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Factors driving Thailand rice farmer decision-making in choice of marketing channel

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Abstract

1 **Purpose:** Since the end of the latest Rice Pledging Scheme, Thai rice farmers have had more

2 freedom in selecting marketing channels. Understanding the determinants of farmers' decision-

3 making associated with these channels is of particular interest to multiple stakeholders in the rice

4 value chain. This study aims to examine how economic, relational, and psychological factors

5 concurrently underpin Thai rice farmers' decision making and influence their marketing channel

6 choice.

7 **Design/methodology/approach**: Drawing on the Theory of Reasoned Action and utility

8 maximization of farmers' decision making, this study used structural equation modelling to

9 examine data collected from a nationwide sample of Thai rice farmers (n=637), focusing on their

10 past and intentional use of the three major marketing channels for paddy rice.

11 **Findings:** The determinants identified include four direct independent variables: attitude,

12 subjective norm (social referents), transaction conditions and economic goals, and two indirect

13 independent variables: past behavior and trust. Multi-group analysis suggests that rice co-

14 operative users were more empowered to consider economic goals and attitude towards the

15 channel, whilst rice miller and local collector users were more likely to be influenced by their

16 social referents and the transaction conditions offered by the channel.

17 **Originality:** Our study makes a unique and substantive contribution to the knowledge of

18 farmers' decision-making about marketing channel choice in Thailand and theoretically the

19 indirect role of past behavior in predicting prospective intention.

20 **Practical implications**: The findings highlight the need for policy to address trust and

21 transparency issues with intermediaries and to empower farmers through improvement of market

22 access.

1. Introduction

Market participation of farmers has been seen as a fundamental part of rural development in developing countries. Marketing channels play a critical role in linking farmers to markets (Fischer and Qaim, 2014). There are different types of market channels for primary producers ranging from informal markets through several business model iterations to formal trading channels such as contract farming or trading through a range of intermediaries (Shepherd, 2007). There is a growing academic interest in farmers' marketing channel use. Within this literature, most tend to focus on high-value products (Tsourgiannis et al., 2008; Milford, 2014), and/or modern channels such as collective sales (Fischer and Qaim, 2014; Zhang et al., 2017), export (Stanton and Burkink, 2008; Arinloye et al., 2015) or contract farming arrangements (Schipmann and Qaim, 2011; Barrett et al, 2012). As suggested by Poole (2017), domestic markets for staple grain crops, characteristically produced by emerging and semi-subsistence farmers in developing countries, contribute more to broad-based rural development due to the scale of such farmers.

Some recent studies have considered marketing channel use and their efficiency associated with paddy and/or milled rice specifically in Tanzania (Mgale and Yunxuan, 2020); India (Kakati and Chakraborty, 2017; Kumar et al., 2019); Vietnam (Pham et al., 2019) and Indonesia (Yonida et al., 2020). Whilst direct selling to consumers or retailers has increasingly been used for milled rice (Kakati and Chakraborty, 2017; Kumar et al., 2019; Yonida et al., 2020), the main marketing channels highlighted for paddy rice were traditional channels such as local agents/collectors, rice millers and wholesale traders (Kakati and Chakraborty, 2017; Kumar et al., 2019; Pham et al., 2019; Mgale and Yunxuan, 2020), and modern farmer organizations (Pham et al., 2019; Mgale and Yunxuan, 2020).

Thailand is one of the main rice producers in the world (FAO, 2018) with 46% of total agricultural land dedicated for rice production (OAE, 2019). Nationally, 79.9% of rice farmers (3.5 million) are small-scale farmers with less than 3.2 hectares of land per household (OAE, 2019). The vast majority of small-scale farmers are located in the Northeast and North regions

49   (83.23% and 77.63% respectively) (OAE, 2019). This sector has experienced successive policy

50   interventions ranging from low-interest loans to rice farmers to fixed higher-than-market price

51   for paddy rice (Poapongsakorn and Pantakua, 2014; Ricks and Laiprakobsup, 2021). Four main

52   marketing channels for paddy rice have been used by rice farmers in Thailand (Srisompun,

53   2014). They are: agricultural cooperatives, local collectors (or middlemen), rice millers and

54   central paddy market. The interventions, most notably, a series of Rice Pledging Schemes (RPS),

55   were often associated with the promotion of particular market channels (Liese *et al.,* 2014).

