

Two energy suppliers are better than one: survey experiments on consumer engagement with local energy in GB

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Abstract

Usually consumers have a single energy supplier. Permitting consumers to take on additional contracts with local suppliers in a multiple-supplier model could support growth of local renewable energy. The aims of this study were to assess the attractiveness of a multiple-supplier model and to understand whether consumers would be more likely to engage with local energy suppliers in a multiple-supplier model or the current single supplier model. An additional aim was to explore the role of default effects and cognitive biases associated with remaining with incumbent suppliers (loss-aversion, cognitive effort and implied endorsement). Two nationally representative survey experiments were conducted in Great Britain (n=1042, n=762). Results showed that participants were significantly more likely to engage with local energy suppliers under a multiple-supplier model than the current single supplier model. In one experiment, consumers' preference for adding a local supplier under a multiple-supplier model was so strong that it overcame default effects. The perception that the supplier has been recommended (i.e. implied endorsement) was the most robust mechanism associated with remaining with default suppliers, suggesting that explicit endorsement of local energy suppliers may encourage

uptake. Results suggest multiple-supplier models are likely to be a promising avenue for increased energy market engagement.

Keywords: Local energy, behavioural economics, cognitive biases, future energy markets, survey experiment, multiple-supplier models

1. Introduction

Decentralised energy is increasingly referenced in UK government strategy documents as having a key role to play in the energy transition (BEIS and Ofgem, 2017; Ofgem, 2017*a*). Similar trends have been observed internationally (Schmid et al., 2019; van der Schoor and Scholtens, 2015; Zhang et al., 2017). Technologies enabling decentralisation are known as distributed energy resources (DERs): renewable generation and storage units that connect directly to distribution networks (Akorede et al., 2010). The increasing prevalence of DERs is resulting in the emergence of new business models in the energy market. One example is local energy suppliers: “an organisational form with either the legal ability, or in partnership with another agency with that ability, to supply electricity to commercial and domestic consumers predominantly within a single distribution network region or group of regions at the sub-national scale” (Hall and Roelich, 2016, p.287). As well as supporting decarbonisation (Chiradeja and Ramakumar, 2004), effective utilisation of local energy can reduce constraints on the grid at peak times, avoiding expensive infrastructure upgrades (BEIS and Ofgem, 2017). Under some conditions, local energy can even help achieve social policy objectives, such as alleviating fuel poverty, strengthening community ties, and empowering consumers (Wiersma and Devine-Wright, 2014; Gui and MacGill, 2018; Ofgem, 2017*a*).

However, local energy is also raising challenges for the regulatory frameworks governing the British energy market. Industry codes and regulations revolve around the ‘supplier hub’ principle: licensed energy suppliers act as an interface between domestic consumers and a centralised energy system, with domestic consumers contracting with one energy supplier at a time (Ofgem, 2018*a*). This model has

been criticised for limiting innovation in the energy market, including opportunities for local suppliers (Elexon, 2019; Nolan, 2017; Spence, 2018). Whilst concerns about regulations stifling new market entrants are not new (Littlechild, 2005), Ofgem, the UK energy regulators, are questioning whether the supplier hub model remains fit for purpose in light of increasing decentralisation (Ofgem, 2018a; Nolan, 2017).

One solution is a multiple-supplier model (MSM) (Spence, 2018). This is a change to the balance and settlement code that would enable domestic consumers to take on more than one energy supplier. Under this model, households could purchase available electricity from local DER via a local supplier, with their current national-level supplier meeting the rest of their demand. As well as supporting local supplier models, this could also facilitate new business models, such as households with solar photovoltaic panels trading energy with neighbours in a peer-to-peer (P2P) system, smart appliances that include the energy to power them as part of a service, or suppliers dealing exclusively with electric vehicle charging (Spence, 2018).

Many industry stakeholders concur that the current supplier hub model is unlikely to deliver the innovations needed for the energy transition (e.g. Bristol Energy, 2018; DLT, 2018; ELEXON, 2018; Limejump, 2018; Northern Powergrid, 2018; Smartest Energy, 2018). However, views on the value of MSMs remain mixed. Some advocate MSMs as a means of integrating DER into a centralised grid (e.g. ELEXON, 2018; Gemserv, 2018). Others, particularly larger incumbents, highlight the benefits that have been delivered by the supplier hub model and the challenges and cost of moving away from a model enshrined in industry codes and regulations (e.g. Centrica, 2018; Energy, 2018; E.ON, 2018; ICoSS, 2018; MRA, 2018). Furthermore, consumer advocacy groups, among others, have expressed concerns that the additional complexity introduced by an MSM could risk increasing consumer disengagement, disadvantage disengaged consumers, or lead to a lack of accountability for suppliers in the absence of a single point of contact (Citizens Advice, 2018; Sustainability First, 2018; Octopus Energy, 2018). Given widespread disengagement in the current energy market (Ofgem, 2018b), if

changes are to be made to the supplier hub model, it is important to understand how consumers might respond to more complex alternatives.

Despite the relevance of understanding attitudes towards MSMs, this topic has received little attention in the academic literature. This paper draws from a behavioural economic framework to offer the first stated-preference evidence on consumer preferences regarding MSMs. The main aims of the study were 1) to assess the attractiveness of a multiple-supplier model for British consumers; and 2) to understand causally whether British adults would be more likely to engage with local energy suppliers in an MSM compared to the current single supplier model. An additional aim of the study was to explore the role of default effects and cognitive biases associated with remaining with incumbent suppliers (loss-aversion, cognitive effort and implied endorsement) in this specific context. Two pre-registered, nationally representative online survey experiments (n=1042, n=762) were conducted in Great Britain.

2. Literature review

2.1. Future energy markets and consumer attitudes

A broad and cross-disciplinary literature is beginning to examine how the increasing prevalence of DERs is changing the energy retail market. Strands of this literature range from the role of technologies in enabling business model innovation (Mazur et al., 2019), to sources of value in the energy markets of the future (Wegner et al., 2017; Hall and Foxon, 2014), and the changing role of utility companies (Hannon et al., 2013; Apajalahti et al., 2015; Richter, 2012; Fuentes-Bracamontes, 2016). Whilst there has been work on how alternatives to the supplier hub model might facilitate the rise of local energy supply models (e.g. Hall and Roelich, 2016; Gui and MacGill, 2018), studies in this vein have tended to take a whole-systems perspective or focus on the role of energy utility companies, rather than consumers.

Although there is a large body of literature across Europe and the UK examining consumer attitudes to local energy more broadly (e.g. Koirala et al., 2018; Walker et al., 2010; Kalkbrenner and Roosen,

2016; Mundaca et al., 2018; Rogers et al., 2008), MSMs have not been explicitly discussed in this literature. Much attention has been devoted to ‘community’ energy: a subset of local energy characterised by involvement of local stakeholders as investors or volunteers (Walker and Simcock, 2012). Although receiving electricity from a community energy scheme could be supported by an MSM, many of these projects centre around a community-driven approach requiring more active citizen participation than switching to or adding on a new local supplier. Regarding local suppliers specifically, a survey (n=418) conducted by consultancy Energy Systems Catapult (2019) found that 58% of participants expressed positive views about the concept of local energy suppliers, although the study did not ask whether participants would switch to such a service themselves. In Germany, where municipal supply models are relatively common, several studies have found evidence of higher willingness to pay for electricity generated locally (Sagebiel et al., 2014; Kalkbrenner et al., 2017). Nonetheless, none of these studies consider alternative energy retail market models such as MSMs. Furthermore, given that the German studies recruited from the German population and Sagebiel et al. (2014) used a non-representative convenience sample, these findings are unlikely to generalise to the British market. The only study in the UK to implicitly address MSMs focuses on P2P energy trading rather than local suppliers. Fell et al. (2019) conducted a nationally representative survey experiment to test willingness to participate in P2P under various conditions, including the proportion of electricity provided by the scheme, implying that the remainder would be provided by another supplier. Stated willingness to participate ranged from 52-66.8% across conditions. Whilst these results suggest consumers are likely to be open to the idea of receiving electricity from multiple providers, engaging in P2P differs from switching to a local supplier.

In short, there is a research gap regarding consumer attitudes towards MSMs and how this might affect engagement with local energy suppliers in the UK. In a competitive market, the success of MSMs in supporting local suppliers will largely depend on the willingness of consumers to engage with them. Despite high stated preferences for energy generated from renewable sources (BEIS, 2019) and alternative models such as P2P trading (Fell et al., 2019), current evidence points towards

widespread disengagement in the UK energy market. 54% of consumers remain on default standard variable tariffs (SVTs) - typically the most expensive (Ofgem, 2018b). Furthermore, consumers who do switch tariff are more likely to switch tariff with the same supplier than switch to a new supplier (Ofgem, 2017b). As the energy market becomes increasingly complex, Ofgem have expressed concerns that disengagement might stifle uptake of new suppliers (Ofgem, 2011, 2018a). To gain insight into how consumers sticking with incumbent suppliers might affect uptake of local energy under an MSM, it is important to understand the mechanisms behind this tendency.

2.2. Behavioural economics and default effects

One commonly applied model to understand consumer decision making in energy demand is a microeconomic model of utility maximisation and rational choice (Wilson and Dowlatabadi, 2007). Classic economic theory assumes that consumers analyse the costs and benefits of every option and make choices that maximise their utility subject to budget constraints. Accordingly, consumers that could maximise financial gains or other forms of utility by switching energy supplier should do so. For classical economists, consumers' failure to switch to suppliers better aligned with their preferences can be best explained by market failures such as incomplete information, transaction costs and imperfect market competition (Gillingham and Palmer, 2014).

However, a large body of experimental and field evidence suggests that individuals systematically fail to make decisions consistent with a rational choice model of decision making, under circumstances that cannot be explained by market failures (Camerer et al., 2003). For instance, in the field of tariff switching, natural experiments in Germany found that when default electricity tariffs were changed to a 'green' tariff less than 5% of customers switched back to the previous 'grey' tariff (Pichert and Katsikopoulos, 2008). In these cases, market failures, such as insufficient information about alternatives or the transaction costs of switching could offer a 'rational choice' explanation for the pervasiveness of the default tariff (Gillingham and Palmer, 2014). However, when Pichert and Katsikopoulos (2008) reproduced the scenario in a hypothetical choice experiment, giving

participants full information and reducing transaction costs to the click of a button, defaults were still found to have profound effects. These findings have been conceptually replicated several times (Momsen and Stoerk, 2014), including a larger, pre-registered experiment (Vetter and Kutzner, 2016) and a large-scale (n=41,952) randomised control trial in the field (Ebeling and Lotz, 2015). This suggests that rational choice cannot fully explain decisions to stick with incumbent suppliers.

Behavioural economics offers an alternative explanation. This theory assumes that consumers' failures to make decisions aligned with their preferences are due to 'bounded' rationality: individuals attempt to make rational decisions, but apply mental short-cuts, known as 'cognitive biases' (Simon, 1997). Under this framework, the tendency to remain with incumbents can be conceptualised in terms of 'the default effect'. Defaults refer to "the choice alternative a consumer receives if he/she does not explicitly specify otherwise" (Brown et al., 2004, p.529). Overwhelmingly, experimental and field evidence suggests that defaults 'stick'; that is, people do not tend to move away from the option that is automatically assigned to them (see Johnson and Goldstein, 2003). In the energy context, defaults include tariffs consumers are automatically enrolled on such SVTs, or contracts taken over after moving to a new house.

Under the behavioural economic framework, three mechanisms have been proposed to explain why individuals stick with defaults: cognitive effort; implied endorsement; and loss-aversion (Johnson and Goldstein, 2003; Jachimowicz et al., 2019). Cognitive effort suggests people stick with defaults because of the mental effort required to process alternatives and form preferences (Tversky and Kahneman, 1974; Samuelson and Zeckhauser, 1988; Thaler and Sunstein, 2008). Implied endorsement suggests that individuals stick with defaults because they are perceived as recommended to them by the default-setter (McKenzie et al., 2006). For instance, if a supplier switched all its customers to a certain tariff unless they opted-out, this might be perceived as an implicit endorsement of that particular tariff. Loss-aversion suggests individuals perform comparative evaluations against the status quo when making decisions and emphasise potential losses more than potential gains - financial or otherwise (Tversky and Kahneman, 1991). Several studies in Europe and qualitative work from Ofgem point to

loss-aversion as a key explanation for consumers remaining with incumbent suppliers (Ek and Söderholm, 2008; Juliusson et al., 2007; Ofgem, 2011). A survey experiment by Nicolson et al. (2017) also found that loss-aversion negatively impacted willingness to switch to a Time-Of-Use tariff in a nationally representative sample of British energy bill payers.

