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Consumer adoption of blockchain food traceability: Effects of innovation-adoption characteristics, expertise in food traceability and blockchain technology, and segmentation

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53 **Consumer adoption of blockchain food traceability: Effects of innovation-**
54 **adoption characteristics, expertise in food traceability and blockchain**
55 **technology, and segmentation**

56
57 **Structured Abstract:**

58
59 **Purpose**

60 This study explores the influence of the following factors on consumer adoption of blockchain food traceability
61 (BFT): innovation-adoption characteristics, segmentation, expertise in food traceability, expertise in blockchain
62 technology, food categorical preferences and perceived important features of BFT.

63
64 **Design/methodology/approach**

65 The data was collected via an online questionnaire with 1,401 participants in New Zealand. Exploratory factor
66 analysis, structural equation modelling and segmentation analysis were undertaken.

67
68 **Findings**

69 Consumer adoption of blockchain food traceability was significantly influenced by two innovation-adoption
70 characteristics – perceived incentives and perceived complexity, as well as their expertise in food traceability.
71 Two consumer segments were identified: Conservatives (48%) and Pioneers (52%). Significant differences were
72 found between these two segments in terms of gender, age, education, occupation, residential area and ethnicity.
73 Consumers are more willing to use BFT for purchasing fresh, imported, staple and normal foods than for processed,
74 domestic and upscale foods. Their perceived important specific features of BFT are product origin, food safety
75 information, quality control, food safety information, hygienic condition and scarcity management.

76
77 **Originality/value**

78 This study contributes knowledge to address the current knowledge gap regarding consumer adoption of
79 blockchain food traceability by using a large sample set. It is also the first study to recognise consumer segments
80 for BFT; to provide information about consumers' important socio-demographic characteristics, food categorical
81 preferences and perceived important features towards BFT; and to explore the influences of consumers'
82 innovation-adoption characteristics, expertise in food traceability and expertise in blockchain technology on their
83 adoption of blockchain food traceability.

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85
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87
88 **Abstract**

89 **Keywords**

90 Consumer; blockchain; food traceability; innovation-adoption characteristics.

91

92 **1. Introduction**

93 Rising consumer demand for safety-assured food products has intensified consumers’ desire to
94 know where their food comes from, including its components and processing history (Akram
95 et al., 2020; Sander et al., 2018). The COVID-19 epidemic has hastened this trend due to the
96 food security and safety issues impacts on food industry labour (Attwood and Hajat, 2020;
97 Galanakis, 2020). A credible traceability system must be able to trace and track food and
98 ingredients through all stages of production, processing and distribution within a supply chain
99 (Van Rijswijk and Frewer, 2008). However, traditional food traceability systems (e.g.
100 alphanumeric codes, barcodes, RFID tags) are problematic: (1) they only consider
101 distribution and warehousing steps; (2) they do not consider if the information shared by supply
102 chain members can be trusted; and (3) they are monopolistic, asymmetric and opaque
103 information systems (Badia-Melis et al., 2015; Tian, 2016). These problems result in further
104 challenges, such as fraud, corruption, tampering and falsification of information (Tian, 2016).

105 In a blockchain-based system, information is supervised by every chain participant,
106 rather than an information centre, thus creating decentralised, collectively maintained and
107 reliable databases (Kamble et al., 2020; Saurabh and Dey, 2021; Tian, 2016). A blockchain-
108 based food traceability system could therefore increase the transparency of a food chain,
109 strengthen information credibility, realise real-time tracking of food products and enhance the
110 assurance of safety along the food chain (Tian, 2016).

111 Research on blockchain food traceability (BFT) is still embryonic with few academic
112 research papers. Most existing studies on the topic are either conceptual or reviews. Few
113 empirical studies include consumer analysis of blockchain technology in association with food
114 traceability. The limited findings show that consumers are unfamiliar with BFT globally, and
115 their attitudes, trust, perceived behaviour control, perceived security and privacy and received
116 extra information about BFT have significantly influenced their adoption of the technology
117 (Cao et al., 2021; Dionysis et al., 2022; Garaus and Treiblmaier, 2021; Kumar et al., 2022;
118 Mazzù et al., 2021; Rainero and Modarelli, 2021; Sander et al., 2018; Shew et al., 2022). There
119 remains a lack of understanding of consumers’ perceptions, choice motives, psychological
120 factors, segmentation, categorical preferences, innovation-adoption characteristics (IACs) and
121 behaviours regarding BFT. There is no literature providing empirical evidence to support
122 actions and policies to enhance BFT adoption.

123 The results of this study will contribute to filling the knowledge gaps concerning
124 consumer adoption of BFT. First, the study presents the theoretical background and research
125 framework shaping the analysis of the influence of the following factors on consumer adoption
126 of BFT: IACs, segmentation, expertise in food traceability (FT), expertise in blockchain
127 technology (BT), food categorical preferences and perceived important features of BFT. Next,
128 the framework is empirically assessed by using exploratory factor analysis, structural
129 equational modelling and segmentation analysis. Last but not the least, the study discusses the
130 academic and practical implications derived from this preliminary and more comprehensive
131 picture of BFT from the perspective of consumers.

132
133 **2. Theoretical background and research objectives**

134
135 Figure 1 shows the research framework, which contextualises the research approaches used in
136 this study. The following subsections introduce the theoretical and empirical background of
137 this research framework.

138
139 >>>>>>>>> Insert Figure 1

142 *2.1. Innovation-adoption characteristics*

143 Hansen (2005) developed a conceptual model with five IAC factors that drive consumers'
144 adoption of innovative products and services: perceived subjective norms (i.e. peer pressure on
145 personal adoption of the innovative products/services); perceived compatibility of the
146 innovative products/services with personal lifestyle and values; perceived complexity of using
147 the innovative products/services; perceived risk regarding the innovative products/services;
148 and perceived relative advantage (i.e. perceived superiority of the innovative products/services
149 over relevant old products/services).

150 However, there is a lack of empirical studies to test if this theoretical five factorial IAC
151 model fits well with real consumer samples regarding the adoption of innovative products and
152 services. A few recent empirical studies have recognised this original model does not fit well
153 with the real consumer samples regarding the adoption of e-commerce food shopping by using
154 factor analysis (Wang and Scrimgeour, 2022; Wang and Somogyi, 2018). These studies have
155 also found a more suitable IAC model for real consumers. This adjusted model has a three-
156 factorial construct for IAC including perceived risk, perceived complexity and perceived
157 incentives (i.e. perceived incentives what drove consumers to adopt the innovative
158 products/services, including original factors- perceived subjective norm, perceived relative
159 advantage and perceived compatibility) (Wang and Scrimgeour, 2022; Wang and Somogyi,
160 2018). Hence, the current study will use factor analysis to explore if an adjusted IAC model
161 can be utilised to identify the adoption of BFT from real consumer samples.