56   Therefore, the share of any specific rice marketing channel would vary under different policy

57   intervention schemes (Poapongsakorn, 2010). For example, central paddy markets, established in

58   1980, had a market share of nearly 24% by 1997 (Isvilanonda, 2010) due to the fact that the

59   initial RPS was implemented by the Bank for Agriculture and Agricultural Cooperatives which

60   operated in central paddy markets. However, the RPS introduced in 2011 favoured rice millers

61   more (Ricks and Laiprakobsup, 2021). By 2013, the share of central paddy markets dropped by

62   near half to 12.59% (Srisompun, 2014). The RPS ended in 2014 (Ricks and Laiprakobsup,

63   2021). It is possible that the channel use may have changed again since then. Thai rice farmers

64   have had more freedom in selecting marketing channels, so understanding the determinants of

65   farmers' decision-making associated with these channels is of particular interest to multiple

66   stakeholders in the rice value chain.

67   In terms of the rationale behind the farmers choice of marketing channel, some studies

68   have taken a socio-economic perspective, focusing on characteristics such as level of education,

69   farm size, location, and social network (Abebe *et al.,* 2016; Pham *et al.,* 2019; Mgale and Yan,

70   2020), transaction cost analysis (Escobal and Cavero, 2012; Mgale and Yan, 2020), utility

71   maximization (Blandon *et al.,* 2010) or asset specificity (Pham *et al.,* 2019). Other studies

72   incorporate relationship dynamics between channel members such as power and trust

73   (Schipmann and Qaim, 2011; Abebe *et al.,* 2016; Mgale and Yunxuan, 2020).

74   Since Gasson's (1973) seminal study, farmers' goals and objectives have featured highly in

75   much empirical research on farmers' decision-making (Beedell and Rehman, 2000; Borges,

76   2015). In the context of economic decision-making, a framework relevant to farmers' goals is

77   expected utility maximization (Nuthall and Old, 2018). If the farmer acts purely as an economic

78   agent they would select a marketing channel by evaluating the expected utility or net benefits of

79   the channel (Blandon *et al.,* 2010; Arinloye *et al.,* 2015). Profit is one aspect of the expected

80   utility, but other aspects include incentives received from buyers (Arinloye *et al.,* 2015), and

81   payment mechanisms and grading (Blandon *et al.,* 2010).

82   The Theory of Reasoned Action (TRA) (Fishbein and Ajzen, 1975) and its later variant the

83   Theory of Planned Behavior (TPB) (Ajzen, 1991; 2011) are used widely to study farmers'

84   decision-making (Burton, 2004; Hansson *et al.,* 2012; Meijer *et al.,* 2015). However, its

85   application to farmers' marketing channel choice has generally been limited (Dunay *et al.,*

86   2018). Dunay *et al.* (2018) found that attitudes and subjective norms, key exogenous factors in

87   the TRA, along with goals and objectives, strongly influenced farmers' decision-making.

88   Considering that farmers do not always make purely economic or economically optimal

89   decisions (Howley, 2015; Howley *et al.,* 2015), we see the need to apply social-psychological

90   perspectives to studies associated to farmers' decision making. Our lens of enquiry is marketing

91   channel choice by rice farmers in Thailand. This study aims to examine how economic,

92   relational, and psychological factors concurrently underpin farmers' decision making and

93   influence marketing channel choice by Thai rice farmers. Our study makes a unique and

94   substantive contribution to existing knowledge of farmers' decision-making about marketing

95   channel choice in Thailand.

96   **2.   Theoretical Framing of the Study**

97   We propose an integrated framework (Fig. 1) illustrating the key decision factors influencing

98   farmers' marketing channel use behavior. As a central premise, TRA identifies that conscious

99   cognition is a causal agent in decision making and choice (Fishbein and Ajzen, 1975; Ajzen,

100 1991; 2011). According to Fishbein and Ajzen (1975), an individual's intention to perform a

101 behavior is an immediate antecedent of that prospective behavior and the intention is influenced

102 by that individual's attitude towards the behavior and subjective norms (social influences from

103 friends, family, experts, policy makers etc.). Our framework extends the TRA model by drawing

104 upon economic goals and transaction conditions to develop a more holistic overview of farmers'

105 decision-making regarding marketing channel choice.