2.3. Default mechanisms and MSMs

Whilst there is evidence for each mechanism in isolation, few studies have attempted to disentangle them (c.f. Dinner et al., 2011) and none in the context of switching to new energy innovations. As evidence suggests that cognitive biases already stifle tariff switching (Ofgem, 2011), it is plausible they would also impact switching to local suppliers. However, it is unclear how an MSM might affect this. For instance, if cognitive effort is the prevailing explanation for consumers' tendency to stick with their current supplier and tariff, the complication of a second supplier may further discourage consumers from engaging with the market. If loss-aversion offers a more convincing explanation, then an MSM might be able to mitigate this, by allowing consumers to 'add on' a local supplier to their current tariff instead of switching entirely. Finally, if implied endorsement can explain consumers' tendency to stay with defaults, this might suggest potential for growing local energy through explicit recommendations. Understanding the mechanisms underlying default effects can therefore provide insight into how MSM might affect uptake of local energy. Specifically, this paper answers the following research questions:

1. Are British adults more likely to engage with a local supplier in a multiple-supplier model than under the current single supplier model?
2. To what extent can cognitive effort, implied endorsement and loss-aversion explain default effects in the context of switching to local energy suppliers in a multiple-supplier model?

The first experiment in this study focuses on assessing the attractiveness of an MSM and also explores the mechanisms underlying consumers' tendency to stay with defaults. Given the large evidence base attesting to the power of defaults (Johnson and Goldstein, 2003; Jachimowicz et al., 2019), defaults would be expected to have a strong effect on consumer willingness to switch in the context of local energy suppliers.

H1: Individuals will be more likely to choose the supplier presented as the default than the supplier presented as the alternative.

Following behavioural economic theory on the mechanisms driving default effects, three hypotheses can also be derived:

H1a: Higher levels of cognitive effort will be correlated with a decreased likelihood of choosing the option presented as the default.

H1b: Higher perceptions of implied endorsement of the default option will be correlated with an increased likelihood of choosing the option presented as the default.

H1c: Loss-aversion will be correlated with an increased likelihood of choosing the option presented as the default.

In H1a, there is a possibility of reverse causality i.e. there may be some people who carefully consider both options and make an active choice to remain with the default. It was assumed that participants will only consider both choices if they put in sufficient cognitive effort. The presence of those who put in a high level of cognitive effort and still choose the default would weaken the overall effect of cognitive effort, but if there is support for this hypothesis, there will be an overall trend towards cognitive effort weakening the default mechanism.

The second experiment in the study focuses on understanding whether consumers would be more likely to engage with local energy suppliers in the current single supplier model or an MSM. It also tests the possibility that an MSM might be able to mitigate the effects of loss-aversion associated with switching energy suppliers (Nicolson et al., 2017; Ofgem, 2011), by enabling consumers to 'add on' a local supplier to their current tariff, without having to give anything up.

H2a: Individuals will be more likely to switch to a local energy supplier in an MSM than a single supplier model.

H2b: In a single supplier model, loss-aversion will be correlated with decreased likelihood of switching to a local energy supplier. In a multiple-supplier model, there will be no relationship between loss-aversion and the likelihood of switching to a local energy supplier.

3. Methods and data

Procedures are reported in accordance with CONSORT guidelines.

3.1. Participants and procedure

Multiple-supplier models are not commercially available in the UK market (Ofgem, 2018a); for this reason, opportunities to conduct experiments in the field are extremely limited. Instead, a stated preferences approach was employed: two population-based, between-participant survey experiments were conducted to test the hypotheses. Stated preference methods are subject to concerns about external validity (Louviere et al., 2000). However, given the novelty of multiple-supplier models, the use of a survey experiment allows insight into whether there is consumer interest in this area before undertaking more resource intensive work in a more ecologically valid environment.

Both experiments were delivered online. Online surveys have several advantages over laboratory, telephone or post: they are less resource intensive, allow access to larger samples, and typically have higher response rates (Wright, 2005). As participants are not directly observed, online delivery may also mitigate the risk of Hawthorne effects and social desirability bias.

As Northern Ireland does not share an electricity network with the rest of the UK, the target population was adults in GB. Market research company Populus was commissioned to recruit a broadly representative sample of 2000 participants from a panel of approximately 165,000 individuals. Inclusion criteria were: 1) aged 18+; 2) living in England, Scotland or Wales. Participants were paid for participating in surveys (£1/5 minutes) and entered into a monthly £250 prize draw. Quota sampling was used to achieve a nationally representative sample, with quotas set on age, gender, region, and social grade. Although the sample was broadly representative in terms of these demographics, some risks of selection bias remain. Participants must have internet access, excluding approximately 10% of the population (ONS, 2018). As participants also voluntarily participate in online surveys, certain characteristics may be disproportionately represented.¹

1200 participants were randomly assigned to the default mechanisms experiment (experiment one) and 800 were assigned to the consumer preferences experiment (experiment two).² Sample sizes were estimated for 95% power at the 5% confidence interval, based on effect sizes found in experiments on defaults in tariff switching (Momsen and Stoerk, 2014; Vetter and Kutzner, 2016; Ebeling and Lotz, 2015; Pichert and Katsikopoulos, 2008). As the first experiment tests more hypotheses, a larger sample was required to achieve sufficient statistical power. Randomisation was carried out by Populus using block randomisation, allocating participants to the least full experimental condition.

¹ Due to the sampling method and use of a topic-blind omnibus survey, response rates do not offer useful information about the representativeness of the experiments and are not reported.

² After exclusion criteria were applied, 13.5% of participants in the first experiment were excluded and 11% of participants in the second, leaving sample sizes of n=1042 and n=762 respectively. Group allocations are discussed in terms of original planned sample sizes in the methods section, sample sizes used in the analysis are reported in full in the results section.

During the survey design process, individual cognitive interviews were conducted with a small convenience sample with no expertise in the energy field (n=8). Cognitive interviewing is a method of field research which seeks to understand the cognitive processes involved in answering survey questions, in order to detect problems in survey questions and minimise response error (Willis, 2005). After interviewing half of the non-expert sample, a focus group (n=8) was conducted with experts in energy and behaviour at the UCL Energy Institute to further refine questions and treatments. The survey was adapted iteratively, incorporating feedback from both samples.

Before any data were collected, a pre-analysis plan (PAP) was registered online at as.predicted.org. PAPs are increasingly being recognised as an important step to improve transparency in the scientific process (Chambers, 2017; Huebner et al., 2017) and should allow “both the researchers and others to have a higher level of confidence that reported results do not reflect statistical noise hand-picked from the data” (Chuang and Wykstra, 2015, p.2).

3.2. *Treatments*

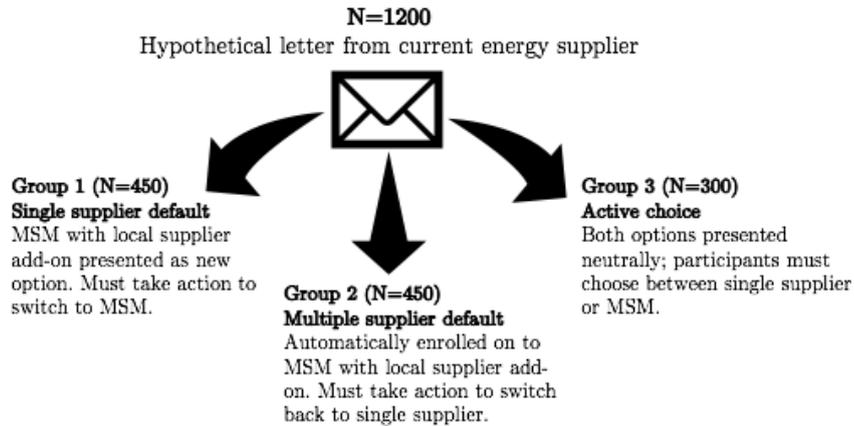
Treatments are summarised in sections 3.2.1 and 3.2.2. and are reproduced in full in Appendix A.

3.2.1. *Experiment one: Why do consumers stay with defaults?*

The first experiment was designed to understand willingness to engage with local suppliers in an MSM and to explore cognitive biases associated with remaining with defaults. The experimental set-up follows that of Pichert and Katsikopoulos (2008) and Vetter and Kutzner (2016).

Participants were randomly assigned to one of three groups. In all groups, participants were asked to imagine that they receive a letter from their current energy supplier. This letter informed them that their supplier is partnering with a new local energy supplier and offered the opportunity to add on the local energy company’s services in an MSM, at no additional cost.

Figure 1: Diagrammatic representation of research design: experiment one



Group one ($n=450$), the ‘single supplier default’ group, is equivalent to an opt-in condition. Participants were told that they are currently enrolled on the ‘single-provider’ model, where all their electricity is provided by their current supplier only. They were offered the opportunity to opt-in to the ‘combined-provider’ model (effectively an MSM), which would allow them to add the local energy company’s services to their current tariff. Participants were informed that they would have to go online and let their energy supplier know if they would like to switch to the MSM. In the survey, the single supplier model was pre-selected to evoke the transaction costs of switching.

Group two ($n=450$), the ‘multiple-supplier default group’ is equivalent to an opt-out condition. Participants were told they had been automatically enrolled on to the new ‘combined-provider’ model and that they would have to go online and inform their energy supplier if they would like to switch back to the single-provider service. In this condition, the MSM option was pre-selected. Although automatically enrolling customers on to an MSM in partnership with a local supplier is not currently a feasible policy option, this hypothetical scenario allows for testing cognitive biases at work in opt-out defaults.

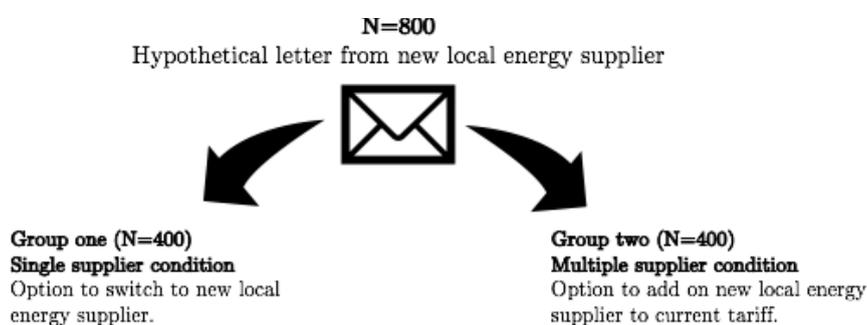
Group three ($n=300$) was an ‘active choice’ group in order to assess participants’ preferences when there is no default. Participants were asked to choose between the two models, which were both presented neutrally, with no pre-selected boxes. As this group was not included in the main analysis, a larger number of participants were allocated to the single and multiple-supplier default groups to

improve statistical power. The order in which the options were presented was randomised and there was no evidence that the choice order affected the decision made.

3.2.2. Experiment two: Consumer preferences for supplier models

The second experiment focuses on whether British adults are more likely to engage with local energy suppliers in an MSM than the current single supplier model and tests the possibility that an MSM might mitigate loss-aversion associated with energy supplier switching.

Figure 2: Diagrammatic representation of research design: experiment two



Participants were randomly assigned to one of two conditions. In both groups, participants were asked to imagine that they have received a letter from a new local energy supplier advertising their services. Participants were also informed that the options offered would cost the same as their current tariff.

In the ‘single supplier’ condition (n=400), the local energy company works under the current supplier hub model, buying from the wholesale market when there is not enough locally generated energy to fill their customers’ needs. Participants were asked whether they would prefer to remain with their current tariff or switch to the local energy company by going online and informing the local supplier that they would like to switch.

In the ‘multiple-supplier’ condition (n=400), the local energy company offers participants the option of adding their services on to their existing contract in an MSM. Participants were asked to choose between staying with their current tariff or adding the local supplier to their current tariff. Again, participants were told that if they wanted to add on the local supplier’s services, they could do so by

going online and informing the supplier. In both conditions, the option of remaining with their current tariff was pre-selected, to elicit a transaction cost associated with switching.

3.3. *Dependent variable*

The dependent variable (willingness to switch from the default) was measured by participants' answer to the hypothetical scenario they were presented with. Five true/false manipulation checks were included to ensure participants had understood key aspects of the scenario.

3.4. *Independent variables*

Participants also completed a closed-ended, eleven item questionnaire, including measures of each default mechanism. These measures are summarised below and reproduced in full in Appendix B.

3.4.1. *Effort*

The physical effort of switching is controlled for; transaction costs of switching are the same across all conditions that have a default: a click of a button. A measure of cognitive effort was drawn from Cooper- Martin (1994), who tested several measures and identified five questions with the greatest validity. These questions have also been used by Heidig et al. (2017) to capture differences in an experimental manipulation of cognitive effort.