162 Prior studies have confirmed the significant effects of one or several IAC factors on
163 consumer adoption of innovative products/services such as business-to-consumer food
164 shopping, online-to-offline food delivery services and autonomous vehicles (Alaimo et al.,
165 2020; Roh and Park, 2019; Wang and Scrimgeour, 2022; Wang and Somogyi, 2018; Yuen et
166 al., 2020). BFT is an innovative and new service to normal consumers. There is still a lack of
167 studies exploring the effects of IACs on consumer adoption of BFT. Hence, this study builds
168 on the prior work by focusing on an unexplored area: the impact of IACs on consumer adoption
169 of BFT. Two hypotheses (H) were developed as follows:

170 **H1a.** Consumers' IACs have significant effects on their attitudes towards BFT.

171 **H1b.** Consumers' IACs have significant effects on their purchase intentions towards BFT.

172

173 *2.2. Expertise in blockchain technology and food traceability*

174 Consumers' expertise is distinguished from their previous product/service-related experiences
175 (Alba and Hutchinson, 1987). Previous studies indicate that consumer expertise pertaining to
176 a product/service has a significant effect on their current or future choice of it (including food
177 traceability systems) (Hu et al., 2021; Lanseng and Sivertsen, 2019; Wang et al., 2017).

178 Given BFT is an innovative product/service based on two categories of existed
179 technologies – food traceability and blockchain technology, consumer expertise in these two
180 technologies may affect the adoption of BFT. The current study thus explores the lack of
181 understanding of the impact of consumers' FT and BT on their adoption of BFT. The current
182 study thus explores the impacts of consumers' FT and BT on their adoption of BFT. Four
183 hypotheses were developed as follows:

184 **H2a.** Consumers' FT have significant effects on their attitudes towards BFT.

185 **H2b.** Consumers' FT have significant effects on their purchase intentions towards BFT.

186 **H3a.** Consumers' BT have significant effects on their attitudes towards BFT.

187 **H3b.** Consumers' BT have significant effects on their purchase intentions towards BFT.

188

189 *2.3. Consumer adoption*

190 Consumer adoption of a product/service is based on a two-stage procedure including their
191 attitudes and consumption behaviours (purchase intention or real purchase) (Bredahl, 2001;

192 Cho et al., 2019; Wang and Scrimgeour, 2022; Wang and Somogyi, 2018). Attitudes are
193 consumers' general evaluation (positive or negative) towards a product/service (Bredahl, 2001;
194 Cho et al., 2019; Wang and Somogyi, 2018; Wang and Scrimgeour, 2022).

195 Previous studies indicate that consumer attitudes have a strongly positive impact on
196 purchase intention/real purchase of a product/service, and are a significant mediator between
197 the purchase intention/real purchase and other significant factors related to the adoption of the
198 product/service (e.g. IACs) (Bartsch et al., 2016; Lim et al., 2017; Wang and Scrimgeour, 2022;
199 Wang and Somogyi, 2018). Hence, this study explores the effect of consumer attitudes on their
200 purchase intention towards BFT and the mediation effects of consumer attitudes between BFT
201 purchase intention and their IACs and FT and BT. Four hypotheses are developed as follows:
202 **H4.** Consumers' attitudes towards BFT have significant effects on their purchase intentions
203 towards BFT.

204 **H5a.** Consumers' attitudes towards BFT have significant mediation effects between their IACs
205 and purchase intentions towards BFT.

206 **H5b.** Consumers' attitudes towards BFT have significant mediation effects between their FT
207 and purchase intentions towards BFT.

208 **H5c.** Consumers' attitudes towards BFT have significant mediation effects between their BT
209 and purchase intentions towards BFT.

210

211 *2.4. Consumer segmentation and profiling*

212 Segmentation analysis is increasingly being conducted in consumer and marketing studies,
213 providing a deeper look at consumer segments and their profiling characteristics (e.g. socio-
214 demographics) towards a product/service (Pallant et al., 2020; Wang & Scrimgeour, 2022), but
215 as far as we know, there has been no empirical study of consumer segmentation and BFT. This
216 study will thus contribute knowledge to filling this gap, by exploring and profiling consumer
217 segments related to BFT.

218 In this study, segmentation variables include those IACs, FT and BT that have
219 significant effects on consumer adoption of BFT and consumers attitudes and purchase
220 intentions towards BFT. The identified segments will be profiled using consumers' socio-
221 demographic characteristics.

222

223 *2.5. Categorical preferences*

224 Scholars have explored consumers categorical preferences when purchasing specific food
225 products by using innovative services (e.g. e-commerce food shopping) (Wang & Somogyi,
226 2018; Wang et al., 2020). Consumers have obviously categorical preferences for different
227 innovative food shopping services (e.g. consumers prefer snack for business-to-consumer food
228 shopping, and restaurant meals for online-to-offline food shopping to other specific food
229 categories) (Wang & Somogyi, 2018; Wang et al., 2020). However, as far as we know, there
230 is not academic empirical study to explore consumers' categorical preferences for FT and BFT.
231 As such, this study will fill this knowledge gap and explore consumer categorical preferences
232 for BFT based on their purchase intentions towards eighteen food categories.

233

234 *2.6. Perceived importance for specific features*

235 A few studies have examined consumers' perceived importance of specific features of FT.
236 These studies recognised most and least important features attached to FT by consumers such
237 as origin, quality control, verifying label information, while attaching less importance to food
238 traceability in allocating liability, manufacturing process information, product sustainability,
239 food scarcity management, manufacturing process information, quality information, fight food
240 counterfeiting and specific information for 'at risk' people (Liu et al., 2018; Myae et al., 2011;
241 Rodriguez-Salvador & Dopico, 2020). However, to our knowledge, there are no relevant

242 empirical consumer studies of BFT adoption. Hence, this study will explore consumers'
243 perceived importance of fifteen specific features relevant to BFT adoption.

244

245 **3. Methods and materials**

246 *3.1. Participants and procedure*

247 The data for this study was collected using an online survey in New Zealand. The questionnaire
248 was developed in English. Before the survey, participants were asked to read a description
249 about FT and BFT summarised from previous studies (Kamble et al., 2020; Saurabh and Dey,
250 2021; Tian, 2016).

