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# Consumers' food cycle and household waste. When behaviors matter

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## Highlights

- Household food waste warrants the analysis of the whole consumer's food cycle.
- The more upstream is the phase the stronger is the influence on household waste.
- The gap between purchasing and outcome leads to additional uncertainty.
- Individuals resort to heuristics and deviate from the standard economic model.
- Situational factors (food retail) influence food waste generation in homes.

**Abstract:** Recent studies have shown that consumers encounter various conflicting motivations that influence the prevention of household food waste. Food choices are rooted in deep-seated judgments, such as emotions, habits, and values, thus raising the cognitive dissonance between motivation and behavior (intention-behavior gap). The complexity of this subject increases when considering that food waste is driven by repetitive, multiple, and hidden individual choices and influenced by a composite set of situational factors. This study argues the presence of a critical distance between food choices and waste generation in homes and this factual interval (behavior-outcome gap) further affects consumer's decision-making when comparing available options. Employing data from a three-year survey of a national representative panel of Italian consumers, this study develops a system of regression models using path analysis methodology. The objective is to measure the relationships between the different phases of the food consumption cycle and rank their contribution to waste. The results suggest that the more upstream the phase, the stronger the influence on food waste generation in homes. Purchasing emerges as the most critical choice of the consumers' food waste cycle. This gap between behavior and outcome adds uncertainty to food decisions, which reverberates on behavioral beliefs and as a result, leads consumers to resort to heuristics. The findings allow for the identification of a set of behavioral patterns with implications on food waste generation. Furthermore, purchasing decisions are exposed to out-of-home contextual factors, suggesting that food retail can affect consumer behaviors relevant to household food waste.

## Keywords

Consumer; attitude; behavior; food waste; purchasing; food retail.

## 46 1. Introduction

47 Why do individuals make and repeat choices that reduce their utility and lead to negative  
48 environmental and social impacts? What are the key behavioral factors explaining these apparently  
49 irrational actions? These questions assume a high complexity in the context of household food  
50 waste.

51 While at the global level, one-third of food production is wasted or lost along the entire food  
52 chain (FAO, 2011), in industrialized nations, the largest share of waste is concentrated at the  
53 household level (Griffin et al., 2009; Parfitt et al., 2010; Gustavsson et al., 2011; Gunders, 2012). In  
54 the EU-28, it amounts to 46% of total and 173 kg food wasted per person each year (Stenmarck et  
55 al., 2016). In the United Kingdom, more specifically, food waste represents an average annual cost  
56 of £680 in household budget (WRAP, 2011).

57 The magnitude and pervasiveness of household food waste suggest a number of implications.  
58 On the one hand, analyses of consumers' food choices (e.g., contingent valuation surveys) should  
59 consider the realm of food waste and the implications on their payoff. On the other hand, the  
60 drivers of food waste should include not only socioeconomic conditions (Andreasen, 2002; Vermeir  
61 and Verbeke, 2006; Evans, 2012, 2011; Farr-Wharton et al., 2014; Setti et al., 2016), but also  
62 individual motivations and behaviors pertaining to the general food domain. In fact, when  
63 consumers deal with various food resources, their evaluations tend to include a series of volitional  
64 factors such as food security, status concerns, time pressure, and food waste. Moreover, food  
65 decisions are influenced by deep-rooted and repeated judgments such as emotions, hunger,  
66 values, and habits (i.e., "visceral factors"; Lowenstein, 1996; Verplanken et al., 1998; Graham-  
67 Rowe et al., 2014). Thus, a high uncertainty level tends to characterize consumers' food choices.

68 These conscious and unconscious behavioral precursors indicate that the performed food  
69 behavior can generate an array of outcomes in terms of their intrinsic characteristics, subjective  
70 relevance, and moment of perceived attainment. This leads consumers to face a set of personal  
71 motivations—for each food behavior—that either compete or cohere with the intention to prevent or  
72 reduce food waste. In this assortment, food waste-related motivation can show a cognitive  
73 dissonance (Festinger et al., 1956) with the relevant food choices causing an intention-behavior  
74 gap (Graham-Rowe et al., 2015). This study argues that a separation between food choices and  
75 expected consequence (food waste), that is a behavior-outcome gap, can further influence  
76 consumers' decision-making and the achievement of the goal.

77 Indeed, while in the broader waste domain, the separation between action and delayed impact  
78 is relatively well-assessed, household food waste is the result of repetitive, multiple, interactive,  
79 and sequential food behaviors (i.e., purchasing, storing, and portioning; Quedsted et al., 2013) that  
80 are not fully covered by existing research. Although the various phases of the consumers' food  
81 cycle have been taken into consideration, and surveys reveal heterogeneities among different  
82 contexts, limited evidence shows major influences of the involved food decisions on household  
83 waste generation. This highlights the need for a systemic approach to analyze the behavioral  
84 process as a whole. Thus, this study attempts to answer the following research question: which  
85 consumer food behavior is crucial in terms of consequences on the generation of household food  
86 waste? The envisaged delay with which this food behavioral effect is perceived or experienced is  
87 expected to entail a series of implications at both the individual and situational level.

88 As for the consumer, a behavior-outcome gap can deliver an additional degree of uncertainty  
89 to food decisions. Considering the behaviors related to food and food waste are private and, thus,  
90 in essence not visible, the inherent risks cannot be limited by simply learning from others'  
91 experiences. This additional source of uncertainty raises the need for new research on the main  
92 non-standard behavioral schemes that can influence the generation of household food waste and,  
93 consequently, cause consumers to experience reduced private (and public) benefit or moral

94 concern. Moreover, the expected deviation from rational and reasoned choices can influence  
95 consumer's overall decision-making with possible repercussions on its earlier precursors (i.e.,  
96 beliefs). This, in turn, can amplify the trade-offs between individuals' motivations when they  
97 compare available food options, triggering a feedback loop that further raises the level of  
98 uncertainty.

99 At the situational level, the identification of food behavior(s) with a higher impact on household  
100 food waste, and their pivotal predictors suggest more precise and effective preventive  
101 interventions. Adopting data from a three-year survey of a statistically representative panel of  
102 Italian consumers, this study develops a system of regression models using path analysis  
103 methodology with three key research objectives. It aims to (i) rank food behaviors in their effect on  
104 the generation of household food waste, (ii) deduce the most affected and, thus, critical food  
105 behavioral precursor, and (iii) identify possible behavioral patterns consumers can adopt to react to  
106 uncertainty. Thus, this study compares the influence of the single phases of a food consumption  
107 cycle on household waste and then, analyzes the interactions among them. By identifying crucial  
108 consumer decisions and their decisive antecedent, this research intends to contribute to a better  
109 understanding of food waste-related behaviors and provide theoretical and operational insights on  
110 possible strategies to prevent and reduce household food waste.

111

## 112 **2. Theoretical background**

113 A well-established behavioral theory that contributes to the understanding of human behaviors  
114 is the Theory of Planned Behavior (TPB) (Ajzen, 1991, 2015). This socio-psychological construct  
115 provides a general description of an individual's decision-making by identifying the determinants of  
116 (food) behavior. With the objective of explaining individual behaviors and their predictors, TPB is  
117 being adopted with success in different research fields—from environmental psychology  
118 (Thøgersen, 2014; Russell et al., 2017; Schluter et al., 2017) to contingent valuation (Borger and  
119 Hattam, 2017) and innovation diffusion (Kiesling et al., 2012)—and particularly, a growing number  
120 of studies focused on household food waste.

121 Even if “TPB doesn't assume rationality on the part of the decision maker” (Ajzen, 2015, p.  
122 126), the proposed framework largely refers to the cognitive antecedents of behavior. *Beliefs*  
123 (considerations) are the earliest behavioral precursors TPB recognizes in individuals' decision-  
124 making process. It distinguishes three types of beliefs (behavioral, normative, and control)  
125 determined by the subjective perception of a specific factor (strength of belief) and a subjective  
126 evaluation. As for behavioral beliefs, which are central to this study's objectives, TPB identifies the  
127 former with the discerned consequences resulting from an action (“the subjective probability that  
128 behavior produces the outcome in question”, Ajzen, 2015, p. 127). Coherently, the latter is  
129 concerned with the normative value of behavioral outcome<sup>1</sup>. The (perceived and evaluated)  
130 relationship between behavior and its specific effect form the behavioral belief in its dual dimension  
131 (awareness and knowledge).

132 In general, the literature on food waste deals with the integration of TPB's beliefs with non-  
133 cognitive determinants such as moral norms (Graham-Rowe et al., 2015; Stancu et al., 2016;  
134 Mondéjar-Jiménez et al., 2016; Diaz-Ruiz et al., 2018), habits (Visschers et al., 2016), and  
135 emotions (Russell et al., 2017). This study, in perspective, analyzes the performance of each food  
136 behavior by measuring the impact on household waste generation to elicit the implications for the  
137 two components of the behavioral beliefs.

138 Along the process of human decision-making, TPB enrolls behavioral beliefs as the immediate

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<sup>1</sup> With reference to normative value, Stern et al. (1999) underline the primordial characteristic of the precursor that implies low adaptability to changes and capacity to explain behavioral performance.