Cronbach's alpha analysis of self-reported items was conducted (0.79 for experiment one and 0.77 for experiment two). Item five ("I found this decision difficult to make") was dropped due to poor correlation with other items. Treating the five-point Likert scale as continuous (Johnson and Creech, 1983), a mean average of the remaining four items was used to capture cognitive effort (1=low cognitive effort, 5=high cognitive effort). "Don't know" responses were excluded; if a participant answered "don't know" for an individual item, a mean average of remaining items was taken.³

³ The PAP stated that cognitive effort would be coded by combining participant rankings for the time taken to respond to the question and rankings of the self-reports. Correlation between rankings of time taken and rankings of the cognitive effort self-reports was statistically significant ($p < 0.01$), but too weak to be considered consistent (23% for experiment one; 12% for

The mean average of cognitive effort was 3.64 (SE=0.02) in experiment one; 3.78 (SE=0.23) in experiment two.

3.4.2. *Implied endorsement*

As implied endorsement is primarily associated with automatic enrolment (McKenzie et al., 2006), this mechanism was only measured in experiment one. Implied endorsement was assessed through a single-item self-report on a five-point Likert scale (McKenzie et al., 2006; Dinner et al., 2011). As implied endorsement pertains to individuals' perception of the default being endorsed, self-reports are appropriate. To avoid heightening the saliency of implied endorsement, the measure was embedded in a series of other statements about the reasons behind participants' decisions. These were drawn from reasons stated by participants in Pichert and Katsikopoulos (2008)'s laboratory study when asked why they made their choice.

3.4.3. *Loss-aversion*

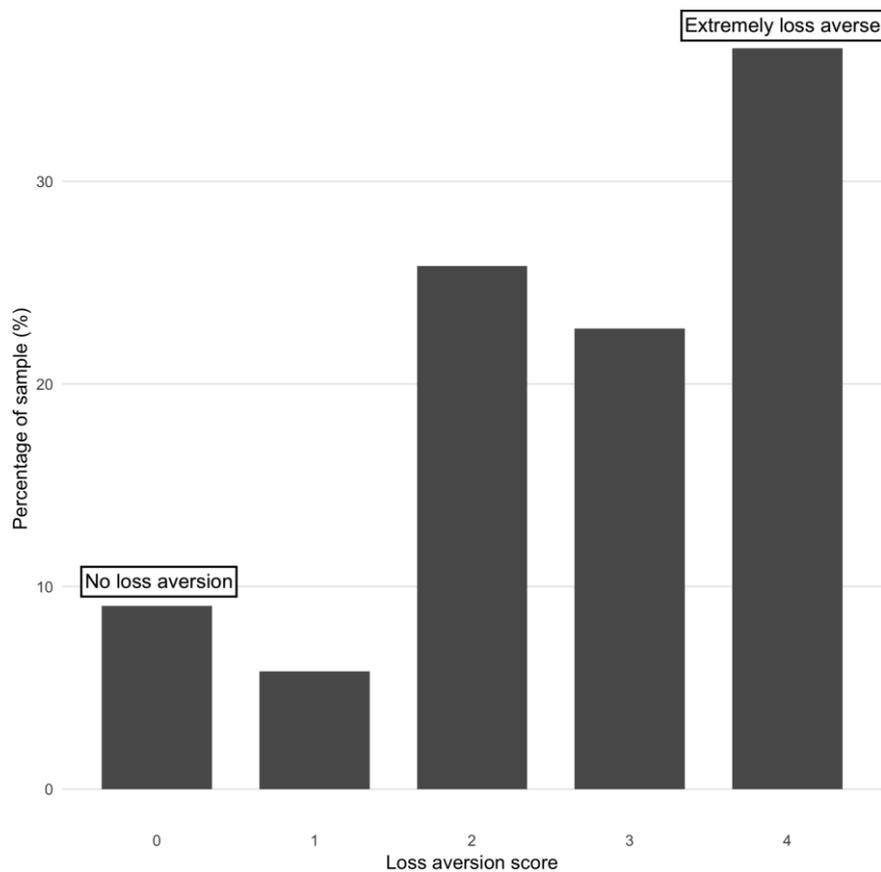
Loss-aversion was measured by asking participants whether they would accept or reject a series of gambles involving hypothetical losses and gains. These questions derive from standard financial decision problems (Gächter et al., 2010) and were adapted by Nicolson et al. (2017) and Fell et al. (2017) to measure loss-aversion in a representative sample of British energy bill payers. As rational choice theory implies that individuals are risk neutral when it comes to small stakes, the gambles involve small sums of money to avoid evoking risk-aversion, rather than loss-aversion. Gambles 1 - 4 are net positive and rejection of any of these gambles was taken to indicate loss-aversion. Rejection of the 50/50 win/lose gamble (5) was not considered to indicate loss-aversion. Coding of loss-aversion has differed across studies. Gächter et al. (2010) coded loss-aversion according to which specific gambles were rejected. However, this assumes that preferences are linear and that participants would not, for

experiment two). For this reason, the measures were not combined and only the self-reported measure of cognitive effort was used in the analysis.

instance, reject gambles involving lower losses and accept gambles involving higher losses. Approximately 14% of the sample exhibited inconsistent preferences. Similar inconsistencies were observed by Nicolson et al. (2017), who coded loss-aversion as a dummy variable indicating whether or not an individual showed any loss-aversion. To enable a greater level of nuance, following Fell et al. (2017), loss-aversion was coded as a discrete scale indicating the number of net positive gambles rejected (0=no loss-aversion, 4=extremely loss-averse).

Across the sample, 8.9% of participants showed no evidence of loss-aversion: 6.1% of the sample acted in accordance with classic economic theory, accepting all net positive gambles and rejecting the net negative gamble, and a further 2.8% accepted all gambles, indicating that they may be risk-positive. 91% of participants were found to have some level of loss-aversion (i.e. rejected at least one net positive gamble). Levels of loss-aversion varied with 35.6% of the sample being extremely loss-averse (i.e. rejected all gambles). This is similar to the levels of loss-aversion observed by Nicolson et al. (2017) in a representative sample of British energy bill payers.

Figure 3: Annotated distribution of loss-aversion scores



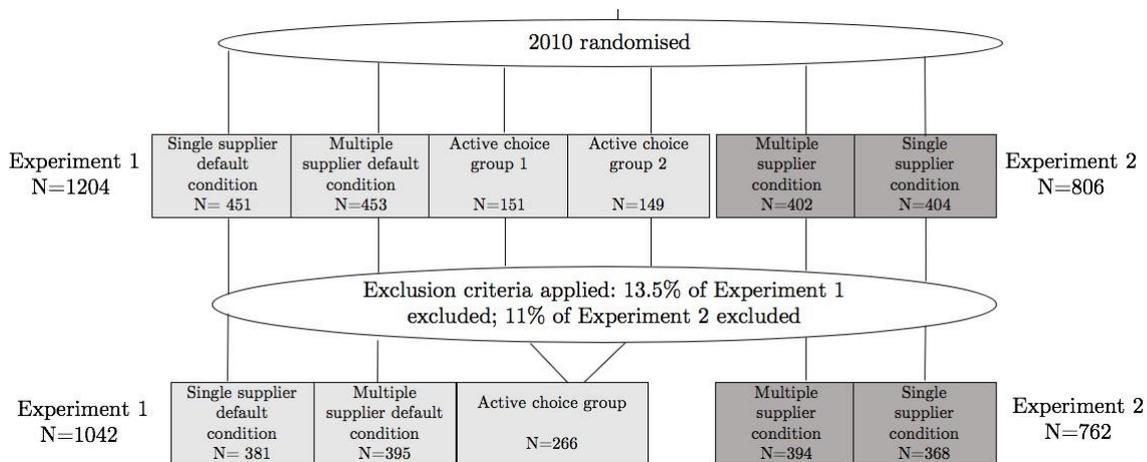
3.4.4. Covariates

Data was collected on covariates shown to be associated with cognitive biases or switching energy supplier or tariff. These included demographic variables age, income, tenure, education and gender (Ofgem, 2018b; Rau, 2014) as well as previous energy tariff and supplier switching behaviour, moving house, and satisfaction with current energy supplier (Ek and Söderholm, 2008; Walsh et al., 2005).

4. Sample

As pre-registered, participants who failed more than 50% of manipulation checks were excluded, as were those who answered “don’t know” to all cognitive effort self-reports and to the implied endorsement measure in experiment one.⁴ In total, 13.5% of participants in experiment one were excluded, 11% of participants in experiment two. Attrition was evenly balanced between groups.

Figure 4: Participant flow



⁴ As time taken was no longer used to construct the cognitive effort measure, the pre-specified criteria of excluding participants who took longer than 3 standard deviations above mean average to respond was not applied.

Figure 5 compares the sample to the Office for National Statistics 2018 population estimates for Great Britain (GB), disaggregated by gender and age group (ONS, 2019). Visual analysis shows that there was a slight under-representation of males 25-34. As the sample is otherwise broadly representative of the population of interest with respect to these socio-demographic characteristics, survey weights were not applied. Table 1 presents summary statistics for additional covariates included in the analysis, after the exclusion criteria had been applied. To test for imbalance of covariates across treatment groups within the same experiment, Chi-squared (for dichotomous or categorical variables) and ANOVA tests (for continuous variables) were run. There were no statistically significant differences between groups (see Appendix C for full results).

Figure 5: Comparison of sample to GB population by age and gender

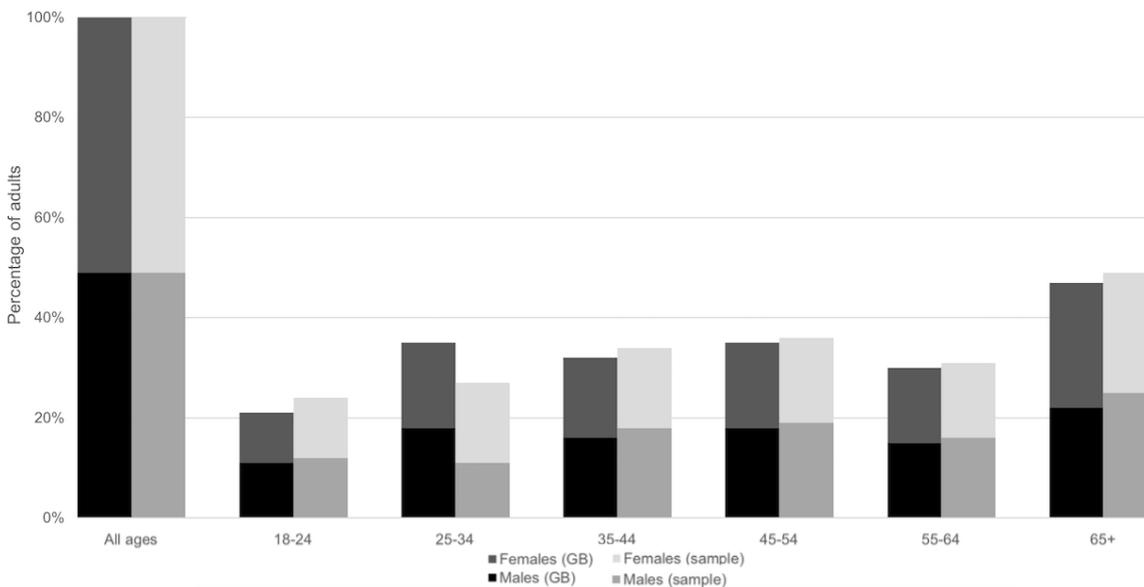


Table 1: Participant characteristics by experiment

Categorical variables	Percentage of sample	
	Experiment 1	Experiment 2
Education		
No secondary school education (reference category)	4%	4%
Secondary	56%	56%
Degree or higher	39%	39%
Gender		
Female (reference category)	51%	50%
Male	49%	50%
Tenure		
Home-owner (reference category)	61%	63%
Private rent	17%	17%
Rent free	3%	2%
Social rent	19%	
Income		
Up to £21,000 (reference category)	36%	38%
£21,001 - £41,000	44%	45%
£41,001 +	12%	10%
Moved house within 12 months		
No (reference category)	91%	91%
Yes	9%	9%
Switched energy tariff or supplier		
Never (reference category)	41%	38%
Don't know	16%	15%
More than 12 months ago	37%	41%
Within last 12 months	5%	4%
Within last 3 months	1%	2%
Continuous variables	Mean (Standard error)	
Age	49 (0.54)	50 (0.65)
Satisfaction with energy supplier (1= extremely dissatisfied, 5 = extremely satisfied)	3.74(0.03)	3.65 (0.04)