251 A pre-test was conducted with 30 New Zealand residents in February 2021. The
252 language expression and measurement design of the questionnaire were adapted based on the
253 pre-test results and participants' feedback. The final questionnaire was distributed among New
254 Zealand-based registered members of the consumer sample panel owned by an internationally
255 reputed research agency from March to April 2021. A total of 1,401 participants completed the
256 survey, and were representative in distribution for age, gender, residential area and ethnicity in
257 New Zealand (population older than 18). Table 1 presents the socio-demographic
258 characteristics of the total sample, including residential area, marital status, age, household size,
259 annual household income, education, occupation, gender and ethnicity.

260

261 >>>>>>>>>> Insert Table 1

262

263 *3.2. Measures*

264 Table 2 indicates the measures and items involved in this study. Participants' IACs towards
265 BFT were measured by fourteen items within the original five factorial constructs
266 conceptualised by Hansen (2005). The measurement items were adjusted into the text
267 expressions suitable for BFT based on a literature review related to (consumer behaviour
268 towards) food traceability, blockchain technology and BFT (Asfarian et al., 2020; Kim and
269 Woo, 2016; Mao et al., 2018; Nie and Luo, 2019; Yeh et al., 2019; Yuan, Wang et al., 2020;
270 Zhao et al., 2019). A seven-point Likert agreement scale was employed, with response
271 categories from 1 = totally disagree to 7 = totally agree (Wang and Scrimgeour, 2022; Wang
272 and Somogyi, 2018).

273 Participants' FT and BT were measured by three items each, also using the seven-point
274 Likert agreement scale mentioned above. The measures were developed from a previous study
275 by Yuan et al. (2020) that explored consumers' FT.

276 Participants' attitudes towards BFT were measured using three items with seven-point
277 semantic differential scales of bipolar adjectives from 1= unhappy/terrible/bad to 7= happy/
278 delightful/good. This design has frequently been used in previous studies to examine
279 consumers' attitudes towards a product/service (e.g. Wang and Scrimgeour, 2022; Wang and
280 Somogyi, 2018).

281 Participants' purchase intentions towards BFT were measured by three items with the
282 seven-point Likert agreement scale. The measurement items were developed from previous
283 studies that examined consumers' purchase intentions towards BFT and food traceability (Kim
284 and Woo, 2016; Nie and Luo, 2019; Yeh et al., 2019; Yuan et al., 2020). Further, participants'
285 purchase intentions towards eighteen food categories with BFT were measured by a single item
286 with the seven-point Likert agreement scale. The food categories were developed from a
287 previous study by Wang and Somogyi (2018) that examined consumers' online purchase
288 intentions towards specific food categories.

289 Participants were asked to evaluate the importance of fifteen features for BFT on the
290 seven-point Likert agreement scale and the measurement design developed from Wang et al.
291 (2017), which explored consumers' perceived importance for beer consumption attributes with

292 questions such as: ‘*It is important that the blockchain-based food traceability system can be*
293 *used [the fifteen important features]*’. The fifteen items were adjusted from previous studies
294 that examined consumers’ perceived important features of food traceability (Liu et al., 2018;
295 Myae et al., 2011; Rodriguez-Salvador and Dopico, 2020).

296 >>>>>>>> Insert Table 2

298 299 3.3. Data analysis

300 Data were analysed using the IBM SPSS 27 and AMOS 27 software package. Figure 2 indicates
301 the data analysis procedure utilised in this study. First, exploratory factor analysis (with a
302 maximum likelihood estimation method with varimax rotation) was conducted to explore the
303 suitable factorial construct for consumers’ IACs towards BFT (Jones et al., 2002). Second, a
304 structural equation model (SEM) was built to associate consumers’ IACs, FT and BT with their
305 attitudes and purchase intentions towards BFT. Path analysis was conducted to identify the
306 significant IACs, FT and BT influencing consumer attitudes and purchase intentions towards
307 BFT (Collier, 2020). A bootstrap sample of 5,000 was used to examine the mediation effect of
308 attitudes on the relationships between the IACs/FT/BT and purchase intention towards BFT
309 (Collier, 2020). Third, two-step hierarchical cluster analysis was conducted to identify
310 consumer segments using the significant IACs, FT and BT for BFT adoption and the BFT
311 attitudes and purchase intentions as segmentation variables (Zhang and Zhao, 2019). Chi-
312 square tests and independent sample T-tests were employed to identify the significant
313 differences between the segments in socio-demographic characteristics (Verbeke and Viaene,
314 1999). Fourth, descriptive analysis (with mean values) were conducted to recognise the most
315 and least important features of BFT and favourite specific food categories with BFT (Myae et
316 al., 2011; Rodriguez-Salvador and Dopico, 2020; Wang and Somogyi, 2018).

317 >>>>>>>> Insert Figure 2

319 320 4. Results

321 4.1. Exploratory factor analysis

322 Table 3 presents the results of the exploratory factor analysis of an adjusted IAC factorial
323 construct for BFT. A new two factor construct was identified. The values of the standardised
324 factor loading for most of the items were within the acceptable limit – higher than 0.5 (Haszard
325 et al., 2013). Internal reliability of the two new IAC factors was acceptable, with all Cronbach’s
326 α values higher than 0.7 (Pieniak et al., 2009).

327 >>>>>>>> Insert Table 3

329 The items for *subjective norm*, *perceived compatibility* and *perceived relative*
330 *advantage* in the original IAC factorial construct loaded on a new factor in the adjusted factorial
331 construct. The three items for *perceived relative advantage* also cross-loaded on a single factor.
332 Previous studies on consumers’ IACs for e-commerce food shopping identified an IAC factor,
333 *perceived incentive*, that included the original IAC factors *subjective norm*, *perceived relative*
334 *advantage* and *perceived compatibility* (Wang and Scrimgeour, 2022; Wang and Somogyi,
335 2018). As such, we finally combined the items of *subjective norm*, *perceived compatibility* and
336 *perceived relative advantage* into a new factor we called ‘*perceived incentive*’ for BFT. One
337 item of the original *perceived compatibility* factor (PCT2) was removed from further analyses
338 due to a low value of the standardised factor loading (less than 0.5).

340 One original *perceived risk* item (PR3) loaded on a new factor with all three items from
341 the original *perceived complexity* factor. This factor was labelled ‘*perceived complexity*’ as

342 item PR3 ('with low understanding of the technology among ordinary consumers, it is difficult
343 for blockchain-based traceability systems to become popular in the real market') semantically
344 dealt with the similar topic related to the complexity of BFT.

345 The other two items of the original *perceived risk* factor (PR1 and PR2) did not load
346 well on any factor found in the exploratory factor analysis, with very low values for
347 standardised factor loading (less than 0.4) for all the factors. Therefore, these two items were
348 removed from further analyses.

