV method (e.g. Santos, 2007; Barrio and Loureiro, 2010; Johnston et al., 2005).

However, meta-analyses using estimates from DCEs for BT seem to be still uncommon. To the best of our knowledge, the only study doing BT through meta-analysis using only estimates from DCEs was recently carried by Rolfe and Brouwer (2012). In order to generate a consistent base for data pooling and analysis, only WTP estimates for one common attribute (river health) across the 19 collected DCEs in Australia were chosen, obtaining in total 145 WTP estimates relating to river health. The implicit prices of different measures and scales of river health were transformed into a common standard of WTP per kilometre of river in good health in order to ensure comparable value estimates across studies.
<table>
<thead>
<tr>
<th>Area of application</th>
<th>Reference</th>
<th>Valuation method(s)</th>
<th>Number of observations</th>
<th>Dependent variable</th>
</tr>
</thead>
<tbody>
<tr>
<td>Wetland</td>
<td>Brouwer et al. (1999)</td>
<td>CV</td>
<td>130 (92)</td>
<td>Log of WTP per household and year in SDR (1995) for the preservation of specific wetland aspects</td>
</tr>
<tr>
<td></td>
<td>Woodward and Wu (2001)</td>
<td>CV, HP, RC, TC, NFI</td>
<td>65</td>
<td>Log of wetland value per acre and year in US$ (1990)</td>
</tr>
<tr>
<td>Recreation</td>
<td>Bateman and Jones (2003)</td>
<td>CV, TC</td>
<td>77</td>
<td>Recreation value per person and visit in £ (1990) for open-access woodland in Great Britain</td>
</tr>
<tr>
<td></td>
<td>Shrestha et al. (2007)</td>
<td>GR, PF, NFI, TC, HP, CV</td>
<td>100 (73)</td>
<td>Log of the coral reef recreation value per visit in US$ (2000)</td>
</tr>
</tbody>
</table>
|                    | Zandwensen and Tol (2009) | TC                        | 251 (189 and 166)     | A) Log of consumer surplus per trip in € (2000) of forest recreation  
B) Consumer surplus per trip in € (2000) normalized for size of forest (per hectare) of forest recreation |
| Water              | Johnston et al. (2003) | CV                      | 29                     | Log of WTP per household and year in US$ (2002) for surface water improvements |
|                    | Johnston et al. (2005) | CV                      | 81                     | Log of WTP per household and year in US$ (2000) for water improvements in aquatic habitat |
|                    | Van Houtven et al. (2007) | CV, CA                  | 131                    | (Log) WTP per household and year in US$ (2000) for water quality change |
| Forest             | Lindhjem (2007) | CV, DCE (only two)      | 72                     | (Log) WTP per household and year in NOK (2005) for protection and multiple use forestry |
B) Log of median WTP in US$ (2008) for forest management programmes |
| Species            | Richardson and Loomis (2009) | CV, DCE (only one)      | 67                     | (Log) WTP per household and year in US$ (2006) for the preservation of threatened, endangered and rare species |
| Ecosystem services | Liu and Stern (2008) | CV                      | 120                    | (Log) WTP per household and year in US$ (2006) for coastal ecosystem services |
| Agricultural landscape | Santos (2007) | CV                      | 61 and 32              | Log of WTP per household and year in £ (1996) for the conservation of agricultural landscapes |

Notes:
In some studies the number of observations is reduced from the total number of value observations because of missing values for explanatory variables. In these cases, the final number of observations used in the meta-regression model is specified in brackets.
When different numbers of observations are specified (e.g. 61 and 32) this is because different meta-models (with different numbers of observations) were estimated.
When Log is in brackets this means that the dependent variable was also initially specified as non-log (i.e. linear specification).

