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Violations of scale compatibility: results from pricing tasks and choice tasks in choice experiments

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Violations of scale compatibility: results from pricing tasks and choice tasks in choice experiments

Abstract

According to rational choice theory, preference orderings should be invariant with respect to the elicitation procedure. The contingent trade-off model (Tversky et al., 1988), however, argues that if an attribute of decision alternatives is also used as response mode (scale compatibility), then the attribute will be weighted more heavily. We propose a two-stage design for a systematic assessment of how scale compatibility impacts willingness to pay (WTP) estimates in choice experiments. At the first stage, a pricing task is implemented. Respondents face two alternative goods. For one good, the price is given. For the alternative good, respondents state the price (WTP) that makes them indifferent between the alternatives. In the second stage, a choice task is implemented: another group of respondents makes pairwise choices between two alternative goods – with the price being one of several attributes. Our empirical findings support the contingent trade-off model, as pricing tasks yield systematically lower WTP estimates than choice tasks. While the trade-offs between the other attributes do not differ between the pricing and the choice task. Thus, in the choice task the weight shifts away from the price attribute but does not change the relative weight of the other attributes.

Key words: willingness to pay, preference heterogeneity, procedure invariance, scale compatibility, stated preference elicitation

JEL classification: D10, D60, Q50

1 Introduction

According to rational choice theory, observed preference orderings should not depend on the underlying preference-elicitation procedure (procedure invariance). The experimental literature, however, revealed systematic violations of procedure invariance. In this literature, one important line of research, following Tversky and Kahneman (1991), studies the phenomenon of reference dependence. According to reference dependence people perceive changes from a given reference point as gains or losses. Another important line, following Slovic (1975) and Tversky et al. (1988), study scale compatibility, meaning that people weigh attributes of decision alternatives more heavily that are also used as response mode (Bleichrodt and Pinto, 2002). As a result, “payoffs are weighted more heavily in pricing than in choice” (Tversky et al., 1990, 204) in lotteries and gambling situations.

The present paper focuses on scale compatibility. Scale compatibility is well-documented in the experimental literature on decisions under risk (e.g., Tversky et al., 1990). We, however, investigate its implications for stated preference methods used for the valuation of non-market and public goods. Particularly, we explore its consequences for deriving willingness to pay (WTP) estimates from choice and pricing tasks. In our choice tasks respondents choose among decision alternatives, with all alternatives being defined by the same set of attributes. In our pricing tasks, respondents are provided with the same alternatives. For one alternative, the default, a price is provided. For the second alternative, respondents are asked to state the price (WTP) so that utility levels for alternatives are equal.¹ Thus, the price serves both as an attribute for the default alternative and as the response mode. According to scale compatibility, this should lead to a heavier weighting of the attribute price in the pricing task compared to the choice task. The observed WTP values from the pricing task should thus be lower than the WTP estimates from the choice task.

We propose a two-stage design for assessing the role of scale compatibility for WTP estimates from different stated preference elicitation modes. In the first stage, the pricing task is implemented, and from the respondents’ answers we derive WTP distributions for all alternatives. In the second stage, we implement the choice task with prices being drawn from the alternative-specific WTP distributions from the pricing task.

The two-step design permits two tests of scale compatibility. The first test builds on comparing WTP estimates from pricing and choices using regressions. The drawback of the first test is that the estimation of WTP from choices requires the specification of a structural utility function (see

¹ This type of pricing task is called ‘matching’ task, as respondents are asked to state the price so that utility levels for alternatives match. Attema and Brouwer (2013), for example, apply such a pricing task to evaluate health outcomes.

Hensher et al., 2014). The appropriateness of the functional form, however, cannot be tested. The second test does not require such a specification. It relies on the inter-connectedness of the prices in the choice task and the observed WTP distributions from the pricing task: Commodity prices in the choice task are taken from particular percentiles p of the WTP distributions from the pricing task. Procedure invariance implies that $100 - p$ percent of the respondents in the choice task should choose the alternative rather than the default product. If more than $100 - p$ percent chose the alternative tariff, this is an indication of scale compatibility.

The present study differs from previous studies on scale compatibility in three dimensions. First, previous estimates typically rely on data from class room experiments with small sample sizes, raising the question of external validity. Instead, our analysis builds on representative samples from two countries. Our broad samples also offer sufficient variability to test for differences in the WTP-distribution between subgroups and assess differences in scale compatibility between sub-groups. Another advantage is the large sample size that allows robust estimation of WTP distributions, required for the aforementioned second test procedure. Second, previous estimates build on intra-subject comparisons, by comparing, for example, data from pricing and choice tasks, where respondents participate in both tasks. A potential drawback of such an intra-subject approach is that institutional learning, i.e., a familiarization with the choice setting, or preference learning could invalidate the comparability of the two settings' results (Czajkowski et al., 2014; Carlsson et al., 2012). Instead, our analysis relies on inter-subject comparisons. Third, we analyze not only the differences in monetary trade-offs between the two elicitation settings but also in the trade-offs between two more attributes of the choices which do not vary across the elicitation modes. Other studies so far have only looked at choice situations where each choice had only two attributes (e.g., Attema and Brouwer, 2013; Bleichrodt and Pinto, 2002; Delquié, 1993; Tversky et al., 1988). We are the first examine contingent weighting and scale compatibility in a choice experiment.

We have implemented the two-step design in Germany and the United Kingdom (UK). As a product, we have selected electricity, a product which all respondents should consume and be familiar with. The electricity tariffs differed with respect to the share of electricity from renewable resources and the reliability of supply (expected number and duration of blackouts). For both countries and both test procedures we find that procedure invariance is violated. WTP estimates from choice tasks are markedly higher than from pricing tasks (up to 100 percent); this strongly indicates that the attribute price is over-weighted in pricing tasks.

The remainder of the paper is organized as follows. Section 2 provides details on the proposed two-step design and tests of scale compatibility. Section 3 outlines the survey design and characteristics of the sample. Our test results on scale compatibility based on regressions and choice probabilities are presented in Section 4. Section 5 concludes.

2 Diagnosis of scale compatibility

Suppose procedure invariance holds and truthful revelation of preferences for decision alternatives is guaranteed. Further assume that two alternatives, $T_c(\cdot)$ and alternative $T_a(\cdot)$, with attribute vectors, \mathbf{A}_c and \mathbf{A}_a , are evaluated.

In each pricing task, a default alternative, $T_c(\cdot)$, with a given price, $P_c > 0$, is provided. For the second alternative, $T_a(\cdot)$, the price is left blank and the respondent is asked to give the price (WTP), WTP_a , that makes her indifferent between the alternatives. Assume $WTP_c \geq P_c$ holds.² Further assume a respondent strictly prefers $T_a(\mathbf{A}_a)$ to $T_c(\mathbf{A}_c)$: $T_a(\mathbf{A}_a) > T_c(\mathbf{A}_c)$. Under these assumptions, the reported WTP for $T_a(\mathbf{A}_a)$ should exceed the price of $T_c(\mathbf{A}_c)$: $WTP_a > P_c$.³

In the choice task, respondents face the same alternatives but prices are assigned to all alternatives. A respondent should prefer $T_a(\mathbf{A}_a)$ to $T_c(\mathbf{A}_c)$ if $WTP_a - P_a \geq WTP_c - P_c$, or equivalently: $P_a \leq WTP_a + P_c - WTP_c$.

Now suppose procedure invariance is violated. In the pricing task, price is included in the attributes of the default alternative, and also serves as response mode. According to Tversky et al. (1988) this scale compatibility leads to a heavier weighing of the attribute price compared to the attributes that do not serve as response mode. Because price is an economic bad, we thus should expect that reported WTP, \overline{WTP}_a , is smaller than the WTP_a under the assumption of procedure invariance. The WTP difference, $WTP_a - \overline{WTP}_a = \Delta_a > 0$, is the monetary effect of scale compatibility on WTP assessments.

