

1 Measuring Consumers' Acceptance in Food Labels: a Cross- 2 Country Investigation on Usefulness, Ease of Use and Trust

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9 **Abstract:** Front-Of-Pack Labels (FOPLs) are being increasingly investigated as a tool to
10 guide customers toward healthier food choices. Existing theoretical models focused on
11 consumers understanding have not been able to provide clear evidences of which labels are
12 more effective on food choices. Drawing on extant studies conducted on primary grocery
13 shoppers from Italy, France and the UK we developed a new framework, the Front-Of-Pack
14 Acceptance Model (FOPAM), to evaluate label effectiveness in terms of usefulness, ease of
15 use, attitude, trust and behavioral intention. Our findings highlight significant relationships
16 between perceived usefulness, ease of use of the labels and trust towards them, and the way
17 consumers form their attitudes and intentions towards buying healthier products.

18 *Introduction*

19 As part of the worldwide effort to help citizens reduce disease connected to food habits, in
20 recent years governments, authorities, and socially responsible businesses and organisations,
21 have been involved in activities that support consumers in making healthier food choices
22 (WHO, 2020).

23 In this perspective, Front-of Pack labels, or FOPLs, have been identified as a fundamental
24 tool to promote a healthier diet. FOPLs are labels placed on the front of food packaging, that
25 provide detailed and simplified information about food composition and nutritional values
26 (van der Bend, 2019). The lack of a streamlined, shared regulation has led to the proliferation
27 of multiple types of labels in various countries, with different designs and underlying
28 categories (such as per the Taxonomy devised by the EU).

29 **Table 1. Recent taxonomy of EU labels**

Nutrient-Specific Labels

Provide detailed information about certain nutrients with the quantities contained in the food.

- **Numerical Labels** (e.g., NutrInform Battery)
- **Colour-Coded Labels** (e.g., Multiple Traffic Light)

Summary Labels

Provide a synthetic assessment of the overall nutritional healthiness through an algorithm

- **Endorsement Logos** (e.g., Keyhole logo)
- **Graded Indicators** (e.g., Nutri-Score)

30 While the approaches are different, an ongoing debate at EU level aims at ensuring better
31 labelling information, and part of its goals is to identify which labels allow to pursue the best
32 outcome in terms of customer education toward informed and healthier diets. In this context,
33 one of the most discussed frameworks (Hersey et al., 2013; Ducrot et al., 2015, 2016, 2016a;
34 Talati et al., 2017; Ares et al., 2018; Finkelstein et al., 2019, Egnell et al., 2019, 2020;
35 Dreano-Trecant et al., 2020; Mazzù et al., 2021a) evaluate label efficacy is the model of
36 Grunert&Wills (2007), which differentiates the phase of understanding the message brought
37 forward by the labels between subjective and objective.

38 Yet, focusing on “understanding” has not been able to provide evidence of the existence of a
39 uniform trait of "superiority" (Mazzù et al., 2021b). Moreover, while the impact of FOPL in
40 identifying healthy food in the context of purchase intention and choice has been widely
41 evaluated, (Ducrot et al., 2016; Talati et al., 2017), effectiveness at the time of purchase still
42 needs to be.

43 Our attempt to close a gap both in literature and in which is the best tool to actually help
44 consumers make better choices for their health, is through an approach based on an alternative
45 evaluation model that could provide a new perspective on the FOPL-driven decision-making
46 process.

47 Our approach has thus concentrated on assessing whether switching the focus to attitude
48 formation, purchase intention, and trust, could positively contribute to the issue. We have
49 developed a sequence of studies, based on Technology Acceptance Model (Davis, 1989;
50 1993), to extend the evaluation of FOPLs.

51 The following paper is divided as follows: after an overview of the theoretical frameworks,
52 we discuss our research methodology and showcase our model, together with the main
53 implications of our studies.