Abbreviations:

Source: Own elaboration.
3. Suitability of discrete choice experiments for meta-analysis regarding landscape values

The data used to discuss the opportunity to perform a meta-analysis concerning landscape values comes from a literature review of DCEs published between 1997 and 2012 for valuing landscape changes. We considered studies that applied DCE in order to value landscapes or landscape elements, such as agricultural landscape, mountain, forest and wetland, or to value a mixture of landscapes (a particular area or region) within Europe. The review considered 87 studies including published articles as well as “grey literature”.

Definition of the dependent variable proved problematic. Our primary DCE studies involved slightly different aspects of landscapes (i.e. different attributes), ranging, for example, from landscape aesthetic, air pollution, or cultural heritage to endangered wildlife, hedgerows and recreational facilities. So, we encounter the first issue when carrying out a meta-analysis using these DCEs: the wide heterogeneity in the attributes used to define the landscape. The most common attributes found across DCEs reviewed are related to biodiversity, landscape in general, forest, recreation, water, grassland, hedgerows, cultural heritage, field boundaries and moorland. Attributes related to biodiversity and to landscape are the most commonly used (used by 39 and 19 studies, respectively).

Richardson and Loomis (2009) conducted a meta-analysis concerning the value of threatened, endangered and rare species (see Table 1). By adding an explanatory variable at the right-hand side of the meta-regression [1], they were able to control for the percentage change in the species population, among other factors. However, transferring values from one landscape to another requires more than just transferring estimates for biodiversity, for example. That is, the value of the landscape is more than transferring the value of the attributes of single landscape elements. But the question is how to conduct a meta-analysis for transferring landscape values using DCEs which involve very specific and different attributes and levels. In the case of Santos (2007), a meta-analysis concerning agricultural landscapes was carried out using only CV methods and characterising the studies with respect to (i) the landscape change (or programme), (ii) the surveyed population and (iii) the method applied, finally having a 61 value observations for the meta-regression model estimation (see Table 1).
But in our case, the DCEs considered involve not only different landscape types but also many specific and different attributes with different descriptions. Thus, using compensating surplus estimates instead of marginal WTP value estimates as the dependent variable was not found to be appropriate because of both the difficulty in finding future management scenarios that were consistent across DCE studies and the variation in attributes between primary studies. In this context, similarly to Rolfe and Brouwer (2012), we decided to analyse the suitability of meta-analysis based on DCEs for landscape valuation focusing on the general landscape attribute as used in 19 studies.

Table 2 reports detailed information about landscape attribute definitions and levels (units of measurement) used in the 19 DCE studies reviewed (see Appendix for a complete listing of these DCE studies along with their code). But even when restricting ourselves to the landscape attribute, we note that it is arduous to ensure commensurability across data sets and that too many (subjective) assumptions would be necessary. First of all, not all of these attributes are measuring the same concept; that is, they are defined in different ways, such as the impact of a project in L1 and L7; landscape view in L2-L6 and L12; landscape desertification level in L10; size of grassland and possibility of surrounding land in L13; the percentage of surface area on which today’s landscape could be seen in the future in L14, or scenic quality in L17 (see Table 2).

In order to make some definitions compatible, one possible approach could be to group together some definitions or terms that may be considered as consistent with indicators, for example, of landscape view or configuration (landscape view in L2-L6 and L12; landscape aesthetic in L8; proportion of native forest and open moor in L9; landscape diversity in L15; intervention in the scenic quality in L17; possible landscapes in L18 and size of forest in the landscape in L19). Then, some dummy variables would be necessary in the meta-regression model to capture the characteristics of the attribute, such as the landscape element to be present (e.g. forest, lake, grassland).