139 antecedents of *attitudes* toward behavior and its outcomes.<sup>2</sup> According to the sign of the behavioral  
140 belief (i.e., subjective evaluation of outcome), attitude expresses positive or aversive thoughts and  
141 feelings about behavioral performance (Graham-Rowe et al., 2015; Russell et al., 2017). While in  
142 the more general context of food consumption, an individual's attitude can be a relevant predictor  
143 of the decision (Ajzen, 2015), findings on household food waste show a high degree of variability.  
144 On the one hand, numerous studies have clearly highlighted unfavorable attitudes toward  
145 behaviors associated with food waste (outcome), often in terms of guilt (Brook Lyndhurst, 2007;  
146 Evans, 2012; Quested et al., 2013; Graham-Rowe et al., 2014, 2015; Parizeau et al., 2015; Stancu  
147 et al., 2016). On the other hand, attitude is considered a minor predictor of behaviors generating  
148 food waste (Vermeir and Verbeke, 2006; Evans, 2012; Watson and Meah, 2013; Farr-Wharton et  
149 al., 2014; Russell et al., 2017; Hebrok and Boks, 2017).

150 Despite the weak role attitude plays in predicting a specific behavior when food waste is the  
151 evaluated consequence, in the TPB framework, an individual's attitude is assumed to be the  
152 precursor of personal *motivation* (intention). Because each food behavior is typically associated  
153 with various outcomes, a cluster of motivations influences consumers' decisions. In particular, the  
154 motivation related to food waste can demonstrate a series of relationships with alternative and  
155 complementary intentions to carry out a certain food behavior. The literature suggests food security  
156 as a major conflicting intention that diverges consumers' attention from food waste reduction to  
157 alternative goals. Indeed, food security is not strictly limited to the individual sphere (self-  
158 gratification), but involves the desire to avoid the lack of food availability for family or guests (i.e.,  
159 risk aversion and status concerns) (Evans, 2012, 2011; Graham-Rowe et al., 2014; Neff et al.,  
160 2015; Visschers et al. 2016). Time constraints or pressure are additional reasons that causing  
161 consumers (e.g., the youth; WRAP, 2014) to not consider food waste prevention among their  
162 priorities. Moreover, some authors find that food safety (i.e., risk aversion and personal  
163 preferences) can drive individuals to the adoption of rigid criteria in the assessment of food quality  
164 and thus, discard edible food (Kriflik and Yeatman, 2005; Watson and Meah, 2013; WRAP, 2014;  
165 Neff et al., 2015; Canali et al., 2017).

166 By contrast, some other studies suggest that the motivations to reduce food waste are mainly  
167 supported by moral and ethical reasons (Evans, 2012; Gjerres and Gaiani, 2013; Stefan et al.,  
168 2013; Graham-Rowe et al., 2015, 2014; Parizeau et al., 2015), whereas environmental concerns  
169 do not appear particularly relevant (Watson and Meah, 2013; Quested et al., 2013; Graham-Rowe  
170 et al., 2014). Given the growing attention paid to food quality and practices, health considerations  
171 represent another factor that can positively stimulate food waste reduction (Quested et al., 2013;  
172 WRAP, 2014; Parizeau et al., 2015).

173 As for economic concerns, recent findings reveal the ambiguity of related motivational  
174 patterns. Price variability and income constraints can not only induce consumers to reduce  
175 household food waste (Quested et al., 2013; Graham-Rowe et al., 2014; Stancu et al., 2016), but  
176 also stimulate over-purchases of discounted, lower-quality foodstuffs that lead to increased  
177 frequencies of in-home waste (Setti et al., 2016). While from the perspective of TPB, motivations  
178 are the antecedent denoting a specific *behavior*, in food-related decision-making, the heterogeneity  
179 of motivations implies, by design, possible inconsistencies with the behavior (cognitive gap). In  
180 addition, when considering household food waste, certain moderators might contribute to the  
181 misalignment between intention and behavior, such as habits (Graham-Rowe et al., 2015) and  
182 perceived behavioral control (Ajzen, 2015). Furthermore, a series of consumers' food behaviors  
183 may be held accountable for the growing disposal of foodstuffs; however, owing to local and  
184 individual diversity, there is no unanimity in literature on the most impactful decisions.

185 Surveys show that consumers' planning and shopping routines significantly affect the

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<sup>2</sup> Instead, normative and control beliefs are the cognitive antecedents of subjective norms and perceived behavioral control (Ajzen, 2015).

186 generation of household food waste (Brook Lyndhurst, 2007; Evans, 2012; Stefan et al., 2013;  
187 Stancu et al., 2016; Mondéjar-Jiménez et al., 2016; Ponis et al., 2017). To this effect, certain  
188 studies find that storage knowledge and procedures are the most influencing factors (Farr-Wharton  
189 et al., 2014; Graham-Rowe et al., 2014; Parizeau et al., 2015; van Holsteijn and Kemna, 2018),  
190 whereas when not directly linked, trained stockpiling practices are considered to play a strong  
191 preventive role (Visschers et al., 2016). Attention toward food preparation habits (Parizeau et al.,  
192 2015) and adoption of appropriate cooking procedures (Evans, 2011; WRAP, 2014; Graham-Rowe  
193 et al., 2014; Stancu et al., 2016) are broadly recognized as effective interventions to reduce waste  
194 generation, whereas leftovers are variably classified as either a major (Ventour, 2008; van der  
195 Horst, 2012; Ponis et al., 2017) or minor factor of domestic food wastage (Farr-Wharton et al.,  
196 2014).

197 The envisaged distance between food-related decisions and food waste generation (behavior-  
198 outcome gap) can cause an additional degree of consumer uncertainty. In a similar conclusion, this  
199 condition would reverberate on the antecedent behavioral beliefs causing a feedback loop (Ajzen,  
200 2015). Understanding this presumed circumstance requires the analysis of individual's possible  
201 responses to uncertainty. When dealing with uncertainty, individuals resort to heuristics and adopt  
202 non-standard behavioral patterns (e.g., time-inconsistent decisions, excess diversification,  
203 dependence on decision framing) that deviate from TPB's assumption of rational and reasoned  
204 choices and the economic core (standard) model (DellaVigna, 2009).

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### 206 **3. Methodology**

#### 207 *3.1 Sampling and data management*

208 This research analyzes data on the food choices of a nationally representative sample of  
209 Italian consumers by region, age, and gender. Data are collected through annual computer-aided  
210 web interviews (the Waste Watcher Observatory) conducted during 2013–2015 (i.e., 1,706, 1,518,  
211 and 1,502 respondents, respectively)<sup>3</sup>. The first group of response or dependent (categorical  
212 dichotomous) variables identifies the individual role of the phases of consumers' choice cycle (i.e.,  
213 planning/purchasing, storing, preparing, and eating) in driving household food waste. These four  
214 "phase variables" are measured in relation to food behaviors influencing waste generation  
215 (respondent selection of at least one item confirms the role of a given phase).

216 The planning/purchasing phase variable as a driver of household food waste is defined using a  
217 five-item scale, including incorrect appraisal of food needs and preferences, perception of food  
218 insecurity, over-purchase, and weekly shopping. The storing phase variable relates to a four-item  
219 scale covering food perishability and expired, moldy, or bad-tasting food. The preparing and eating  
220 phase variables are assessed on a one-item scale comprising of over-cooking and plate-leftovers.  
221 Next, two sets of explanatory variables are then selected. On the one hand, proximal factors  
222 (directly linked to food use) describing consumers' specific food routines and behaviors are  
223 associated with each related phase. On the other hand, distal factors depicting socioeconomic  
224 conditions that can influence consumers along the entire food choice cycle are included in the  
225 dataset (see tables 6–7 in the Appendix).

226 To compare and analyze the contribution of each phase of the consumers' choice cycle to the  
227 food waste generation, this research focuses on a second group of response variables. The  
228 frequency of household discarded edibles measures the self-reported monthly occurrence of

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<sup>3</sup> Italy, where consumers are the main responsible for food waste, shows a high level of socioeconomic and cultural heterogeneity among areas, which offers to this study the opportunity to analyze a spectrum of consumers' food behaviors. Additional details on sources of items and data collection process are described in the Appendix.

229 individuals' food waste behavior on an ordinal scale (i.e., never, sometimes, or often). To account  
230 for heterogeneity in consumer behaviors when dealing with variety in food resources, the  
231 comparative analysis is extended to six product categories—fresh bread, cheese, fruits and  
232 vegetables, milk and yoghurt, cold cuts, and eggs—which are selected in relevance to the  
233 observed frequency of food waste.

### 234 *3.1.1 Limitations*

235 Two key issues limit this study's results. Firstly, the gathered variables may be subject to an  
236 underestimation. This cognitive bias can be attributed to the formal participation in a survey using  
237 computer-assisted web interviewing and the self-assessed measures of food waste drivers and  
238 frequency (Ventour, 2008; Parizeau et al., 2015). Despite this limitation, the use of a survey is  
239 considered appropriate because the work does not aim at food waste quantification and compared  
240 to other methodologies (i.e., food waste diaries), it induces a smaller bias on consumers' food  
241 waste behaviors and reports finding as though they are not observed or judged by researchers.

242 Secondly, the adopted data management process assigns heterogeneous sets of constitutive  
243 items to the adopted phase variables and, in particular, a higher number to the  
244 planning/purchasing and storing stages. However, this can imply an underestimation of the role of  
245 the other phases (i.e., preparing and eating) in stimulating the generation of household food waste.  
246 Nevertheless, it should be emphasized that the choice of at least one item is assumed to be  
247 sufficient to consider the related phase relevant. Moreover, this does not hinder the storing phase  
248 from having the weakest influence on the waste frequency, while food-preparing activities remain a  
249 significant determinant of household waste.