54 ***Conceptual Framework and Extant Literature***

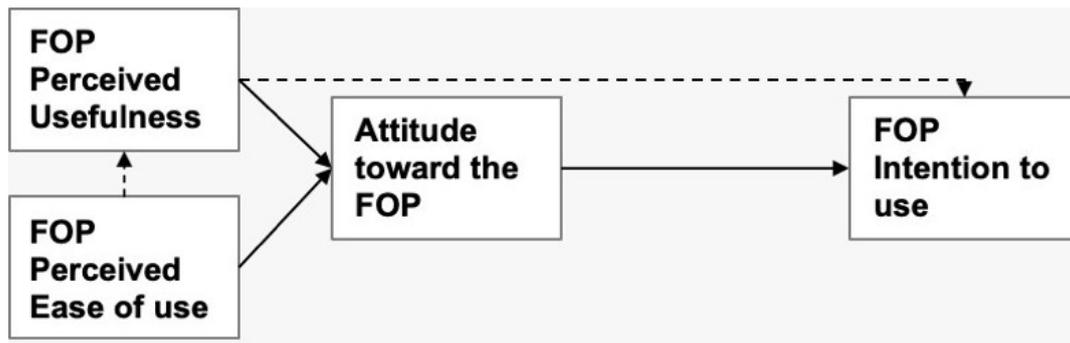
55 As labels could be seen as a Decision Support System (DSS), their acceptance as a technology
56 could be evaluated through the Technology Acceptance Model (Davis, 1989) which
57 establishes “perceived usefulness” (PU) and “perceived ease of use” (PEOU), as antecedents
58 to a customers’ attitude toward using the system and purchase/usage intention.

59 Over the years, the TAM has proven to be flexible in its applications, in sectors that go
60 beyond technology, thanks to its parsimony and predictive power (Venkatesh, 2000).
61 Examples are the acceptance of new policies (Pierce et al., 2014), food traceability (Kim &
62 Woo, 2016), online food purchasing (Nguyen et al., 2019). It could thus be leveraged to
63 understand whether FOPLs can be an easy-to-use system for consumers to make healthier
64 choices when buying food.

65 Moreover, investigating further mediating variables could strengthen the effects produced in
66 the relationship. Trust is a factor that majorly affects the decision process in terms of purchase
67 (Hong and Cha, 2013) and repurchase (Trivedi and Yadav, 2019) intention, technology
68 adoption (Van et al., 2020); it could thus increase the predictive power of the TAM (Gefen,
69 2004). When consumers feel trust, they are more likely to buy or use a product or service
70 (Aydin et al., 2014), because of a perceived risk reduction (Handi et al., 2018). Conversely, a
71 reduction in trust might happen when the amount of information consumers have access to is
72 reduced (Hong and Cha, 2013), as they might perceive an increase in uncertainty.

73 It is thus possible that consumers may feel less trust in a label providing less information
74 about the food content than in a detailed one. Considering that some labels, such as the Nutri-
75 Score (NS), employ algorithms to provide a measure of food healthiness, there is also a need
76 to understand the impact that these calculations, and the trust that consumers feel towards
77 them, have on behavior in the context of FOPLs.

78 **Figure 1. Adapted Technology Acceptance Model tested in studies 1 and 2**



79
80

81 In study 1, we tested the acceptance of two different labels, the Multiple Traffic Light (MTL),
82 a Nutrient-Specific Colour-coded Label and the Nutri-Score, a Graded-Indicator Summary
83 Label, while assessing the effectiveness of PEOU, PU and attitude as explanatory variables of
84 consumers’ intentions. In study 2, we assessed the relationships among PU, PEOU, Attitude
85 and Behavioral Intention in a sample of 940 Italian Primary Grocery Shopper divided into
86 three subgroups: (1) one for the NutrInform Battery (n=319), (2) one for MTL (n=318), and
87 (3) one for Nutri-Score (n=303). In Study 3 we assess whether trust is a relevant mediator of
88 behavioral intention under conditions of exposure to a Nutrient-Specific Label and a
89 Summary Label, also evolving the Front-of-Pack-Acceptance-Model (FOPAM); in Study 3.b,
90 we evaluate trust dimensions, with focus on the perception of the presence of an algorithm
91 behind FOPLs nutritional information, while explaining the computational details of the
92 algorithm and preventing potential biases deriving from previous label exposure.

