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

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## 14 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.
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Abstract: Recent studies have shown that consumers encounter various conflicting motivations 22 23 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 24 motivation and behavior (intention-behavior gap). The complexity of this subject increases when 25 26 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 27 distance between food choices and waste generation in homes and this factual interval (behavior-28 outcome gap) further affects consumer's decision-making when comparing available options. 29 Employing data from a three-year survey of a national representative panel of Italian consumers, 30 this study develops a system of regression models using path analysis methodology. The objective 31 32 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 33 stronger the influence on food waste generation in homes. Purchasing emerges as the most critical 34 choice of the consumers' food waste cycle. This gap between behavior and outcome adds 35 36 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 37 patterns with implications on food waste generation. Furthermore, purchasing decisions are 38 exposed to out-of-home contextual factors, suggesting that food retail can affect consumer 39 40 behaviors relevant to household food waste.

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## 43 Keywords

- 44 Consumer; attitude; behavior; food waste; purchasing; food retail.
- 45

### 46 **1. Introduction**

Why do individuals make and repeat choices that reduce their utility and lead to negative environmental and social impacts? What are the key behavioral factors explaining these apparently irrational actions? These questions assume a high complexity in the context of household food 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. On the one hand, analyses of consumers' food choices (e.g., contingent valuation surveys) should 58 consider the realm of food waste and the implications on their payoff. On the other hand, the 59 drivers of food waste should include not only socioeconomic conditions (Andreasen, 2002; Vermeir 60 and Verbeke, 2006; Evans, 2012, 2011; Farr-Wharton et al., 2014; Setti et al., 2016), but also 61 individual motivations and behaviors pertaining to the general food domain. In fact, when 62 consumers deal with various food resources, their evaluations tend to include a series of volitional 63 factors such as food security, status concerns, time pressure, and food waste. Moreover, food 64 decisions are influenced by deep-rooted and repeated judgments such as emotions, hunger, 65 values, and habits (i.e., "visceral factors"; Lowenstein, 1996; Verplanken et al., 1998; Graham-66 Rowe et al., 2014). Thus, a high uncertainty level tends to characterize consumers' food choices. 67

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

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

As for the consumer, a behavior-outcome gap can deliver an additional degree of uncertainty to food decisions. Considering the behaviors related to food and food waste are private and, thus, in essence not visible, the inherent risks cannot be limited by simply learning from others' experiences. This additional source of uncertainty raises the need for new research on the main non-standard behavioral schemes that can influence the generation of household food waste and, consequently, cause consumers to experience reduced private (and public) benefit or moral concern. Moreover, the expected deviation from rational and reasoned choices can influence
consumer's overall decision-making with possible repercussions on its earlier precursors (i.e.,
beliefs). This, in turn, can amplify the trade-offs between individuals' motivations when they
compare available food options, triggering a feedback loop that further raises the level of
uncertainty.

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

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# 112 **2. Theoretical background**

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

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

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

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

<sup>&</sup>lt;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.

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

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

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

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

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

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

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

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

205

# 206 **3. Methodology**

## 207 3.1 Sampling and data management

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

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

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

<sup>&</sup>lt;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.

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

## 234 3.1.1 Limitations

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

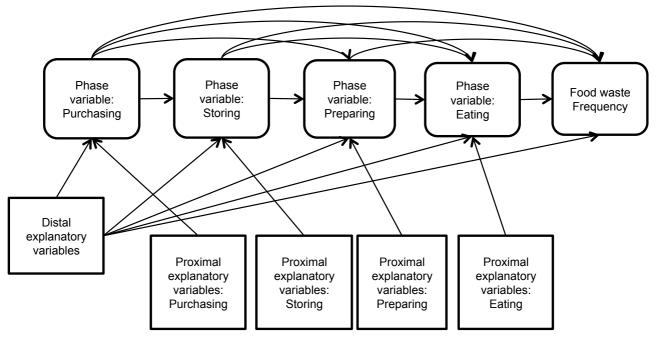
Secondly, the adopted data management process assigns heterogeneous sets of constitutive 242 items to the adopted phase variables and, in particular, a higher number to the 243 planning/purchasing and storing stages. However, this can imply an underestimation of the role of 244 the other phases (i.e.,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

### 250 3.2 Methodological approach

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

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

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



267 268

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

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).