Another possibility could be to group the landscape attributes consistent with landscape protection/conservation (protection of the landscape in L1; landscape desertification level in L10; conservation of forest and/or grassland landscape in L11 and today’s landscape which could be seen in the future in L14). This concept reconciliation, however, would only be possible with some of the studies.
### TABLE 2
DCEs with landscape attribute

<table>
<thead>
<tr>
<th>Code</th>
<th>Reference</th>
<th>Named as</th>
<th>Definition</th>
<th>Levels</th>
</tr>
</thead>
<tbody>
<tr>
<td>L1</td>
<td>Álvarez-Farizo and Hanley (2002)</td>
<td>Landscape protection</td>
<td>Impact of a wind farm project</td>
<td>Protection; Loss</td>
</tr>
<tr>
<td>L2, L3</td>
<td>Baumgart (2005)</td>
<td>Landscape view</td>
<td>Forest land (trees)</td>
<td>A lot; Medium; As today</td>
</tr>
<tr>
<td>L4</td>
<td>Baumgart (2005)</td>
<td>Landscape view</td>
<td>Lake for storing water</td>
<td>Large; Small; As today</td>
</tr>
<tr>
<td>L5, L6</td>
<td>Baumgart (2005)</td>
<td>Landscape view</td>
<td>Hotels and apartments</td>
<td>A lot; Medium; As today</td>
</tr>
<tr>
<td>L7</td>
<td>Bergmann et al. (2006)</td>
<td>Landscape impact</td>
<td>Visual impact of the renewable energy project</td>
<td>None; Low; Moderate; High</td>
</tr>
<tr>
<td>L8</td>
<td>Borresch et al. (2009)</td>
<td>Landscape aesthetic</td>
<td>Landscape scenarios</td>
<td>Status Quo; Multifunctionality scenario; Grassland-dominated scenario; Intensity scenario; High price scenario</td>
</tr>
<tr>
<td>L9</td>
<td>Bullock et al. (1998)</td>
<td>Landscape</td>
<td>Proportion of native forest and open moor</td>
<td>Deer in high open mountain scenery; Deer in both open mountain areas and Caledonian Pine forest; Deer only in low-density Caledonian Pine forest</td>
</tr>
<tr>
<td>L10</td>
<td>Colombo et al. (2005)</td>
<td>Landscape desertification</td>
<td>Desertification level</td>
<td>Deterioration due to desertification; Small improvement: reducing desertification risks in high erosion areas; Moderate improvement: reducing risks in all areas</td>
</tr>
<tr>
<td>L11</td>
<td>Domínguez-Torreiro and Solís (2011)</td>
<td>Rural landscape</td>
<td>Conservation level of forest and/or grassland landscape</td>
<td></td>
</tr>
<tr>
<td>L12</td>
<td>Elsasser et al. (2010)</td>
<td>Landscape view</td>
<td>Typical landscapes of the region in the summer and winter aspect</td>
<td>Situations with different forest types (pine forest, broadleaved forest or mixed forest) with low or high structural diversity; Situation without any forest (grassland only)</td>
</tr>
<tr>
<td>L13</td>
<td>Hasund et al. (2011)</td>
<td>Surrounding landscape and size of grassland</td>
<td>Size of grassland and possibility of surrounding land</td>
<td>Grassland small and isolated; Grassland large and surrounded by other similar land</td>
</tr>
<tr>
<td>L14</td>
<td>Hoyos et al. (2009)</td>
<td>Landscape</td>
<td>Percentage of surface area on which today's landscape could be seen in the future</td>
<td>40%; 60%; 80%; 100%</td>
</tr>
<tr>
<td>L15</td>
<td>Meyerhoff et al. (2009)</td>
<td>Landscape diversity</td>
<td>Diversity level</td>
<td>Low; Medium; High</td>
</tr>
<tr>
<td>L16</td>
<td>Moran et al. (2004)</td>
<td>Landscape and access</td>
<td>Changes regarding landscape and access</td>
<td>Current practices; Enhance landscape appearance; Enhance public access</td>
</tr>
<tr>
<td>L17</td>
<td>Rulz et al. (2008)</td>
<td>Landscape</td>
<td>Intervention in the scenic quality</td>
<td>Clear-cutting without replanting; Bush replanting; Sample replanting</td>
</tr>
<tr>
<td>L18</td>
<td>Schmitz (2006)</td>
<td>Landscape</td>
<td>Possible landscapes</td>
<td>Only forest; Grassland dominated; Mulch dominated; Arable land dominated; Arable and grassland balanced</td>
</tr>
<tr>
<td>L19</td>
<td>Schmitz et al. (2003)</td>
<td>Landscape</td>
<td>Percentage of forest in the landscape</td>
<td>Only forest; 75%; 55%; 35%; 5%</td>
</tr>
</tbody>
</table>