349

350 4.2. Structural equation modelling

351 Figure 3 presents an SEM that associates consumers' IACs, FT and BT with their attitudes and
352 purchase intentions towards BFT, including six latent variables and twenty-three observed
353 variables. The observed variables for the latent variables FT, BT and attitudes and purchase
354 intentions towards BFT had good internal reliability, with high Cronbach α values: 0.879 for
355 the attitudes toward BFT, 0.815 for the purchase intentions towards BFT, 0.783 for BT and
356 0.835 for FT. There was no severe multi-collinearity among the independent variables in the
357 later path analysis, as correlation coefficients were below 0.8, as shown in Table 3 (Pieniak et
358 al., 2009).

359

360 >>>>>>>>> Insert Figure 3

361

362 Path analysis performed well based on the SEM, as the values for the goodness-of-fit
363 indices were within acceptance limits: lower than 0.08 for RMSEA and higher than 0.9 for CFI
364 (Pieniak et al., 2009). The SEM was modified by correlating the error variables among the
365 three observed variables – PRA1, PRA2 and PRA3. This was done because the values for
366 modification indices were much higher than the other modifications suggested by the path
367 analysis outputs, and the three measurement items for the observed variables from the original
368 IAC factor *perceived relative advantage* and referred to similar issues – that is, the relative
369 advantages related to BFT (Collier, 2020; Wang et al., 2015; Wu, 2009).

370

371 >>>>>>>>>>>>>>> Insert Table 4

372

373 Table 4 summarises the findings from the path analysis. Consumers' FT and *perceived*
374 *incentive* were significantly and positively linked to their *attitudes* and *purchase intentions*
375 *towards BFT*. While consumers' *perceived complexity* was significantly and negatively linked
376 to their *attitudes towards BFT*. Therefore, H1a and H1b are partially supported. H2a and H2b
377 are supported. In other words, those consumers who had more prior experience with food
378 traceability and perceived more incentives for BFT would have more positive attitudes and
379 stronger purchase intention for food shopping with BFT. Those consumers who considered
380 BFT as being relatively complicated would have negative attitudes towards food shopping with
381 it. Further, consumers' BT had no significant effects on either their attitudes or purchase
382 intention towards BFT. As such, H3a and H3b are not supported.

383 Consumers' attitudes had a significantly positive impact on their purchase intention
384 towards BFT. H4 is supported. However, as shown in Table 4, it was not significant for the
385 mediation effects of attitudes on the relationships between IACs/FT/BT and purchase intention
386 towards BFT. H5a, H5b and H5c are not supported. In other words, those IACs/FT/BT that
387 had significant effects on BFT attitudes would have no significant indirect effects on BFT
388 purchase intention through attitudes.

389

390

391

491 reliability in the measures for *perceived risk* and the items do not load well on the factors in
492 the IAC factorial construct for BFT in this study.

493 This study explored the influence of consumers' FT and BT on their adoption of BFT.
494 Interestingly, although the major innovation of BFT compared to traditional FT systems is to
495 decentralise databases of food traceability using BT, consumers' BT has no significant
496 influence on their adoption of BFT. By contrast, consumers' FT has a significantly positive
497 influence on both their attitudes and purchase intentions for shopping food with BFT. Shew et
498 al. (2021) also had similar findings: consumers place little or no additional value in blockchain
499 technology-specific labels for food purchases compared to traditional food traceability labels
500 authorised by government departments without any technological claims.

501 This study identified two consumer segments: *Pioneers* and *Conservatives*.
502 Consumers in the *Pioneers* segment have much more positive general attitudes, perceived
503 incentives and stronger purchase intentions towards BFT, are more experienced in FT and are
504 less likely to consider BFT as complicated compared to their counterparts in the *Conservatives*
505 segment. The BFT *Pioneers* are more likely to be male, young, of high educational level, have
506 a high- or medium-level position and live in a big city than BFT *Conservatives*. This is in line
507 with previous findings that young people, males, people with a high level of occupation and
508 education and/or people lived in big cities are more likely to become pioneers in accepting food
509 innovation (e.g. e-commerce food shopping and cultured meat) (Barrena et al., 2015; Bryant &
510 Barnett, 2018; Wang & Scrimgeour, 2022; Wang & Somogyi, 2018). The BFT *Pioneers*
511 segment also contains a higher percentage of consumers of Chinese and Indian ethnicity than
512 the BFT *Conservatives* segment. This might be caused by the Asian demographic in New
513 Zealand having a higher educational level on average, than other ethnicities in the country
514 (Meehan et al., 2019).

515 Regarding the specific features of BFT that are perceived as important, consumers
516 attach more importance to product origin, food safety information, quality control, food safety
517 information, hygienic condition and scarcity management. While they attach less importance
518 to fighting food counterfeiting, manufacturing process information, product sustainability
519 allocating liability and healthy food selection. Our findings are partly in line with previous
520 findings regarding consumers' perceived important features of food traceability. Rodriguez-
521 Salvador and Dopico (2020) suggested that consumers attach more importance to food
522 traceability in terms of product origin, quality control and verifying label information, while
523 attaching less importance to food traceability in allocating liability, manufacturing process
524 information and product sustainability. Myae et al. (2011) pointed out that consumers attach
525 more importance to food scarcity management, manufacturing process information, quality
526 information, fight food counterfeiting and specific information for 'at risk' people.

527 According to the food categorical preferences for BFT, consumers are more willing
528 to use BFT for purchasing fresh, imported, staple and normal foods than for processed,
529 domestic and upscale foods. This corresponds with the reality and tendency of food
530 consumption indicated by previous studies. Consumers are increasingly worried about the
531 safety of imported food, particularly during the COVID-19 pandemic (Armstrong & Reynolds,
532 2020; Ben Hassen et al., 2020). Fresh food products are more difficult to transport, store and
533 trace than processed and bulky food products, which have more widely established safety and
534 quality control standards and systems (Golan et al., 2004; Unnevehr, 2000). Consumers eat
535 staple and normal foods every day: much more frequently than the luxury and upscale foods
536 that they often consume on special occasions (Hastorf, 2003; Wargenau & Che, 2006).
537 Consumers may therefore care more about the safety and quality of and have a stronger
538 willingness to trace (using BFT) fresh, imported, staple and normal foods than processed,
539 domestic and upscale foods.

540

6. Conclusion

This study contributes to filling an important research gap in the understanding of consumer adoption of BFT using a large sample representative for New Zealand. As far as we know, this is the first study to identify consumer segments for BFT; to provide information about consumers' important socio-demographic characteristics, food categorical preferences and perceived important features towards BFT; and to explore the influences of consumers' IACs, FT and BT on their adoption of BFT. These findings are exploratory in nature but do provide foundational methods and insights that should inform ongoing research into the adoption of BFT.