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

1. implement the PPOM that includes non-proportional odds for all explanatory variables;

identify explanatory variables by means of statistical hypothesis tests to assess the equivalence
 of odds ratios that satisfy the proportional odds assumption; and

282 3. develop the PPOM considering only non-proportional odds explanatory variables identified in
 283 the previous stage.

For all the regression models, the significant results indicate odds ratio (OR) values with 95% 284 confidence intervals (CI). For the logistic models, the odds ratios represent multiplicative 285 coefficients on the odds of the related phase variable generating household food waste. In the 286 case of PPOMs, the odds ratios describe multiplicative coefficients on the odds of the higher level 287 of food waste frequency in the home (e.g., often). In particular, two odds ratios are considered 288 when PPOMs measure non-proportional odds explanatory variables. The first OR compares food 289 290 waste frequency (response level) "often" with "sometimes" or "never," while the second OR compares "often" or "sometimes" with "never". 291

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 chisquare statistic and associated p-values, which ascertain the adherence to the stated assumption and significance of the results.

<sup>269</sup> 

# 298 **4. Results**

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

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

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

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

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

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

### Table 1. Logistic model: "Planning/Purchasing" phase's drivers of household food waste

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

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

Bolded odds ratio parameters are statistically significant (0.05).

# Table 2. Logistic model: "Storing" phase's drivers of household food waste.

# Table 3. Logistic model: "Preparing" phase's drivers of household food waste.

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

Bolded odds ratio parameters are statistically significant (0.05).

### Table 4. Logistic model: "Eating" phase's drivers of household food waste.

Variable	Category	Reference	OR	95% CI	P- value
Preparing phase	Yes	No	2.27	(1.58 - 3.25)	<.0001
Storing phase	Yes	No	0.50	(0.37 - 0.69)	<.0001
Planning/Purchasing phase	Yes No		0.79	(0.58 - 1.09)	0.1566
Frequency of lunch at home	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
Frequency of dinner at home	Often	Sometimes /	1.10	(0.50 - 2.40)	0.8169
	Nearly always	Rarely / Never	1.06	(0.50 - 2.22)	0.8832
	Always	·	0.83	(0.39 - 1.78)	0.6361
Gender	Female	Male	0.80	(0.59 - 1.09)	0.1550
Macro region	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
Household size	2	1	0.50	(0.30 - 0.83)	0.0071
	3	·	0.70	(0.42 - 1.14)	0.1519
	4	·	0.55	(0.32 - 0.93)	0.0265
	>4	·	0.84	(0.44 - 1.60)	0.5901
Perception of monetary wealth	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
Education level	High school	Primary / middle	0.76	(0.51 - 1.13)	0.1694
	Master degree	school	0.81	(0.50 - 1.34)	0.4147
	PHD	·	0.50	(0.20 - 1.24)	0.1345
Age	18-34	35-64	1.09	(0.75 - 1.59)	0.6548
	65+	·	1.51	(1.00 - 2.30)	0.0524

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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		
Planning/Purchasing phase	Both	2.62	2.80	3.19	2.59	2.81	2.92		
Storing phase	Often vs. Sometimes/Never	-	0.55	-	0.65	0.53	0.53		
	Often/Sometimes vs. Never	-	1.54	2.97	1.30	1.32	-		
Preparing phase	Both	2.56	1.86	1.81	1.57	2.05	1.47		
Eating phase	Both	1.41	-	-	-	-	-		