One way to diagnose scale compatibility is by means of consistency checks of stated WTPs from a pricing task and choices from a choice task, where each subject participates in both tasks (Attema and Brouwer 2013; Bleichrodt and Pinto 2002; Delquíé 1993). There is a potential drawback of such an intra-subject diagnosis: In repeated tasks subjects learn about their preferences as well as about the institutional setting of the choice situation. Czajkowski et al. (2014) observe up until about the eighth task in a choice experiment an improvement in accuracy i.e., a decrease in variance in responses to choice experiments. Carlsson et al. (2012) find - in addition to a decrease in the variance - a change in preferences for the price attribute in repeated choice sets. In an intra-subject design, these learning effects might carry over from the pricing task to the choice task despite the differences between the two.

To avoid the potential bias from learning in intra-subject comparisons, we have implemented an inter-subject design for assessing the role of scale compatibility in stated preference proce-

² The assumption that $WTP_c \geq P_c$ is mild, because P_c is in line with actual consumer prices, and basically the entire population purchases electricity for this price.

³ It is assumed that the subjects do not deliberately misreport WTP.

dures. The basic idea is to compare pricing and choice data from two independent representative samples. One sample participates in the pricing task. The derived alternative-specific WTP distributions determine the given prices in the choice task. Particularly, we have determined three prices for each alternative: P_a^{10} , P_a^{50} , and P_a^{90} , defined by the 10th, 50th and 90th percentile, $p \in \{10, 50, 90\}$, of the alternative-specific WTP distributions.

Our inter-subject design, allows two tests of scale compatibility. The first test relies on a comparison of regression estimates of WTP using data from pricing vs. choices. The advantage of the test is that it allows comparisons with results from previous literatures. Its drawback is that for the estimation of WTP from choices a structural form of the utility function must be imposed (see Hensher et al., 2014). Our second test does not suffer from this drawback. The test exploits the fact that the pre-determined prices in the choice task are derived from particular percentiles of the WTP distributions from the pricing task. Under procedure invariance the probability that respondents choose $T_a (A_a)$ rather than $T_c (A_c)$ in the choice task should be inversely related to the percentile of the WTP distribution determining the price of $T_a (A_a)$. As an example, suppose the price for $T_a (A_a)$ is drawn from the p^{th} percentile of the WTP distribution. According to procedure invariance, $100 - p$ percent of the respondents should choose $T_a (A_a)$ rather than $T_c (A_c)$. If scale compatibility is fulfilled, more than $100-p$ percent should choose $T_a (A_a)$.

3 Survey design, sample characteristics and methodology

3.1 Survey design

The pricing and choice surveys in Germany and the UK have been conducted with an online panel, recruited by a professional survey institute. All data were collected between October and December 2013.

In the first part of the field work, the collection of the pricing data, each respondent from a representative sample was offered a set of electricity tariffs, differing in the share of electricity from renewable resources, and in the reliability of electricity supply (expected frequency and average duration of blackouts). Two tariffs were presented on a vertically-split screen (see Figure A-1 in the Appendix). The left-hand part of the screen provided a tariff with attributes (share of renewables, reliability, price) reflecting those of a standard tariff in the respective country. The price was adjusted for the household size of the respective respondent's household. We call this tariff the default tariff. As an example, the default tariff for a 2-person household in Germany (the UK) had the following attributes: 20 (20) percent of renewable energy, 1 blackout per year with an expected duration of 10 (80) minutes, for EUR ct. 28.73 (EUR ct. 17.47)⁴ per kWh.

⁴ Corresponding to about GBp 14.81 per kWh.

The right-hand part of the screen provided an alternative tariff. The alternative tariff differed from the default tariff with respect to at least one of the two attributes share of renewables and/or reliability. Respondents were asked to provide us with their maximum monthly WTP for the alternative tariff in a blank field at the bottom of the right-hand panel. The difference between the stated WTP for the alternative tariff and the given price for the default tariff is the price premium or deduction associated with the specific variation in the attributes. In total, 50 different tariffs were provided: the default tariff plus 49 alternative tariffs. In the alternative tariffs, the share of renewables varied between 20 and 100 percent, the annual duration of blackouts between 5 and 80 minutes (in the UK between 5 and 160 minutes),⁵ and the number of blackouts between 1 and 5 per year.⁶ To keep respondents' cognitive burden low, each respondent was provided with five alternative tariffs. Since the ordering in which the alternative tariffs are presented might affect respondents' WTP statements, we randomized the order of appearance of the alternative tariffs.⁷

After having gathered the data from the pricing task, the second part of field work began: the collection of the choice data. Data were collected from another independent representative sample. Each respondent was, again, provided with a sequence of five pairs of tariffs. Like in the pricing task, a pair consisted of the default tariff and an alternative tariff. Unlike in the pricing task, a price was provided for both the default *and* the alternative tariff. Respondents had to choose between the default and the alternative tariff. Prices of the alternative tariff were derived from the distribution of WTPs from the pricing task. Particularly, we assigned prices that reflect the 10th, 50th, and 90th percentile of the tariff-specific WTP distributions. These percentiles were randomly assigned to the alternative tariff.

3.2 Sample characteristics

The socio-demographics of the four samples (two for each country) are described in Table B-1 in the Appendix. In Germany, 1,300 respondents participated in the pricing task, and 1,800 in the choice task. The derived working samples are slightly smaller (pricing: 1,121; choice: 1,679) due to missing information on individual variables. In both German samples, the distributions for household size (HHSIZE), equivalent income (INCOME), gender (FEMALE), and education (EDU)

⁵ Longer durations of blackouts in the UK compared to the German survey reflect the fact that the electricity supply in the UK is less reliable than in Germany (CEER 2012).

⁶ We used five levels for the share of renewables (20, 40, 60, 80 and 100), five levels for the duration of blackouts (Germany: 5, 10, 20, 40 and 80; UK: 5, 20, 40, 80 and 160) and two levels for the number of blackouts per year (1 and 5). For an overview of the attribute combinations see Table A-1 in the Appendix.

⁷ The phenomenon is called 'ordering effect' (Bateman and Langford 1997, Clark and Friesen 2008). Empirical evidence on the ordering effect in WTP studies such as ours is mixed (Boyle et al. 1993).

are quite similar,⁸ whereas the age (AGE) distributions are different. In both samples, average household size is slightly above two members, average monthly equivalent income is about EUR 1,500, and shares of female and male respondents are about equal. Slightly more than 20 percent has tertiary education, while the remaining share of respondents is in the category low and medium level of education.⁹

Turning to the two UK samples, in the UK, 1,301 respondents participated in the pricing task and 1,824 in the choice task. The working sample sizes are 1,215 and 1,647 respectively. For the UK samples distributions of all the individual characteristics are quite similar. Compared with the German samples, average household size is slightly larger, average age is similar to the average age in both German samples, the fraction of female respondents is slightly higher, and a higher share of respondents is assigned to the highest educational category.

3.3 Methodology

Our first test of scale compatibility relies on WTP estimates from regressions using data from the pricing vs. the choice task. In our regression analyses we model preference heterogeneity among respondents with respect to the tariff attributes using random-parameter techniques. These techniques allow the estimation of respondent-specific regression coefficients. Such a coefficient, $\beta_{ix} := \beta_x + u_{ix}$, is the sum of two components: a mean coefficient, β_x , for the attribute x , and a random respondent-specific deviation, u_{ix} , with i denoting the respondent. It is assumed that the respondent-specific deviations are normally distributed in the sample with zero mean and unknown standard deviation.

For the pricing task, we model WTP for tariffs $a = 1, \dots, 49$ in linear form as,

$$(1) \text{WTP}_{ia} = \alpha + \sum_x (\beta_x + u_{ix}) A_a + \sum_c \gamma_c C_{ic} + v_i + \varepsilon_{ia}.$$

The levels of the attributes of tariff a are contained in A_a . These attributes are the share of renewables in electricity generation (RENEW; in percent), the expected annual duration of blackouts (DUR-AN), and the average duration per blackout (DUR-AV). An attribute's regression coefficient gives the (marginal) change in WTP if the respective attribute changes by a single unit. Respondents' characteristics c are contained in C_{ic} . These characteristics are log of household equivalent income (LN-INCOME), number of household members (HHSIZE), age (AGE) and age

⁸ Income is measured in equivalent units to adjust for differences in needs across household types. The equalization is done using the OECD equivalence scale. The scale is defined as: $S^{OECD} = 1 + 0.5(n_A - 1) + 0.3n_C$, with n_A denoting the number of adults in the household (persons age 15 or older), and with n_C denoting the number of children.