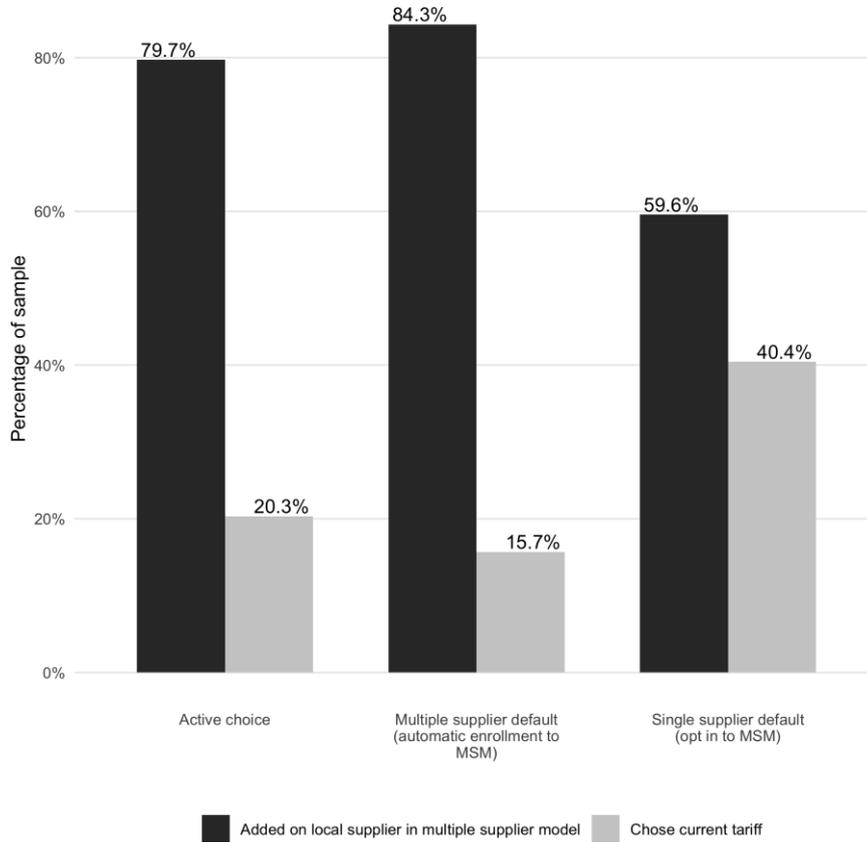
5. Results

5.1. Experiment one: Why do consumers stay with defaults?

5.1.1. Supplier model choice

H1: Individuals will be more likely to choose the energy supplier model presented as the default than the energy supplier model presented as the alternative.

Figure 6: Experiment one: Supplier model chosen by condition



Contrary to expectations, the majority of participants chose to add on a local supplier in an MSM in all conditions, including the single supplier default group. Nonetheless, the proportion of the sample that chose the MSM was greater in the active choice (79.7%) and multiple-supplier default (84.3%) groups compared to the single supplier default group (59.6%).

A 2X3 Chi-squared test showed the overall effect of the condition on participants' choices was large and statistically significant ($\chi^2 = 67.6, p < 0.01$). Pairwise Chi-squared tests were applied to find out which groups were statistically significantly different from one another.⁵ Only the single supplier default group was found to be statistically significantly different from both the multiple-supplier default group ($\chi^2 = 57.79, p < 0.01$) and the active choice group ($\chi^2 = 28.15, p < 0.01$). In other words,

⁵ The Bonferroni correction for multiple comparisons was applied in order to reduce the risk of false positives, meaning that a significance level of $\alpha = 0.016$ ($0.5/3$) was used for the multiple comparisons (Dunn, 1961).

when participants were not given a default option, they were as likely to choose the MSM as those who were automatically enrolled on to it. Although the majority of participants in all conditions chose the MSM, participants who had to take action to choose this model were less likely to engage with it.

Logistic regression models (table 2) showed that, when holding other covariates constant, participants assigned to the single supplier default group were approximately 8 times more likely to switch to the MSM than those in the multiple-supplier default group were to switch back to their current tariff ($p < 0.01$). Overall, these results suggest a strong preference for the MSM and H1 is only supported in the multiple-supplier default group.

5.1.2. *Default mechanisms*

The next set of analyses reports the associations between each cognitive bias and the likelihood of staying with the assigned default, for the multiple-supplier default and single supplier default groups (n=776).

Table 2 presents the results of logistic regression models. Coefficients are presented in odds ratios (OR) with confidence intervals in parentheses. An OR of 1 indicates that there is no association between the predictor and the outcome (switching away from the default: 1=switch, 0=no switch); an OR >1 indicates a positive association; and an OR <1 indicates a negative association. Model 1 includes the treatment effect and all default mechanisms; model 2 includes covariates specified in the PAP; model 3 adds an interaction term between the each default mechanism and treatment group assignment, in order to isolate the effects within each group.⁶

⁶ This analysis was not included in the PAP. However, given the large and statistically significant effect of the treatment group suggests that the effect of the default differed between groups, it seemed likely that the default mechanisms also were also playing a different role in each group.

Table 2: Results of logistic regression models - experiment one (default mechanisms)

	<i>mechanisms)</i>		
	<i>Dependent variable</i>		
	Switching away from default		
	(1)	(2)	(3)
Single supplier group	7.795 (5.57-11.038)	8.273 (5.789- 11.983)	4.007 (0.325- 54.205)
Cognitive effort	1.243 (0.981-1.581)	1.318 (1.018-1.714)	1.760 (1.158-2.748)
Implied endorsement	0.912 (0.768-1.082)	0.947 (0.788-1.138)	0.496 (0.358-0.680)
Loss-aversion	1.038 (0.910-1.185)	1.042 (0.900-1.209)	1.308 (1.009-1.730)
Single supplier group: cognitive effort			0.654 (0.376-1.118)
Single supplier group: implied endorsement			2.700 (1.833-4.020)
Single supplier group: loss-aversion			0.719 [*] (0.518-0.982)
Controls		✓	✓
Observations	776	714	714
Log Likelihood	-426.388	-385.028	-368.747
Akaike Inf. Crit.	862.776	810.055	783.493

Note: For brevity, odds ratios are only presented for the main independent variables. See appendix C for full results.

H1a: Higher levels of cognitive effort will be correlated with a decreased likelihood of choosing the option presented as the default.

Cognitive effort only had a statistically significant effect ($p < 0.05$) in the multiple-supplier default group, with an OR of 1.76. In other words, in the multiple-supplier default group, those who exercised greater levels of cognitive effort were more likely to take action to switch back to their current tariff. However, in the single supplier default group, cognitive effort had no statistically significant effect. H1a is only supported in the multiple-supplier default group. This result was not robust to corrections for multiple hypothesis testing (see section 6.3).

H1b: Implied endorsement of the default option will be correlated with an increased likelihood of choosing the default option.

In the multiple-supplier default group, implied endorsement had a negative and statistically significant effect ($p < 0.01$) with an OR of 0.496: as expected in H1b, the more strongly a participant felt that the option they chose had been endorsed to them, the less likely they were to switch back to the single supplier model. Conversely, in the single supplier default group, implied endorsement was associated with an increased likelihood of choosing the MSM ($OR = 0.496 * 2.7 = 1.339, p < 0.01$). In other words, in the single supplier group, the more strongly a participant felt that the option they chose was endorsed, the more likely they were to switch to the MSM. H1b is only supported in the multiple-supplier default group.

H1c: Loss-aversion will be correlated with an increased likelihood of choosing the default option.

Loss-aversion was significant at the $p < 0.05$ level in both groups, in opposite directions. In the single supplier default group, loss-aversion was negatively associated with switching to the MSM, although this effect was very small with an OR of 0.94 ($1.308 * 0.719$). For participants in the multiple-supplier default group, loss-aversion had a positive and statistically significant association with switching back to the single supplier model, with a moderate OR of 1.308. In other words, the more loss-averse an individual is, the more likely they were to stay with the status quo, even when this involves taking action to switch back. This is in support of H1c. However, this result failed several robustness checks and should be treated with caution (see section 6.3).

5.2. Experiment two: consumer preferences for supplier models

H2: Participants will be more likely to switch to a local energy supplier in an MSM than in a single supplier model.

In the second experiment, where participants were hypothetically approached by a new local supplier, the majority of participants chose to stay with their current supplier in both the single and multiple-supplier model conditions. This is in contrast to the default mechanisms experiment, where the majority chose the MSM regardless of their assigned default. Nonetheless, in the multiple-supplier condition, 47.6% of participants chose to add on the local energy company's services, compared to

37.5% in the single supplier condition. The treatment effect was statistically significant ($p < 0.05$), with an OR of 1.52. In other words, participants were approximately one and a half times more likely to engage with local energy suppliers in a multiple-supplier model than the current single supplier model. This is in support of H2.

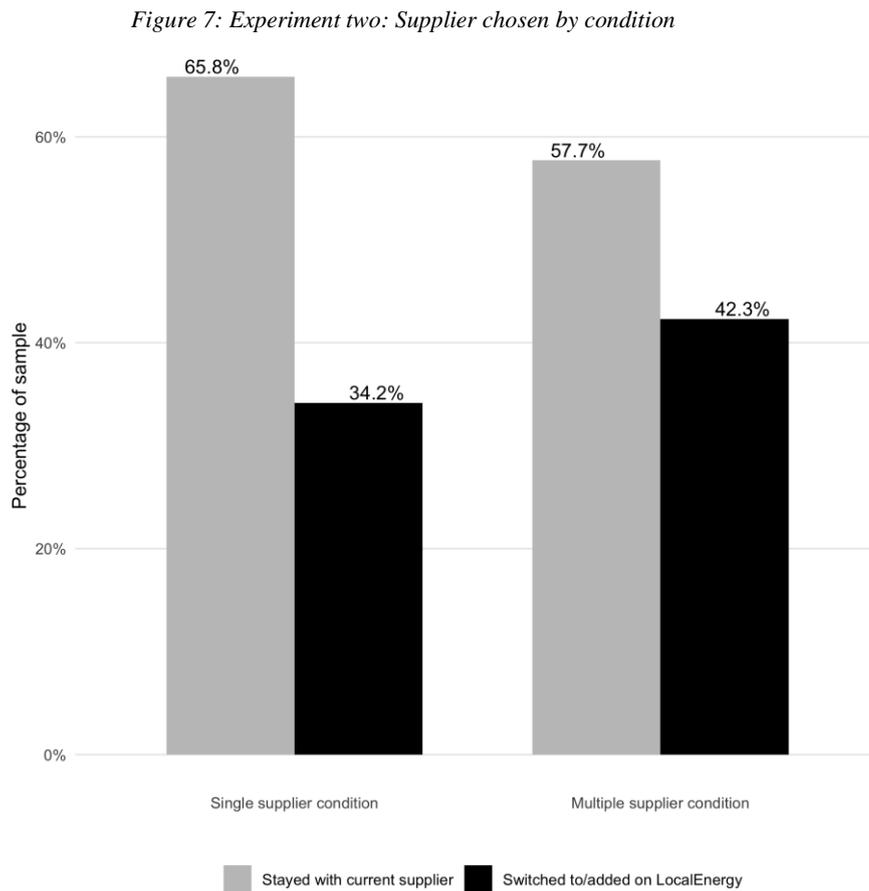


Table 3 presents the results of logistic regression models. Model one represents the effect of the treatment group when controlling for covariates. Model two includes an interaction term between the multiple-supplier condition and loss-aversion.

H2a: In a single local supplier model, loss-aversion will be correlated with decreased likelihood of switching to the local supplier. In a multiple-supplier model, there will be no relationship between loss-aversion and likelihood of adding on the local supplier.

As expected, loss-aversion did not have a statistically significant effect in the multiple-supplier condition. However, there was no effect of loss-aversion on likelihood of switching in the local energy supplier group either. Contrary to expectations, loss-aversion did not appear to play a role in switching to new local suppliers, whether in an MSM or under the current supplier hub model. H2a is therefore not supported.

Table 3: Results of logistic regression models - experiment two (consumer preferences)

	<i>Dependent variable</i>	
	Switching to/adding on local energy supplier (1)	(2)
Multiple-supplier condition	1.568*** (1.135, 2.167)	1.258 (0.582, 2.726)
Loss-aversion	0.970 (0.849, 1.109)	0.933 (0.778, 1.119)
Multiple-supplier condition: loss-aversion		1.085 (0.837, 1.406)
Controls	✓	✓
Observations	664	664
Log Likelihood	-429.538	-429.348
Akaike Inf. Crit.	897.075	898.697

Note: For brevity, odds ratios are only presented for the main independent variables. See appendix C for full results.

5.3. Robustness

Robustness tests were conducted to determine the sensitivity of the results to exclusion criteria and model specification. With regards to the supplier models chosen and the effect of the treatment groups, the results of both experiments were robust to all alternative specifications tested. Results of robustness checks for the findings on default mechanisms are summarised below and presented in full in Appendix C.