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

### 372 **5. Discussion**

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

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

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

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

A second class of deviations is represented by non-standard preferences. Since food waste is conditioned by a considerable delay with respect to its driving choice (provisioning), when expected alternative rewards are compared and evaluated, individuals may demonstrate *timeinconsistent preferences*. This can be the case of ensuring food security against avoiding wastes in 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, when the most influential decisions are made, consumers are exposed to diverse information and 429 complex choices. To simplify their decision-making process, individuals could use different types of 430 sub-optimal heuristics such as orienting their choice to familiarity and identity criteria (habits; 431 Verplanken and Orbell, 2003), preselecting available information (limited attention), or referring to 432 the most noticeable elements of alternative options (salience). To this effect, the weight of the 433 434 relationship between "person responsible for shopping" and purchasing phase (source of household food waste) suggests that this typology of consumers could be inclined to resort to non-435 standard routines when buying food. By contrast, larger-sized households report higher food waste 436 437 frequency, which appears mainly due to mismanagement of food practices in the downstream phases (e.g., storing and preparing ones) rather than to adoption of heuristics at the provisioning. 438

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

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

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

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

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

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

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

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

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### 718 Appendix

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

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

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

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## 732 (Tables 6 – 14)

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

<b>6.</b> Distai explanatory vari			ooponiac
Distal variable	Value	Tot	%
Macro region	North-West	1,289	27.3%
	North-East	896	19.0%
	Center	919	19.5%
	South	1,079	22.8%
	Islands	543	11.5%
Household size	1	516	10.9%
	2	1,365	28.9%
	3	1,260	26.7%
	4	1,202	25.4%
	>4	383	8.1%
Perception of monetary	Feeling Poor	306	6.5%
wealth	Many	677	14.3%
	Difficulties		
	Some	1,959	41.5%
	Difficulties		
	Safely	1,784	37.8%
Education level	Primary /	964	20.4%
	middle school		
	High school	2,551	54.0%
	Master degree	968	20.5%
	PHD	243	5.1%
Yearly respondents	2013	1,706	36.1%
	2014	1,518	32.1%
	2015	1,502	31.8%
Gender	Male	2,286	48.4%
	Female	2,440	51.6%
Age	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	%
Use of shopping list	Never	396	8.4%
	Sometimes	1,933	40.9%
	Often	2,397	50.7%
Frequency of food	Every 15 days or	274	5.8%
shopping	more		
	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%
Weekly household food	0 - 50 €	544	11.5%
expenditure	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%
Frequency of non-	Never	369	7.8%
seasonal products	Rarely	2,121	44.9%
purchasing	Sometimes	1,941	41.1%
	Often	295	6.2%
Frequency of non-local	Never	351	7.4%
food purchasing	Rarely	1,954	41.4%
	Sometimes	2,081	44.0%
	Often	340	7.2%
Person responsible for	Different	1,022	21.6%
shopping	Always the same	3,704	78.4%
Take advantage of special	Sometimes/Never	826	17.5%
offers	Often	2,107	44.6%
	Always	1,793	38.0%
Shopping strategy	Low cost	281	6.0%
	Pragmatic	3,232	68.4%
	High quality	1,113	23.6%
	MISSING	100	2.1%