⁹ A high level of education corresponds to a completed degree at a higher education institution (ISCED-5A, -6).

squared (AGE^2), gender (FEMALE), level of education (EDU-HIGH) and having experienced a blackout during the last five years (EXP-BO) (the latter three as dummy variables). The coefficient v_i shifts the regression line for each respondent individually. Finally, ε_{ia} is the error term.

Regression equation (1), SPEC1, is our most flexible specification. We test the sensitivity of the estimates from SPEC1 against two less flexible nested specifications. In SPEC2, we assume that the random respondent-specific deviations, u_{ix} , are zero for all respondents and all product attributes. Hence, all respondents' preferences equal the mean preference. In SPEC3, we further exclude the personal characteristics, i.e., the vector C_{ic} .

For the choice task, the estimation of marginal WTP requires two modifications in the estimation procedure. First, the dependent variable is discrete: A respondent chooses either the default or the alternative tariff. Thus, a linear regression is not appropriate. Instead, we construct the dummy variable $Switch_{it}$ that equals 1 if the respondent chooses the alternative tariff – otherwise it is zero – and estimate the mixed logit model for SPEC1 to SPEC3,

$$(2) \Pr(Switch_{ia} = 1) = F\left(\alpha^l + \sum_x (\beta_x^l + u_{ix}^l)A_a^l + \sum_c \gamma_c^l C_{ic} + \beta_p P_a + v_i^l + \varepsilon_{ia}^l\right).$$

$F(\cdot)$ denotes the cumulative logistic distribution, and the superscript l distinguishes the regression coefficients of the logit from the linear model (equation 1). The attribute vector for the logit model is denoted A_a^l . Since price is an exogenous variable in the choice task, the equation is expanded by the attribute price per month, P_a (PRICE). By definition, the marginal WTP of consumer i for attribute x is, $\frac{dU_i/dx}{dU_i/dP}$, with U_i denoting the utility derived from the product in question. According to equation (2), the marginal utility of attribute x is $\frac{\partial \Pr(Switch_{ia}=1)}{\partial x} = (\beta_x^l + u_{ix}^l)F'(\cdot)$, where $F'(\cdot)$ is the derivative of the cumulative logistic distribution $F(\cdot)$, and the marginal derivative with respect to price is $\frac{\partial \Pr(Switch_{ia}=1)}{\partial P_a} = (\beta_p^l)F'(\cdot)$ (for details see Hensher et al. 2014). Both derivatives, of course, depend on the particular functional form of the utility function. Modifying this functional form will change the derivatives and thus the WTP estimates. The models were estimated with maximum likelihood using Halton draws with 500 repetitions.

Our second test of scale compatibility relies on observed choice probabilities as explained in Section 2 above. We use binomial probability tests to assess whether the observed shares of switchers from the choice task comply with the percentiles of the WTP distribution from the pricing task. The H0 is that the probability to switch from the default tariff, $T_c (A_c)$, to the alternative tariff $T_a (A_a, P_a^p)$, $\Pr(Switch_{ia} = 1, P_a^p)$, equals $1 - p$, with p denoting the percentile of the WTP distribution defining the price of the alternative contract.

4 Empirical results

4.1 Comparison of WTP estimates and marginal rates of substitution from regressions (Test 1)

Table 1 presents the marginal WTPs for the product attributes “share of renewables” and “reliability” (number and duration of blackouts) from the grand means obtained from the SPEC1-regressions for pricing and choice tasks. The regression results are provided in Appendix C.

For all attributes, the signs of marginal WTP estimates from pricing and choice task are equal. In both countries, WTP increases in the share of renewables but decreases if energy provision is less reliable. The size of the estimates from pricing and choice, however, systematically differs. According to the estimates from the German choice task data, the average WTP for an additional percentage point of renewable energy in the electricity mix is EUR ct. 0.101. According to the German pricing task data, it is only about half of that (EUR ct. 0.053). For the UK, we find a similar divide: again the estimate from the choice task is about twice as large as the estimate from the pricing task (EUR ct. 0.028 vs. 0.044). For the attributes blackout duration and number of blackouts the result is similar: again we find that estimates of marginal WTP (in absolute terms) from choices are substantially larger than from pricing (1.5 to 2 times).

Table 1: Marginal WTP in Eur ct. per kWh based on grand mean coefficients

	Germany		United Kingdom	
	pricing	choice	pricing	choice
renewables	0.053	0.101	0.028	0.044
average blackout duration for 1 blackout per year	-0.059	-0.114	-0.024	-0.037
average blackout duration for 5 blackouts per year	-0.295	-0.570	0.120	-0.185

While our results indicate substantial discrepancies in WTP estimates from choice and pricing tasks, it is interesting to note that the marginal rates of substitutions between attributes, captured by the ratios of marginal WTPs, are very similar. These trade-offs are reported in Table 2: The trade-offs are derived in the same way as the marginal WTP (compare above). They are the marginal rate of substitution of one attribute by another. This means that respondents in Germany would be willing to accept an additional minute of blackouts per year (for one expected blackout) if the share of renewables was 0.898 percentage points higher. Table 2 shows first that the magnitude of the trade-offs is very similar in pricing and choice. This means that the occurrence of scale compatibility does not affect the relative non-monetary valuation of the other attributes. Second, also the values are a lot more similar in Germany and the UK. For comparative research, this means that not just marginal WTP but also tradeoffs between attrib-

utes should be analyzed to gain a comprehensive picture of the differences and communalities in the valuation of the attributes. Whether these results are a coincidence or a regularity, in our view, requires further research.

Table 2: Marginal rates of substitution (based on grand mean coefficients)

	Germany		United Kingdom	
	pricing	choice	pricing ¹	choice
Blackout / renewable at 1 blackout per year	1.113	1.133	0.857	0.833
Blackout / renewable at 5 blackouts per year	5.566	5.667	4.290	4.167
Renewables / blackout at 1 blackout per year	0.898	0.882	1.167	1.200
Renewables / blackout at 5 blackouts per year	0.180	0.180	0.233	0.240

¹ Note: for the UK pricing task the coefficient for DUR-AV was included into the calculations because it was significant for this sample.

In sum, the regression results indicate that the elicitation procedure has a sizeable effect on estimates of marginal WTP. While the estimates from pricing and choice have the same sign and yield the same tradeoffs between attributes, the marginal WTPs markedly differ in quantitative terms, with absolute values being about two times larger for the choice task, indicating the presence of scale compatibility, i.e., overpricing in the pricing task. However, as explained above, the WTP estimates from the choice task hinge upon the assumed functional form of the utility function. Our second test is not based on such a specification.