Table 4: Results of robustness tests

Robustness test	Experiment one (default mechanisms)	Experiment two (consumer preferences for supplier models)
1. OLS regression models instead of logistic to test for sensitivity to model specification	Effect of loss-aversion no longer significant at $P < 0.05$ level.	Robust (no effect of loss-aversion in either group)

2. Alternative specification of loss-aversion as a dichotomous variable	Effect of loss-aversion no longer significant at $P < 0.05$ level.	Robust (no effect of loss-aversion in either group)
3. Benjamini Hochberg correction for multiple hypotheses	The effects of cognitive effort and loss-aversion both no longer significant at $P < 0.05$ level. Implied endorsement was the only default mechanism that remained statistically significant	N/A
4. Regional fixed effects to control for potential omitted variables	Robust: strengthened the effect and statistical significance of all findings on default mechanisms	Robust
5. No exclusion criteria applied	Robust	Robust

6.

Discussion

6.1. Consumer engagement with local energy in an MSM

The main aim of this study was to understand whether consumers would be willing to engage with local energy suppliers in an MSM. Overall, the results indicate strong interest from consumers in adding a local supplier through a multiple-supplier model. The results of the consumer preferences experiment (experiment two) suggest that British adults would be more likely to engage with a local energy supplier in an MSM than under the current single supplier model. More strikingly, the default mechanisms experiment finds that consumers' preference for adding a local supplier under an MSM is so strong that it overrides default effects for all groups - this contradicts much of the experimental literature on default effects in energy supplier switching (e.g. Pichert and Katsikopoulos, 2008; Vetter and Kutzner, 2016; Momsen and Stoerk, 2014). This is particularly unexpected, given the vast literature on default effects, which overwhelmingly finds that defaults 'stick' (Jachimowicz et al., 2019). Due to the strength of this result, a replication study would be beneficial.

It must be noted that preferences expressed in a survey experiment may not translate into choices made in reality (Louviere et al., 2000). The transaction costs of switching are greatly reduced in a hypothetical environment and it may be that participants felt obliged to choose the MSM and local energy option due to perceived social desirability bias or experimental demand effects. However,

other stated preference studies on switching to ‘green’ energy suppliers have found large and persistent default effects, even in hypothetical scenarios with minimal transaction costs (e.g. Pichert and Katsikopoulos, 2008; Vetter and Kutzner, 2016; Momsen and Stoerk, 2014). This suggests that the overcoming of default effects in this study is more than an artefact of experimental demand effects. Nonetheless, further research is required to test whether these findings hold in a more ecologically valid environment. As MSMs are a pre-commercial area of energy research, opportunities for research into consumer behaviour in the field remain limited. Future research could involve partnerships with energy suppliers in an online environment (e.g. Ebeling and Lotz, 2015) to test how well stated preferences translate into consumer behaviour in a commercial environment. Further research is also needed to understand how additional factors such as price or receiving multiple bills might affect consumer willingness to engage with MSMs. Nonetheless, this study supports a growing body of evidence that consumers are showing an interest in local energy and alternative ways of interacting with the energy market (Fell et al., 2019; Energy Systems Catapult, 2019).

However, it is worth highlighting that in the second experiment, whilst participants were more likely to switch to a local energy supplier under an MSM than under the current supplier hub model, the majority of participants chose to stay with their current supplier. One key difference between the two experiments was the actor the participant receives the letter from: in the first experiment it is their current energy supplier, in the second it is the new local supplier. It must be emphasised that this was not a focus of this study and was not included in the PAP; in these experiments, the identity of the entity contacting the consumer arose from the need to construct plausible scenarios to test the research question relevant to each experiment. However, the difference in results suggests consumers’ relationships with actors endorsing MSM are also likely to influence willingness to engage with them. Future work could focus on experimentally manipulating the identity of the entity contacting consumers to understand how consumer relationships might affect willingness to engage with local suppliers in MSMs.

6.2. *Default mechanisms*

A secondary aim of this study was to explore the mechanisms behind consumers' tendency to stick with defaults. In the default mechanisms experiment, implied endorsement emerged as the most robust mechanism in both groups. In the multiple-supplier default group, where participants were automatically enrolled on to the MSM, implied endorsement was associated with an increased likelihood of staying with the MSM default. This supports work by McKenzie et al. (2006), who found experimental evidence that participants view defaults as 'endorsed' by the default-setter, as well as a meta-analysis of default effects, which suggested that defaults are most effective when they work through the implied endorsement mechanism (Jachimowicz et al., 2019). However, in the single supplier default group, the stronger a participant's feeling of implied endorsement for the option they chose, the more likely they were to switch to the MSM. This could suggest that participants perceived the MSM as being endorsed simply by virtue of being the newer option offered in the hypothetical letter.

Cognitive effort only appeared to have an effect in the multiple-supplier default group: those who invested more cognitive effort were more likely to switch back to the single supplier model, whereas in the single supplier default group, cognitive effort had no effect on the likelihood of individuals switching to the MSM. One possible explanation for the differing effects of cognitive effort could be the presence of 'rational non-switchers' in the single supplier group: individuals who carefully consider both options and decide that the default is the best option for them. However, there is no reason to expect rational non-switchers only to be present in the single supplier group. It is also possible that there were individuals in the multiple-supplier default group who resented automatic enrolment and made a conscious choice to reject this default (Lodge and Wegrich, 2016).

In the default mechanisms experiment (experiment 1), participants who were more loss-averse were more likely to choose the single supplier option, regardless of their default. In other words, loss-aversion was associated with a return to the status quo. This aligns with previous literature on

the effect of loss-aversion on supplier switching (Ek and Söderholm, 2008; Nicolson et al., 2017; Juliusson et al., 2007). However, this result did not withstand several robustness checks and further analysis revealed that this effect was driven almost entirely by participants who were extremely loss averse (i.e. rejected all gambles).

One explanation for the non-robust effect of loss aversion may be participants making differing assumptions regarding who their default supplier is. Some may have assumed it is the supplier assigned to them in the hypothetical scenario (the MSM), while others may have assumed it is their current single energy supplier. This could occur if some of those participants who were asked to imagine having been automatically enrolled on to the MSM still considered their current single energy supplier to be the status quo. This would have caused effects in opposite directions, which could account for the non-robust effect. As loss aversion also did not play a role in the second experiment, it is possible that loss was less salient in the framing of scenarios presented here compared to other studies on energy supplier switching.

Finally, it must be noted that this study established only correlations and not causation with regards to the default mechanisms. Future research could use additional treatments to evoke varying levels cognitive effort, loss aversion, and implied endorsement to establish causal effects.

7. Conclusions and policy implications

This study set out to understand whether consumers would be more likely to engage with local energy suppliers in a multiple-supplier model compared to the current single supplier model and to explore default mechanisms associated with remaining with incumbent suppliers. The results suggested a strong preference for the MSM; in the first experiment, the majority of participants in all conditions switched to the MSM, even when they had to take action to switch. The second experiment found that participants were statistically significantly more likely to switch to a local energy supplier in an MSM than under the current single supplier model. The most robust finding on default mechanisms was that

a greater perception of implied endorsement was associated with an increased likelihood of choosing the MSM in both default conditions. Contrary to fears that MSMs would increase consumer disengagement, these findings suggest that MSMs are likely to be a promising avenue for driving the growth of local energy and opening up opportunities for innovation in the British energy market.

Whilst the studies were not explicitly designed to test this, willingness to switch was much higher when participants were hypothetically contacted by their current supplier in the first experiment as opposed to a new local supplier in the second. This exploratory finding suggests that the cooperation and support of existing suppliers is likely to be important. Whilst many of the largest suppliers in the UK have expressed concerns about dissolving the supplier hub model, many of them are engaging with innovative business models that could be enabled by an MSM - for instance Centrica's Local Energy Market in Cornwall and EDF Energy's P2P demonstrator projects (Centrica, n.d.; Energy, n.d.). If MSMs are successful, they could allow incumbents to explore new business models whilst avoiding loss of customers.

Regarding consumer engagement, work in the field of energy efficiency home retrofits points to the importance of using trusted actors to engage with consumers and convince them to make a change (Fuller et al., 2010; Rosenow and Porter, 2015); the literature on consumer acceptance of local energy highlights the role of trust in public support for a project (Kalkbrenner and Roosen, 2016; Walker et al., 2010; Koirala et al., 2018); and research on energy supplier switching suggests that rapport with existing suppliers plays an important role for incumbents retaining their customer base (Ek and Söderholm, 2008). Similarly, these findings highlight the importance of engaging trusted figures in promoting MSMs. Explicit recommendations of MSMs may also help to drive supplier switching and engagement with innovative offers in the energy market.

Data availability

The full dataset, survey questions and replication code can be downloaded from:
<https://doi.org/10.5522/04/12808406.v1>

The PAP for experiment one is available at: <https://aspredicted.org/zk7zt.pdf>. The PAP for experiment two is available at: <https://aspredicted.org/nj7nv.pdf>.

Declaration of competing interest

The authors declare no conflicts of interest.

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Author contributions

Conceptualization, N.W., G.H., M.J.F and D.S.; methodology, N.W., G.H., M.J.F and D.S.; software, N.W; formal analysis, N.W.; investigation, N.W.; data curation, N.W.; writing—original draft preparation, N.W; writing—review and editing, G.H, M.J.F., D.S. and N.W; visualization, N.W; supervision, G.H, M.J.F., and D.S., project administration, N.W.; funding acquisition, D.S.

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Appendix A: Treatments

Figure 1: Treatments for default mechanisms experiment (experiment one), single supplier default group (top) and multiple-supplier default group (bottom)

Imagine that you receive the following letter from your current electricity supplier:

Dear customer,

We are working in partnership with LocalEnergy, a new local energy company.

You are currently enrolled on our **single-provider service**. This means that all your electricity will continue to be provided by us.

Alternatively, you can switch to the new **combined-provider service**, which is offered in partnership with LocalEnergy. Customers of the combined-provider service will get most of their electricity from locally produced, renewable sources, such as solar and wind farms. If there is not enough local energy to cover your needs, we will act as a back-up, so that you will always have the power you need.

The price of each service will be the same as your current tariff.

If you would like to stay with the single-provider service, you do not need to do anything.

If you would rather change to the combined-provider service, you need to go to our website, log in to your account, and check a box indicating that you would like to switch. It will take a few days to switch. You can switch back at any time.

What would you choose to do in this scenario?
Please select one only

Stay with the single-provider service

Switch to the combined-provider service

Imagine that you receive the following letter from your current electricity supplier:

Dear customer,

We are working in partnership with LocalEnergy, a new local electricity company.

You have automatically been switched to our new **combined-provider service**, which is offered in partnership with LocalEnergy. Customers of the combined-provider service will get most of their electricity from locally produced, renewable sources, such as solar and wind farms. If there is not enough local energy to cover your needs, we will act as a back-up, so that you will always have the power you need.

Alternatively, you can switch back to our **single-provider service**. This means that all your energy will continue to be provided by us.

The price of each service will be the same as your current tariff.

If you would like to stay with the combined-provider service, you do not need to do anything.

If you would rather change back to the single-provider service, you need to go to our website, log in to your account, and check a box indicating that you would like to switch. It will take a few days to switch. You can switch back at any time.

What would you choose to do in this scenario?
Please select one only

Stay with the combined-provider service

Switch back to the single-provider service

Figure 2: Treatments for default mechanisms experiment (experiment one), active choice group (order randomised)

The figure displays two screenshots of a survey interface, each enclosed in a rounded rectangular frame. On the left side of each frame is a vertical grey button with a double left-pointing arrow (<<). The top screenshot shows a scenario where the 'Single-provider' service is listed first, followed by the 'Combined-provider' service. The bottom screenshot shows the opposite order, with the 'Combined-provider' service listed first, followed by the 'Single-provider' service. Both screenshots include a simulated letter from an electricity supplier and two radio button options for the respondent to choose.

Imagine that you receive the following letter from your current electricity supplier:

Dear customer,

We are working in partnership with LocalEnergy, a new local energy company.

You now have the choice of two services:

1) Single-provider: This means that all your electricity will be provided by us.

2) Combined-provider: This service is offered in partnership with LocalEnergy. Customers of the combined-provider service will get most of their electricity from locally produced, renewable sources, such as solar and wind farms. If there is not enough local energy to cover your needs, we will act as a back-up, so that you will always have the power you need.

The price of each service will be the same as your current tariff.

You need to go to our website, log in to your account, and check a box indicating which service you would like to choose. You can switch at any time.

Which service would you choose in this scenario?
Please select one only

The single-provider service

The combined-provider service

Imagine that you receive the following letter from your current electricity supplier:

Dear customer,

We are working in partnership with LocalEnergy, a new local energy company.

You now have the choice of two services:

1) Combined-provider: This service is offered in partnership with LocalEnergy. Customers of the combined-provider service will get most of their electricity from locally produced, renewable sources, such as solar and wind farms. If there is not enough local energy to cover your needs, we will act as a back-up, so that you will always have the power you need.