The findings have vital managerial and policy implications. They highlight the need for consumer education regarding BFT. Further, this education needs to be focused on the different segments. These results will help stakeholders in the food industry to develop effective promotion strategies and policies for BFT, allowing stakeholders to clearly identify different consumer segments for BFT based on socio-demographic profiles: *Conservatives* (e.g. a low level of education and occupation, female, older and living in small cities) and *Pioneers* (e.g. a high level of education and occupation, male, younger and living in big cities). Consumer education that positively impact the "Conservatives" will need to be different to that which meets the needs of "Pioneers". BFT promotion strategies and policies designed specifically for the different consumer segments based on their different IACs, food categorical preferences and perceived important features of BFT have the potential to enhance the rate of BFT adoption. In addition, citizens should also be more broadly educated regarding food traceability, rather than blockchain technology, as governments and companies seek to socialise the benefits of BFT.

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Table 1 Socio-demographics of the sample in this study

Total sample (n=1401)			
Gender		Household size	
Male	47.7%	1	15.8%
Female	52.3%	2	35.8%
Age		3	18.9%
Mean value	45.42	4	16.7%
18- 30	25.1%	≥5	11.8%
31- 40	18.0%	Education	
41- 50	18.2%	Secondary school or below	27.3%
51- 60	15.3%	Trade training or equivalent	26.1%
61- 70	13.6%	Bachelor degree or equivalent	34.0%
≥70	9.8%	Master degree, equivalent or above	12.6%
Household income		Occupation	
0- \$50,000	31.7%	Managing employee	6.1%
\$50,001- \$100,000	33.3%	Salaried employee	32.8%
≥ \$100,000	35.0%	Student	5.9%
Marital status		Worker	13.4%
Married	45.8%	Self-employed	10.7%
No, but has a partner	22.3%	Other (Unemployed, Retired, Farmer, Housewife/houseman or On leave)	31.1%
Single	31.8%	Ethnicity	
Residential area		NZ European	62.5%
Auckland	32.8%	Maori	7.8%
Wellington	13.3%	Pacific Island	2.6%
Waikato	11.2%	European	5.8%
Canterbury	7.3%	Chinese	4.5%
Bay of Plenty	8.3%	Indian	5.0%
Otago	5.8%	Other ethnicities (Sri Lankan, South American, South African, Middle East, Filipino, Korean, Japanese etc.)	11.8%
Other regions (Whangarei, Wanganui, Tasman, Taranaki, Southland, Palmerston North, Northland, Nelson, Napier, Hawkes Bay, Manawatu, Marlborough, Gisborne etc.)	21.3%		

Note: \$= New Zealand dollars

Table 2 Measurement items in the study

Code	Factor and measurement items
	(Innovation-adoption characteristics)
SN	Subjective norm
SN1	My family members probably think that it is a good idea to track food information using blockchain-based traceability systems.
SN2	My friends and acquaintances probably think that tracking food information through blockchain-based traceability systems is a good idea.
PCL	Perceived complexity
PCL1	It will be complex to use the blockchain-based food traceability system.
PCL2	The blockchain-based food traceability system is hard to understand.
PCL3	It will be difficult to learn how to use the blockchain-based food traceability system.
PCT	Perceived compatibility
PCT1	Tracking food information by using blockchain-based traceability systems is attractive to me.
PCT2	In general, tracking food information using blockchain-based traceability systems is problem free.
PCT3	The blockchain-based food traceability system will be very useful in my life.
PRA	Perceived relative advantage
PRA1	The blockchain-based traceability system is favorable as it can eliminate fraud and false information in the food supply chain.
PRA2	Blockchain-based traceable food products are safety- and quality-assured.
PRA3	The blockchain-based food traceability system can help people to obtain trustworthy information about food products.
PR	Perceived risk
PR1	Using blockchain-based traceability systems will increase food production costs that may be passed on to consumers.
PR2	A risk with the high level of information transparency of blockchain-based food traceability systems is leaking my privacy.
PR3	With low understanding of the technology among ordinary consumers, it is difficult for blockchain-based traceability systems to become popular in the real market.
FT	Expertise towards food traceability
FT1	I understand standards related to traceable food.
FT2	I have a lot of experience in purchasing traceable food.
FT3	I am an expert in the purchase of traceable food.
BT	Expertise towards blockchain technology
BT1	I understand the rationale of blockchain technology.
BT2	I know more about blockchain technology than my friends.
BT3	I have more experience than my friends in purchasing or using blockchain-based products e.g. Bitcoin or other blockchain coins.
A	Attitudes towards blockchain food traceability (When I think about blockchain-based food traceability system, I feel ...)
A1	Unhappy/Happy
A2	Terrible/Delightful
A3	Bad/Good
PI	Purchase intentions towards blockchain food traceability
PI1	I will use blockchain-based food traceability tools in the future.
PI2	I am willing to buy blockchain-based traceable food products.
PI3	I will recommend blockchain-based food traceability tools to my friends.

Table 2 Continued

Factor and measurement items
Perceived important aspects for blockchain food traceability (It is important that the blockchain-based food traceability system can be used...)
To know the origin of the food product.
To know the quality of the food product.
To make sure that the food product is safe and risk free.
To make sure that the food product has been prepared, transported and stored in hygienic conditions.
To know if the food product has been maintained within quality control specifications.
To know if it is a sustainable food product.
To allocate liability in case the food product has deteriorated or is in a bad condition.
To know the information about every stage of the food manufacturing process.
To know all the ingredients of the food product.
To manage a food scare and to identify and recall products in bad condition.
To verify the information that appears on the food label.
To ensure authenticity of food products.
To fight food counterfeiting.
To provide confidence when selecting healthy foods.
To provide specific information for “at risk” individuals e.g. with weakened immune system.
Purchase intentions towards using blockchain food traceability systems for specific food categories (I am willing to buy blockchain-based traceable...)
Fresh meats
Processed meat products (e.g. canned meat, smoked meat, dried meat, sausage etc.)
Eggs
Dairy products
Fresh vegetables
Processed vegetable products (e.g. dried and pickled vegetables etc.)
Fresh fruits
Processed fruit products (e.g. dried and canned fruits etc.)
Staple food (e.g. rice and flour products etc.)
Fresh seafood
Processed seafood (e.g. canned seafood, smoked seafood, dried seafood etc.)
Soft drink
Alcoholic drink
Snack
Domestic food
Imported food
Normal food
Upscale food

Table 3. Outputs of the exploratory factor analysis and the correlation matrix of the independent variables in the path analysis based on the SEM (n=1401)

Factor and item	Standardized factor loading		Cronbach's α	
Perceived incentive	Code: IN		0.861	
PCT1		0.745		
SN1		0.721		
SN2		0.704		
PCT3		0.692		
PRA1		0.615		
PRA2		0.566		
PRA3		0.554		
Perceived complexity			0.775	
PCL3		0.733		
PCL2		0.681		
PR3		0.521		
PCL1		0.514		
Factor	1	2	3	4
1. BT	1			
2. FT	0.716	1		
3. PCT	-0.141	-0.040	1	
4. IN	-0.449	0.310	-0.260	1

Note: For the codes of measurement items, please see Table 2 and Table 3; Goodness-of-fit Test ($\chi^2=466.676, p=0.000$).