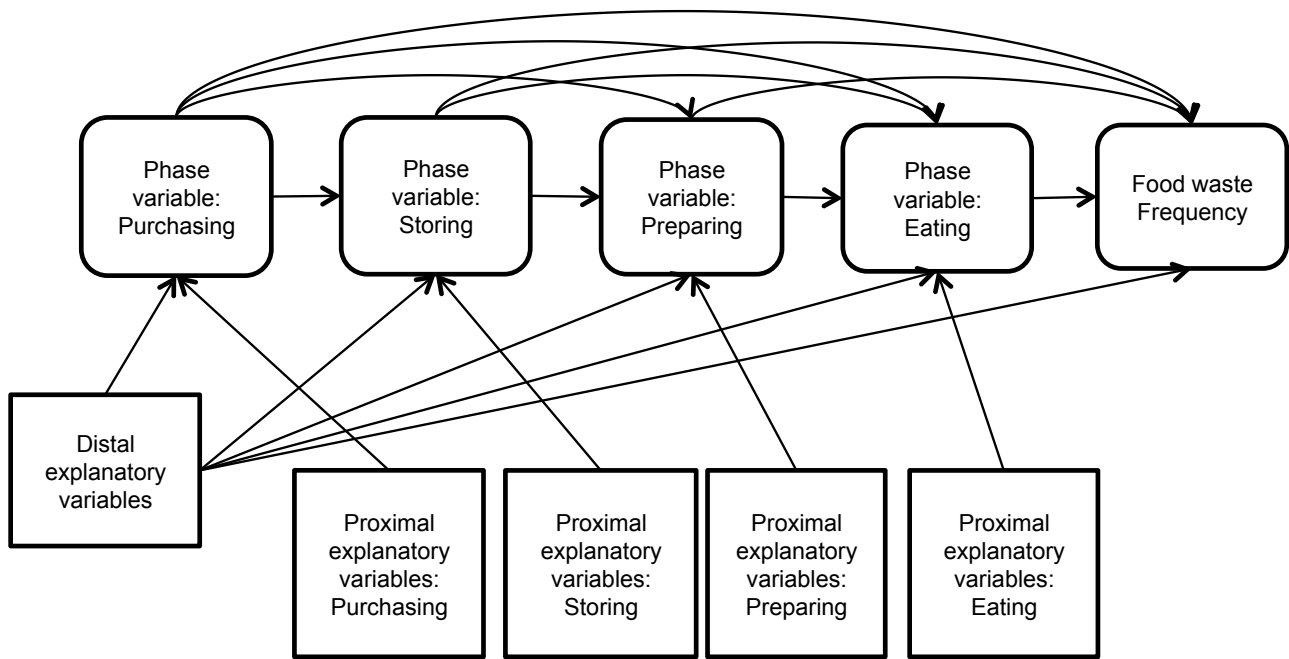
### 250 *3.2 Methodological approach*

251 A system of regression models is implemented according to the path analysis methodology  
252 (Wright, 1934; Mueller, 1996) for the three cumulated years and following a two-step approach.

253 Firstly, a set of logistic models is developed to identify the main drivers of the single phases of  
254 the consumer's food choice cycle that contribute to food waste generation (Hosmer et al., 2013).  
255 For each of the four analyzed segments of the cycle, the regression procedure assesses which  
256 covariates, among the proximal and distal explanatory factors, are significantly related to the stage  
257 responsibility for food waste (dependent phase variable). At each step, the influences of the  
258 previous phases on the role of the currently modeled phase are included as potentially additional  
259 explanatory variables.

260 Secondly, the study further develops six specific partial proportional odds regression models  
261 (PPOM) (Peterson and Harrell, 1990) to rank the phases of consumers' food choice cycle  
262 according to their contribution to household waste for each selected food item. The models identify  
263 and measure the relationships between a set of covariates (e.g., phase variables and distal  
264 explanatory factors) and the dependent level of (ordinal frequency of) household food waste (Fig.  
265 1).

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**Fig. 1.** Consumer's food waste cycle and path diagram: regression models framework

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The use of the PPOM concepts is motivated by their coherence with the odds assumption and nature of the modeled variables. Conversely, the standard statistical modeling technique for ordinal response variables, namely the proportional odds model (POM or ordinal logistic model), may not ensure the coherence of the results with the major proportional odds assumption. The PPOM framework is preferred since odds ratios related to certain explanatory variables are not restricted to being equal for all the increasing levels of the ordinal response variable (Peterson and Harrell, 1990).

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According to these considerations, a PPOM is developed for each food category by adopting a three-stage procedure:

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1. implement the PPOM that includes non-proportional odds for all explanatory variables;
2. identify explanatory variables by means of statistical hypothesis tests to assess the equivalence of odds ratios that satisfy the proportional odds assumption; and
3. develop the PPOM considering only non-proportional odds explanatory variables identified in the previous stage.

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For all the regression models, the significant results indicate odds ratio (OR) values with 95% confidence intervals (CI). For the logistic models, the odds ratios represent multiplicative coefficients on the odds of the related phase variable generating household food waste. In the case of PPOMs, the odds ratios describe multiplicative coefficients on the odds of the higher level of food waste frequency in the home (e.g., often). In particular, two odds ratios are considered when PPOMs measure non-proportional odds explanatory variables. The first OR compares food waste frequency (response level) "often" with "sometimes" or "never," while the second OR compares "often" or "sometimes" with "never".

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An OR value of less than one indicates that the associated category of explanatory variables has a lower frequency of household food waste than that of the reference category. For each logistic and partial proportional odds model, the goodness-of-fit is measured using general chi-square statistic and associated p-values, which ascertain the adherence to the stated assumption and significance of the results.



#### 298 4. Results

299 The respondents are equally distributed by gender and mainly belong to the central classes in  
300 terms of household size and age (e.g., 2—4 family members and 35—64 years; see Table 6 in the  
301 Appendix). Approximately 54% of the respondents hold a secondary education diploma, while 26%  
302 completed a primary education, and the remaining 20% had a university degree. Although  
303 perceived monetary wealth marginally improves during the observed period, about 60% of the  
304 respondents mention difficult economic conditions.

305 Several patterns emerge for main food routines and behaviors (proximal variables; Table 6,  
306 Appendix): shopping lists represent a tool regularly adopted by about 50% of the respondents,  
307 frequency of shopping is relatively high (2—3 times per week and every 2 days for about 50%),  
308 seasonal and local food products are largely preferred, and special offers are often prioritized.  
309 Although the respondents attest—in their large majority—a general concern regarding generation  
310 of food waste in the home, the frequency of waste increases during the analyzed three-year period  
311 and reaches high values for a large share of the respondents and different food products (see  
312 Table 8 in the Appendix). These results reveal that 65% of respondents (sometimes or often)  
313 discard fruits and vegetables. With a similar frequency of waste, 35%—38% respondents throw  
314 away their dairy products and bread and about 30% wasted cold cuts and eggs.

315 Considering the connection between the purchasing phase and food waste generation, the  
316 logistic model emphasizes the role of both distal and proximal explanatory factors of provisioning  
317 food play (Table 1). On the one hand, a more active attitude toward food waste prevention  
318 characterizes the members of larger households. This is revealed by the adoption of specific  
319 measures during the provisioning phase such as the use of shopping lists. On the other hand, the  
320 concentration of shopping responsibility on a single person and preference accorded to special  
321 offers and out-of-season products are the main constitutive variables that make the purchasing  
322 stage a driver of food waste.

323 The findings for the storing phase highlight a positive relationship between in-home food  
324 stockpiling and waste generation: the higher the amount of food stored, the greater the waste  
325 (Table 2). This is influenced by different categories of distal factors such as household size and  
326 education level. Put differently, other storing routines and behaviors, such as the use of expired  
327 foodstuffs and adoption of sustainable food management practices, constitute concrete actions in  
328 the fight against food waste. In the preparing phase, food waste drivers present significant regional  
329 differences: for instance, consumers living in the central and southern regions tend to discard food  
330 more frequently while cooking (Table 3). In addition, there is a strong relationship between large  
331 household groups and food waste generation during food preparation, whereas family size does  
332 not have a considerable effect on final consumption (eating phase; Table 4).

333 The different stages of consumers' food cycle present well-identifiable characteristics with  
334 specific food waste drivers and overall, show a clear pattern of inter-relationships (phase variables;  
335 see Tables 2—4). As for food waste generation within households, food purchasing significantly  
336 influences the contribution of the preparation phase. In turn, the preparation phase affects eating  
337 activities, whereas storing is deemed an independent phase with limited linkages with other  
338 segments of the food cycle.

339 Considering both the phase and distal variables, the series of PPOMs developed by this study  
340 identifies the main determinants of the in-home waste frequency of six selected foodstuffs, and  
341 enables comparisons among the roles each behavioral phase plays (Table 5; see also Tables 9—  
342 14 in the Appendix). According to some recent studies (Section 2), the results show that  
343 purchasing followed by preparing are the key phases of household food waste generation for all  
344 the analyzed products. Storing, on the other hand, limits its influence to lower frequency cases of

345 waste (i.e., sometimes) and acts as a waste-preventing phase for the more perishable products  
346 (i.e., dairies, fruits and vegetables, and cold cuts). Finally, eating is a significant driving phase only  
347 for the waste of fresh bread in the home.

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**Table 1.** Logistic model: “Planning/Purchasing” phase's drivers of household food waste.