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

Table 8. Frequency of waste per food category and year (number and % of respondents).								
Year	Never	Sometimes	Often	Missing				
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%)				
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%)				
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%)				
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%)				
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%)				
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%)				
	Year           Total           2013           2014           2015           Total           2013           2014           2013           2014           2015           Total           2013           2014	Year         Never           Total         2,895 (61.3%)           2013         1,050 (61.6%)           2014         976 (64.3%)           2015         869 (57.9%)           Total         2,929 (62.0%)           2013         1,102 (64.6%)           2014         962 (63.4%)           2015         865 (57.6%)           Total         1,637 (34.6%)           2013         642 (37.6%)           2014         529 (34.9%)           2015         466 (31.0%)           2014         529 (34.9%)           2015         466 (31.0%)           Total         2,967 (62.8%)           2013         1,219 (71.5%)           2014         917 (60.4%)           2015         831 (55.3%)           Total         3,184 (67.4%)           2013         1,197 (70.2%)           2014         1,063 (70.0%)           2015         924 (61.5%)           Total         3,391 (71.8%)           2013         1,270 (74.4%)           2014         1,093 (72.0%)	Year         Never         Sometimes           Total         2,895 (61.3%)         1,327 (28.1%)           2013         1,050 (61.6%)         469 (27.5%)           2014         976 (64.3%)         387 (25.5%)           2015         869 (57.9%)         471 (31.4%)           Total         2,929 (62.0%)         1,502 (31.8%)           2013         1,102 (64.6%)         526 (30.8%)           2014         962 (63.4%)         445 (29.3%)           2015         865 (57.6%)         531 (35.3%)           2015         865 (57.6%)         531 (35.3%)           2013         642 (37.6%)         915 (53.6%)           2014         529 (34.9%)         860 (56.6%)           2015         466 (31.0%)         905 (60.3%)           Total         2,967 (62.8%)         1,402 (29.7%)           2013         1,219 (71.5%)         384 (22.5%)           2014         917 (60.4%)         468 (30.8%)           2015         831 (55.3%)         550 (36.6%)           2015         831 (55.3%)         550 (36.6%)           2014         917 (70.2%)         378 (22.2%)           2015         924 (61.5%)         442 (29.4%)           2015         924 (61.	YearNeverSometimesOftenTotal2,895 (61.3%)1,327 (28.1%)284 (6.0%)20131,050 (61.6%)469 (27.5%)76 (4.4%)2014976 (64.3%)387 (25.5%)109 (7.2%)2015869 (57.9%)471 (31.4%)99 (6.6%)Total2,929 (62.0%)1,502 (31.8%)211 (4.5%)20131,102 (64.6%)526 (30.8%)56 (3.3%)2014962 (63.4%)445 (29.3%)79 (5.2%)2015865 (57.6%)531 (35.3%)76 (5.1%)Total1,637 (34.6%)2,680 (56.7%)382 (8.1%)2013642 (37.6%)915 (53.6%)142 (8.3%)2014529 (34.9%)860 (56.6%)121 (8.0%)2015466 (31.0%)905 (60.3%)119 (7.9%)Total2,967 (62.8%)1,402 (29.7%)240 (5.1%)2015831 (55.3%)550 (36.6%)72 (4.2%)2014917 (60.4%)468 (30.8%)96 (6.3%)2015831 (55.3%)550 (36.6%)72 (4.2%)20131,197 (70.2%)378 (22.2%)72 (4.2%)20131,197 (70.2%)378 (22.2%)72 (4.2%)20141,063 (70.0%)335 (22.1%)64 (4.2%)2015924 (61.5%)442 (29.4%)63 (4.2%)20141,063 (70.0%)336 (22.1%)185 (3.9%)20131,270 (74.4%)336 (19.7%)68 (4.0%)20141,093 (72.0%)336 (22.1%)52 (3.4%)				