93 **Research Hypotheses**

94 **Table 2. Research Hypotheses**

Study	Description
Studies 1 and 2	<ul style="list-style-type: none"> • H1: Front-of-Pack Perceived Usefulness positively affects the attitude toward the usage • H2: Front-of-Pack Perceived Ease of Use positively affects the Front-of-PacPerceived Usefulness • H3: Front-of-Pack Perceived Ease positively affects the attitude toward the usage • H4: Front-of-Pack Perceived Usefulness positively affects the behavioral intention • H5: Attitude toward the usage positively affects the behavioral intention
Study 3	<ul style="list-style-type: none"> • H6: Trust mediates the relationship between attitude toward using the FOP and behavioral intention.
Study 3.b	<ul style="list-style-type: none"> • H7: The attitude toward the usage of NutrInform Battery is higher than the attitude toward the usage of the Nutri-Score also in contexts where the FOPL computational methods are made transparent • H8: Trust toward FOPL mediates the relationship between attitude toward using the FOPL and behavioral intention also in contexts where the FOPL computational methods are made transparent • H9: Behavioral intention enhanced by Nutri-Score is higher than the behavioral intention enhanced by the NutrInform Battery also in contexts where the FOPL computational methods are made transparent • H10: The trust toward NutrInform Battery is higher than the trust toward the Nutri-Score also in contexts where the FOPL computational methods are made transparent

95 **Research Objectives:** To establish the role of acceptance in the evaluation of the
96 performance of FOPLs, we validate the “Front-of-Pack Acceptance Model” (FOPAM), a new
97 theoretical model applied to food purchase and healthy diet decision making.

98 Our approach began by defining item scales and then testing the FOPAM on them. We
99 decided to apply our model to three labels representative of the different blocks of the EU
100 Taxonomy, and that follow a logical continuum.

101 We thus selected two Nutrient-specific labels, the Multiple Traffic Light, a color coded label,
 102 and the NutrInform Battery, a numerical indicator, and one graded Summary Label, the Nutri-
 103 Score, as they reflect a continuum in the recent classification outlined by the EU Commission
 104 (2020).

105 We then assessed the robustness and goodness-of-fit of the model, by surveying 3001
 106 purchase responsables from three markets (UK, France, Italy) characterised by different
 107 situations in terms of obesity rates, introduction date of the market, exposure and awareness
 108 levels, (Talati et al, 2017). In particular, the UK has adopted the Multiple Traffic Light since
 109 2013 while France has used the Nutri-Score since 2017 in France. Italy has also been selected
 110 for the lack of previous exposure to FOPL. We also focused on the role of trust in the model,
 111 to assess its relevance as a mediating factor.

112 **Methods and Materials**

113 **Study Population:** We worked on a cumulative sample of 3,001 purchase responsible on
 114 multiple studies. All surveys were performed in each respondent’s language, and sample sizes
 115 were computed in accordance with Kline’s (2015) optimal ratios of observation : parameter
 116 (20:1, with a lower limit of 10:1)

117 **Table 3. Sample**

Study	Population	Description
Preliminary Phase Early Stage	61 Italian MBA Students	<ul style="list-style-type: none"> • Assessment of main label associations with PU and PEOU • Could have been influenced by previous exposure • Awarded with extra points on the final grade
Preliminary Phase Confirmatory	288 (France + UK)	<ul style="list-style-type: none"> • Understanding of PU and PEOU associations with FoPL • Confirmation of the validity in two countries in which FoPL are already in use
Study 1	1523 (France + UK)	<ul style="list-style-type: none"> • Identification of construct relationships in FOPL-mediated contexts
Study 2	191 (Italy)	<ul style="list-style-type: none"> • Three subgroups: NutrInform, MTL, Nutri-Score
Study 3 and 3.b	189 (Italy)	<ul style="list-style-type: none"> • Manipulation check + analysis on mediation effect