2) Single-provider: This means that all your electricity will be provided by us.

The price of each service will be the same as your current tariff.

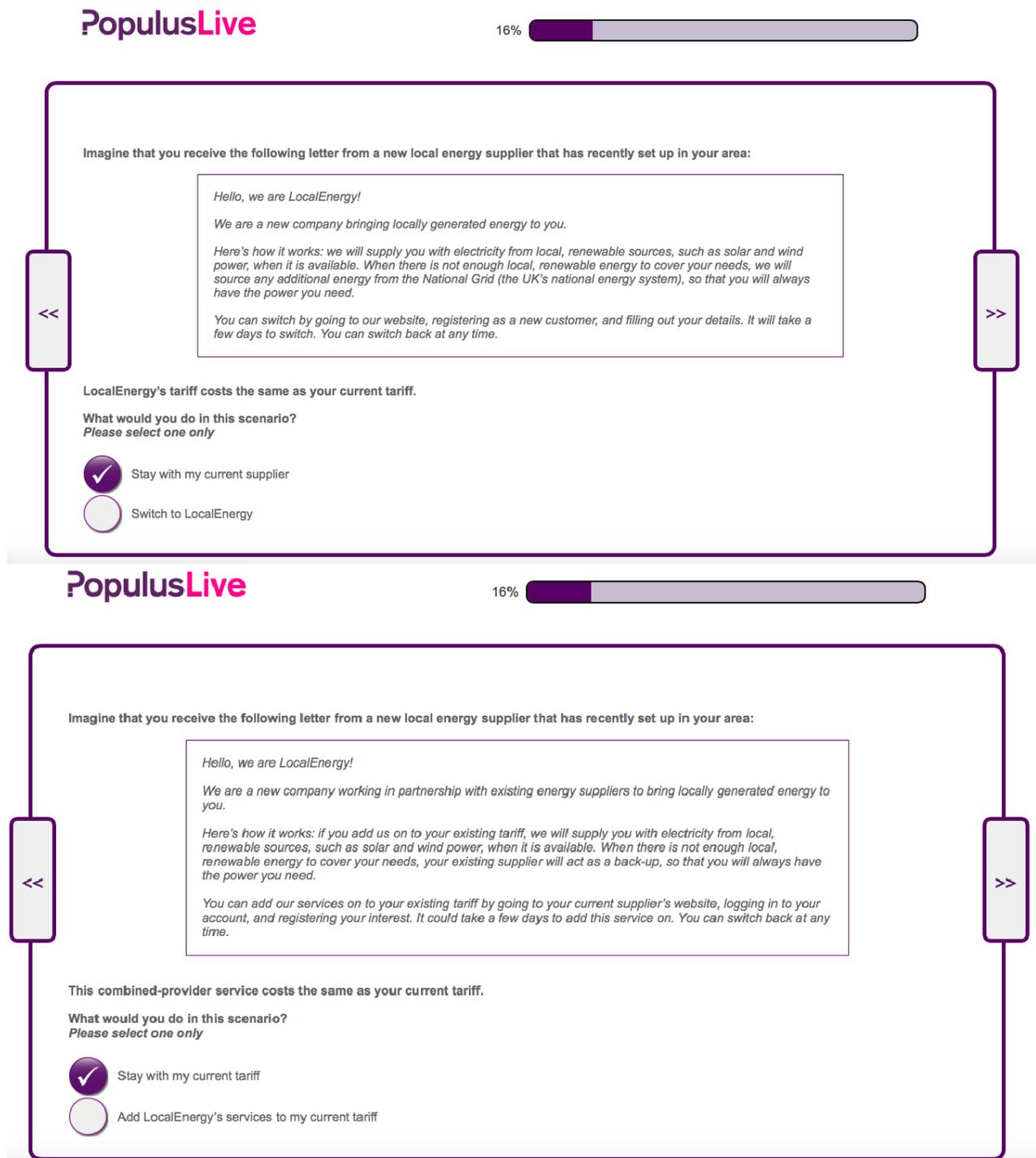
You need to go to our website, log in to your account, and check a box indicating which service you would like to choose. You can switch at any time.

Which service would you choose in this scenario?
Please select one only

The combined-provider service

The single-provider service

Figure 3: Treatments for consumer preferences experiment (experiment two), single supplier condition (top), multiple-supplier condition (bottom)



Appendix B: Measures of cognitive biases

Figure 1: Cognitive effort measure

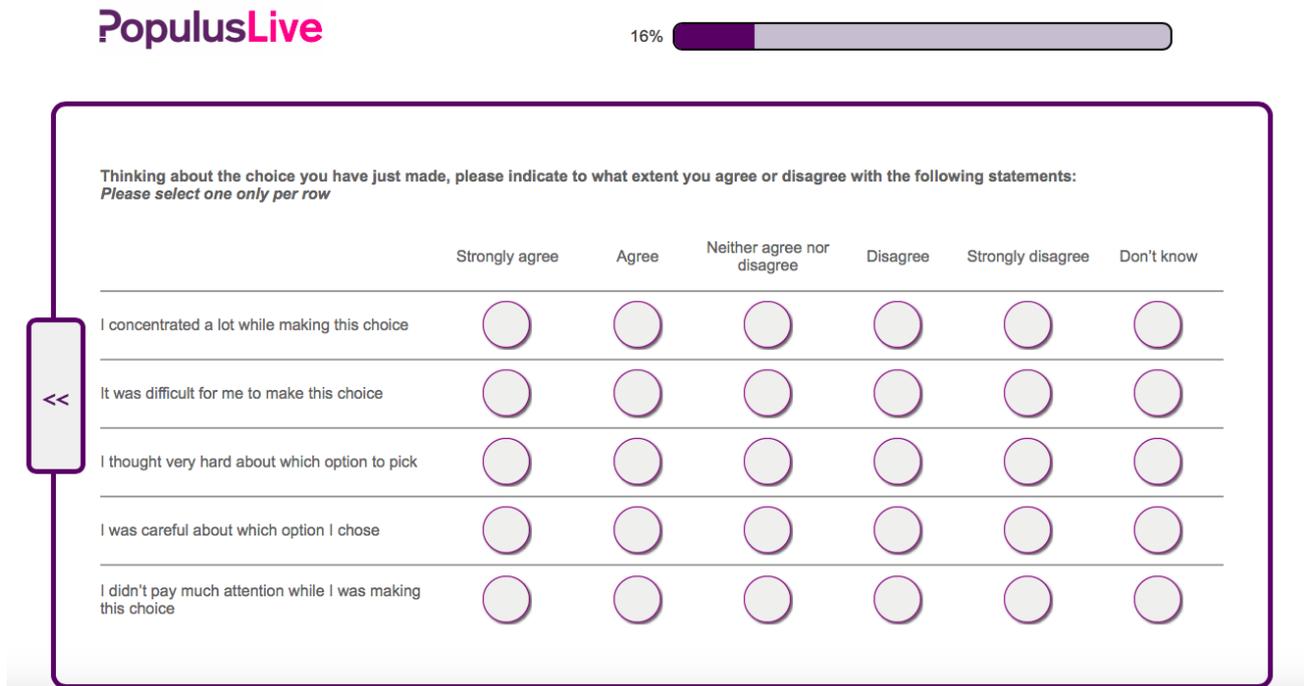


Figure 2: Implied endorsement measure (highlighted) with buffer questions

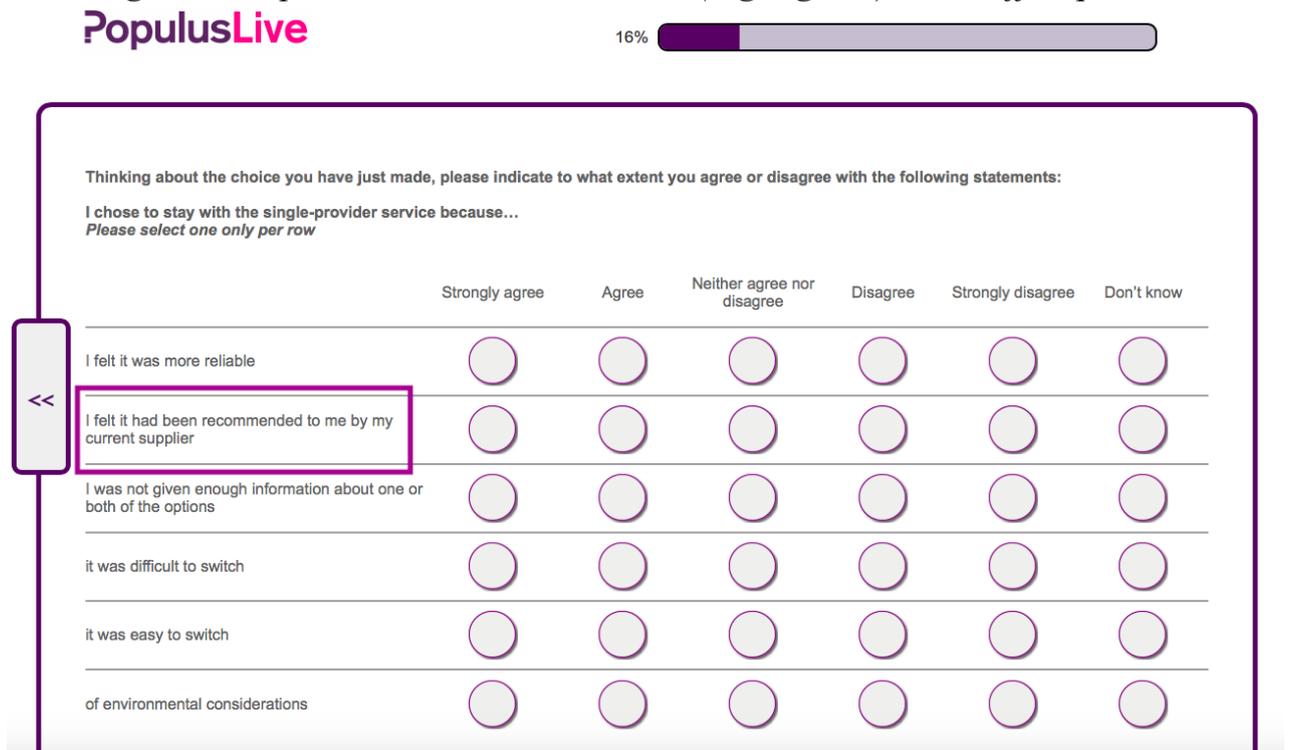


Figure 3: Loss-aversion measure

Please imagine that you are offered the opportunity to take a series of coin flipping 'heads or tails' gambles and then tell us whether you would take the gamble in each case.

- if you take the gamble, then you could hypothetically win or lose money
- if you don't take the gamble, then you can't win or lose anything

Please remember that there are no right or wrong answers for these questions - it is just about your preferences.
Please select one only per row



	Yes I would take this gamble	No I would not take this gamble
If the coin turns up heads then you lose £2; if the coin turns up tails then you win £6	<input type="radio"/>	<input type="radio"/>
If the coin turns up heads then you lose £3; if the coin turns up tails then you win £6	<input type="radio"/>	<input type="radio"/>
If the coin turns up heads then you lose £4; if the coin turns up tails then you win £6	<input type="radio"/>	<input type="radio"/>
If the coin turns up heads then you lose £5; if the coin turns up tails then you win £6	<input type="radio"/>	<input type="radio"/>
If the coin turns up heads then you lose £6; if the coin turns up tails then you win £6	<input type="radio"/>	<input type="radio"/>
If the coin turns up heads then you lose £7; if the coin turns up tails then you win £6	<input type="radio"/>	<input type="radio"/>

Appendix C: Covariates by treatment group

This appendix presents covariates by treatment group and the results of statistical tests to check for imbalance of covariates across treatment groups within the same experiment. Chi-squared (for dichotomous or categorical variables) and ANOVA tests (for continuous variables) were run. There were no statistically significant differences between groups.