Note: Baumgart (2005) described five different DCEs (with varying attributes) in the same publication, so each DCE has been treated in the analysis as an individual reference (L2, L3, L4, L5 and L6).

Source: Own elaboration.
The next step required, if we are to continue discussing the suitability of carrying out a meta-analysis with DCEs valuing landscapes, is to identify for each DCE study the attributes’ levels used to obtain the WTP estimates. At this point, the task of defining a common dependent variable becomes even more laborious, given the use of different attributes’ levels measuring very specific changes in the landscape. Thus, the main issue to address is that the changes in landscapes are measured differently in different DCE studies.

The attributes concerning landscape view or configuration present very specific levels. For instance, as can be seen in Table 2, studies L2-L3 and L5-L6 measure the presence of trees and hotels, respectively as a lot, medium and as today; study L8 describes different landscape scenarios (multifunctionality, grassland, intensity scenario, high price scenario); study L12 shows different pictures with either a pine forest, a broadleaved forest, a mixed forest with low or high structural diversity or grassland only; study L17 presents different replanting levels (clear-cutting, bush, sample) and study L19 presents the percentage of landscape covered by forests. Furthermore, some of these studies show images or pictures in the choice cards to describe visually the possible changes in landscapes (studies L2-L6, L12, L18, L19). Finally, in one of these studies related to landscape view, linear coding is used to address changes in landscape diversity in the econometric model (study L15), compared with other studies that use dummy variables in their econometric model specification (e.g. L8, L12, L17, L18, L19).

Taking into account all the above stated issues, the estimate values are usually incommensurable and hence incomparable across studies. Based on certain assumptions, it might be possible to apply some restrictions. For example, regarding studies on landscape view, we could take into account only those WTP estimate values that have been measured qualitatively according to some landscape scenario, such as forest, lake or grassland. However, this would be feasible only with a few studies (L2-L6, L8, L12, L18). All in all, after several exclusions to reconcile first of all the commodity definitions and then the WTP units of measurement, a limited number of WTP value observations remain for the estimation stage of our data. Moreover, given that many independent variables would have to be included in the meta-regression model distinguishing the characteristics of the change under valuation (i.e. one for each
type of landscape to be present) and regarding methodological and socio-economic factors, degrees of freedom would be lost.

In addition, as stated by Bergstrom and Taylor (2006), on the one hand, accounting for commodity inconsistency by adjusting the dependent variable of the meta-regression model before the estimation may be inadequate if it essentially alters the nature of what people valued. On the other hand, addressing commodity inconsistency by introducing explanatory variables may also be problematic because it inserts restrictions into the model that may or may not be plausible. As pointed out by Colombo and Hanley (2008), the analyst should check beforehand whether the study site is suitable or not for the corresponding analysis.

4. Discussion and conclusions

There is growing interest in the potential for producing meta-analyses in the environmental valuation field which do not rely upon expensive and time-consuming survey work and also help to statistically explain the variation in the values obtained in different studies. The increasing number of landscape DCE publications in recent years shows that the use of DCEs is seen as a promising method for landscape valuation. However, we could not identify any meta-analysis using compensating surplus measure estimates derived from the DCE. Due to the inconsistency in potential management scenarios, as well as the variation in attributes across DCE studies, the meta-analysis studies focused on just one common attribute to collect value estimates from DCEs, therefore losing information (e.g. Rolfe and Brouwer, 2012).