118 Surveys were administered via Prolific, a web panel provider with high recruitment standards,
 119 a proper response rate and high reliability and replicability of studies (Palan & Schitter,
 120 2018). We used SPSS Statistics for reliability tests, and SPSS AMOS for CFA and SEM.

121 **Research design:** In the preliminary phase, MBA students were tasked with composing a
 122 healthy meal by considering information derived from labels applied to different alternatives
 123 of specific products. They then performed prioritization, to rank items according to how well
 124 the meaning of each statement matched the given definition of PEOU and PU (Sherif &
 125 Sherif, 1967), and categorization, to allocate items into clusters according to their similarity
 126 (Davis, 1989; Sherif & Sherif, 1967), leveraging spontaneous associations.

127 In Study 1, we tested the French and UK samples, to verify whether the relationships among
 128 constructs were observable in Front-of-Pack labels mediated contexts (i.e., Nutri-Score and
 129 Multiple Traffic Light). We assessed the validity of PU, PEOU, Attitude Toward the Using
 130 (ATT) and Behavioral Intention (BI) items through a CFA. Subsequently, in study 2 we tested
 131 the validity of FOPAM in Italy, a country where no FoPL system has been adopted, on
 132 NutrInform Battery, on the Multiple Traffic Light, and on the Nutri-Score. Study 3 poised to
 133 understand, via a manipulation check, the degree to which consumers perceive the nutritional
 134 information of FOPLs as deriving from an algorithm, nutritional experts, consumers’
 135 associations, firms or institutions, as well as the mediating effect of trust.

136 **Analysis:** After a de-duplication process to reduce redundancies, a combined 33 items
 137 concerning PU and PEOU were selected using scales from studies regarding technology

138 (Davis, 1989, 1993; Venkatesh et al., 2003) and food (Kim & Woo, 2016; Ma et al., 2017),
139 plus items for “Attitude toward using the label” and “Behavioral Intention”, adapted to the
140 context of food shopping in FoP-mediated settings. Mirroring Davis’s approach (1989), we
141 obtained three main clusters for both scales. A combination of Python code, Gephi and R
142 code was employed to represent the clusters as part of a weighted network graph. Modularity
143 was employed to create communities (Langfelder et al., 2008) of clusters, prioritising items
144 according to a scoring model which considered the times an item was ranked in a position.
145 Items were then ranked based on their normalized score. The scales were confirmed by
146 surveying French and UK grocery shoppers, already exposed to FoPLs, applying Cronbach’s
147 Alpha, Principal Component Analysis (PCA) and Confirmatory Factor Analysis (CFA) and
148 resulting in 4 items for PU and 3 for PEOU.

149 Study 1 was employed to (1) validate the effectiveness of the scales tested in the preliminary
150 phase in a larger sample to confirm that PU and PEOU function as prominent constructs in
151 forming attitudes and behavioural intention in FOP-mediated contexts, and (2) develop and
152 introduce FOPAM through a Structural Equation Model, adapting TAM to food contexts.

153 In study 2, a CFA was followed by the evaluation of the goodness of fit of the model. A
154 multi-group analysis evaluated the invariance across the groups exposed to the three different
155 labels.