4.2 Comparison of WTP distributions and choice probabilities (Test 2)

Our second test of scale compatibility involves the comparison of choice probabilities. Most importantly, it does not depend on the definition of a particular form of household utility function. Table 3 provides WTP percentiles from the pricing task and the percentage of respondents choosing the alternative tariff (at a given price) rather than the default tariff (switchers). The left hand side of the table provides the results for the German samples, the right hand side for the UK samples. As we have 49 tariffs in total, we condense the findings. This is justified because WTP responses systematically vary along with the share of renewables in the energy mix and the total annual duration of blackouts, but not with the average duration or the frequency of blackouts. Hence, Table 3 provides responses for a particular combination of the attributes renewable share (RENEW) and total annual duration of blackouts (DUR-AN) over all attribute

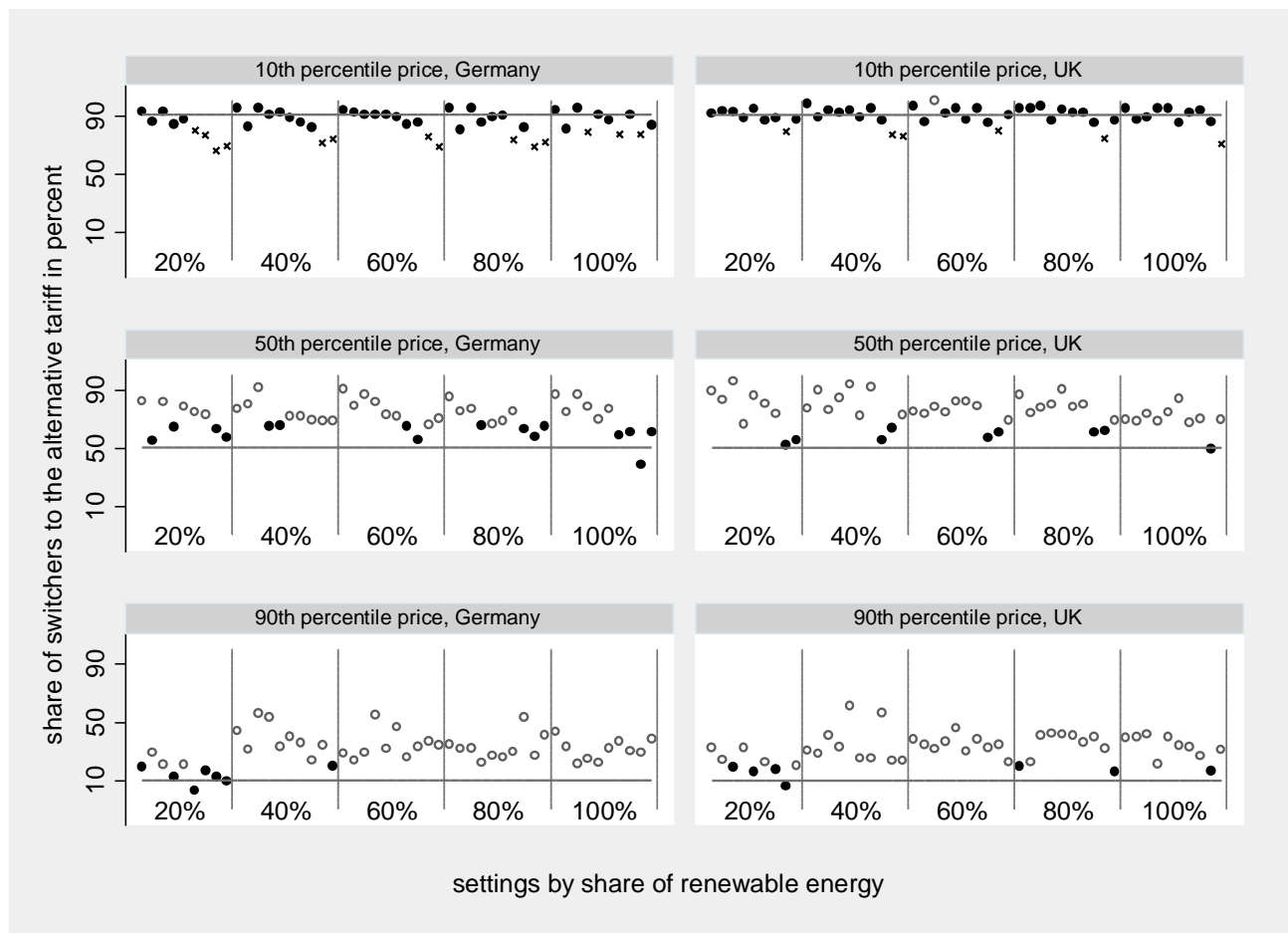
Table 3: WTP percentiles from the pricing task in Eur. cts. and share of switchers in choice tasks at 10th, 50th and 90th percentile price level, full sample

Germany								United Kingdom							
pricing task				choice task				pricing task				choice task			
WTP percentiles in Eur. Cts.				share of switchers				WTP percentiles in Eur. Cts.				share of switchers			
renew-ables	duration	10 th pct	50 th pct	90 th pct	10 th pct	50 th pct	90 th pct	renew-ables	duration	10 th pct	50 th pct	90 th pct	10 th pct	50 th pct	90 th pct
20	5	17.05	28.73	31.70	0.89	0.68	0.24	20	5	12.04	17.47	20.42	0.92	0.86	0.28
20	10	12.56	25.24	30.85	0.92	0.82	0.21	20	20	11.62	17.47	21.37	0.90	0.81	0.26
20	20	15.73	25.24	31.58	0.85	0.71	0.17	20	40	9.67	17.47	21.37	0.90	0.83	0.19
20	40	13.27	21.47	29.58	0.78	0.73	0.10	20	80	9.60	16.89	19.42	0.88	0.74	0.18
20	80	10.59	22.07	29.58	0.67	0.60	0.11	20	160	7.72	13.79	19.42	0.83	0.54	0.13
40	5	18.90	28.73	34.56	0.89	0.78	0.38	40	5	13.50	17.47	23.32	0.93	0.83	0.29
40	10	16.85	28.73	32.36	0.93	0.78	0.55	40	20	13.50	17.47	21.37	0.93	0.81	0.37
40	20	14.22	26.83	32.36	0.90	0.69	0.37	40	40	10.53	17.47	20.20	0.91	0.83	0.43
40	40	14.22	25.24	32.36	0.83	0.70	0.29	40	80	11.62	17.47	20.12	0.91	0.74	0.40
40	80	12.61	22.07	31.70	0.72	0.68	0.27	40	160	7.43	15.06	19.44	0.76	0.68	0.24
60	5	17.12	28.73	35.99	0.93	0.84	0.26	60	5	12.04	18.07	21.90	0.91	0.74	0.37
60	10	16.85	28.73	35.99	0.90	0.84	0.42	60	20	11.62	18.07	23.32	0.96	0.76	0.34
60	20	15.67	28.73	35.94	0.89	0.72	0.39	60	40	10.19	17.47	21.90	0.91	0.82	0.38
60	40	14.22	27.46	35.99	0.84	0.60	0.29	60	80	10.19	17.47	20.12	0.90	0.68	0.36
60	80	12.61	25.34	31.93	0.71	0.68	0.36	60	160	9.02	15.52	20.12	0.85	0.65	0.28
80	5	21.09	28.73	35.99	0.87	0.80	0.33	80	5	12.82	18.62	27.22	0.95	0.80	0.21
80	10	18.90	29.79	37.92	0.90	0.71	0.27	80	20	10.21	18.07	23.32	0.92	0.79	0.42
80	20	14.73	28.73	35.99	0.89	0.68	0.26	80	40	11.62	18.07	23.32	0.93	0.84	0.41
80	40	14.22	27.46	34.59	0.77	0.69	0.41	80	80	11.83	17.47	21.90	0.88	0.70	0.38
80	80	12.61	25.34	33.82	0.69	0.61	0.34	80	160	9.60	17.47	21.55	0.80	0.66	0.24
100	5	17.85	29.58	41.41	0.87	0.81	0.38	100	5	12.06	19.42	25.27	0.91	0.69	0.40
100	10	20.67	29.58	38.06	0.87	0.83	0.23	100	20	11.62	19.42	25.44	0.92	0.71	0.31
100	20	15.94	28.73	38.06	0.88	0.73	0.27	100	40	11.62	18.07	23.32	0.90	0.79	0.37
100	40	14.22	28.73	38.06	0.83	0.60	0.33	100	80	11.62	17.47	23.32	0.93	0.69	0.30
100	80	14.22	28.73	35.99	0.80	0.50	0.34	100	160	9.02	17.47	23.32	0.78	0.59	0.24

Note: WTP in Euro cents.

levels of frequency of blackouts. If procedure invariance held, and the price for the alternative tariff represents the 10th, 50th, 90th percentile of the WTP distribution, respondents should choose the alternative contract with the converse probability (90%, 50% and 10% respectively). Table 3 already gives a first impression of a violation of procedure invariance. For example in the last column – UK, choice task, 90th percentile price – the expected share of switchers would be 10%, i.e., a share of 0.10. It lies, however, consistently across all attribute combinations above this value with a maximum share of 0.42 switchers.