Variable	Category	Reference	OR	95% CI	P-value
<b>Use of shopping list</b>	Sometimes	Never	0.98	(0.77 - 1.25)	0.8903
	Always		<b>0.67</b>	<b>(0.53 - 0.86)</b>	<b>0.0013</b>
<b>Weekly household food expenditure</b>	50 - 100 €	0 - 50 €	<b>1.39</b>	<b>(1.11 - 1.73)</b>	<b>0.0041</b>
	100 - 200 €		<b>1.63</b>	<b>(1.28 - 2.07)</b>	<b>&lt;.0001</b>
	200 - 300 €		<b>1.75</b>	<b>(1.27 - 2.41)</b>	<b>0.0006</b>
	More than 300 €		<b>1.94</b>	<b>(1.19 - 3.14)</b>	<b>0.0074</b>
<b>Take advantage of special offers</b>	Often	Sometimes / Never	<b>1.27</b>	<b>(1.06 - 1.53)</b>	<b>0.0088</b>
	Always		1.13	(0.94 - 1.37)	0.1983
<b>Person responsible for shopping</b>	Always the same	Different	<b>1.23</b>	<b>(1.05 - 1.45)</b>	<b>0.0104</b>
<b>Shopping strategy</b>	Pragmatic	Low-cost	1.02	(0.77 - 1.34)	0.8886
	High quality		1.18	(0.88 - 1.59)	0.2676
<b>Frequency of food shopping</b>	Every 2 days	Every day	1.22	(0.90 - 1.65)	0.1978
	2/3 times per week		1.09	(0.81 - 1.48)	0.5766
	1 time per week		<b>1.42</b>	<b>(1.02 - 1.99)</b>	<b>0.0394</b>
	Every 15 days or more		1.09	(0.78 - 1.51)	0.6199
<b>Frequency of non-seasonal products purchasing</b>	Rarely	Never	<b>1.66</b>	<b>(1.24 - 2.22)</b>	<b>0.0006</b>
	Sometimes		<b>2.43</b>	<b>(1.80 - 3.29)</b>	<b>&lt;.0001</b>
	Often		<b>3.28</b>	<b>(2.21 - 4.85)</b>	<b>&lt;.0001</b>
<b>Frequency of non-local food purchasing</b>	Rarely	Never	0.93	(0.70 - 1.23)	0.5882
	Sometimes		0.95	(0.71 - 1.27)	0.7125
	Often		0.90	(0.62 - 1.31)	0.5969
<b>Gender</b>	Female	Male	0.88	(0.78 - 1.00)	0.0523
<b>Macro region</b>	Center	North-West	1.05	(0.87 - 1.26)	0.6364
	Islands		0.96	(0.77 - 1.20)	0.7322
	North-East		0.94	(0.78 - 1.13)	0.4839
	South		1.19	(0.99 - 1.42)	0.0656
<b>Household size</b>	2	1	0.80	(0.64 - 1.02)	0.0681
	3		0.84	(0.65 - 1.07)	0.1535
	4		<b>0.67</b>	<b>(0.52 - 0.87)</b>	<b>0.0023</b>
	>4		<b>0.52</b>	<b>(0.38 - 0.72)</b>	<b>&lt;.0001</b>
<b>Perception of monetary wealth</b>	Many difficulties	Feeling poor	1.36	(0.98 - 1.89)	0.0664
	Some difficulties		<b>1.62</b>	<b>(1.20 - 2.20)</b>	<b>0.0018</b>
	Safely		<b>1.70</b>	<b>(1.24 - 2.31)</b>	<b>0.0008</b>
<b>Education level</b>	High school	Primary / middle school	1.04	(0.87 - 1.23)	0.6895
	Master degree		1.20	(0.97 - 1.48)	0.0937
	PHD		1.18	(0.85 - 1.62)	0.3228
<b>Age</b>	18-34	35-64	1.04	(0.90 - 1.22)	0.5936
	65+		0.91	(0.77 - 1.09)	0.3171

Goodness-of-fit of the model: Chi-square statistic = 253.0, df=40, p-value < 0.0001

Bolded odds ratio parameters are statistically significant (0.05).

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**Table 2.** Logistic model: “Storing” phase’s drivers of household food waste.

Variable	Category	Reference	OR	95% CI	P-value
<b>Planning/Purchasing phase</b>	Yes	No	<b>0.51</b>	<b>(0.45 - 0.59)</b>	<b>&lt;.0001</b>
<b>Person responsible for storing</b>	Always the same	Different	<b>0.83</b>	<b>(0.69 - 1.00)</b>	<b>0.0438</b>
<b>Use of expired products</b>	Reuse	Throw away	<b>0.79</b>	<b>(0.66 - 0.95)</b>	<b>0.0104</b>
	Other destinations		<b>0.52</b>	<b>(0.41 - 0.66)</b>	<b>&lt;.0001</b>
<b>Gender</b>	Female	Male	0.94	(0.82 - 1.08)	0.3812
<b>Macro region</b>	Center	North-West	0.90	(0.74 - 1.11)	0.3210
	Islands		0.86	(0.67 - 1.10)	0.2263
	North-East		0.93	(0.76 - 1.15)	0.5184
	South		<b>0.78</b>	<b>(0.64 - 0.95)</b>	<b>0.0123</b>
<b>Household size</b>	2	1	<b>1.31</b>	<b>(1.03 - 1.66)</b>	<b>0.0283</b>
	3		<b>1.38</b>	<b>(1.08 - 1.76)</b>	<b>0.0112</b>
	4		1.08	(0.84 - 1.38)	0.5629
	>4		<b>1.67</b>	<b>(1.20 - 2.34)</b>	<b>0.0027</b>
<b>Perception of monetary wealth</b>	Many difficulties	Feeling poor	<b>1.39</b>	<b>(1.00 - 1.92)</b>	<b>0.0493</b>
	Some difficulties		1.18	(0.88 - 1.58)	0.2586
	Safely		1.12	(0.83 - 1.51)	0.4624
<b>Education level</b>	High school	Primary / middle school	<b>1.32</b>	<b>(1.10 - 1.58)</b>	<b>0.0024</b>
	Master degree		<b>1.54</b>	<b>(1.23 - 1.94)</b>	<b>0.0002</b>
	PHD		<b>1.68</b>	<b>(1.17 - 2.41)</b>	<b>0.0048</b>
<b>Age</b>	18-34	35-64	0.97	(0.82 - 1.14)	0.6932
	65+		0.94	(0.77 - 1.14)	0.5038

Goodness-of-fit of the model: Chi-square statistic = 189.1, df=23, p-value < 0.0001

Bolded odds ratio parameters are statistically significant (0.05).

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**Table 3.** Logistic model: “Preparing” phase’s drivers of household food waste.

Variable	Category	Reference	OR	95% CI	P-value
<b>Storing phase</b>	Yes	No	0.93	(0.77 - 1.13)	0.4637
<b>Planning/Purchasing phase</b>	Yes	No	<b>1.37</b>	<b>(1.15 - 1.63)</b>	<b>0.0005</b>
<b>Person responsible for preparing</b>	Always the same	Different	0.98	(0.76 - 1.27)	0.8946
<b>Gender</b>	Female	Male	0.90	(0.76 - 1.07)	0.2303
<b>Macro region</b>	Center	North-West	<b>1.60</b>	<b>(1.22 - 2.10)</b>	<b>0.0007</b>
	Islands		<b>1.66</b>	<b>(1.22 - 2.26)</b>	<b>0.0014</b>
	North-East		1.22	(0.91 - 1.62)	0.1807
	South		<b>2.11</b>	<b>(1.64 - 2.71)</b>	<b>&lt;.0001</b>
<b>Household size</b>	2	1	1.34	(0.91 - 1.99)	0.1439
	3		<b>2.05</b>	<b>(1.39 - 3.02)</b>	<b>0.0003</b>
	4		<b>2.62</b>	<b>(1.77 - 3.86)</b>	<b>&lt;.0001</b>
	>4		<b>2.72</b>	<b>(1.74 - 4.24)</b>	<b>&lt;.0001</b>
<b>Perception of monetary wealth</b>	Many difficulties	Feeling poor	1.03	(0.66 - 1.60)	0.9076
	Some difficulties		1.40	(0.94 - 2.10)	0.0988
	Safely		1.17	(0.78 - 1.77)	0.4496
<b>Education level</b>	High school	Primary / middle school	1.22	(0.95 - 1.55)	0.1177
	Master degree		1.22	(0.91 - 1.65)	0.1845
	PHD		0.98	(0.61 - 1.58)	0.9482
<b>Age</b>	18-34	35-64	1.16	(0.95 - 1.42)	0.1381
	65+		1.20	(0.93 - 1.55)	0.1546

Goodness-of-fit of the model: Chi-square statistic = 148.3, df=22, p-value < 0.0001

Bolded odds ratio parameters are statistically significant (0.05).

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**Table 4.** Logistic model: “Eating” phase’s drivers of household food waste.

Variable	Category	Reference	OR	95% CI	P-value
<b>Preparing phase</b>	Yes	No	<b>2.27</b>	<b>(1.58 - 3.25)</b>	<b>&lt;.0001</b>
<b>Storing phase</b>	Yes	No	<b>0.50</b>	<b>(0.37 - 0.69)</b>	<b>&lt;.0001</b>
<b>Planning/Purchasing phase</b>	Yes	No	0.79	(0.58 - 1.09)	0.1566
<b>Frequency of lunch at home</b>	Sometimes	Rarely / Never	1.31	(0.57 - 3.00)	0.5289
	Often		1.46	(0.63 - 3.35)	0.3754
	Nearly always		1.27	(0.57 - 2.84)	0.5533
	Always		1.34	(0.61 - 2.95)	0.4667
<b>Frequency of dinner at home</b>	Often	Sometimes / Rarely / Never	1.10	(0.50 - 2.40)	0.8169
	Nearly always		1.06	(0.50 - 2.22)	0.8832
	Always		0.83	(0.39 - 1.78)	0.6361
<b>Gender</b>	Female	Male	0.80	(0.59 - 1.09)	0.1550
<b>Macro region</b>	Center	North-West	0.66	(0.39 - 1.11)	0.1169
	Islands		1.52	(0.92 - 2.51)	0.1035
	North-East		1.05	(0.66 - 1.68)	0.8405
	South		1.30	(0.84 - 2.01)	0.2382
<b>Household size</b>	2	1	<b>0.50</b>	<b>(0.30 - 0.83)</b>	<b>0.0071</b>
	3		0.70	(0.42 - 1.14)	0.1519
	4		<b>0.55</b>	<b>(0.32 - 0.93)</b>	<b>0.0265</b>
	>4		0.84	(0.44 - 1.60)	0.5901
<b>Perception of monetary wealth</b>	Many difficulties	Feeling poor	1.04	(0.52 - 2.12)	0.9059
	Some difficulties		0.80	(0.42 - 1.54)	0.5076
	Safely		1.07	(0.56 - 2.07)	0.8339
<b>Education level</b>	High school	Primary / middle school	0.76	(0.51 - 1.13)	0.1694
	Master degree		0.81	(0.50 - 1.34)	0.4147
	PHD		0.50	(0.20 - 1.24)	0.1345
<b>Age</b>	18-34	35-64	1.09	(0.75 - 1.59)	0.6548
	65+		1.51	(1.00 - 2.30)	0.0524

Goodness-of-fit of the model: Chi-square statistic = 75.2, df=29, p-value < 0.0001

Bolded odds ratio parameters are statistically significant (0.05).