106   **Take in Figure 1 here**

107 *2.1 Past behavior and intention*

108 It is important to note that there are three aspects of behavior: retrospective behavior, intention

109 and prospective behavior. According to Fishbein and Ajzen (2010), empirically, many studies

110 based on self-report surveys have assessed intentions and past (retrospective) behavior at the

111 same time. Based on the conclusions of several meta-analyses of the relationship between

112 intention, prospective behavior (Armitage and Conner, 2001) and/or retrospective behavior

113 (Albarracin *et al.,* 2001), Fishbein and Ajzen (2010) concluded that "intentions are found to

114 predict behavior quite well" (p.51) whilst acknowledging that the findings from the meta

115 analyses were inconsistent in that "intentions sometimes predict past behavior better than future

116 behavior but at other times predict future behavior better than past behavior" (p. 50). We argue

117 that current intention cannot be used as a predictor of past behavior because prediction is about

118 the future not the past and temporal precedence is one of the criteria for prediction (Kenny, 1979;

119 Hair *et al.,* 2013). Ajzen (1991; 2011) explicitly points out that past behavior does not constitute

120 a causal antecedent of intention due to its lack of regularity. We, therefore, propose that past

121 behavior exerts influence on intention indirectly, through the outcomes of the appraisal of the

122 channel used, typically including attitude towards and trust in the channel. Past behavior itself is

123 influenced by subjective norm, economic goal and transaction condition offered.

124 **2.2 Attitude and subjective norm**

125 Attitude is a latent disposition defined as the decision maker's favorable or unfavorable

126 evaluation of the performance of a particular behavior or as a response to a given behavior

127 (Fishbein and Ajzen, 1975; 2010). Commitment to a given marketing channel is empirically

128 linked to two evaluative attitudinal constructs: satisfaction, informed by past experience (Selnes,

129 1998; Schirmer *et al.,* 2018) and trust. Trust entails a channel member's belief in an exchange

130 partners' integrity or honesty (Morgan and Hunt, 1994). In line with the TRA framework which

131 sees attitude as the antecedent of behavior, we delineate causal paths from past behavior to trust

132 (H1a), trust to attitude (H1b), past behavior to attitude (H1c), and from attitude to channel use

133 intention (H1d) as shown in Figure 1.

134   Subjective norm is narrowly defined in the TRA framework as perceived social pressure to

135 perform (or not to perform) a particular behavior (Fishbein and Ajzen, 2010) that can inform

136 channel use intention. Social pressure to perform a given behavior can arise in the context of this

137 study from extension officers, neighbors and peer groups, and family members (Meijer *et al.,*

138 2015; van Dijk *et al.,* 2016). We therefore propose causal paths from subjective norm to past

139 behavior (H2a) and from subjective norm to intention (H2b) as shown in Figure 1.

2.2 Economic goal and transaction condition

141 We delineate two categories of utility maximization intrinsic to market exchange: economic

142 goals held by farmers and consideration of transaction conditions offered by the channel. These

143 can be achieving higher price, lower costs of selling or better cash flow, depending on individual

144 farmers' circumstances. Price per se is rarely the sole determining factor (Tsourgiannis *et al.,*

145 2012), particularly when the heterogeneity in socioeconomic characteristics of farmers and farm

146 is taken into consideration (Hansson *et al.,* 2012). Cost-focused farmers are more motivated by

147 reducing costs than achieving a higher price (Tsourgiannis *et al.,* 2012) or having a healthy cash

148 flow (Blandon *et al.,* 2010). Whilst economic goals are about what farmers want to achieve

149 through a transaction, the other side of the coin is the transaction conditions offered by the

150 channel (Shepherd, 2007). Indeed, farmers are concerned not only with the price offered but also

151   channel accessibility, mode and speed of payment, grading and standard, purchase volumes of

152   buying and other costs of selling such as transportation (Blandon *et al.,* 2010; Tsourgiannis *et al.,*

153   2008; 2012; Arinloye *et al.,* 2015). In summary, we propose causal paths from economic goal to

154   past behavior (H3a), from economic goal to intention (H3b), from transaction condition to past

155   behavior (H4a) and from transaction condition to intention (H4b) as shown in Figure 1.