<b>Table 9.</b> Partial Proportional Odds Model: drivers of fresh bread waste.
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Phase and distal	Category	Reference	Response level	OR	95% CI	P
variable		Category				value
Planning/Purchasing phase	Yes	No	Both	2.62	(2.30 - 2.98)	<.000
Storing phase	Yes	No	Often vs. Sometimes/Never	0.80	(0.62 - 1.04)	0.0886
			Often/Sometimes	1.15	(0.99 - 1.34)	0.061
Preparing phase	Yes	No	Both	2.56	(2.16 - 3.04)	<.000
Eating phase	Yes	No	Both	1.41	(1.03 - 1.94)	0.032
Gender	Female	Male	Both	0.85	(0.74 - 0.96)	0.009
Macro region	Center	North-West	Both	1.35	(1.12 - 1.64)	0.001
	Islands	-	Both	1.18	(0.94 - 1.48)	0.149
	North-East	-	Both	1.04	(0.86 - 1.27)	0.6917
	South	-	Both	1.45	(1.21 - 1.75)	<.000
Household size	2	1	Both	1.01	(0.79 - 1.29)	0.926
	3	-	Both	1.28	(1.00 - 1.63)	0.047
	4	-	Both	1.36	(1.06 - 1.75)	0.015
	>4	-	Both	1.38	(1.02 - 1.88)	0.038
Perception of	Many difficulties	Feeling poor	Both	0.92	(0.68 - 1.24)	0.575
monetary wealth	Some difficulties	-	Both	0.82	(0.62 - 1.09)	0.171
	Safely	-	Both	1.14	(0.86 - 1.51)	0.370
Education level	High school	Primary /	Both	1.00	(0.84 - 1.2)	0.975
	Master degree	middle school	Both	1.21	(0.97 - 1.49)	0.087
	PHD	-	Both	1.07	(0.78 - 1.47)	0.663
Yearly respondents	2014	2013	Often vs. Sometimes/Never	1.57	(1.16 - 2.14)	0.003
			Often/Sometimes vs. Never	0.99	(0.84 - 1.15)	0.8513
	2015	-	Often vs. Sometimes/Never	1.41	(1.03 - 1.93)	0.032
			Often/Sometimes vs. Never	1.24	(1.05 - 1.45)	0.009 <sup>,</sup>
Age	18-34	35-64	Both	1.07	(0.92 - 1.24)	0.395
	65+	-	Both	0.69	(0.57 - 0.84)	0.000

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	<b>Table 10.</b> Partial Proportional Odds Model: drivers of cheese waste.
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Phase and distal	Category	Reference	Response level	OR	95% CI	P-
variable		Category				value
Planning/Purchasing phase	Yes	No	Both	2.80	(2.46 - 3.18)	<.0001
Storing phase	Yes	No	Often vs. Sometimes/Never	0.55	(0.42 - 0.74)	<.0001
			Often/Sometimes vs. Never	1.54	(1.33 - 1.78)	<.0001
Preparing phase	Yes	No	Both	1.86	(1.57 - 2.22)	<.0001
Eating phase	Yes	No	Both	0.76	(0.54 - 1.07)	0.1109
Gender	Female	Male	Both	1.02	(0.90 - 1.16)	0.7396
Macro region	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
Household size	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
Perception of	Many difficulties	Feeling poor	Both	1.07	(0.78 - 1.46)	0.6783
monetary wealth	Some difficulties	—	Both	1.17	(0.88 - 1.56)	0.2706
	Safely	—	Both	1.38	(1.03 - 1.84)	0.0291
Education level	High school	Primary /	Both	0.97	(0.82 - 1.16)	0.7521
	Master degree	middle school	Both	1.09	(0.88 - 1.34)	0.4300
	PHD		Both	1.14	(0.83 - 1.55)	0.4209
Yearly respondents	2014	2013	Both	1.10	(0.95 - 1.29)	0.2013
	2015	_	Both	1.34	(1.15 - 1.56)	0.0002
Age	18-34	35-64	Both	1.44	(1.24 - 1.66)	<.0001
	65+	_	Both	0.73	(0.61 - 0.87)	0.0007

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).

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

Phase and distal	Category	Reference	Response level	OR	95% CI	P-
variable		Category				value
Planning/Purchasing phase	Yes	No	Both	3.19	(2.80 - 3.63)	<.0001
Storing phase	Yes	No	Often vs. Sometimes/Never	0.91	(0.72 - 1.15)	0.4224
			Often/Sometimes vs. Never	2.97	(2.57 - 3.44)	<.0001
Preparing phase	Yes	No	Both	1.81	(1.51 - 2.16)	<.0001
Eating phase	Yes	No	Both	1.01	(0.74 - 1.39)	0.9334
Gender	Female	Male	Both	0.95	(0.84 - 1.07)	0.3825
Macro region	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
Household size	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
Perception of	Many difficulties	Feeling poor	Both	1.00	(0.76 - 1.33)	0.9763
monetary wealth	Some difficulties		Both	1.06	(0.82 - 1.37)	0.6627
	Safely	_	Both	1.22	(0.94 - 1.59)	0.1340
Education level	High school	Primary /	Both	1.11	(0.95 - 1.30)	0.2014
	Master degree	middle school	Both	1.27	(1.04 - 1.55)	0.0204
	PHD	_	Both	1.42	(1.05 - 1.93)	0.0221
Yearly respondents	2014	2013	Both	1.14	(0.99 - 1.32)	0.0719
	2015		Both	1.21	(1.05 - 1.40)	0.0099
Age	18-34	35-64	Both	1.09	(0.94 - 1.25)	0.2516
	65+	—	Both	0.75	(0.64 - 0.89)	0.0009