In this paper we discuss whether meta-analysis is an option for transferring landscape values from DCEs. We found that the opportunities to use the results from DCE studies in meta-analysis concerning landscape values in Europe are quite limited, given the issues related to commodity inconsistency and incommensurability across the data sets reviewed. We also discarded the use of compensating surplus estimates, given both the difficulty in establishing future management scenarios in agreement with the DCEs reviewed and the use of specific and different attributes across primary studies. Even focusing on the landscape specific attribute, we found that the use of different definitions and levels to describe the landscape under valuation makes it hard to compare WTP values across studies.
Many meta-analysts do not pay enough attention to ensuring comparable commodity definitions and metrics, and consequently often pool inconsistent measures (Smith and Pattanayak, 2002; Nelson and Kennedy, 2009). Furthermore, in many of the meta-analyses reviewed it is not clearly specified how commodity definitions and measurements reconciliations are achieved. In our case, the definition and measurement of the dependent variable of the meta-regression model is problematic and several subjective assumptions and explanatory variables for possible reconciliation would be required, leading to the loss of specific information. It is important to reflect on the extent to which the assumptions made for meta-analysis are in line with what the primary studies had in mind.

As one exception, a reasonable reconciliation between attribute definitions and units of measurement was possible in Rolfe and Brouwer (2012) because they collected DCE studies valuing rivers in Australia that were conducted by a small community of researchers using similarly defined attributes and levels. It might be the case that once one study started to cover river health in its DCE (and usually measured per kilometre), others tried to follow the same pattern. However, this is not the case with European DCE studies valuing landscapes, where each study seems to select specific attributes and levels depending on the particular landscape to be valued and main objective of valuation.

There is still no consensus on the concept through which the quantity (or quality) of certain landscape changes may be defined and measured. In the words of Johnston and Rosenberger (2010), “without consensus, decisions made during benefit transfer research or policy applications to reconcile attribute definitions will probably continue to be perceived as ad hoc”. Some advances have been made to find a common way to address water quality in Europe by the introduction of the Water Framework Directive (WFD) in the European Union (Martin-Ortega, 2012; Martin-Ortega et al., 2012). So, with enough DCEs addressing water quality following the WFD, meta-analysis might be possible for water quality improvements.

However, there is still no agreement on a set of criteria for the selection of certain landscape attributes to be used in DCEs, their definition and units of measurement. An additional point is whether landscape values are comparable across different studies at all. The term landscape is subjective and becomes a mental and social construct with different values and meanings linked to it. Although from the meta-analysis point of
view a consensus could be useful to ensure a common dependent variable, such standardization may not be desirable for specific landscape changes under valuation, since landscape is closely related to concepts such as identity, character and place-making, which also carry many symbolic meanings (Antrop et al., 2013). As pointed out by Brouwer and Spaninks (1999), the historical context in which estimate values have emerged may be one of the causes behind the difficulty of establishing an unequivocally valid BT in the literature, even in the ideal case where the environmental goods and the sites where they are found and the development scenarios are identical.

In conclusion, currently estimates from DCEs seem to be unsuitable for transferring landscape values through meta-analysis in spite of their increasing popularity. A limitation of this study might be its generic aim of transferring landscape values without focusing on a particular type, and consequently, the large heterogeneity in the studies considered, all of which are site specific. The main consequences for transferring values when meta-analysis might not be an option are that fewer results could be incorporated in the analysis, the possibility of controlling for effects such as sample size and heterogeneity would be restricted and methodological effects on values (e.g. number of choice cards, alternatives or attributes, payment vehicle) in DCE studies may not be distinguishable. Unlike the BT function approach, where extensive research has been carried out on improving and testing the accuracy of its transfer (Morrison and Bergland, 2006), further work is needed to evaluate the applicability of the meta-analysis BT approach based on DCEs.
Appendix

List of DCE studies reviewed with landscape related attribute(s):


References