156 **Findings**

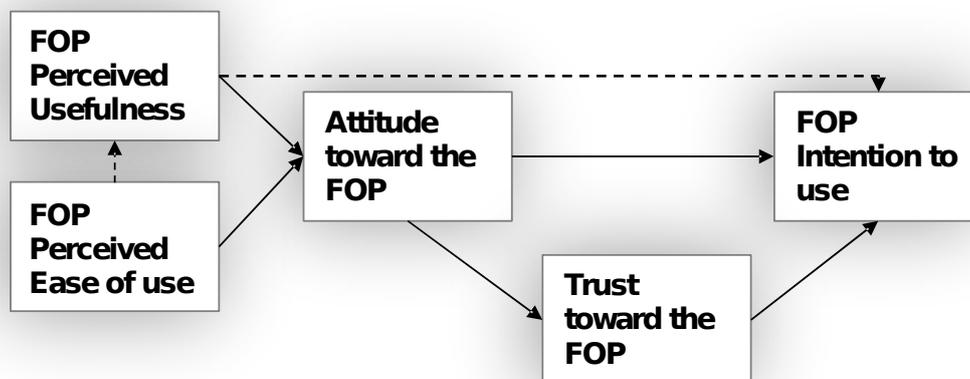
157 *Study 1:* The CFA suggested a high model fit (UK: $\chi^2 = 229.651$, $p=0.000$; GFI=0.953;
158 CFI=0.981; NFI = 0.977; SRMR=0.028; RMSEA=0.069) (France: $\chi^2 = 130.197$, $p=0.000$;
159 GFI=0.969; CFI=0.989; NFI = 0.982; SRMR=0.0214; RMSEA=0.051). Also, for the SEM
160 (UK: $\chi^2 = 308.673$, $p=0.000$; GFI=0.937; CFI=0.973; NFI = 0.969; SRMR=0.0621;
161 RMSEA=0.080) (France: $\chi^2 = 137.312$, $p=0.000$; GFI=0.968; CFI=0.988; NFI = 0.981;
162 SRMR=0.0623; RMSEA=0.023) all the relationships among latent constructs were
163 significant, and PEOU significantly predicted PU (UK: $\beta=0.689$, $p<0.001$; France: $\beta=0.522$,
164 $p<0.001$) and attitude toward usage ($\beta=0.230$, $p<0.001$; France: $\beta=0.305$, $p<0.001$).
165 Specifically, in a FoPL context, PU significantly affects attitude toward usage (UK: $\beta=0.746$,
166 $p<0.001$; France: $\beta=0.653$, $p<0.001$), which in turn affects BI (UK: $\beta=0.386$, $p<0.001$; France:
167 $\beta=0.564$, $p<0.001$). Attitude significantly mediates the relationship between PU and BI (UK:
168 $\beta=0.287$, $p<0.001$; France: $\beta=0.259$, $p < 0.001$). This study validates the FOPAM model,
169 connecting PU and PEOU to Attitude Formation and BI.

170 *Study 2:* The CFA showed an acceptable fit of the sample data of Italian customers to the
171 model (NS: $\chi^2 = 151.077$, $p=0.000$; GFI=0.923 CFI=0.977; NFI = 0.966; SRMR=0.0220;
172 RMSEA=0.084; MTL: $\chi^2 = 102.822$, $p=0.000$; GFI=0.951; CFI=0.988; NFI = 0.977;
173 SRMR=0.0210; RMSEA=0.060; NI: $\chi^2 = 109.207$, $p=0.000$; GFI=0.945; CFI=0.985; NFI =
174 0.974 SRMR=0.0260; RMSEA=0.063), and a high reliability and validity of the constructs. It
175 indicates that the measurements are discriminating properly among themselves and converge
176 within the assumed construct. Convergent validity suggests that all loadings are measuring the
177 construct properly whilst, as per the discriminant validity, are distinctive among them ($\lambda>0.70$
178 and $SMC>0.50$). For all the three labels, PEOU significantly predicts PU (NI: $\beta=0.723$,
179 $p<0.001$; MTL: $\beta=0.697$, $p<0.001$; NS: $\beta=0.650$, $p<0.001$) and Attitude (NI: $\beta=0.287$, $p<0.001$;
180 MTL: $\beta=0.324$, $p<0.001$; NS: $\beta=0.184$, $p<0.001$) which in turn positively affects BI (NS:
181 $\beta=0.291$, $p<0.001$; MTL: $\beta=0.267$, $p<0.01$; NI: $\beta=0.547$, $p<0.001$). BI is also predicted by PU
182 (NI: $\beta=0.363$, $p<0.001$; MTL: $\beta=0.267$, $p<0.001$; NS: $\beta=0.291$, $p<0.001$). The invariance test
183 showed that the configural invariance model across front-of-packs demonstrated an
184 acceptable model fit. In the configural model all parameters are estimated for the calibration
185 and validation groups simultaneously; that is, no parameter is constrained equally across