Goodness-of-fit of the model: Chi-square statistic = 651.3, df=24, p-value < 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 y	ognun waste.

Phase and distal	Category	Reference	Response level	OR	95% CI	Р
variable		Category				valu
Planning/Purchasing phase	Yes	No	Both	2.59	(2.27 - 2.94)	<.0001
Storing phase	Yes	No	Often vs. Sometimes/Never	0.65	(0.49 - 0.85)	0.0019
			Often/Sometimes vs. Never	1.30	(1.12 - 1.50)	0.0006
Preparing phase	Yes	No	Both	1.57	(1.32 - 1.87)	<.000
Eating phase	Yes	No	Both	1.23	(0.89 - 1.69)	0.2090
Gender	Female	Male	Both	0.91	(0.80 - 1.03)	0.146
Macro region	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)	0.591
	South	_	Both	1.15	(0.96 - 1.38)	0.1278
Household size	2	1	Both	0.87	(0.69 - 1.10)	0.242
	3		Both	1.17	(0.93 - 1.48)	0.189
	4		Both	1.27	(1.00 - 1.61)	0.054
	>4	_	Both	1.17	(0.87 - 1.59)	0.296
Perception of	Many difficulties	Feeling poor	Both	0.99	(0.72 - 1.37)	0.9446
monetary wealth	Some difficulties		Both	1.20	(0.90 - 1.61)	0.219
	Safely	_	Both	1.36	(1.01 - 1.83)	0.045
Education level	High school	Primary /	Both	1.10	(0.92 - 1.32)	0.290
	Master degree	middle school	Both	1.29	(1.04 - 1.59)	0.021
	PHD	_	Both	1.47	(1.08 - 2.00)	0.015
Yearly respondents	2014	2013	Often vs. Sometimes/Never	1.58	(1.15 - 2.17)	0.0044
			Often/Sometimes vs. Never	1.76	(1.50 - 2.06)	<.000 <sup>,</sup>
	2015	_	Often vs. Sometimes/Never	1.10	(0.79 - 1.54)	0.578
			Often/Sometimes vs. Never	1.99	(1.70 - 2.33)	<.000 <sup>,</sup>
Age	18-34	35-64	Both	1.48	(1.28 - 1.71)	<.000
	65+	_	Both	0.71	(0.59 - 0.86)	0.0004

Bolded odds ratio parameters are statistically significant (0.05).