3.   Materials and methods

3.1 Study area, sampling and data collection

158   This study focuses on understanding the factors that influence rice farmers' choice of marketing

159   channel. The study area covered three of the four geographical regions in Thailand: North,

160   Northeast and Central regions with a collective share of 98.5% of the total rice production in

161   Thailand (OAE, 2019). A two-phase sequential approach was adopted for data collection. Phase

162   one involved preliminary in-depth interviews conducted in 2015 with 33 rice farmers from three

163   provinces as shown in Figure 2a. The interviews aimed to explore marketing channel choice and

164   validate/inform the development of the measures of the key concepts as depicted in Figure 1.

165   The findings of the interviews were used to inform the phase-two survey in relation to

166   questionnaire design, sampling and the actual data collection process. Phase two cross-sectional

167   survey questionnaires were collected in person in 2016 from nine provinces as shown in Figure

168   2b and produced the main data for this study.

Take in Figure 2 here

170   To ensure a representative sample, the selection of the provinces and villages took into

171   consideration three main factors: the number of crops, farm size and rice varieties. For phase-one

172   interviews, a theoretical sampling was used to select one representative province for each region.

173   Two villages in each province were selected at the recommendation of the sub-district

174   administrative organization (SAO) in each region. Individual rice farmers were identified by the

175   Agricultural Extension Officers (AEO) based on the criteria provided by the researchers. For the

176 second stage survey, a three-step sampling procedure was adopted. Three representative

177 provinces in each region were firstly selected based on the agricultural census data for each

178 province. This is followed by selecting villages from the nine chosen provinces with the

179 assistance of the administration officers of the SAO and AEOs in each province, leading to the

180 identification of a total of 21 villages (four in the North, seven in the Northeast and ten in the

181 Central region). Finally, the respective SAO or AEO helped send requests for assistance to the

182 head of villages, or government officers attached to the village, who called an assembly in the

183 village and promoted the survey to the individual rice farmers.

184 Phase-one interviews were conducted face-to-face and fully recorded by the first author.

185 Phase two questionnaires were distributed and collected in person by the first author and three

186 assistants under the supervision of the first author. Prior to the data collection, the assistants were

187 all trained by the first author. A total of 661 valid questionnaires were collected, 24 of which

188 reported using channels which were excluded for detailed analysis due to small sizes of sub-

189 groups. Therefore, the main data analysis was based on responses from 637 rice farmers, selling

190 rice to any of the three main marketing channels: millers, local collectors and cooperatives. The

191 specific number of respondents from each province can be found in Figure 2b.

3.2 Questionnaire design and measures of analytical variables

193 The questionnaire was designed to collect information for analytical variables as specified in the

194 proposed model (Figure 1) and also relevant socio-demographic information. Particular attention

195 was paid to specific and precise wording. Findings from the preliminary in-depth interviews

196 were used to contextualize the measures for Thai rice farmers where appropriate. Measures for

197 the variables in the proposed model were also developed through synthesis of the scales

198 established from previous studies (Selnes, 1998; Fishbein and Ajzen, 2010; Hernández-

199 Espallardo *et al.,* 2012).

200 Two channel use behavioral variables were proposed in the conceptual framework: past

201 behavior and intention. Past behavior was measured by asking respondents to indicate how much

202   rice they sold to any of the three channels between February 2014 (end of last RPS) and June

203   2016 (when survey conducted). The scale ranged from "never", then "less than 10%" to "always

204   - over 90%" against each channel. Intention, defined as the likelihood of a farmer selling the

205   next rice crop to use any particular marketing channel, was measured using the statement "Next

206   crop, I intend to sell to this marketing channel" scaling from most unlikely (1) to most likely (7).

207   The influencing variables considered were attitude, trust, subjective norms and economic

208   goal and transaction conditions. All variables were measured with 7-point scale. For the variable

209   economic goal, respondents were asked to indicate the level of importance from 'not important at

210   all' (1) to 'extremely important' (7). All other variables were measured using Likert scale (from

211   1=strongly disagree to 7=strongly agree) against each relevant statement.