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**Table 5.** Partial Proportional Odds Model: phase variables and waste frequencies per food category.

Phase variable	Response level	Odds Ratio					
		Fresh bread	Cheese	Vegetables and Fruits	Milk and Yoghurt	Cold cuts	Eggs
<b>Planning/Purchasing phase</b>	Both	2.62	2.80	3.19	2.59	2.81	2.92
<b>Storing phase</b>	Often vs. Sometimes/Never	-	0.55	-	0.65	0.53	0.53
	Often/Sometimes vs. Never	-	1.54	2.97	1.30	1.32	-
<b>Preparing phase</b>	Both	2.56	1.86	1.81	1.57	2.05	1.47
<b>Eating phase</b>	Both	1.41	-	-	-	-	-

Notes: non-statistically significant Odds Ratios (> 0.05) are not reported.

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## 372 5. Discussion

373 According to recent studies (Section 2), the results emphasize the complex structure of  
374 consumers' food cycle while suggesting a potential ranking for its behavioral phases in their  
375 contribution to the frequency of household food waste. The most upstream stage, provisioning,  
376 exerts the highest influence on the generation of household food waste. The distance between  
377 choice and outcome offers key implications at both the individual and situational level. At the  
378 individual level, this gap delivers an additional element of uncertainty to consumers' decision-  
379 making. Because of the early and crucial food choice made during the purchasing stage, there is a  
380 time lapse in the generation of household food waste. The measure of this behavior-outcome gap  
381 is a subjective perception rather than an ordinary temporal scale (Lapinski et al. 2005). The highest  
382 odd ratio value reached by the planning/purchasing stage for the most perishable foodstuffs (i.e.,  
383 wasted fruits and vegetables; see Table 11 in the Appendix) supports this assumption.

384 The time interval between food-provisioning behavior and the resulting waste emphasizes the  
385 challenge consumers face when making their decision in response to their priorities. This limited  
386 capacity to deal with this ultimate effect of the food behavior adds risk to the related process of  
387 choice as a whole. According to TPB, the uncertainty that emerges from the behavior-outcome gap  
388 is in fact envisaged to affect the cognitive determinants of an individual's decision-making.  
389 Consumers perceive food waste as a weak effect of the purchasing choice. Thus, the first  
390 component of the behavioral beliefs (the subjective probability that a decision produces the  
391 consequence) is rendered enfeebled. The identification of this early behavioral precursor leads to  
392 two main considerations. Firstly, it offers a possible explanation for the debated role of an  
393 unfavorable attitude—its immediate descendant—toward food waste in food behaviors (Section 2).  
394 While numerous studies ascertain the potential contribution that these attitudes can deliver to the  
395 prevention and reduction of food waste, it is recognized that they are not sufficient to fight food  
396 waste (i.e., contextual factors are pivotal to this goal) and they often do not denote food behavior.  
397 Secondly, because the identified factual gap squeezes the first component of the behavioral  
398 beliefs—perception of consequence—it appears necessary to boost the second element,  
399 subjective evaluation of outcome. According to this objective, initiatives aimed at enhancing  
400 knowledge about the food waste domain can determine a positive cascade effect on the related  
401 behavioral belief, attitude, motivation, and behavior (food purchasing).

402 Without adequate interventions, the uncertainty generated from the behavior-outcome (factual)  
403 gap not only merges with that caused by the intention-behavior (cognitive) gap but also propagates  
404 (through behavioral beliefs) along an individual's decision-making like a constitutive condition.  
405 Accounting for the hidden and repetitive characteristics of food waste-related choices, a possible  
406 personal strategy that consumers can implement to face this cumulated uncertainty is the greater  
407 use of heuristics. In the field of behavioral economics, the literature shows a large spectrum of  
408 evidence and variables explaining the deviation of individuals' choices from the standard economic  
409 model. To this effect, the emerged primary role of purchasing in the consumers' food waste cycle  
410 allows for the identification of the main coherent, non-standard food behavioral patterns.

411 As for the classification proposed by DellaVigna (2009), the first group of deviations is related  
412 to non-standard beliefs. Because of the (factual) gap between the dominant early-stage food  
413 choice (purchasing) and expected outcome (food waste), consumers are inclined to identify their  
414 future preferences with the current ones. This, however, forms *projection biases* (Read and van  
415 Leeuwen, 1998).

416 A second class of deviations is represented by non-standard preferences. Since food waste is  
417 conditioned by a considerable delay with respect to its driving choice (provisioning), when  
418 expected alternative rewards are compared and evaluated, individuals may demonstrate *time-*  
419 *inconsistent preferences*. This can be the case of ensuring food security against avoiding wastes in  
420 the home when consumers are inclined to prioritize food security and consequently, to purchase

421 commensurate volumes of food (Stephen and Loewenstein, 1991; Frederick et al., 2002).  
422 Moreover, consumers reveal *risk preferences* when they evaluate a food choice by comparing  
423 losses and gains. The purchasing stage can influence the framing of possible outcomes such as  
424 food-related status concerns (“reference point”), which is an alternative to the household waste  
425 prevention target. The repartition of utility into substitutive components can lead individuals to  
426 adopt probabilistic schemes that overweight losses (status erosion), subordinate alternatives (food  
427 waste reduction), and favor risk-aversion behaviors (Kahneman and Tversky, 1979).

428 Non-standard decision-making characterizes the third class of deviations. During provisioning,  
429 when the most influential decisions are made, consumers are exposed to diverse information and  
430 complex choices. To simplify their decision-making process, individuals could use different types of  
431 sub-optimal heuristics such as orienting their choice to familiarity and identity criteria (*habits*;  
432 Verplanken and Orbell, 2003), preselecting available information (*limited attention*), or referring to  
433 the most noticeable elements of alternative options (*saliency*). To this effect, the weight of the  
434 relationship between “person responsible for shopping” and purchasing phase (source of  
435 household food waste) suggests that this typology of consumers could be inclined to resort to non-  
436 standard routines when buying food. By contrast, larger-sized households report higher food waste  
437 frequency, which appears mainly due to mismanagement of food practices in the downstream  
438 phases (e.g., storing and preparing ones) rather than to adoption of heuristics at the provisioning.

439 At the situational level, since in the purchasing stage individuals make the most critical choices  
440 in terms of generation of food waste within the household, understanding consumers’ uncertainty  
441 requires extending the analysis to influences exerted in out-of-home contexts (Evans, 2012, 2011).  
442 In particular, throughout provisioning, individuals make subjective judgments contingent on the  
443 conditions at the food retail level and (potential) mutual exposure to other consumers’ behavior.  
444 Since the purchasing phase is crucial for the food waste pattern, its quasi-social dimension should  
445 contribute to moderating the related behavioral-outcome gap. The literature focused on household  
446 food waste and the motivation-behavior gap shows that social norms could affect food waste-  
447 related intentions; however, the complexity of the subject reveals additional specific elements.  
448 While Quested et al. (2013) assume that social norms exert a limited influence on the—scarcely  
449 visible—household food waste, Graham-Rowe et al. (2014) identify “a lack of perceived social  
450 pressure” as a barrier to the consumers’ intention to reduce the food waste. Further, there is no  
451 evidence on the relationship between awareness about other individuals’ food waste  
452 countermeasures (descriptive norms) and individual’s motivation to replicate it. Nevertheless, there  
453 is a significant association with the perception of what the others think (injunctive norms; Graham-  
454 Rowe et al., 2015; Stancu et al., 2016).

455 According to Aschemann-Witzel et al. (2017) and Lee (2018), the findings of this study  
456 demonstrate that the contextual factors (e.g., retailers’ special offers and purchase frequency of  
457 non-seasonal products) are significant when modeling the contribution of the provisioning phase to  
458 the generation of food waste within the household. In particular, this is proved by the significant  
459 contradiction between the respondents’ self-assessed general concern for food waste and the  
460 increase in waste frequency during the study period. Moreover, this positive trend and long-lasting  
461 pre-eminence of the purchasing stage as the major source of waste behavior in homes (factual  
462 gap) suggest a persistency in consumers’ uncertainty.

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## 464 **6. Conclusions**

465 Purchasing emerges as the most crucial consumer choice for the generation of food waste  
466 in the household and it became increasingly important during the three-year period covered by the  
467 survey.