Table 4 Standardized regression weights of the path analysis the Bootstrap test for indirect effects based on the SEM:

Factor	Path	Factor	Total sample (n=1401)
Expertise towards blockchain technology	→	Attitude	ns
Expertise towards food traceability	→	Attitude	0.068*
Perceived complexity	→	Attitude	-0.089***
Perceived incentive	→	Attitude	0.819***
Expertise towards blockchain technology	→	Purchase intention	ns
Expertise towards food traceability	→	Purchase intention	0.135***
Perceived complexity	→	Purchase intention	ns
Perceived incentive	→	Purchase intention	0.820***
Attitude	→	Purchase intention	0.094*
Bootstrap test for indirect effects			Total sample (n=1401)
Expertise towards blockchain technology → Attitude → Purchase intention			ns
Expertise towards food traceability → Attitude → Purchase intention			ns
Perceived complexity → Attitude → Purchase intention			ns
Perceived incentive → Attitude → Purchase intention			ns

*Note: ***= $p < 0.001$, *= $p < 0.05$, ns = no significant; Goodness-of-fit indices: RMSEA=0.077, CFI=0.902, Chi-square=1995.001, Degrees of freedom=213, $p=0.0000$.*

Table 5 Outputs of the segmentation and profiling analyses

Segmentation variable	Segment 1	Segment 2
	Pioneers	Conservatives
Share of the total sample (n=1401)	52%	48%
Expertise towards food traceability***	4.09	2.86
Perceived complexity***	4.50	4.87
Perceived incentive***	5.54	4.21
Attitude***	5.32	3.93
Purchase intention***	5.51	3.89
Gender***		
Male	51.4%	43.7%
Female	48.6%	56.3%
Marital status		
Married	47.9%	43.5%
No, but has a partner	22.4%	22.3%
Single	29.7%	34.2%
Age*		
Mean value	44.35	46.58
Household Income		
0- \$50,000	29.9%	33.6%
\$50,001- \$100,000	32.6%	34.0%
≥ \$100,000	37.5%	32.4%
Education***		
Secondary school or below	23.1%	31.8%
Trade training or equivalent	25.0%	27.3%
Bachelor degree or equivalent	37.8%	30.0%
Master degree, equivalent or above	14.1%	10.8%
Household size		
1	14.8%	16.9%
2	34.1%	37.7%
3	20.2%	17.5%
4	17.6%	15.8%
≥5	13.3%	12.0%
Occupation***		
Managing employee	8.0%	4.2%
Salaried employee	35.9%	29.6%
Student	6.3%	5.3%
Worker	12.4%	14.6%
Self-employed	10.9%	10.5%
Other	26.6%	35.8%
Residential area**		
Auckland	36.7%	28.7%
Wellington & Canterbury & Waikato	31.0%	32.5%
Otago & Bay of Plenty	14.3%	13.8%
Other regions	18.0%	25.0%
Ethnicity***		
NZ European	58.8%	66.6%
Maori	8.4%	7.1%
Pacific Island	3.0%	2.2%
European	5.9%	5.6%
Chinese	5.8%	3.1%
Indian	7.4%	2.4%
Other ethnicities	10.7%	12.9%

*Note: ***= $p < 0.001$, **= $p < 0.01$, *= $p < 0.05$ (Chi-square tests or Independent sample T-tests); \$= New Zealand dollars.*