212   Attitude towards a particular channel used was seen as a latent predisposition as shown in

213   either a favorable or unfavorable manner (Fishbein and Ajzen, 2010). Two evaluative statements

214   (adapted from Hernández-Espallardo et al., 2012) were used to measure attitude: "This channel

215   is a good choice for me" and, "Overall, I am happy with this channel". Trust was measured using

216   two items in relation to honesty/integrity and reliability (Morgan and Hunt, 1994). The two

217   items were: "I choose this channel because I don't have to worry about being cheated on: 1) the

218   weighing scale and 2) rice quality grading assessment".

219   Subjective norms were measured against five normative referents: friends and/or

220   neighbors, family members, government officers, mass media and harvest machine drivers (all

221   validated through phase-one interviews). The statement used were adapted from Fishbein and

222   Ajzen (2010): "Most of my friends and neighbors sell their rice to this channel"; and "My family

223   member/Government officer/Rice harvest machine driver recommend that I should sell to this

224   channel".

225   The construct, economic goal, was measured with three items identified from preliminary

226   interviews and extant literature (Blandon et al., 2010; Tsourgiannis et al., 2012). The three goals

227   were: selling at a higher price, minimizing cost, and enhancing cash flow. Transaction conditions

228 offered by the channel were measured with items adapted from Blandon *et al.* (2010) and

229 Tsourgiannis *et al.* (2012) and contextualized after the preliminary interviews. Transaction

230 condition was conceptualized to include mode of payment, buying capacity, costs of selling and

231 accessibility offered by channels. The items for transaction conditions provided by a specific

232 channel included: cash payment, confidence in being paid, buying any quantity, easiness to

233 access, price offer and cost of transportation.

234   All measures of the variables in the proposed model were tabulated in the questionnaire

235 against each specified marketing channel previously used or where there was an intention to use.

236 Common method bias was checked by using Harman's single factor test. Constraining the

237 number of factors extracted to one, the total variance explained by all indicators of the

238 independent variables was 27.31%, which showed that common method bias was not an issue for

239 the observed items of the determinant factors.

3.3 Analytical procedure

241 Socio-demographic characteristics of the respondents were first summarized. Descriptives of the

242 analytical variables were explored. ANOVA test was used to compare the differences of socio-

243 demographic attributes across the sub-groups of different channel users.

244   The proposed model was tested based on the main survey data using covariance-based

245 structural equation modelling (SEM) with AMOS 26. SEM tests a series of regression equations

246 simultaneously, encompassing the modelling of correlated independents, measurement error,

247 multiple latent independent and dependent variables with single or multiple observed indicators,

248 path analysis and analysis of covariance (Blunch, 2013; Hair *et al.,* 2017). Maximum likelihood

249 estimation was used to infer the value of the unobserved, or latent variables. This method makes

250 use of full information or all data points available (Arbuckle, 2017). A two-step strategy

251 (Blunch, 2015) for SEM was adopted, followed by multi-group analysis.

252   The first step was a confirmatory factor analysis (CFA) or measurement modelling of the

253   observed and latent variables. Model fit was assessed using standard model fit indices. To assess

254   the model fit, the criteria provided by Hair *et al.* (2013) were adopted. The indices suggested by

255   Hair *et al.* (2013) vary slightly according to sample size (N) and number of measures or

256   indicators (m). They suggested that if N > 250 and 12 < m <30, the significant p-values for

257   likelihood ratio chi-square expected should be less than .05 (p < .05), comparative fit index (CFI)

258   should be greater than .92, room mean square residual (RMR) should be less than.08, and the

259   root mean square error of approximation (RMSEA) be less than .07. Minor modifications were

260   made according to the modification indices for the covariances produced by AMOS. As a result,

261   two indicators ('government officer' and 'rice harvest machine driver') for the latent variable

262   'subjective norm', and two for the latent variable 'transaction condition' ('Price offer' and 'cost

263   of transportation') were deleted.