Table 1: Summary of participant characteristics by treatment

Categorical variables	Percentage of sample						
	Experiment 1			Experiment 2			
	Total	Single supplier default	Multiple supplier default	Active choice	Total	Single supplier condition	Multiple supplier condition
Education							
No secondary school education (reference category)	4%	4%	5%	3%	4%	3%	5%
Secondary	56%	54%	57%	58%	56%	56%	56%
Degree or higher	39%	41%	38%	39%	39%	40%	39%
				<i>P value = 0.78</i>			<i>P value = 0.56</i>
Gender							
Female (reference category)	51%	52%	51%	51%	50%	51%	49%
Male	49%	48%	49%	49%	50%	49%	51%
				<i>P value = 0.97</i>			<i>P value = 0.63</i>
Tenure							
Home-owner (reference category)	61%	60%	61%	62%	63%	63%	63%
Private rent	17%	17%	16%	20%	17%	16%	18%
Rent free	3%	3%	3%	2%	2%	2%	2%
Social rent	19%	19%	20%	17%	18%	9%	16%
				<i>P value = 0.69</i>			<i>P value = 0.79</i>
Income							
Up to £21,000 (reference category)	36%	33%	35%	42%	38%	39%	37%

£21,001 - £41,000	44%	46%	44%	42%	45%	44%	47%
£41,001 +	12%	14%	13%	11%	10%	11%	9%
	<i>P value = 0.27</i>			<i>P value = 0.64</i>			
Moved house within 12 months							
No (reference category)	91%	92%	92%	88%	91%	90%	93%
Yes	9%	8%	8%	11%	9%	10%	7%
	<i>P value = 0.20</i>			<i>P value = 0.16</i>			
Switched energy tariff or supplier							
Never (reference category)	41%	39%	42%	43%	38%	39%	38%
Don't know	16%	17%	16%	14%	15%	15%	14%
More than 12 months ago	37%	40%	36%	36%	41%	41%	42%
Within last 12 months	5%	4%	6%	6%	4%	4%	4%
Within last 3 months	1%	1%	1%	1%	2%	1%	3%
	<i>P value = 0.91</i>			<i>P value = 0.45</i>			
Continuous variables				Mean (Standard error)			
Age	49.68	49.44	49.91	49.68	50	49.67	51.69
	(0.54)	(0.88)	(1.04)	(1.05)	(0.65)	(0.90)	(0.92)
	<i>P value = 0.93</i>			<i>P value = 0.12</i>			
Satisfaction with energy supplier	3.74	3.70	3.76	3.78	3.65	3.58	3.72
	(0.032)	(0.054)	(0.052)	(0.059)	(0.04)	(0.056)	(0.053)
	<i>P value = 0.51</i>			<i>P value = 0.06</i>			

Appendix D: Full models

Experiment 1: Model 1 includes the treatment effect and all default mechanisms; model 2 includes covariates specified in the PAP; model 3 adds an interaction term between each default mechanism and treatment group assignment, in order to isolate the effects within each group.

Table 1: Full odds ratios - experiment 1 (default mechanisms)

	<i>Dependent variable</i>		
	Switched to/added on local energy supplier		
	(1)	(2)	(3)
Single supplier group	7.795 (5.57-11.038)	8.273 (5.789-11.983)	4.007 (0.325- 54.205)
Cognitive effort	1.243 (0.981-1.581)	1.318 (1.018-1.714)	1.760 (1.158-2.748)
Implied endorsement	0.912 (0.768-1.082)	0.947 (0.788-1.138)	0.496 (0.358-0.680)
Loss-aversion	1.038 (0.910-1.185)	1.042 (0.900-1.209)	1.308 (1.009-1.730)
Single supplier group: cognitive effort			0.654 (0.376-1.118)
Single supplier group: implied endorsement			2.700 (1.833-4.020)
Single supplier group: loss-aversion			0.719 (0.518-0.982)
Male		1.009 (0.706-1.442)	1.039 (0.721-1.498)
Age		0.961 (0.851-1.085)	0.956 (0.843-1.085)
Education: secondary school		1.259 (0.448-3.824)	1.517 (0.518-4.782)

Education: degree or higher	1.135	1.313
	(0.402-3.459)	(0.447-4.147)
Private renter	0.774	0.737
	(0.463-1.281)	(0.434-1.243)
Social renter	0.872	0.878
	(0.542-1.397)	(0.539-1.425)
Rent free	0.949	0.844
	(0.299-2.845)	(0.269-2.510)
Moved house within last 12 months	1.743	1.878
	(0.862-3.495)	(0.894-3.946)
Satisfaction with energy supplier	1.057	1.112
	(0.892-1.254)	(0.935-1.325)
Don't know if switched tariff/supplier	1.558	1.702
	(0.907-2.676)	(0.974-2.981)
Switched tariff/supplier more than 12 months ago	1.364	1.389
	(0.917-2.033)	(0.925-2.090)
Switched tariff/supplier within last 12 months	1.038	1.045
	(0.435-2.39)	(0.433-2.458)
Switched tariff/supplier within last 3 months	0.211	0.110
	(0.010-1.750)	(0.005-1.013)
Income £21,001-£41,000	1.110	1.127
	(0.753-1.635)	(0.757-1.679)
Income £41,000+	0.654	0.668
	(0.359-1.177)	(0.363-1.215)

Experiment 2: Model 1 represents the effect of the treatment group when controlling for covariates. Model 2 includes an interaction term between the multiple-supplier condition and loss-aversion.

Table 2: Full odds ratios: Experiment 2 (consumer preferences)

	<i>Dependent variable</i>	
	Switched to/added on local energy supplier	
	(1)	(2)
Multiple-supplier condition	1.567 (1.135-2.167)	1.258 (0.582-2.726)
Loss-aversion	0.970 (0.849-1.109)	0.933 (0.778-1.119)
Multiple-supplier condition: loss-aversion		1.085 (0.837-1.406)
Cognitive effort	1.107 (0.856-1.435)	1.108 (0.857-1.436)
Male	0.868 (0.624-1.206)	0.865 (0.622-1.202)
Age	0.974 (0.866-1.095)	0.973 (0.866-1.094)
Education: secondary school	0.613 (0.245-1.551)	0.612 (0.244-1.547)
Education: degree or higher	0.907 (0.359-2.306)	0.901 (0.357-2.291)
Private renter	1.864 (1.173-2.976)	1.877 (1.180-3.000)
Social renter	0.937 (0.589-1.478)	0.937 (0.589-1.478)
Rent free	2.293	2.313

	(0.718-8.015)	(0.724-8.092)
Moved house within last 12 months	0.929	0.926
	(0.506-1.689)	(0.505-1.684)
Satisfaction with energy supplier	0.858	0.859
	(0.734-1.002)	(0.735-1.002)
Don't know if switched tariff/supplier	1.147	1.136
	(0.676-1.936)	(0.669-1.920)
Switched tariff/supplier more than 12 months ago	1.362	1.349
	(0.949-1.958)	(0.939-1.943)
Switched tariff/supplier within last 12 months	1.278	1.288
	(0.550-2.902)	(0.555-2.922)
Switched tariff/supplier within last 3 months	0.153	0.151
	(0.008-0.827)	(0.008-0.817)
Income £21,001-£41,000	1.084	1.075
	(0.764-1.539)	(0.757-1.527)
Income £41,000+	1.437	1.458
	(0.807-2.564)	(0.817-2.608)

Appendix E: Robustness

Please note, coefficients are presented in odds ratios in the main text and logistic regression coefficients in Appendix D. For odds ratios for all robustness tests, please see replication code available at <https://doi.org/10.5522/04/12808406.v1>

E1. Model specification: OLS regression models and logistic regression models with regional fixed effects

Experiment 1: Model 1 includes the treatment effect and all default mechanisms; models 2 and 4 include covariates specified in the PAP; models 3 and 5 add an interaction term between each default mechanism and treatment group assignment, in order to isolate the effects within each group. Models 1, 2 and 3 are OLS models to test for robustness to alternative model specification. Models 4 and 5 are the logistic regression models presented in the main paper with regional fixed effects.

Table 1: OLS Regression Models and regional fixed effects - experiment 1 (default mechanisms)

	<i>Dependent variable:</i>				
	Switched away from default			<i>Logistic regression models with regional</i>	
	<i>OLS</i>	<i>OLS</i>	<i>OLS</i>	<i>fixed effects</i>	<i>fixed effects</i>
	(1)	(2)	(3)	(4)	(5)
Single supplier group	0.433*** p = 0.000	0.438*** p = 0.000	0.142 p = 0.510	2.175*** p = 0.000	1.539 p = 0.247
Cognitive effort	0.039* p = 0.077	0.047** p = 0.046	0.066** p = 0.029	0.276** p = 0.042	0.584*** p = 0.010
Implied endorsement	-0.017 p = 0.284	-0.011 p = 0.518	-0.093*** p = 0.0002	-0.066 p = 0.484	-0.716*** p = 0.00002
Loss aversion	0.007	0.009	0.032* p = 0.0002	0.044	0.278** p = 0.0002

	p = 0.576	p = 0.517	p = 0.096	p = 0.569	p = 0.047
Single supplier group: cognitive effort			-0.034		-0.449
			p = 0.460		p = 0.116
Single supplier group: implied endorsement			0.159***		1.001***
			p = 0.00001		p = 0.00000
Single supplier group: Loss aversion			-0.043*		-0.339**
			p = 0.097		p = 0.040
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Male	-0.001	0.005	-0.004	0.032	
	p = 0.984	p = 0.868	p = 0.983	p = 0.867	
Age	-0.008	-0.011	-0.036	-0.042	
	p = 0.467	p = 0.346	p = 0.562	p = 0.524	
Education: secondary school	0.051	0.084	0.166	0.354	
	p = 0.602	p = 0.384	p = 0.764	p = 0.538	
Education: degree or higher	0.033	0.059	0.064	0.202	
	p = 0.734	p = 0.546	p = 0.909	p = 0.726	
Private renter	-0.049	-0.056	-0.338	-0.412	
	p = 0.296	p = 0.225	p = 0.206	p = 0.137	
Social renter	-0.022	-0.043	-0.052	-0.137	
	p = 0.836	p = 0.679	p = 0.929	p = 0.811	
Rent free	-0.025	-0.026	-0.138	-0.136	
	p = 0.566	p = 0.553	p = 0.573	p = 0.589	
Moved house within last 12 months	0.090	0.092	0.614*	0.705*	
	p = 0.169	p = 0.155	p = 0.092	p = 0.069	
Satisfaction with energy supplier	0.010	0.016	0.067	0.115	
	p = 0.536	p = 0.309	p = 0.453	p = 0.204	
Don't know if switched tariff/supplier	0.085*	0.094*	0.437	0.522*	
	p = 0.095	p = 0.061	p = 0.118	p = 0.071	

Switched tariff/supplier more than 12 months ago	0.059	0.061*	0.318	0.344
	p = 0.114	p = 0.096	p = 0.124	p = 0.107
Switched tariff/supplier within last 12 months	0.013	0.011	-0.034	-0.013
	p = 0.862	p = 0.883	p = 0.939	p = 0.978
Switched tariff/supplier within last 3 months	-0.194	-0.277*	-1.565	-2.252*
	p = 0.253	p = 0.099	p = 0.201	p = 0.074
Income £21,001-£41,000	0.020	0.016	0.109	0.130
	p = 0.576	p = 0.656	p = 0.594	p = 0.533
Income £41,000+	-0.075	-0.068	-0.446	-0.421
	p = 0.172	p = 0.208	p = 0.148	p = 0.179
Regional fixed effects			✓	✓
Observations	776	714	714	714
R ²	0.211	0.230	0.258	
Adjusted R ²	0.207	0.209	0.235	
Log Likelihood			-379.900	-363.744
Akaike Inf. Crit.			819.801	793.487
Residual Std. Error	0.431 (df = 771)	0.431 (df = 694)	0.424 (df = 691)	
F Statistic	51.422*** (df = 4; 771)	10.928*** (df = 19; 694)	10.940*** (df = 22; 691)	

Experiment 2: Models 1 and 3 represent the effect of the treatment group when controlling for covariates. Models 2 and 4 include an interaction term between the multiple-supplier condition and loss-aversion. Models 1 and 2 are OLS models to test for robustness to alternative model specification. Models 3 and 4 are the logistic regression models presented in the main paper with regional fixed effects.