468 The first theoretical contribution derived from this main findings is the detection of a  
469 distance between behavior and outcome. While the literature has largely focused on a motivation-  
470 behavior (cognitive) gap, this study shows an additional (factual) gap in consumer's decision-  
471 making, which raises the complexity of the food waste domain. Unclear and deferred  
472 consequences, such as to avoid discarding edible products at home, can be a detrimental to the  
473 originating food decision when compared to alternatives leading to clearly recognizable and near  
474 results. This is particularly significant when consumers deal with food waste-related choices  
475 influenced by concurrent outcomes, such as timing, risk, and social factors. In other words, it is a  
476 set of converging elements that deliver uncertainty and weaken waste-adverse attitudes.

477 Indeed, the identification of the purchasing action as the most influential for the generation  
478 of household food waste underlines that this choice is part of a complex series of interrelated  
479 behavioral precursors. To the effect of TPB, this study suggests that the behavior-outcome gap  
480 and resultant uncertainty can have a feedback influence on individuals' behavioral beliefs. The  
481 second theoretical contribution of this study is the detection of the most relevant behavioral  
482 precursor in tackling food waste generation. In particular, the revealed factual gap compresses the  
483 subjective perception of the behavioral effect (i.e., awareness of food waste). Therefore, it is the  
484 other component of the behavioral beliefs—the subjective evaluation of outcome (i.e.,  
485 knowledge)—the earliest behavioral antecedent that can lead to more robust behavioral beliefs  
486 and attitudes against food waste. Without initiatives supporting knowledge of the food field,  
487 uncertainty can further diffuse across an individual's overall decision-making process and become  
488 a constitutive driver of food waste. Moreover, because the food-related experiences of other  
489 individuals are, generally, not visible and therefore, not helpful in orienting or modifying the  
490 decisions and reducing uncertainty, consumers are more likely to resort to heuristics.

491 The third important theoretical contribution of this work is the analysis of household food  
492 waste represented by the extension of the TPB approach to a series of individuals' not rational  
493 food-related decisions through the lens of behavioral economics. Considering the role of  
494 provisioning, this study suggests the main non-standard behavioral patterns (e.g., projection  
495 biases, risk preferences, and salience) that can influence food waste generation. The centrality of  
496 the purchasing stage and the deriving behavior-outcome gap highlight that consumers' uncertainty  
497 is the foremost obstacle to preventing household food waste. This leads to a two-fold order of  
498 implications.

499 At the individual level, the need to limit uncertainty and use of heuristics require the  
500 enhancing of consumers' knowledge of food waste and, in general, the boosting of food culture.  
501 These leverages should favor the correct assessment of food waste repercussions and, in turn,  
502 form responsible behavioral beliefs and mature attitudes toward food waste prevention and  
503 reduction. With this aim, concrete interventions should prioritize educational programs targeted at  
504 students and informative campaigns at the food retail level by operators and experts of the supply  
505 chains. Furthermore, given the importance of provisioning behaviors demonstrated by consumers  
506 in out-of-home contexts (food retail), additional research efforts should be oriented toward  
507 analyzing the role of social norms (normative beliefs) in influencing the individuals' purchase  
508 decisions in relation to food waste. Given this objective, laboratory and field experiments focused  
509 on food-provisioning situations should investigate consumers' deviations from standard behaviors  
510 with reference to the role of both *social preferences* (e.g., responsible behaviors; Forsythe et al.,  
511 1994) and *social pressures* (e.g., other individuals' choices and retailers' options; Abrahamse and  
512 Steg, 2013; Young et al., 2017) in food decision-making.

513 Finally, at food retail level, it is necessary to transform related situational features from the  
514 source of uncertainty into opportunities of strengthening customers' understanding of food values  
515 and a guiding framework for coherent food behaviors. Negotiated voluntary agreements engaging  
516 retailers and consumers could stimulate multi-stakeholder governance and corporate social

517 responsibility patterns. They could also represent the structural condition to favor sustainable food  
518 purchasing choices and joint solutions to address household food waste.  
519

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718 **Appendix**

719 Residents in Italy aged between 18 and 74 years represent the concerned population. The  
 720 parameters of this population refer to four study domains (i.e., national territory, five geographical  
 721 areas, geographical regions, and municipal typologies in their effect on socioeconomic and  
 722 demographic characteristics).

723 The population has been partitioned into four strata: age, gender, macro-area (Nielsen) of  
 724 residence, and qualification. All the strata's parameters are conform to ISTAT's categories (the  
 725 National Institute for Statistics).

726 The surveyed sample has been derived from a panel of 60,000 individuals—components have  
 727 been recruited by means of online and phone sourcing techniques—and selected through stratified  
 728 random sampling. Finally, a post-stratification survey weighting has been adopted.

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**Table 6.** Distal explanatory variables (number and % of respondents).

Distal variable	Value	Tot	%
<b>Macro region</b>	North-West	1,289	27.3%
	North-East	896	19.0%
	Center	919	19.5%
	South	1,079	22.8%
	Islands	543	11.5%
<b>Household size</b>	1	516	10.9%
	2	1,365	28.9%
	3	1,260	26.7%
	4	1,202	25.4%
	>4	383	8.1%
<b>Perception of monetary wealth</b>	Feeling Poor	306	6.5%
	Many Difficulties	677	14.3%
	Some Difficulties	1,959	41.5%
	Safely	1,784	37.8%
<b>Education level</b>	Primary / middle school	964	20.4%
	High school	2,551	54.0%
	Master degree	968	20.5%
	PHD	243	5.1%
<b>Yearly respondents</b>	2013	1,706	36.1%
	2014	1,518	32.1%
	2015	1,502	31.8%
<b>Gender</b>	Male	2,286	48.4%
	Female	2,440	51.6%
<b>Age</b>	18-34	1,262	26.7%
	35-64	2,541	53.8%
	65+	923	19.5%

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**Table 7.** “Planning/Purchasing” phase: proximal explanatory variables (number and % of respondents).

Proximal variable	Value	Tot	%
<b>Use of shopping list</b>	Never	396	8.4%
	Sometimes	1,933	40.9%
	Often	2,397	50.7%
<b>Frequency of food shopping</b>	Every 15 days or more	274	5.8%
	1 time per week	1,456	30.8%
	2/3 times per week	1,707	36.1%
	Every 2 days	582	12.3%
	Every day	707	15.0%
<b>Weekly household food expenditure</b>	0 - 50 €	544	11.5%
	50 - 100 €	2,012	42.6%
	100 - 200 €	1,556	32.9%
	200 - 300 €	320	6.8%
	More than 300 €	91	1.9%
	MISSING	203	4.3%
<b>Frequency of non-seasonal products purchasing</b>	Never	369	7.8%
	Rarely	2,121	44.9%
	Sometimes	1,941	41.1%
	Often	295	6.2%
<b>Frequency of non-local food purchasing</b>	Never	351	7.4%
	Rarely	1,954	41.4%
	Sometimes	2,081	44.0%
	Often	340	7.2%
<b>Person responsible for shopping</b>	Different	1,022	21.6%
	Always the same	3,704	78.4%
<b>Take advantage of special offers</b>	Sometimes/Never	826	17.5%
	Often	2,107	44.6%
	Always	1,793	38.0%
<b>Shopping strategy</b>	Low cost	281	6.0%
	Pragmatic	3,232	68.4%
	High quality	1,113	23.6%
	MISSING	100	2.1%

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**Table 8.** Frequency of waste per food category and year (number and % of respondents).

Food category	Year	Never	Sometimes	Often	Missing
<b>Fresh bread</b>	Total	2,895 (61.3%)	1,327 (28.1%)	284 (6.0%)	220 (4.7%)
	2013	1,050 (61.6%)	469 (27.5%)	76 (4.4%)	111 (6.5%)
	2014	976 (64.3%)	387 (25.5%)	109 (7.2%)	46 (3.0%)
	2015	869 (57.9%)	471 (31.4%)	99 (6.6%)	63 (4.2%)
<b>Cheese</b>	Total	2,929 (62.0%)	1,502 (31.8%)	211 (4.5%)	84 (1.8%)
	2013	1,102 (64.6%)	526 (30.8%)	56 (3.3%)	22 (1.3%)
	2014	962 (63.4%)	445 (29.3%)	79 (5.2%)	32 (2.1%)
	2015	865 (57.6%)	531 (35.3%)	76 (5.1%)	30 (2.0%)
<b>Fruits and vegetables</b>	Total	1,637 (34.6%)	2,680 (56.7%)	382 (8.1%)	27 (0.6%)
	2013	642 (37.6%)	915 (53.6%)	142 (8.3%)	7 (0.4%)
	2014	529 (34.9%)	860 (56.6%)	121 (8.0%)	8 (0.5%)
	2015	466 (31.0%)	905 (60.3%)	119 (7.9%)	12 (0.8%)
<b>Milk and yoghurt</b>	Total	2,967 (62.8%)	1,402 (29.7%)	240 (5.1%)	117 (2.5%)
	2013	1,219 (71.5%)	384 (22.5%)	72 (4.2%)	31 (1.8%)
	2014	917 (60.4%)	468 (30.8%)	96 (6.3%)	37 (2.4%)
	2015	831 (55.3%)	550 (36.6%)	72 (4.8%)	49 (3.3%)
<b>Cold cuts</b>	Total	3,184 (67.4%)	1,155 (24.4%)	199 (4.2%)	188 (4.0%)
	2013	1,197 (70.2%)	378 (22.2%)	72 (4.2%)	59 (3.5%)
	2014	1,063 (70.0%)	335 (22.1%)	64 (4.2%)	56 (3.7%)
	2015	924 (61.5%)	442 (29.4%)	63 (4.2%)	73 (4.9%)
<b>Eggs</b>	Total	3,391 (71.8%)	1,046 (22.1%)	185 (3.9%)	104 (2.2%)
	2013	1,270 (74.4%)	336 (19.7%)	68 (4.0%)	32 (1.9%)
	2014	1,093 (72.0%)	336 (22.1%)	52 (3.4%)	37 (2.4%)
	2015	1,028 (68.4%)	374 (24.9%)	65 (4.3%)	35 (2.3%)

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**Table 9.** Partial Proportional Odds Model: drivers of fresh bread waste.