264   Secondly, structural equation modelling was run based on the modified measurement

265   model and structure of the proposed relationship between the latent variables. For the two single-

266   indicator channel choice variables (i.e. past behavior and intention), Hayduk and Littvay's

267   (2012) approach was used to fix the measurement error variances of the two items. Therefore,

268   0.1 was assigned to the error of past behavior and 0.3 to the error of intention, assuming less

269   error for actioned practice than predictive actions. One modification was done to improve the

270   structural model fit. Details are provided in the results section. Thresholds for model fit

271   assessments (Hair *et al.,* 2013) can be found in Table 3.

272   Finally, multi-group analysis (MGA) for different user groups of rice marketing channels

273   was conducted based on the modified structural model. The MGA compared the differences of

274   the model structure and individual path coefficients (standardized regression weights) across the

275   subgroups.

276   **4. Results**

277   4.1 Socio-demographic characteristics of the respondents and marketing channel used

278   The socio-economic characteristics of the sample population of rice farmers across the three

279   regions in Thailand are presented in Table 1. Of the 637 rice farmers who sold rice to any of the

280   three main marketing channels (i.e. millers, local collectors or co-ops), 27 percent were (170

281   farmers) were from the North region, 46 percent (293 farmers) were from the Northeast and 27

282   percent (174 farmers) were from the Central region. There were 384 female farmers (60% of the

283   total). The average age of the respondents were 52 years. On average, the farmers received about

284   4 years of formal education. The average farm size was 8.7 Rais (1.4 ha) in the North region, 9.7

285   Rais (1.5 ha) in the Northeast and 26.2 Rais (4.2 ha) in the Central region. When compared with

286   the agricultural census conducted by the National Statistical Office, Thailand (OAE, 2019), the

287   sample is largely representative of rice farmers in Thailand in terms of education and farm size.

288   Females and older farmers were slightly over represented in this sample.

Take in Table 1 here

290   In terms of the marketing channel used for selling rice, 369 farmers (57.9% of the 637

291   respondents) sold rice to a miller, 120 farmers (18.8%) to a cooperative and 201 farmers (31.6%)

292   to a local collector. The majority of the farmers only sold rice to one channel (91.9%) and 49

293   respondents used two marketing channels and two used all three channels. For respondents who

294   selected more than one channel, their responses for each channel were treated separately. This

295   means the final sample for the SEM analysis was a pooled sample with a total of 690 channel-

296   specific responses.

297   Statistically significant differences in channel used were found when region, gender,

298   education and farm size were considered (Table 1). Chi-Square test of independence indicated

299   that millers and local collectors were used more by farmers in the Northeast region whilst

300   cooperatives were used more by those in the North region ($X^2 = 56.065$, $p < .001$). Female

301   farmers were more likely to use local collectors ($X^2 = 6.65$, $p = .036$). When compared by

302   education, those who had completed more than seven years of education were more likely to use

303   cooperatives ($X^2 = 29.06$, p <.001). Those who farmed more than 12 Rais (or 1.92 ha) of rice were

304   more likely to have used millers ($X^2 = 20.16$, p < .001).

305   4.2 Modelling results of determinants of paddy rice marketing channel use

306   SEM analysis involved confirmatory factor analysis (CFA) of the measurement model, structural

307   modelling and multi-group comparisons. The good model fit thresholds and indices were

308   explained in section 3.3 and also presented in Table 3.

309   The initial CFA was based on the original 19 observed indicators of the seven latent

310   variables of the proposed model as shown in Figure 1. The results showed poor model fit

311   (X^2=630.59 with df =126 and probability level =.00; X^2/df=5.01; CFI=.859; RMSEA=.078,

312   PCLOSE =.000, N=690). Based on the modification indices and regression weights, four items

313   with large modification indices were removed as explained in section 3.3. The modified CFA

314   model (Model 2) with the remaining 15 indicators was improved to a good fit (X^2=171.597 with

315   df=71 and probability level=.00; X^2/df=2.417; GFI=.967; CFI=.964; RMSEA=.045 and

316   PCLOSE=.804, N=690).

317   Convergent and discriminant validity of the five latent variables based on the modified

318   measurement model were then examined. The validity test results for all five latent variables

319   were shown in the last section of Table 2. The average variance extracted (AVE) is used to test

320   convergent validity. The recommended level is greater than 0.50. The AVE of attitude, trust and

321   subjective norm were above 0.5. Transaction condition and economic goals were below 0.5.