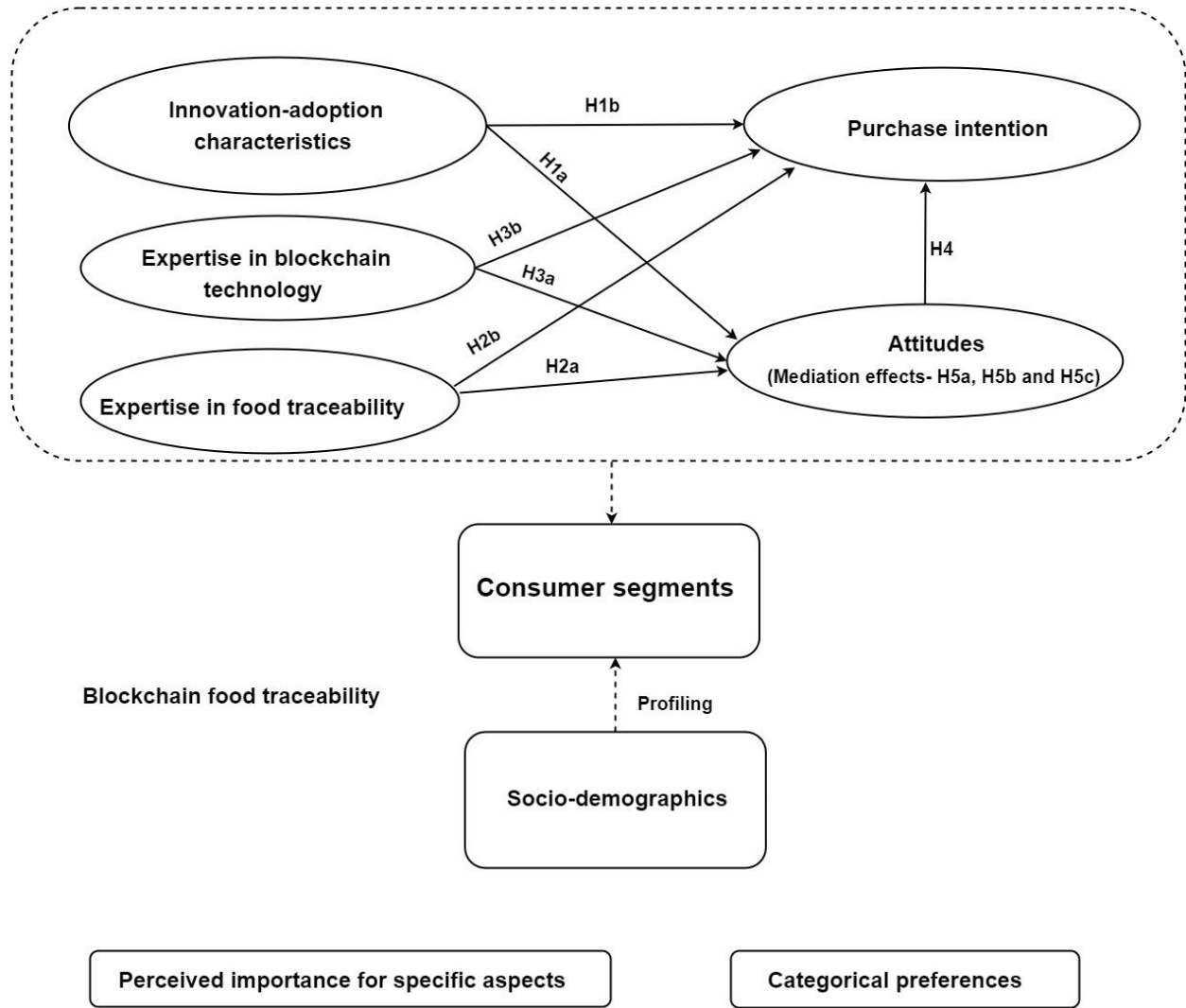


Figure 1 Research framework of the study

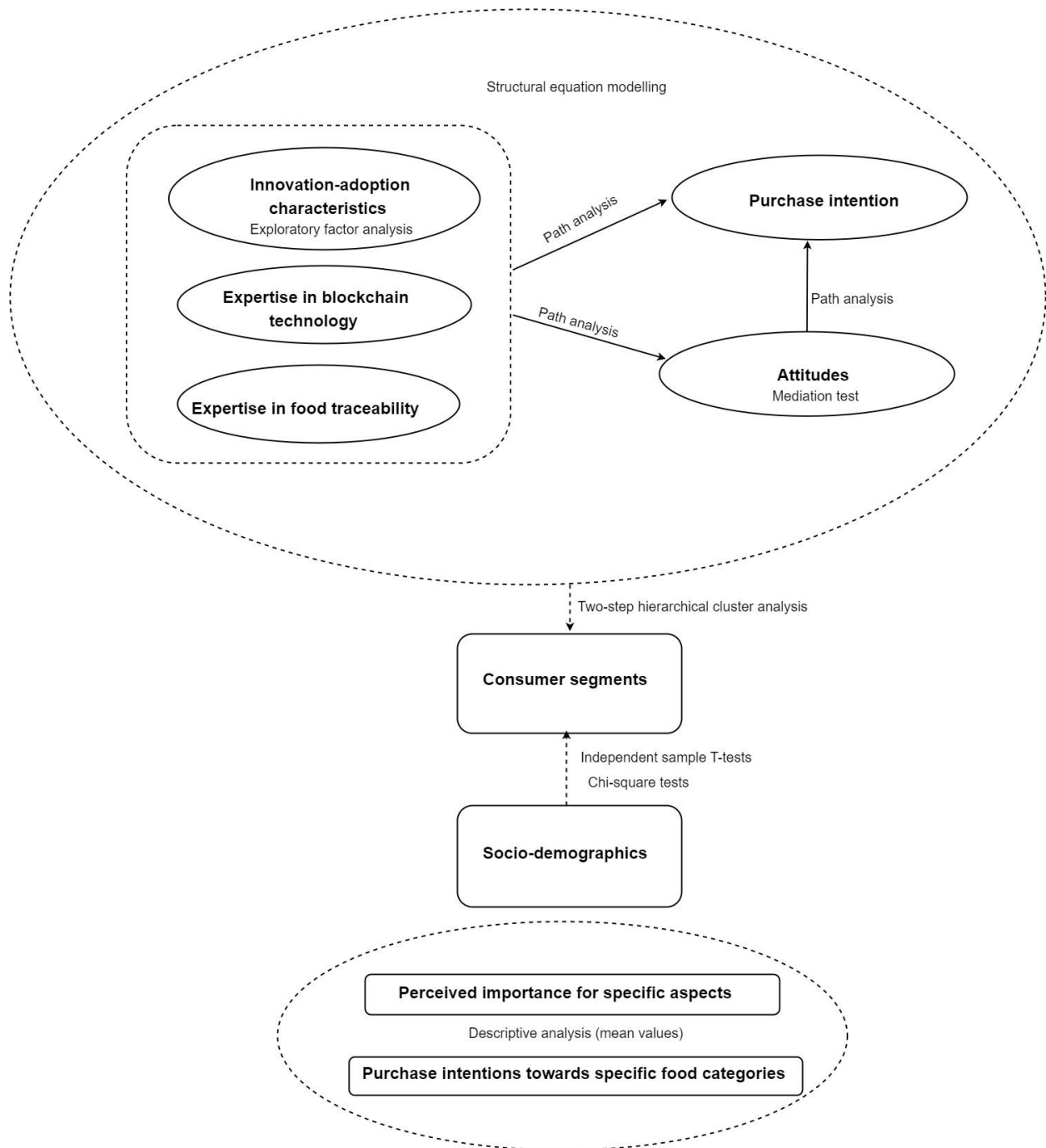


Figure 2 Procedure of data analysis

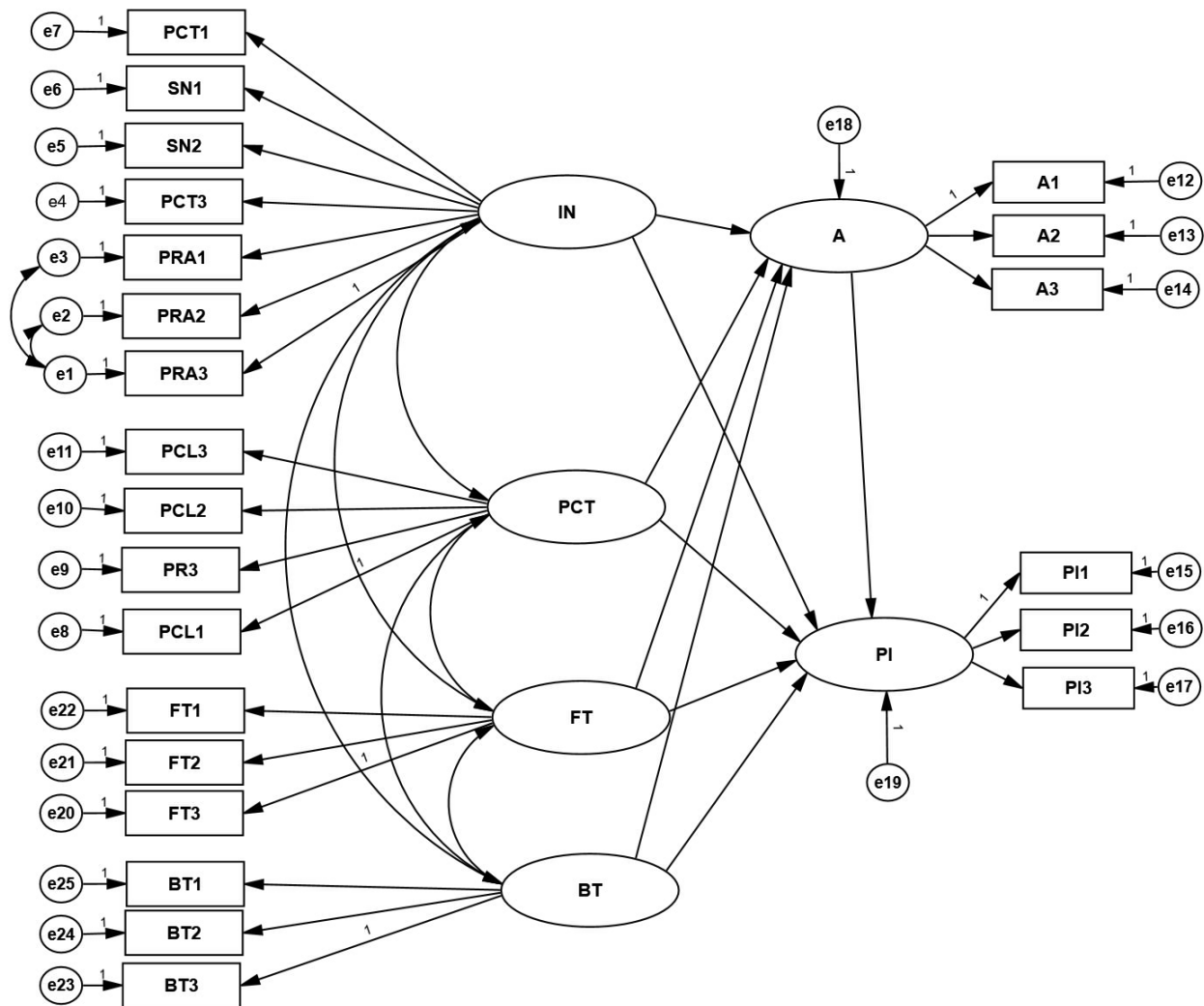


Figure 3 Structural equation model in the study

Note: For the codes of variables, please refer to Table 2 and Table 3; e1-e25: error variables.