Phase and distal variable	Category	Reference Category	Response level	OR	95% CI	P-value
<b>Planning/Purchasing phase</b>	Yes	No	Both	<b>2.62</b>	<b>(2.30 - 2.98)</b>	<b>&lt;.0001</b>
<b>Storing phase</b>	Yes	No	Often vs. Sometimes/Never	0.80	(0.62 - 1.04)	0.0886
			Often/Sometimes vs. Never	1.15	(0.99 - 1.34)	0.0610
<b>Preparing phase</b>	Yes	No	Both	<b>2.56</b>	<b>(2.16 - 3.04)</b>	<b>&lt;.0001</b>
<b>Eating phase</b>	Yes	No	Both	<b>1.41</b>	<b>(1.03 - 1.94)</b>	<b>0.0323</b>
<b>Gender</b>	Female	Male	Both	<b>0.85</b>	<b>(0.74 - 0.96)</b>	<b>0.0093</b>
<b>Macro region</b>	Center	North-West	Both	<b>1.35</b>	<b>(1.12 - 1.64)</b>	<b>0.0016</b>
	Islands		Both	1.18	(0.94 - 1.48)	0.1498
	North-East		Both	1.04	(0.86 - 1.27)	0.6917
	South		Both	<b>1.45</b>	<b>(1.21 - 1.75)</b>	<b>&lt;.0001</b>
<b>Household size</b>	2	1	Both	1.01	(0.79 - 1.29)	0.9267
	3		Both	<b>1.28</b>	<b>(1.00 - 1.63)</b>	<b>0.0475</b>
	4		Both	<b>1.36</b>	<b>(1.06 - 1.75)</b>	<b>0.0156</b>
	>4		Both	<b>1.38</b>	<b>(1.02 - 1.88)</b>	<b>0.0380</b>
<b>Perception of monetary wealth</b>	Many difficulties	Feeling poor	Both	0.92	(0.68 - 1.24)	0.5759
	Some difficulties		Both	0.82	(0.62 - 1.09)	0.1717
	Safely		Both	1.14	(0.86 - 1.51)	0.3701
<b>Education level</b>	High school	Primary / middle school	Both	1.00	(0.84 - 1.2)	0.9757
	Master degree		Both	1.21	(0.97 - 1.49)	0.0871
	PHD		Both	1.07	(0.78 - 1.47)	0.6637
<b>Yearly respondents</b>	2014	2013	Often vs. Sometimes/Never	<b>1.57</b>	<b>(1.16 - 2.14)</b>	<b>0.0038</b>
			Often/Sometimes vs. Never	0.99	(0.84 - 1.15)	0.8513
	2015		Often vs. Sometimes/Never	<b>1.41</b>	<b>(1.03 - 1.93)</b>	<b>0.0327</b>
			Often/Sometimes vs. Never	<b>1.24</b>	<b>(1.05 - 1.45)</b>	<b>0.0091</b>
<b>Age</b>	18-34	35-64	Both	1.07	(0.92 - 1.24)	0.3956
	65+		Both	<b>0.69</b>	<b>(0.57 - 0.84)</b>	<b>0.0001</b>

Goodness-of-fit of the model: Chi-square statistic = 541.2, df=26, p-value < 0.0001

Bolded odds ratio parameters are statistically significant (0.05).

751 **Table 10.** Partial Proportional Odds Model: drivers of cheese waste.

Phase and distal variable	Category	Reference Category	Response level	OR	95% CI	P-value
<b>Planning/Purchasing phase</b>	Yes	No	Both	<b>2.80</b>	<b>(2.46 - 3.18)</b>	<b>&lt;.0001</b>
<b>Storing phase</b>	Yes	No	Often vs. Sometimes/Never	<b>0.55</b>	<b>(0.42 - 0.74)</b>	<b>&lt;.0001</b>
			Often/Sometimes vs. Never	<b>1.54</b>	<b>(1.33 - 1.78)</b>	<b>&lt;.0001</b>
<b>Preparing phase</b>	Yes	No	Both	<b>1.86</b>	<b>(1.57 - 2.22)</b>	<b>&lt;.0001</b>
<b>Eating phase</b>	Yes	No	Both	0.76	(0.54 - 1.07)	0.1109
<b>Gender</b>	Female	Male	Both	1.02	(0.90 - 1.16)	0.7396
<b>Macro region</b>	Center	North-West	Both	1.14	(0.95 - 1.37)	0.1556
	Islands		Both	1.09	(0.87 - 1.36)	0.4459
	North-East		Both	0.96	(0.80 - 1.16)	0.6976
	South		Both	0.94	(0.79 - 1.13)	0.5204
<b>Household size</b>	2	1	Both	0.80	(0.64 - 1.01)	0.0589
	3		Both	1.16	(0.92 - 1.45)	0.2138
	4		Both	1.04	(0.82 - 1.31)	0.7759
	>4		Both	1.16	(0.87 - 1.56)	0.3178
<b>Perception of monetary wealth</b>	Many difficulties	Feeling poor	Both	1.07	(0.78 - 1.46)	0.6783
	Some difficulties		Both	1.17	(0.88 - 1.56)	0.2706
	Safely		Both	<b>1.38</b>	<b>(1.03 - 1.84)</b>	<b>0.0291</b>
<b>Education level</b>	High school	Primary / middle school	Both	0.97	(0.82 - 1.16)	0.7521
	Master degree		Both	1.09	(0.88 - 1.34)	0.4300
	PHD		Both	1.14	(0.83 - 1.55)	0.4209
<b>Yearly respondents</b>	2014	2013	Both	1.10	(0.95 - 1.29)	0.2013
	2015		Both	<b>1.34</b>	<b>(1.15 - 1.56)</b>	<b>0.0002</b>
<b>Age</b>	18-34	35-64	Both	<b>1.44</b>	<b>(1.24 - 1.66)</b>	<b>&lt;.0001</b>
	65+		Both	<b>0.73</b>	<b>(0.61 - 0.87)</b>	<b>0.0007</b>

Goodness-of-fit of the model: Chi-square statistic = 524.4, df=24, p-value < 0.0001  
 Bolded odds ratio parameters are statistically significant (0.05).

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**Table 11.** Partial Proportional Odds Model: drivers of fruits and vegetables waste.

Phase and distal variable	Category	Reference Category	Response level	OR	95% CI	P-value
<b>Planning/Purchasing phase</b>	Yes	No	Both	<b>3.19</b>	<b>(2.80 - 3.63)</b>	<b>&lt;.0001</b>
<b>Storing phase</b>	Yes	No	Often vs. Sometimes/Never	0.91	(0.72 - 1.15)	0.4224
			Often/Sometimes vs. Never	<b>2.97</b>	<b>(2.57 - 3.44)</b>	<b>&lt;.0001</b>
<b>Preparing phase</b>	Yes	No	Both	<b>1.81</b>	<b>(1.51 - 2.16)</b>	<b>&lt;.0001</b>
<b>Eating phase</b>	Yes	No	Both	1.01	(0.74 - 1.39)	0.9334
<b>Gender</b>	Female	Male	Both	0.95	(0.84 - 1.07)	0.3825
<b>Macro region</b>	Center	North-West	Both	0.89	(0.75 - 1.06)	0.1834
	Islands		Both	0.85	(0.69 - 1.05)	0.1368
	North-East		Both	0.97	(0.81 - 1.15)	0.6885
	South		Both	0.86	(0.72 - 1.02)	0.0843
<b>Household size</b>	2	1	Both	1.02	(0.83 - 1.26)	0.8613
	3		Both	1.08	(0.87 - 1.34)	0.4765
	4		Both	1.14	(0.91 - 1.42)	0.2574
	>4		Both	1.05	(0.79 - 1.39)	0.7345
<b>Perception of monetary wealth</b>	Many difficulties	Feeling poor	Both	1.00	(0.76 - 1.33)	0.9763
	Some difficulties		Both	1.06	(0.82 - 1.37)	0.6627
	Safely		Both	1.22	(0.94 - 1.59)	0.1340
<b>Education level</b>	High school	Primary / middle school	Both	1.11	(0.95 - 1.30)	0.2014
	Master degree		Both	<b>1.27</b>	<b>(1.04 - 1.55)</b>	<b>0.0204</b>
	PHD		Both	<b>1.42</b>	<b>(1.05 - 1.93)</b>	<b>0.0221</b>
<b>Yearly respondents</b>	2014	2013	Both	1.14	(0.99 - 1.32)	0.0719
	2015		Both	<b>1.21</b>	<b>(1.05 - 1.40)</b>	<b>0.0099</b>
<b>Age</b>	18-34	35-64	Both	1.09	(0.94 - 1.25)	0.2516
	65+		Both	<b>0.75</b>	<b>(0.64 - 0.89)</b>	<b>0.0009</b>

Goodness-of-fit of the model: Chi-square statistic = 651.3, df=24, p-value &lt; 0.0001

Bolded odds ratio parameters are statistically significant (0.05).

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761 **Table 12.** Partial Proportional Odds Model: drivers of milk and yoghurt waste.