Figure 1: Scale compatibility (Germany and United Kingdom); share of switchers to the alternative tariff by price level



Legend:

- x significantly below the expected share of switchers ($p < .01$)
- not significantly different from the expected share of switchers
- o significantly above the expected share of switchers ($p < .01$)

Note: separate panels indicate the different price levels of the alternative tariff. The tariffs are sorted ascending by share of renewables and total blackout duration

The results of Table 3 are visualized in Figure 1. The graphs in the left (right) column provide the results for Germany (UK). In Figure 1, each dot represents the share of respondents who chose

the alternative tariff over the default tariff in one of the 49 settings. The settings are sorted like in Table 3 from a low to a high share of renewable energy on the horizontal axis. The grouping of tariffs with the same share of renewables is indicated by the vertical lines. The graphs in the first/second/third row refer to alternative tariffs with prices taken from the 10th/50th/90th percentile of WTP distributions respectively. Within these groups, the tariffs are sorted by total blackout duration from the lowest on the left end of the grouping to the highest on the right end of the grouping. For example, the graphs in the first row shows the pattern for all choice situations where the alternative tariff was offered at the 10th percentile price, i.e., a very low price. If procedure invariance held, we would expect 10 % to choose the default tariff while 90% should switch to the alternative tariff. The expected switching rate is indicated by the horizontal line in each graph.

For a vast majority of scenarios, the observed probability to switch exceeds the expected probability under procedure invariance. Take, for example the results from row two, a tariff with 20% renewables and expected blackout duration of 5 minutes. If the price is taken from the 50th percentile, we should expect that about 50% of the respondents prefer the alternative over the default tariff. The observed fraction, however, is much higher: 82% in Germany and 81% in the UK. Thus, for being in line with the expected probability to switch, the price for the alternative tariff should have been higher. In other words: WTP estimates from the pricing task, consistent with the overpricing hypothesis, are too low. Only for the tariffs where the price is taken from the 10th percentile, observed switching probabilities are not systematically larger than expected. This is not surprising given a reverse probability of 90 percent.

Formal tests of differences between expected and observed switching probabilities are reported in Figure 1 as well. Altogether, results from 147 binomial probability tests (49 settings at 3 price levels) are reported. A filled-in black dot indicates that there is no significant difference between expected and the observed switching probabilities. A circle indicates that the observed probability to switch is higher than the expected – indicating overpricing in the pricing task. A cross indicates the opposite. If the price is taken from the 10th percentile, the vast majority of binomial probability tests indicate no differences between expected and observed switch probabilities. As explained above, this is not unexpected. If, however, the price is taken from the 50th or 90th percentile, almost all tests indicate that observed switching probabilities are significantly higher than expected,¹⁰ indicating a violation of procedure invariance and the presence of overpricing in pricing tasks.

¹⁰ These results are also available in Table D-1 in the Appendix D.

5 Conclusion

We show in an inter-personal two-stage design that prices weigh heavier in pricing than in choice, i.e., when the response mode is in the same format as one of the attributes. First, we implemented a pricing task where the respondents had to state their WTP for a product compared to another product for which the price information was available. Second, a new group of respondents fulfilled a choice task where they choose between two products with a price given for each. For choice experiments, this means that WTP is higher when the price is given than when respondents have to state her WTP for a product compared to another. This is a violation of scale compatibility. Our results confirm previous findings from experimental studies where every respondent had to do both tasks (Attema and Brouwer 2013; Bleichrodt and Pinto 2002; Delquié 1993). The choice experiments were done for two representative samples in Germany and the UK respectively. The product was electricity tariffs. With our inter-personal design administered for representative samples, we can rule out learning effects between the two tasks and we can thus better assess the magnitude of the differences between pricing and choice.

While the monetary trade-offs are inconsistent between the two tasks, we find a surprisingly high consistency for the tradeoffs between the other attributes. The marginal rates of substitution for the share of renewable energy and blackout risk are the same in pricing and in choice. Previous studies have only looked at choices which consisted of only two attributes.

Future work should concentrate on the external validity of the WTP estimates derived from stated preference elicitation. The question is whether the results from pricing or from choice in choice experiments are closer to the real WTP. Previous literature on contingent valuation suggests that choice should yield more realistic estimates because respondents are more used to the choice between two goods with a given price than to proposing a price (Arrow et al., 1993).