Figure 4 Mean values of perceived important features of blockchain food traceability

Note: Sample size (n=1022), Missing values (n=379).

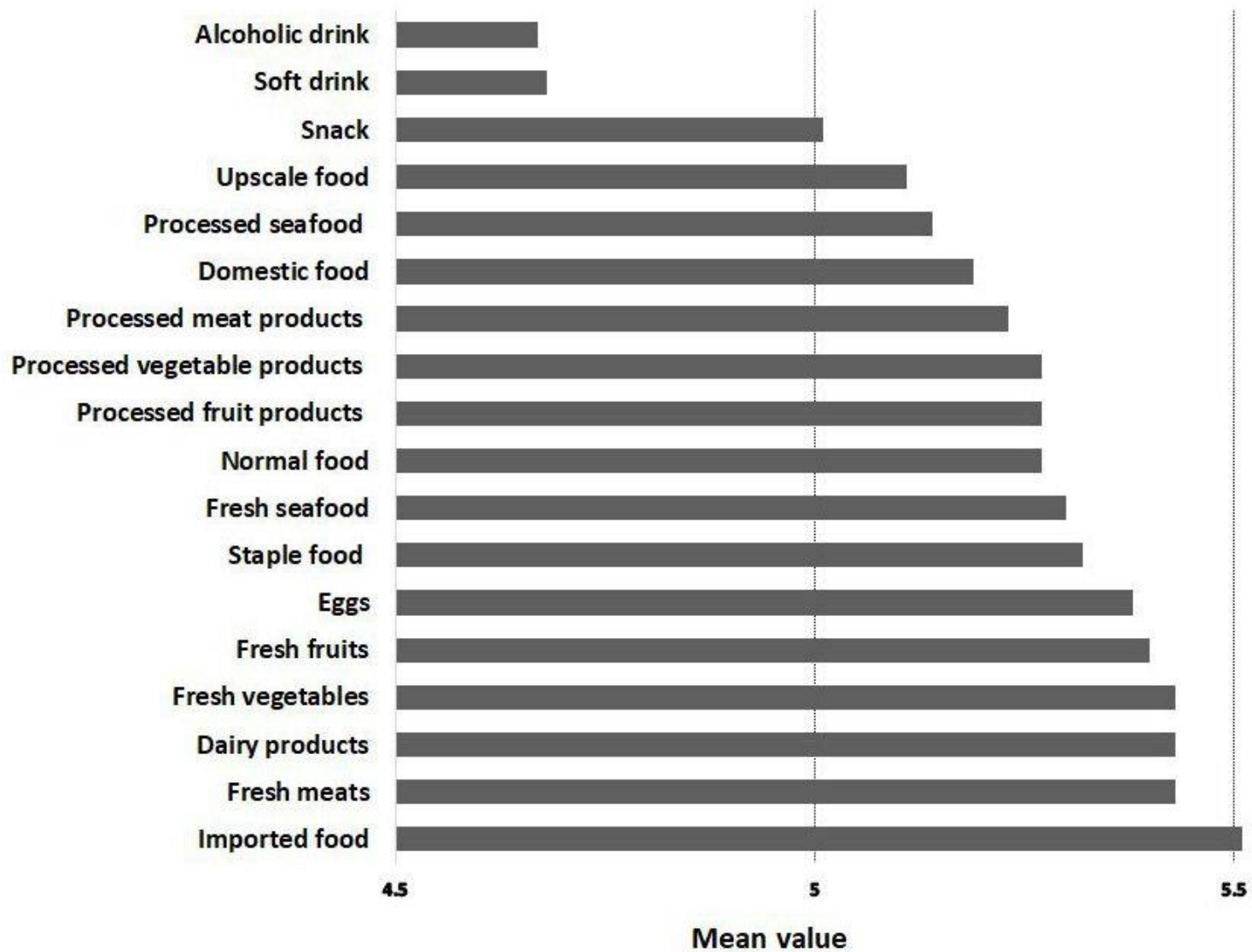


Figure 5 Mean values of purchase intentions towards using blockchain food traceability systems for specific food categories

Note: Sample size (n=1022), Missing values (n=379).