Table 2: OLS Regression Models and regional fixed effects - experiment 2 (consumer preferences)

	<i>Dependent variable:</i>			
	Switching to/adding Local energy supplier			
	<i>OLS</i>	<i>Logistic regression models with regional fixed effects</i>		
	(1)	(2)	(3)	(4)
Multiple-supplier condition	0.104*** p = 0.007	0.054 p = 0.559	0.447*** p = 0.008	0.254 p = 0.528
Loss-aversion	-0.007 p = 0.638	-0.016 p = 0.453	-0.026 p = 0.711	-0.060 p = 0.527
Multiple-supplier condition: loss-aversion		0.018 p = 0.549		0.071 p = 0.597
Cognitive effort	0.024 p = 0.638	0.024 p = 0.425	0.078 p = 0.562	0.080 p = 0.597
Male	-0.032 p = 0.404	-0.033 p = 0.397	-0.158 p = 0.354	-0.160 p = 0.347
Age	-0.006 p = 0.665	-0.006 p = 0.658	-0.023 p = 0.710	-0.023 p = 0.704
Education: secondary school	-0.112 p = 0.303	-0.113 p = 0.301	-0.584 p = 0.224	-0.585 p = 0.222
Education: degree or higher	-0.021 p = 0.848	-0.023 p = 0.837	-0.239 p = 0.622	-0.244 p = 0.615
Private renter	0.147*** p = 0.009	0.149*** p = 0.008	0.608** p = 0.012	0.615** p = 0.012

Social renter	0.195	0.197	0.825	0.833
	p = 0.161	p = 0.157	p = 0.177	p = 0.173
Rent free	-0.015	-0.016	-0.071	-0.071
	p = 0.773	p = 0.770	p = 0.765	p = 0.767
Moved house within last 12 months	-0.015	-0.015	-0.066	-0.069
	p = 0.837	p = 0.830	p = 0.832	p = 0.825
Satisfaction with energy supplier	-0.035*	-0.035*	-0.161**	-0.160**
	p = 0.055	p = 0.056	p = 0.047	p = 0.048
Don't know if switched tariff/supplier	0.031	0.029	0.110	0.103
	p = 0.623	p = 0.646	p = 0.685	p = 0.707
Switched tariff/supplier more than 12 months ago	0.071*	0.069	0.332*	0.323*
	p = 0.098	p = 0.110	p = 0.078	p = 0.087
Switched tariff/supplier within last 12 months	0.055	0.057	0.238	0.246
	p = 0.574	p = 0.564	p = 0.576	p = 0.564
Switched tariff/supplier within last 3 months	-0.288**	-0.291**	-1.927*	-1.930*
	p = 0.049	p = 0.047	p = 0.071	p = 0.070
Income £21,001-£41,000	0.018	0.016	0.082	0.074
	p = 0.657	p = 0.693	p = 0.649	p = 0.684
Income £41,000+	0.085	0.088	0.347	0.360
	p = 0.219	p = 0.203	p = 0.249	p = 0.234
Regional fixed effects			✓	✓
Observations	664	664	664	664
R ²	0.061	0.062		
Adjusted R ²	0.035	0.034		
Residual Std. Error	0.485 (df = 645)	0.485 (df = 644)		
F Statistic	2.332*** (df = 18; 645)	2.226*** (df = 19; 644)		
Log Likelihood			-425.285	-425.145

E2. Exclusion criteria: no exclusion criteria applied

Experiment 1: Model 1 includes the treatment effect and all default mechanisms; model 2 includes covariates specified in the PAP; model 3 adds an interaction term between each default mechanism and treatment group assignment, in order to isolate the effects within each group.

Table 3: Results of logistic regression models with no exclusion criteria applied - experiment 1 (default mechanisms)

	Dependent variable:		
	Switched away from default		
	(1)	(2)	(3)
Single supplier group	1.956***	2.031***	0.852
	p = 0.000	p = 0.000	p = 0.492
Cognitive effort	0.232**	0.252**	0.419**
	p = 0.047	p = 0.047	p = 0.043
Implied endorsement	-0.058	-0.029	-0.604***
	p = 0.489	p = 0.745	p = 0.0001
Loss aversion	0.064	0.056	0.271**
	p = 0.318	p = 0.432	p = 0.041
Single supplier group: cognitive effort			-0.225
			p = 0.389
Single supplier group: implied endorsement			0.887***
			p = 0.00001
Single supplier group: Loss aversion			-0.317**
			p = 0.043
Male		-0.076	-0.047

		p = 0.662	p = 0.790
Age		0.018	0.012
		p = 0.762	p = 0.840
Education: secondary school		-0.005	0.202
		p = 0.991	p = 0.684
Education: degree or higher		-0.092	0.092
		p = 0.850	p = 0.854
Private renter		-0.265	-0.299
		p = 0.290	p = 0.244
Social renter		-0.111	-0.205
		p = 0.840	p = 0.706
Rent free		-0.148	-0.163
		p = 0.517	p = 0.483
Moved house within last 12 months		0.578*	0.643*
		p = 0.092	p = 0.074
Satisfaction with energy supplier		0.009	0.053
		p = 0.917	p = 0.530
Don't know if switched tariff/supplier		0.538**	0.618**
		p = 0.042	p = 0.023
Switched tariff/supplier more than 12 months ago		0.356*	0.370*
		p = 0.068	p = 0.063
Switched tariff/supplier within last 12 months		-0.179	-0.197
		p = 0.663	p = 0.636
Switched tariff/supplier within last 3 months		-1.476	-2.009*
		p = 0.212	p = 0.100
Income £21,001-£41,000		0.154	0.147
		p = 0.417	p = 0.447
Income £41,000+		-0.246	-0.236
		p = 0.399	p = 0.425
Observations	838	768	768

Log Likelihood	-463.280	-415.152	-401.532
Akaike Inf. Crit.	936.560	870.304	849.065

Experiment 2: Model 1 represents the effect of the treatment group when controlling for covariates. Model 2 includes an interaction term between the multiple-supplier condition and loss-aversion.

Table 4: Results of logistic regression models with no exclusion criteria applied - experiment 2 (consumer preferences)

	<i>Dependent variable:</i>	
	Switched away from default	
	(1)	(2)
Multiple-supplier condition	0.398** p = 0.013	0.381 p = 0.311
Loss aversion	-0.018 p = 0.787	-0.021 p = 0.816
Multiple-supplier condition: loss aversion		0.006 p = 0.960
Cognitive effort	0.231* p = 0.059	0.231* p = 0.059
Male	-0.137 p = 0.400	-0.137 p = 0.400
Age	0.013 p = 0.816	0.013 p = 0.816
Education: secondary school	-0.389 p = 0.370	-0.389 p = 0.369
Education: degree or higher	-0.019 p = 0.965	-0.020 p = 0.964
Private renter	0.528** p = 0.019	0.529** p = 0.019
Social renter	0.848	0.849

	p = 0.142	p = 0.141
Rent free	-0.139	-0.139
	p = 0.540	p = 0.539
Moved house within last 12 months	0.126	0.126
	p = 0.672	p = 0.672
Satisfaction with energy supplier	-0.161**	-0.161**
	p = 0.037	p = 0.037
Don't know if switched tariff/supplier	0.120	0.119
	p = 0.637	p = 0.640
Switched tariff/supplier more than 12 months ago	0.325*	0.324*
	p = 0.071	p = 0.072
Switched tariff/supplier within last 12 months	0.076	0.077
	p = 0.847	p = 0.845
Switched tariff/supplier within last 3 months	-1.773*	-1.774*
	p = 0.095	p = 0.095
Income £21,001-£41,000	0.192	0.191
	p = 0.268	p = 0.271
Income £41,000+	0.337	0.338
	p = 0.224	p = 0.224
Observations	722	722
Log Likelihood	-460.368	-460.367
Akaike Inf. Crit.	958.737	960.734

E3. Benjamini Hochberg correction for multiple hypotheses (default mechanisms experiment only - model with all covariates and interaction effects)

Table 5: Benjamini Hochberg correction for multiple hypotheses (default mechanisms experiment only - model with all covariates and interaction effects)

Dependent variable:

Switched away from default

Single supplier group	1.388
	p = 0.440
Cognitive effort	0.565*
	p = 0.077
Implied endorsement	-0.700***
	p = 0.0003
Loss aversion	0.268
	p = 0.192
Single supplier group: cognitive effort	-0.425
	p = 0.263
Single supplier group: implied endorsement	0.993***
	p = 0.00002
Single supplier group: Loss aversion	-0.330
	p = 0.192
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Male	0.038
	p = 0.876
Age	-0.045
	p = 0.659
Education: secondary school	0.417
	p = 0.659
Education: degree or higher	0.273
	p = 0.723
Private renter	-0.305
	p = 0.421
Social renter	-0.170
	p = 0.836
Rent free	-0.130
	p = 0.723

Moved house within last 12 months	0.630
	p = 0.243
Satisfaction with energy supplier	0.106
	p = 0.408
Don't know if switched tariff/supplier	0.532
	p = 0.204
Switched tariff/supplier more than 12 months ago	0.328
	p = 0.263
Switched tariff/supplier within last 12 months	0.044
	p = 0.922
Switched tariff/supplier within last 3 months	-2.206
	p = 0.235
Income £21,001-£41,000	0.120
	p = 0.710
Income £41,000+	-0.404
	p = 0.363
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Observations	714
Log Likelihood	-368.747
Akaike Inf. Crit.	783.493
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E4. Alternative specification of loss aversion as dichotomous variable

Experiment 1: Model 1 includes the treatment effect and all default mechanisms; model 2 includes covariates specified in the PAP; model 3 adds an interaction term between each default mechanism and treatment group assignment, in order to isolate the effects within each group.

Table 6: Logistic regression coefficients with alternative specification of loss-aversion - experiment 1 (default mechanisms)

	<i>Dependent variable:</i>		
	Switched away from default		
	(1)	(2)	(3)
Single supplier group	2.047***	2.107***	1.485
	p = 0.000	p = 0.000	p = 0.283
Cognitive effort	0.220*	0.274**	0.548**
	p = 0.070	p = 0.039	p = 0.012
Implied endorsement	-0.093	-0.052	-0.715***
	p = 0.290	p = 0.577	p = 0.00002
Loss aversion	-0.004	-0.190	0.671
	p = 0.989	p = 0.560	p = 0.321
Single supplier group: cognitive effort			-0.410
			p = 0.136
Single supplier group: implied endorsement			1.022***
			p = 0.00000
Single supplier group: Loss aversion			-1.281
			p = 0.107
Male		-0.018	0.012
		p = 0.924	p = 0.951
Age		-0.029	-0.034
		p = 0.633	p = 0.587
Education: secondary school		0.270	0.473
		p = 0.619	p = 0.403
Education: degree or higher		0.162	0.332

		p = 0.767	p = 0.558
Private renter		-0.269	-0.336
		p = 0.301	p = 0.213
Social renter		0.011	-0.094
		p = 0.985	p = 0.868
Rent free		-0.122	-0.095
		p = 0.614	p = 0.701
Moved house within last 12 months		0.562	0.657*
		p = 0.114	p = 0.079
Satisfaction with energy supplier		0.054	0.100
		p = 0.532	p = 0.264
Don't know if switched tariff/supplier		0.442	0.529*
		p = 0.108	p = 0.063
Switched tariff/supplier more than 12 months ago		0.308	0.336
		p = 0.129	p = 0.106
Switched tariff/supplier within last 12 months		0.040	0.052
		p = 0.927	p = 0.907
Switched tariff/supplier within last 3 months		-1.657	-2.426*
		p = 0.182	p = 0.068
Income £21,001-£41,000		0.104	0.113
		p = 0.600	p = 0.577
Income £41,000+		-0.431	-0.417
		p = 0.154	p = 0.175
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Observations	776	714	714
Log Likelihood	-426.544	-385.010	-369.428
Akaike Inf. Crit.	863.089	810.019	784.857
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Experiment 2: Model 1 represents the effect of the treatment group when controlling for covariates. Model 2 includes an interaction term between the multiple-supplier condition and loss-aversion.

Table 7: Logistic regression coefficients with alternative specification of loss-aversion - experiment 2 – consumer preferences)

	<i>Dependent variable:</i>	
	Switched to/added on local energy supplier	
	(1)	(2)
Multiple-supplier condition	0.446***	0.202
	p = 0.007	p = 0.712
Loss aversion	0.256	0.131
	p = 0.385	p = 0.741
Multiple-supplier condition: loss aversion		0.269
		p = 0.639
Cognitive effort	0.096	0.094
	p = 0.465	p = 0.476
Male	-0.123	-0.127
	p = 0.460	p = 0.446
Age	-0.038	-0.038
	p = 0.522	p = 0.516
Education: secondary school	-0.500	-0.515
	p = 0.285	p = 0.271
Education: degree or higher	-0.081	-0.099
	p = 0.864	p = 0.833
Private renter	0.630***	0.631***
	p = 0.008	p = 0.008
Social renter	0.827	0.826
	p = 0.169	p = 0.170
Rent free	-0.072	-0.070

	p = 0.760	p = 0.767
Moved house within last 12 months	-0.040	-0.041
	p = 0.896	p = 0.894
Satisfaction with energy supplier	-0.153*	-0.152*
	p = 0.053	p = 0.056
Don't know if switched tariff/supplier	0.123	0.124
	p = 0.648	p = 0.644
Switched tariff/supplier more than 12 months ago	0.308*	0.307*
	p = 0.095	p = 0.097
Switched tariff/supplier within last 12 months	0.232	0.234
	p = 0.581	p = 0.578
Switched tariff/supplier within last 3 months	-1.871*	-1.889*
	p = 0.078	p = 0.076
Income £21,001-£41,000	0.083	0.081
	p = 0.642	p = 0.652
Income £41,000+	0.371	0.379
	p = 0.208	p = 0.199
Observations	664	664
Log Likelihood	-429.255	-429.146
Akaike Inf. Crit.	896.511	898.291