Phase and distal variable	Category	Reference Category	Response level	OR	95% CI	P-value
<b>Planning/Purchasing phase</b>	Yes	No	Both	<b>2.59</b>	<b>(2.27 - 2.94)</b>	<b>&lt;.0001</b>
<b>Storing phase</b>	Yes	No	Often vs. Sometimes/Never	<b>0.65</b>	<b>(0.49 - 0.85)</b>	<b>0.0019</b>
			Often/Sometimes vs. Never	<b>1.30</b>	<b>(1.12 - 1.50)</b>	<b>0.0006</b>
<b>Preparing phase</b>	Yes	No	Both	<b>1.57</b>	<b>(1.32 - 1.87)</b>	<b>&lt;.0001</b>
<b>Eating phase</b>	Yes	No	Both	1.23	(0.89 - 1.69)	0.2090
<b>Gender</b>	Female	Male	Both	0.91	(0.80 - 1.03)	0.1461
<b>Macro region</b>	Center	North-West	Both	1.10	(0.92 - 1.33)	0.2966
	Islands		Both	1.04	(0.83 - 1.30)	0.7420
	North-East		Both	1.05	(0.87 - 1.27)	0.5915
	South		Both	1.15	(0.96 - 1.38)	0.1278
<b>Household size</b>	2	1	Both	0.87	(0.69 - 1.10)	0.2425
	3		Both	1.17	(0.93 - 1.48)	0.1899
	4		Both	1.27	(1.00 - 1.61)	0.0545
	>4		Both	1.17	(0.87 - 1.59)	0.2960
<b>Perception of monetary wealth</b>	Many difficulties	Feeling poor	Both	0.99	(0.72 - 1.37)	0.9446
	Some difficulties		Both	1.20	(0.90 - 1.61)	0.2192
	Safely		Both	<b>1.36</b>	<b>(1.01 - 1.83)</b>	<b>0.0458</b>
<b>Education level</b>	High school	Primary / middle school	Both	1.10	(0.92 - 1.32)	0.2905
	Master degree		Both	<b>1.29</b>	<b>(1.04 - 1.59)</b>	<b>0.0213</b>
	PHD		Both	<b>1.47</b>	<b>(1.08 - 2.00)</b>	<b>0.0154</b>
<b>Yearly respondents</b>	2014	2013	Often vs. Sometimes/Never	<b>1.58</b>	<b>(1.15 - 2.17)</b>	<b>0.0044</b>
			Often/Sometimes vs. Never	<b>1.76</b>	<b>(1.50 - 2.06)</b>	<b>&lt;.0001</b>
	2015		Often vs. Sometimes/Never	1.10	(0.79 - 1.54)	0.5780
			Often/Sometimes vs. Never	<b>1.99</b>	<b>(1.70 - 2.33)</b>	<b>&lt;.0001</b>
<b>Age</b>	18-34	35-64	Both	<b>1.48</b>	<b>(1.28 - 1.71)</b>	<b>&lt;.0001</b>
	65+		Both	<b>0.71</b>	<b>(0.59 - 0.86)</b>	<b>0.0004</b>

Goodness-of-fit of the model: Chi-square statistic = 552.2, df=26, p-value < 0.0001

Bolded odds ratio parameters are statistically significant (0.05).

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765 **Table 13.** Partial Proportional Odds Model: drivers of cold cuts waste.

Phase and distal variable	Category	Reference Category	Response level	OR	95% CI	P-value
<b>Planning/Purchasing phase</b>	Yes	No	Both	<b>2.81</b>	<b>(2.45 - 3.22)</b>	<b>&lt;.0001</b>
<b>Storing phase</b>	Yes	No	Often vs. Sometimes/Never	<b>0.53</b>	<b>(0.40 - 0.72)</b>	<b>&lt;.0001</b>
			Often/Sometimes vs. Never	<b>1.32</b>	<b>(1.13 - 1.54)</b>	<b>0.0006</b>
<b>Preparing phase</b>	Yes	No	Both	<b>2.05</b>	<b>(1.72 - 2.45)</b>	<b>&lt;.0001</b>
<b>Eating phase</b>	Yes	No	Both	1.29	(0.92 - 1.81)	0.1422
<b>Gender</b>	Female	Male	Both	1.03	(0.90 - 1.18)	0.6852
<b>Macro region</b>	Center	North-West	Both	<b>1.35</b>	<b>(1.10 - 1.65)</b>	<b>0.0034</b>
	Islands		Both	<b>1.92</b>	<b>(1.53 - 2.42)</b>	<b>&lt;.0001</b>
	North-East		Both	0.96	(0.78 - 1.18)	0.6697
	South		Both	<b>1.47</b>	<b>(1.21 - 1.78)</b>	<b>0.0001</b>
<b>Household size</b>	2	1	Often vs. Sometimes/Never	0.60	(0.35 - 1.03)	0.0632
			Often/Sometimes vs. Never	0.98	(0.75 - 1.27)	0.8779
	3		Often vs. Sometimes/Never	1.05	(0.64 - 1.73)	0.8553
			Often/Sometimes vs. Never	1.29	(1.00 - 1.68)	0.0546
	4		Often vs. Sometimes/Never	1.00	(0.61 - 1.66)	0.9954
			Often/Sometimes vs. Never	<b>1.43</b>	<b>(1.10 - 1.87)</b>	<b>0.0081</b>
	>4		Often vs. Sometimes/Never	0.66	(0.32 - 1.35)	0.2527
			Often/Sometimes vs. Never	<b>1.72</b>	<b>(1.25 - 2.38)</b>	<b>0.0010</b>
<b>Perception of monetary wealth</b>	Many difficulties	Feeling poor	Both	1.07	(0.76 - 1.50)	0.6980
	Some difficulties		Both	1.09	(0.80 - 1.48)	0.5922
	Safely		Both	1.33	(0.98 - 1.82)	0.0710
<b>Education level</b>	High school	Primary / middle school	Both	0.87	(0.73 - 1.05)	0.1562
	Master degree		Both	1.06	(0.85 - 1.32)	0.6242
	PHD		Both	1.07	(0.77 - 1.49)	0.6744
<b>Yearly respondents</b>	2014	2013	Both	1.03	(0.87 - 1.21)	0.7317
	2015		Both	<b>1.43</b>	<b>(1.22 - 1.68)</b>	<b>&lt;.0001</b>
<b>Age</b>	18-34	35-64	Both	<b>1.18</b>	<b>(1.01 - 1.38)</b>	<b>0.0346</b>
	65+		Both	<b>0.78</b>	<b>(0.64 - 0.95)</b>	<b>0.0125</b>

Goodness-of-fit of the model: Chi-square statistic = 534.5, df=28, p-value < 0.0001  
 Bolded odds ratio parameters are statistically significant (0.05).

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770 **Table 14.** Partial Proportional Odds Model: drivers of eggs waste.

Phase and distal variable	Category	Reference Category	Response level	OR	95% CI	P-value
<b>Planning/Purchasing phase</b>	Yes	No	Both	<b>2.92</b>	<b>(2.55 - 3.36)</b>	<b>&lt;.0001</b>
<b>Storing phase</b>	Yes	No	Often vs. Sometimes/Never	<b>0.53</b>	<b>(0.39 - 0.72)</b>	<b>&lt;.0001</b>
			Often/Sometimes vs. Never	1.02	(0.87 - 1.19)	0.8437
<b>Preparing phase</b>	Yes	No	Both	<b>1.47</b>	<b>(1.22 - 1.77)</b>	<b>&lt;.0001</b>
<b>Eating phase</b>	Yes	No	Both	1.11	(0.78 - 1.56)	0.5677
<b>Gender</b>	Female	Male	Both	<b>0.86</b>	<b>(0.75 - 0.99)</b>	<b>0.0360</b>
<b>Macro region</b>	Center	North-West	Both	1.01	(0.83 - 1.24)	0.9176
	Islands		Both	1.06	(0.83 - 1.36)	0.6236
	North-East		Both	0.95	(0.78 - 1.17)	0.6503
	South		Both	1.07	(0.88 - 1.30)	0.4835
<b>Household size</b>	2	1	Both	<b>0.67</b>	<b>(0.53 - 0.85)</b>	<b>0.0009</b>
	3		Both	<b>0.76</b>	<b>(0.60 - 0.96)</b>	<b>0.0217</b>
	4		Both	<b>0.63</b>	<b>(0.49 - 0.81)</b>	<b>0.0003</b>
	>4		Both	<b>0.57</b>	<b>(0.41 - 0.79)</b>	<b>0.0007</b>
<b>Perception of monetary wealth</b>	Many difficulties	Feeling poor	Both	0.87	(0.62 - 1.22)	0.4089
	Some difficulties		Both	0.82	(0.60 - 1.11)	0.1961
	Safely		Both	1.00	(0.73 - 1.36)	0.9792
<b>Education level</b>	High school	Primary / middle school	Both	0.92	(0.77 - 1.12)	0.4145
	Master degree		Both	1.08	(0.86 - 1.35)	0.5305
	PHD		Both	<b>1.65</b>	<b>(1.20 - 2.27)</b>	<b>0.0023</b>
<b>Yearly respondents</b>	2014	2013	Both	1.11	(0.94 - 1.31)	0.2078
	2015		Both	<b>1.27</b>	<b>(1.07 - 1.50)</b>	<b>0.0051</b>
<b>Age</b>	18-34	35-64	Both	<b>1.33</b>	<b>(1.14 - 1.56)</b>	<b>0.0004</b>
	65+		Both	<b>0.71</b>	<b>(0.58 - 0.87)</b>	<b>0.0010</b>

Goodness-of-fit of the model: Chi-square statistic = 403.7, df=24, p-value < 0.0001  
 Bolded odds ratio parameters are statistically significant (0.05).

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