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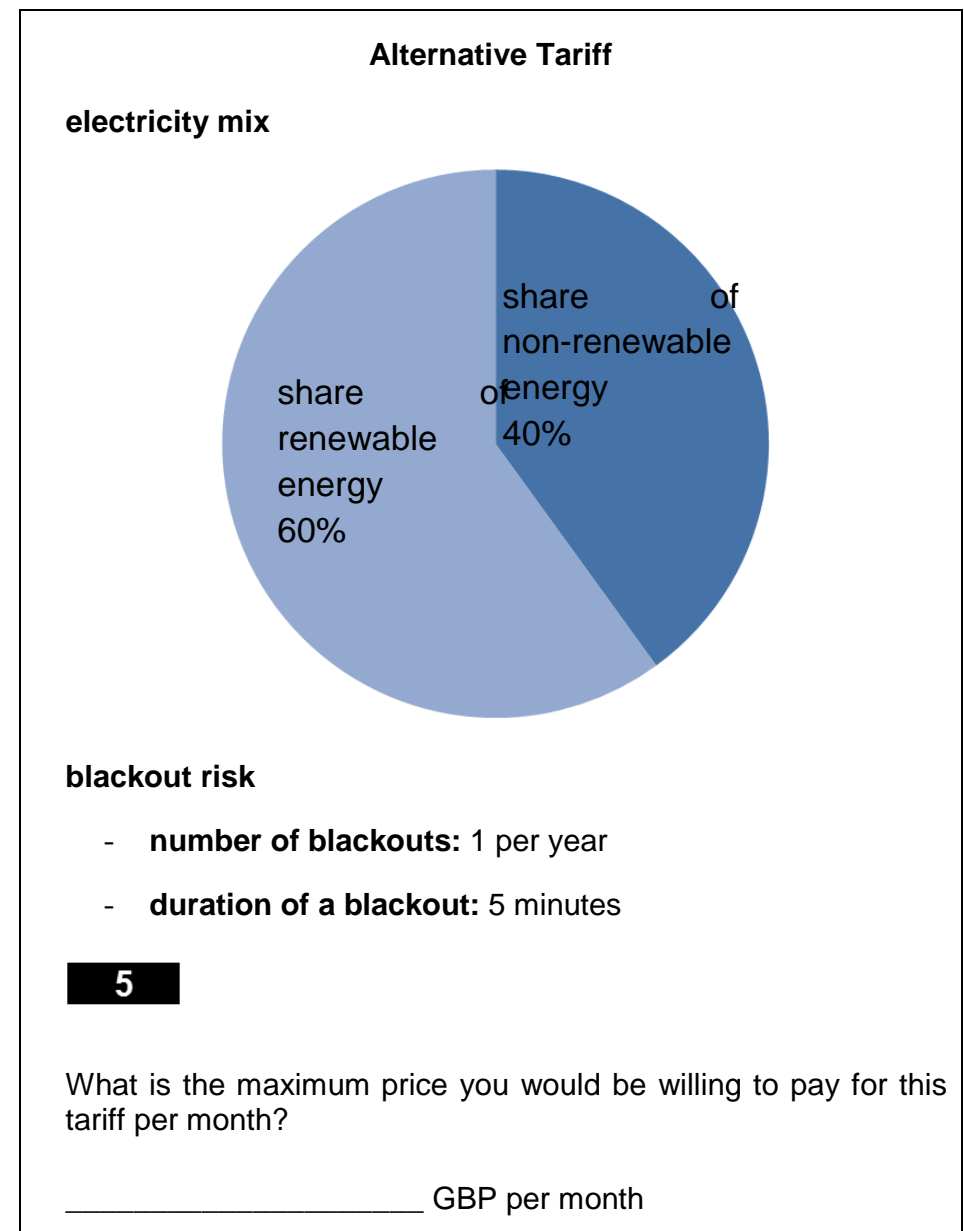
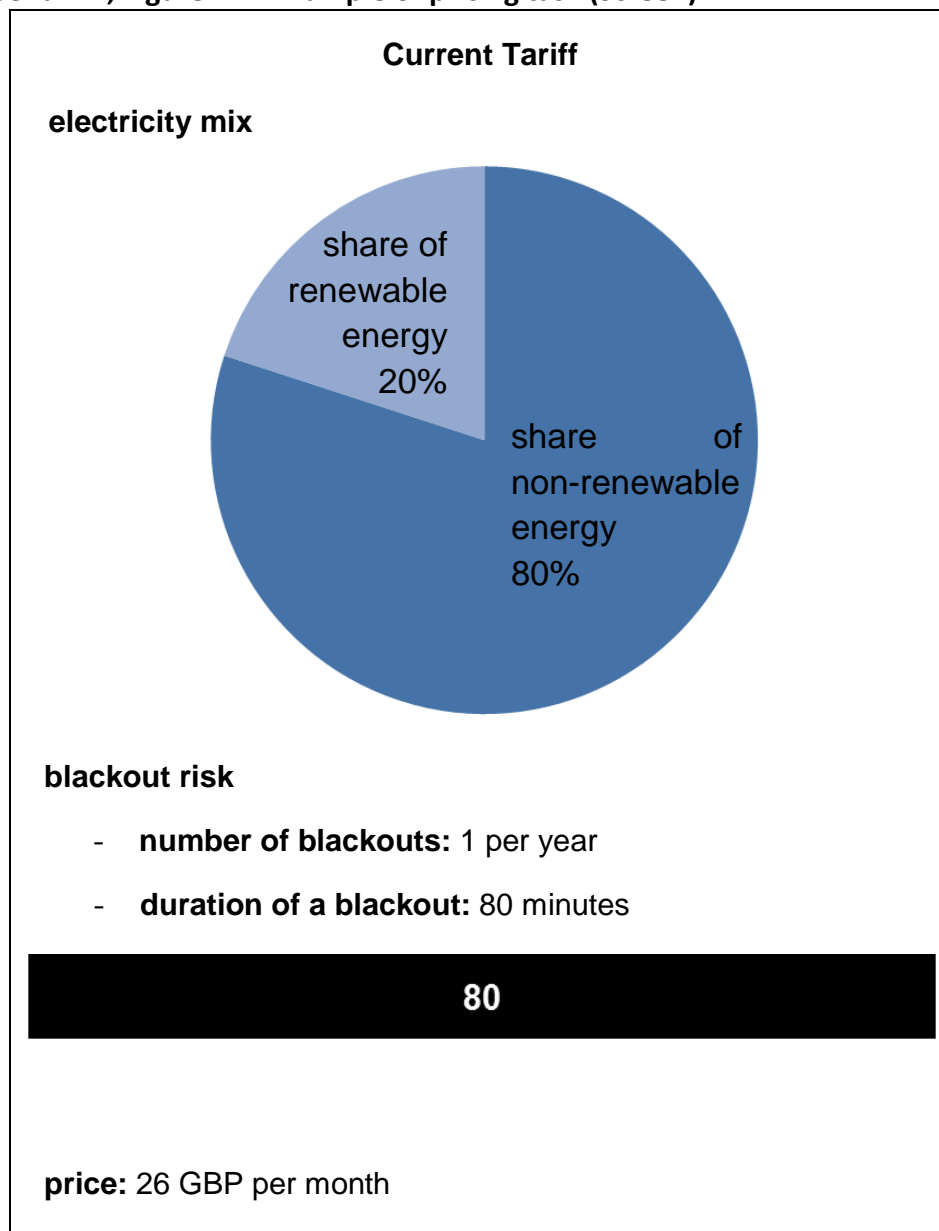
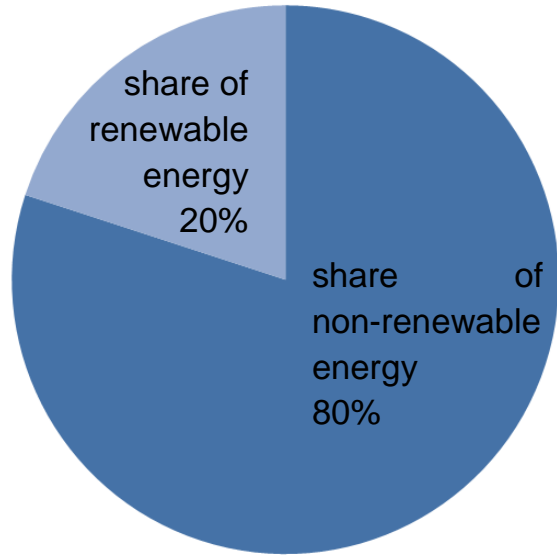


Figure A-2: Example of choice task (Screen)

Which of the two tariffs would you choose?

Current Tariff

electricity mix



blackout risk

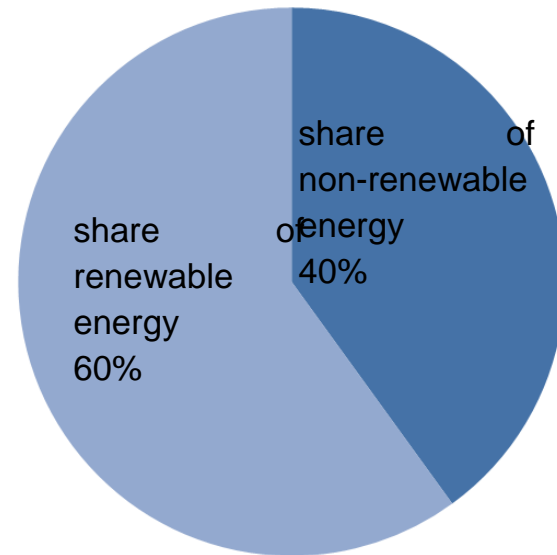
- **number of blackouts:** 1 per year
- **duration of a blackout:** 80 minutes

80

price: 26 GBP per month

Alternative Tariff

electricity mix



blackout risk

- **number of blackouts:** 1 per year
- **duration of a blackout:** 5 minutes

5

price: 37 GBP per month

UK Questionnaire

Screen 1

Dear survey participant,

First, we would like to thank you for your willingness to participate in the online surveys of the Ipsos-Access-Panel. The survey is part of a scientific, publicly funded research project. We will be asking you questions about electricity supply in the future.

It is very important that <name of participant> is completing the survey. If you are not this person we ask you not to answer the survey in his or her name.

It will take about 20 minutes to answer the survey. Your answers are anonymous and we will handle your data in accordance with privacy protection regulations.

Please only use the “next” button below to get to the next page of the questionnaire,

To start the survey, please click on “next”.

Screen 2

What is your date of birth (year and month)?

What is your gender?

- male
- female

How many people are there currently living in your household, including yourself?

Number of household members _____

And how many people in your household are younger than 14 years?

Number of children under 14 in your household _____

UK only: **How many bedrooms are there in your house/flat?**

Number of bedrooms _____

Screen 3

What is the current total net income of your household from all sources per week, per day or per year? Net income is income after tax payments and compulsory deductions. The sources include earnings from employment and self-employment, income from social benefits and pensions and income from other sources such as property, interest on savings etc.

Please use the dropdown menu you find easiest to answer (weekly, monthly or annual income [in Euro]).