186 groups (Byrne, 2013). This multigroup model yielded a χ^2 value of 474.073 with 210 degrees
 187 of freedom and served as the baseline referent against which all subsequent models were
 188 compared. In the second model tested (measurement weights), all factor loadings of the
 189 indicator variables were constrained equally across groups, hypothesizing no variance in the
 190 way constructs are formed. In the models (structural weights, covariances and residuals) all
 191 betas, covariances and residuals were constrained equally across groups, hypothesizing no
 192 variance in the way relationships between constructs were formed. Observing the fact that the
 193 ΔCFI across all models never exceeded the cutoff of 0.01, according to Cheung & Rensvold
 194 (2002), we concluded that the model is completely and totally invariant across the three
 195 groups. In summary, this conclusion advises that all factor loadings, structural paths, factor
 196 covariances, factor residual variances, and measurement error variances are operating
 197 equivalently across calibration and validation samples.

198 *Study 3:* Study 3 revealed that an increase of attitude toward using the FOPL led to a higher
 199 trust toward the label ($b=0.73$; $t(189)=16.37$; $p<0.001$); the trust, in turn, positively affected
 200 consumers' Behavioral Intention ($b=0.36$; $t(189)=6.26$; $p<0.001$). Whereas, since the
 201 algorithm aversion does not vary significantly across the sample, we found it to be not
 202 significant ($b=0.25$; $t(189)=0.92$; $p=0.35$). When considering attitude toward using FOPL in
 203 the regression model with Behavioral Intention (BI) as the dependent variable, the variable
 204 shows a positive significant effect on BI ($b=0.53$; $t(189)=9.691$; $p<0.001$). More importantly,
 205 the indirect effect of trust variable on BI was positive and significant ($b=0.2607$; 95%
 206 CI:0.1659 – 0.3675), thus confirming that trust partially mediates the effect of attitude toward
 207 using the label on Behavioral Intention.

208 **Figure 2. Front-of-Pack Acceptance Model adopted in studies 3 and 3.b**



209
 210 *Study 3.b:* Subsequently, to assess whether the Front-of-Pack labels cued with computational
 211 details significantly vary the perception of the two Front-of-Pack labels, we ran a sequence of
 212 independent two-sample t-test. As regards the comparison between Nutri-Score and
 213 NutrInform Battery cued with computational information, the Nutrient-specific label showed
 214 higher means in terms of attitude (MNutri-Score=5.30; SD = 1.56 vs. MNutrInform=6.03;
 215 SD=0.84; $t(1.189)=-3.778$; $p<0.001$) and trust (MNutri-Score=4.94; SD=1.47 vs.
 216 MNutrInform=5.58; SD=0.90; $t(1.189)=-3.410$; $p<0.001$). Similarly, the behavioral intention
 217 varies significantly between the two labels (MNutri-Score=5.30; SD=1.47 vs.
 218 MNutrInform=5.89; SD=0,94; $t(1189)=-3.134$; $p<0.001$). Hence, NutrInform Battery showed

219 significant and positive mean differences in terms of Trust, Attitude and Behavioral Intention
 220 also when the two labels are cued with details referring to the computational information. By
 221 focusing on trust, we offered a possible analysis of the extent to which the presence, even if
 222 indirect, of algorithmic elements can bring changes to decision-making. In addition, looking
 223 at the results on the individual variables, there was evidence that the Nutri-Score was less
 224 effective than the NutrInform Battery on Attitude, Behavioral Intention, and Trust.
 225 Furthermore, the results indicated that respondents do not vary their Attitude toward the label,
 226 Trust and BI if the Front-of-Pack labels are combined with computational details.