765 <b>Table 13.</b> Partial Proportional Odds Model: drivers of cold cuts was
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Phase and distal	Category	Reference	Response level	OR	95% CI	P value
variable	Yes	Category	Both	2.04	(0.45 2.00)	
Planning/Purchasing phase	res	No	Both	2.81	(2.45 - 3.22)	<.000′
Storing phase	Yes	No	Often vs.	0.53	(0.40 - 0.72)	<.000
			Sometimes/Never			
			Often/Sometimes	1.32	(1.13 - 1.54)	0.0006
			vs. Never			
Preparing phase	Yes	No	Both	2.05	(1.72 - 2.45)	<.000
Eating phase	Yes	No	Both	1.29	(0.92 - 1.81)	0.1422
Gender	Female	Male	Both	1.03	(0.90 - 1.18)	0.685
Macro region	Center	North-West	Both	1.35	(1.10 - 1.65)	0.0034
	Islands	_	Both	1.92	(1.53 - 2.42)	<.000
	North-East		Both	0.96	(0.78 - 1.18)	0.669
	South		Both	1.47	(1.21 - 1.78)	0.000
Household size	2	1	Often vs.	0.60	(0.35 - 1.03)	0.063
			Sometimes/Never			
			Often/Sometimes	0.98	(0.75 - 1.27)	0.877
			vs. Never			
	3		Often vs.	1.05	(0.64 - 1.73)	0.855
			Sometimes/Never			
			Often/Sometimes	1.29	(1.00 - 1.68)	0.054
			vs. Never			
	4		Often vs.	1.00	(0.61 - 1.66)	0.9954
			Sometimes/Never			
			Often/Sometimes	1.43	(1.10 - 1.87)	0.008
			vs. Never			
	>4		Often vs.	0.66	(0.32 - 1.35)	0.252
			Sometimes/Never			
			Often/Sometimes	1.72	(1.25 - 2.38)	0.001
			vs. Never			
Perception of	Many difficulties	Feeling poor	Both	1.07	(0.76 - 1.50)	0.698
monetary wealth	Some difficulties	_	Both	1.09	(0.80 - 1.48)	0.592
	Safely		Both	1.33	(0.98 - 1.82)	0.071
Education level	High school	Primary / middle school	Both	0.87	(0.73 - 1.05)	0.156
	Master degree		Both	1.06	(0.85 - 1.32)	0.624
	PHD	_	Both	1.07	(0.77 - 1.49)	0.674
Yearly respondents	2014	2013	Both	1.03	(0.87 - 1.21)	0.731
	2015		Both	1.43	(1.22 - 1.68)	<.000
Age	18-34	35-64	Both	1.18	(1.01 - 1.38)	0.034
	65+		Both	0.78	(0.64 - 0.95)	0.012

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).

770	<b>Table 14.</b> Partial Proportional Odds Model: drivers of eggs waste.

Phase and distal	Category	Reference	Response level	OR	95% CI	P-
variable		Category				value
Planning/Purchasing phase	Yes	No	Both	2.92	(2.55 - 3.36)	<.0001
Storing phase	Yes	No	Often vs.	0.53	(0.39 - 0.72)	<.0001
			Sometimes/Never			
			Often/Sometimes	1.02	(0.87 - 1.19)	0.8437
			vs. Never		(1.55. (	
Preparing phase	Yes	No	Both	1.47	(1.22 - 1.77)	<.0001
Eating phase	Yes	No	Both	1.11	(0.78 - 1.56)	0.5677
Gender	Female	Male	Both	0.86	(0.75 - 0.99)	0.0360
Macro region	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
Household size	2	1	Both	0.67	(0.53 - 0.85)	0.0009
	3		Both	0.76	(0.60 - 0.96)	0.0217
	4		Both	0.63	(0.49 - 0.81)	0.0003
	>4	_	Both	0.57	(0.41 - 0.79)	0.0007
Perception of	Many difficulties	Feeling poor	Both	0.87	(0.62 - 1.22)	0.4089
monetary wealth	Some difficulties		Both	0.82	(0.60 - 1.11)	0.1961
	Safely		Both	1.00	(0.73 - 1.36)	0.9792
Education level	High school	Primary /	Both	0.92	(0.77 - 1.12)	0.4145
	Master degree	middle school	Both	1.08	(0.86 - 1.35)	0.5305
	PHD	_	Both	1.65	(1.20 - 2.27)	0.0023
Yearly respondents	2014	2013	Both	1.11	(0.94 - 1.31)	0.2078
	2015		Both	1.27	(1.07 - 1.50)	0.0051
Age	18-34	35-64	Both	1.33	(1.14 - 1.56)	0.0004
	65+	_	Both	0.71	(0.58 - 0.87)	0.0010

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).