UK

approximate weekly	approximate monthly	approximate annual
Less than £190	Less than £820	Less than £9,850
£190 to under £250	£820 to under £1,100	£9,850 to under £13,190
£250 to under £310	£1,100 to under £1,360	£13,190 to under £16,320
£310 to under £380	£1,360 to under £1,640	£16,320 to under £19,650
£380 to under £450	£1,640 to under £1,960	£19,660 to under £23,520
£450 to under £540	£1,960 to under £2,330	£23,520 to under £28,000
£540 to under £650	£2,330 to under £2,820	£28,000 to under £33,790
£650 to under £790	£2,810 to under £3,450	£33,790 to under £41,350
£790 to under £1,050	£3,450 to under £4,580	£41,350 to under £54,910
£1,050 or more	£4,580 or more	£54,910 or more

Germany

approximate weekly	approximate monthly	approximate annual
0 to 230	0 to 980	0 to 11770
231 to 310	981 to 1350	11771 to 16140
311 to 380	1351 to 1660	16141 to 19920
381 to 460	1661 to 1990	19921 to 23880
461 to 540	1991 to 2340	23881 to 28070
541 to 630	2341 to 2730	28071 to 32780
631 to 730	2731 to 3200	32781 to 38340
731 to 880	3201 to 3820	38341 to 45830
881 to 1110	3821 to 4840	45831 to 58040
1111 or more	4841 or more	58041 or more

[Pricing task]

In the UK [Germany], the future of the electricity supply is a widely discussed subject. Three issues seem to be of particular interest:

- The composition of the electricity mix from renewable and non-renewable resources. Please note:
 - non-renewable resources are coal, gas, oil and uranium (for nuclear energy)
 - renewable resources are wind, solar, water and biomass
- The blackout risk:
this is defined as the number of blackouts per year and the average duration of a blackout.
- The price households have to pay for electricity every month.

In the following, we will be referring to two different electricity tariffs for your household: a current tariff and an alternative tariff. The tariffs differ on only two points: the electricity mix and the blackout risk (or both). Everything else – payment scheme, cancellation period, etc. – is the same.

Assume that the blackout risk is the number and duration of unplanned outages you can expect in the course of one year. This excludes blackouts caused by technical problems in your house and planned outages with prior notification. You do not know when the blackouts will happen, they can occur at any time of the year and day.

A monthly price is quoted for the current tariff. This is a typical price for a household such as yours (with respect to the number of bedrooms). Please assume that this tariff is your current tariff.

The question we are asking is: What is the maximum price per month you would be willing to pay for the alternative tariff.

Here is an example:

Example for the respondent's household size

5 pricing situations

[Choice task]

In the UK [Germany], the future of the electricity supply is a widely discussed subject. Three issues seem to be of particular interest:

- The composition of the electricity mix from renewable and non-renewable resources. Please note:
 - non-renewable resources are coal, gas, oil and uranium (for nuclear energy)
 - renewable resources are wind, solar, water and biomass.
- The blackout risk:
this is defined as the number of blackouts per year and the average duration of a blackout.
- The price households have to pay for electricity every month.

In the following, we will be referring to two different electricity tariffs for your household: a current tariff and an alternative tariff. The tariffs can differ on the electricity mix, the blackout risk and the price. Everything else – payment scheme, cancellation period, etc. – is the same.

Assume that the blackout risk is the number and duration of unplanned outages you can expect in the course of one year. This excludes blackouts caused by technical problems in your house and planned outages with prior notification. You do not know when the blackouts will happen, they can occur at any time of the year and day.

A monthly price is quoted. This is a typical price for a household such as yours (with respect to the number of bedrooms). Please assume that the current tariff is your current tariff.

In the following, you can choose between the current tariff and an alternative tariff. Which one would you choose?

First, we will show you an example.

Here is an example:

Example for the respondent's household size

5 choice situations

Have you personally experienced one or more blackouts during the last 5 years?

Please count planned and unplanned blackouts.

- no [*→ skip next 2 questions*]
- one
- more than one but fewer than five
- 5 or more
- don't know [*→ skip next 2 questions*]

For UK

What is the highest general educational qualification you have obtained so far?

- primary school
- secondary school (age under 15 years old)
- general National Vocational Qualification Foundation or Intermediate Level (GNVQ, GSVQ) / GCSE/ SCE standard
- NVQ1, NVQ2
- NVQ3 / SCE Higher Grade / Scottish Certificate of Sixth Year Studies / General National Vocational Qualification Advanced Level / GCE Advanced Level (GCE A/AS)
- other _____
- refuse to answer

What is the highest level of continuing you have obtained so far?

- NVQ4 / Higher National Certificate (HNC) / Higher National Diploma (HND) / Diploma in HE (including nurses training) / Bachelor's degree (BA, BSc, BEd, BEng, MB, BDS, BV, etc.)
- NVQ5 / Master's degree (MSc, MA, MBA, etc.) / Post-graduate diplomas and certificates / Doctorate (Ph.D.)
- none
- other _____
- refuse to answer

Table A-1: Attribute combinations (default tariff: shaded grey)

Germany			United Kingdom		
share of renewable energy	number of blackouts	duration of blackouts in minutes	share of renewable energy	number of blackouts	duration of blackouts in minutes
20	1	10	20	1	80
20	1	5	20	1	5
20	5	1	20	5	1
20	5	2	20	1	20
20	1	20	20	5	4
20	5	4	20	1	40
20	1	40	20	5	8
20	5	8	20	5	16
20	1	80	20	1	160
20	5	16	20	5	32
40	1	5	40	1	5
40	5	1	40	5	1
40	1	10	40	1	20
40	5	2	40	5	4
40	1	20	40	1	40
40	5	4	40	5	8
40	1	40	40	1	80
40	5	8	40	5	16
40	1	80	40	1	160
40	5	16	40	5	32
60	1	5	60	1	5
60	5	1	60	5	1
60	1	10	60	1	20
60	5	2	60	5	4
60	1	20	60	1	40
60	5	4	60	5	8
60	1	40	60	1	80
60	5	8	60	5	16
60	1	80	60	1	160
60	5	16	60	5	32
80	1	5	80	1	5
80	5	1	80	5	1
80	1	10	80	1	20
80	5	2	80	5	4
80	1	20	80	1	40
80	5	4	80	5	8
80	1	40	80	1	80
80	5	8	80	5	16
80	1	80	80	1	160
80	5	16	80	5	32
100	1	5	100	1	5
100	5	1	100	5	1
100	1	10	100	1	20
100	5	2	100	5	4
100	1	20	100	1	40
100	5	4	100	5	8
100	1	40	100	1	80
100	5	8	100	5	16
100	1	80	100	1	160
100	5	16	100	5	32

Appendix B

Table B-1: Summary statistics of personal characteristics

		Germany				United Kingdom			
		Pricing task		Choice task		Pricing task		Choice task	
Variable names	Definition	Mean	Std.dev.	Mean	Std.dev.	Mean	Std.dev.	Mean	Std.dev.
FEMALE	female=1; zero otherwise	0.47		0.50		0.49		0.53	
AGE	in years	37	12	46	13	42	13	41	14
EXP-BO	experienced an unplanned blackout during the past 5 years=1; zero otherwise	0.66		0.63		0.62		0.60	
EDU-HIGH	completed tertiary education (ISCED-5A, -6)=1; zero otherwise	0.25		0.21		0.35		0.35	
INCOME	monthly OECD-equivalence income per household member in EURO	1526	885	1532	920	1277	791	1300	812
HHSIZE	Number of household members	2.4	1.25	2.2	1.1	2.7	1.3	2.6	1.3
Sample size		1,300		1,800		1,301		1,824	
Working sample size		1,121		1,679		1,215		1,647	

Appendix C: Regression results

Turning to the results of the pricing task first, Table C-1 summarizes the results for specifications SPEC1 to SPEC3 of the linear regression model (1). The upper part of the Table shows the coefficients and their standard errors, the middle part shows the standard deviations of the random parameters, and the lower part shows the likelihood-ratio statistics. The reported likelihood-ratio statistics reveal that the flexibility of the random parameter specification SPEC1 boosts the model fit. This can also be seen in the likelihood-ratio model tests in Table C-3.