227 **Table 4. Summary of results**

Study	Description
Studies 1 and 2	<ul style="list-style-type: none"> • H1: not rejected • H2: not rejected • H3: not rejected • H4: not rejected • H5: not rejected
Study 3	<ul style="list-style-type: none"> • H6: not rejected
Study 3.b	<ul style="list-style-type: none"> • H7: not rejected • H8: not rejected • H9: not rejected • H10: not rejected

228

229 **General Discussion**

230 The aim of our studies was to contribute to the FOPL debate with a new approach, oriented at
 231 understanding the role of consumer acceptance of labelling systems and their usefulness and
 232 ease of use in directing towards informed and healthier choices, as mediated by trust.

233 We thus expanded the TAM, a very flexible and tested model, to measure the efficacy of
 234 labelling schemes. As labels play a prominent role in informing the consumer decision-
 235 making process at the point of sale, it is important to understand not only which kind is more
 236 informative, but also which ones overthrow acceptance barriers and guide consumers towards
 237 healthier behaviours.

238 The FOPAM contributes to the current debate by proposing an alternative and complementary
 239 framework that provides additional information about acceptance and efficacy: this can be of
 240 significant advantage to policymaking, and to citizens as well. Building on the fact that when
 241 consumers feel trust, their purchase intention increases (Aydin et al., 2014), consumers may
 242 feel less trust towards a less detailed food label. Considering that some labels, as the Nutri-
 243 Score, are the result of algorithmic calculations, this study shows statistically significant
 244 differences in the effects of algorithms on consumer behavior and investigates the
 245 effectiveness of the mediating role of trust towards the label in these contexts.

246 Thus, the upgraded FOPAM allows for an alternative theoretical model to move forward in
 247 the EU debate to assess the relative superiority of different FOPLs and by preventing the
 248 possible introduction labels that are less trusted, and then less effective, from a customer
 249 acceptance point of view.

250 Our model suggests that the effectiveness of labels in affecting attitudes and Behavioral
 251 Intentions towards food purchase is predicted by their usefulness and ease of use: labelling
 252 systems support buying decisions and contribute to the definition of healthier choices when
 253 they are perceived as useful and easy to use. We have shown trust to be a significant factor in
 254 the BI of consumers towards food labels, and we have built on it a model capable of
 255 measuring the different performances of two divergent labels - the NutrInform Battery and the

256 Nutri-Score - leading to clear and incontrovertible results on the superiority of the NutrInform
257 Battery across all dimensions of the FOPAM.

258

259 **Conclusion**

260 Our research aims at contributing to the European Debate on Front-of-Pack labelling by
261 introducing acceptance and trust as metrics to evaluate label efficacy on improving people's
262 decision toward healthier food choices, through a systematic model that can provide
263 actionable insights to multiple stakeholders and to policymakers in particular

264 Yet, our study is not exempt from limitations. As it was decided to study Italian, French and
265 English respondents, future research should try to target individuals from other countries in
266 order to understand whether the results are generalizable in a broader way. Furthermore, our
267 research has given relevance to trust, but the model could be further integrated by considering
268 other variables, such as actual use versus perceptual use, user experience level or type of users
269 and type of use (King & He, 2006; Venkatesh, 2000). In addition to that, our work was
270 performed as a survey-based experiment: implementing a field test could provide a real-world
271 validation of the results and behavioral values.

272 We believe study such as ours should have the chance to encourage policy makers to promote
273 nutrition education among consumers, and can help identify the right way to do so to benefit
274 citizens.

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