Our results suggest that, as expected, respondents are willing to pay for a more reliable energy supply, but the WTP for a greening of the electricity mix would decrease if greening would decrease the reliability of electricity supply. The positive significant coefficient for average duration of a blackout (DUR-AV) for the UK indicates that respondents in the UK for a given total duration of blackouts prefer few longer over many short blackouts. We further find that personal characteristics (SPEC 1 and SPEC2) can explain some variation in the WTP values, in particular LN-INCOME. The reported standard deviations for the random parameter distributions (SPEC1) are highly significant for both countries, suggesting that the individual slopes u_{ia} differ across respondents and thus indicate heterogeneity in the preferences for tariff attributes. In general, the results for the tariff attributes are robust across all three specifications.

Turning to the results of the choice task, Table C-2 summarizes the results for specifications SPEC1 to SPEC3 of the logit regression models (equation 2). It has the same structure as Table C-1 for the linear regression models. Again, the likelihood-ratio statistics in the lower panel show that the flexible random parameter specification in SPEC1 yields the best model fit (compare also Table C-3). This specification, additionally, accounts for the individual heterogeneity, which is not captured by the demographic variables. The individual heterogeneity is significant as the standard deviations in the lower panel show.

The results of choice task are in general very similar to those of the pricing task.¹¹ However, compared to Table 2, the results for the personal characteristics (SPEC 1 and SPEC2) are more mixed and the frequency of blackouts (DUR-AV) does not play a significant role for the switching decision in the UK. As expected, the coefficient PRICE is negative and significant; a higher price for the alternative reduces the probability of switching.

¹¹ The logit-coefficients cannot be as easily interpreted as in the linear model, tendencies, however, become apparent. The demographic variables in the logit model should not be interpreted with respect to their influence on the valuation of the attributes as the dependent variable is whether or not the alternative tariff is chosen. The alternative tariff does not necessarily have a higher share of renewables or lower blackout duration than the default tariff.

Table C-1: Results of the pricing task

	Germany						UK					
	SPEC 1		SPEC 2		SPEC 3		SPEC 1		SPEC 2		SPEC 3	
WTP	coeff.	std. error	coeff.	std. error	coeff.	std. error	coeff.	std. error	coeff.	std. error	coeff.	std. error
Attributes (random coefficients)												
RENEW	0.053***	(0.003)	0.053***	(0.003)	0.053***	(0.003)	0.028***	(0.002)	0.029***	(0.002)	0.029***	(0.002)
DUR-AN	-0.059***	(0.004)	-0.058***	(0.004)	-0.058***	(0.004)	-0.024***	(0.001)	-0.024***	(0.001)	-0.024***	(0.001)
DUR-AV	-0.004	(0.005)	-0.000	(0.005)	-0.000	(0.005)	0.006***	(0.002)	0.006***	(0.002)	0.006***	(0.002)
Constant	25.951***	(2.939)	26.608***	(3.044)	24.889***	(0.285)	10.652***	(2.142)	12.175***	(2.380)	16.772***	(0.202)
Individual characteristics (non-random coefficients)												
FEMALE	-0.508	(0.421)	-0.197	(0.435)			0.681*	(0.294)	0.388	(0.327)		
AGE	-0.293*	(0.115)	-0.311**	(0.118)			-0.054	(0.074)	-0.115	(0.082)		
Age ²	0.004*	(0.001)	0.004*	(0.001)			0.000	(0.001)	0.001	(0.001)		
EXP-BO	-0.035	(0.443)	-0.069	(0.458)			0.508	(0.299)	0.287	(0.332)		
EDU-HIGH	0.974*	(0.491)	1.234*	(0.507)			-0.057	(0.311)	0.439	(0.345)		
LN-INCOME	0.899**	(0.344)	0.864*	(0.356)			0.966***	(0.225)	1.002***	(0.249)		
HHSIZE	-0.821***	(0.168)	-0.865***	(0.173)			-0.053	(0.119)	-0.104	(0.132)		
S.D. RENEW	0.055***	(0.003)					0.065***	(0.002)				
S.D. DUR-AN	0.053***	(0.005)					0.021***	(0.001)				
S.D. DUR-AV	0.073***	(0.005)					0.022***	(0.002)				
S.D. Constant	6.121***	(0.178)	7.082***	(0.162)	7.228***	(0.165)	4.138***	(0.128)	5.290***	(0.119)	5.363***	(0.120)
N	6105		6105		6105		6075		6075		6075	
chi ²	717.844		880.871		835.587		521.276		717.935		687.232	
LL	-20414.309		-20612.813		-20634.953		-17690.071		-18289.596		-18304.7	

* $p < 0.10$, ** $p < 0.05$, *** $p < 0.01$

Table C-2: Results of the choice task

SWITCH	Germany						UK					
	SPEC 1		SPEC 2		SPEC 3		SPEC 1		SPEC 2		SPEC 3	
	coeff.	std. error	coeff.	std. error	coeff.	std. error	coeff.	std. error	coeff.	std. error	coeff.	std. error
Attributes	random		non-random		non-random		random		non-random		non-random	
RENEW	0.015***	(0.001)	0.010***	(0.001)	0.010***	(0.001)	0.012***	(0.001)	0.008***	(0.001)	0.008***	(0.001)
DUR-AN	-0.017***	(0.001)	-0.014***	(0.001)	-0.014***	(0.001)	-0.010***	(0.001)	-0.009***	(0.001)	-0.009***	(0.001)
DUR-AV	-0.002	(0.001)	-0.002	(0.001)	-0.002	(0.001)	0.001	(0.001)	0.001	(0.001)	0.001	(0.001)
Price and individual characteristics (not random)												
WTP	-0.149***	(0.003)	-0.121***	(0.003)	-0.121***	(0.003)	-0.271***	(0.006)	-0.232***	(0.008)	-0.232***	(0.004)
FEMALE	0.048	(0.046)	0.057	(0.042)			0.165***	(0.046)	0.209***	(0.043)		
AGE	-0.029**	(0.012)	-0.023**	(0.011)			0.018	(0.011)	0.015	(0.010)		
Age ²	0.000	(0.000)	0.000	(0.000)			-0.000**	(0.000)	-0.000**	(0.000)		
EXP-BO	0.101**	(0.047)	0.094**	(0.043)			0.099**	(0.046)	0.091**	(0.044)		
EDU-HIGH	0.023	(0.058)	0.047	(0.053)			0.197***	(0.049)	0.185***	(0.046)		
LN-INCOME	0.069*	(0.036)	0.033	(0.033)			0.026	(0.032)	0.037	(0.031)		
HHSIZE	-0.042	(0.021)	-0.033*	(0.020)			-0.041**	(0.019)	-0.037**	(0.018)		
Constant (random)	4.516***	(0.347)	4.034***	(0.315)	3.534***	(0.090)	4.859***	(0.329)	4.229***	(0.310)	4.729***	(0.108)
S.D. RENEW	0.027***	(0.001)					.024***	(0.001)				
S.D. DUR-AN	0.030***	(0.001)					.010***	(0.000)				
S.D. DUR-AV	0.025***	(0.001)					.003***	(0.001)				
S.D. Constant	0.256***	(0.032)	1.495***	(0.038)	1.517***	(0.039)	.348***	(0.032)	1.261***	(0.036)	1.290***	(0.037)
N	8395		8395		8395		8235		8235		8235	
chi ²	889.154		674.141		699.267		532.862		393.363		420.629	
LL	-4269.545		-4377.051		-4390.115		-3915.736		-3985.486		-4005.815	

* $p < 0.1$, ** $p < 0.05$, *** $p < 0.01$,

Table C-3: Likelihood-ratio-tests for the specifications

		SPEC 3 vs. SPEC 2	SPEC 2 vs. SPEC 1	SPEC 3 vs. SPEC 1
Germany	pricing	44.28 (0.00)	397.01 (0.00)	441.29 (0.00)
	choice	26.13 (0.00)	215.01 (0.00)	241.14 (0.00)
UK	pricing	30.21 (0.00)	1199.05 (0.00)	1229.26 (0.00)
	choice	40.66 (0.00)	139.50 (0.00)	180.16 (0.00)