322   However, considering the exploratory nature of this study and other conditions being met, we

323   decided to retain the constructs. Discriminant validity is confirmed because all square root of

324   AVE (diagonals in the table) is greater than inter-construct correlations. Maximum shared

325   variances (MSVs) were all less than AVEs. Construct/composite reliability (CR) measures the

326   inherent consistency of the indicators of a construct. A CR coefficient of greater than 0.6 is

327   considered acceptable (Hair *et al.,* 2017). CRs of the five latent variables in the present study

328   ranged from 0.601 to 0.842.

329   **Take in Table 2 here**

330   Table 2 presents the descriptive statistics for the two behavioral variables (intention and

331   past behavior) and the five AMOS-imputed factor scores of the five determinant latent variables.

332   It can also be seen in Table 2 that statistically significant differences in the means value of

333   intention (p < .05), past behavior (p < .01), attitude (p < .10), trust (p < .01) and transaction

334   conditions (p < .0) were found across different channel user groups. Local collector users

335   reported the highest frequency of past use of this channel and highest level of intention to sell to

336   this channel whilst co-op users reported the lowest of both. Local collector users also had the

337   highest rating on attitude towards this channel, trust in this channel, and transaction condition

338   offered by the channel. Miller was the least trusted channel and the transaction condition of the

339   co-op was rated the lowest by their users. Subjective norm and economic goals showed no

340   statistically significant differences across the three channels (p = .26 and .71 respectively).

341   The CFA model fitting retained 15 observed items which were subjected to structural

342   equation modelling with AMOS. The proposed model (Model 1) was first tested, and the model

343   fit indices and results are presented in column 3 of Table 3. Model 1 had poor model fit indices

344   with none meeting the standard threshold (see column 2 of Table 3). Examination of the

345   modification indices suggested that a covariance should be added between the error terms of past

346   behavior (e14) and intention (e15) (M.I.=129.468). Adding the covariance between the two error

347   terms led to much improved model fit indices as shown in Model 2 (column 4 of Table 3) with

348   all model fit indices better than the thresholds shown in column 2. A Chi Square difference test

349   showed statistically significant difference between model 1 and model 2 (p <.0001). The

350   modified model with path coefficients is presented in Figure 3.

351   **Take in Table 3 here**

352   **Take in Figure 3 here**

353   All causal paths in model 2 apart from the one from H3b (economic goal to intention) were

354   statistically significant and the statistical estimates can be found in Table 4. The factors

355   identified in the model account for 48.2% of the variance of farmers' intentional channel use

356   (R^2=.482). Subjective norm, economic goal and transaction condition have statistically

357   significant and strong influence on past behavior (R^2=.75) with transaction condition being the

358   strongest influencing factor for past behavior (std. β = .476, $p < .001$).

359   **Take in Table 4 here**

360   Given the statistically significant differences found in intention, past behavior, attitude,

361   trust and transaction condition amongst the three marketing channels, it is important to conduct a

362   multiple group analyses (MGA) on the structural weights of model 2 based on channel used.

363   Table 4 presents the MGA results. This includes the structural weights coefficients (standardized

364   regression), the significance probability (p value) for each structural path and the results of

365   comparison of each individual path. The MGA showed statistically significant differences in

366   model structural weights amongst the three channels (X^2= 92.338; df=36; $p < .0001$). For rice

367   miller users, all but one hypothesized causal path were supported. The exception was H3b

368   (economic goal to intention). For local collector users, three hypothesized causal paths not

369   supported were: H1d (attitude to intention), H3a (economic goal to past behavior), and H3b

370   (economic goal to intention). For cooperative users, four causal paths were not supported. They

371   were: H1b (trust to attitude), H3a (economic goal to past behavior), H2b (subjective norm to

372   intention) and H4b (transaction condition to intention). Interestingly, cooperative users' intention

373   to continue using this channel seemed to be mainly motivated by economic goals, which was in

374   direct contrast with those selling to millers and local collectors whose intention was mainly

375   influenced by subjective norm and transaction condition offered by the channel.

5. Discussion

Thai rice farmers have mainly relied on rice millers, local collectors and cooperatives to take paddy rice to market. Over 90 percent of the farmers used one channel only. This is extremely high compared to the proportion (47.9 percent) found in Tanzania rice famers (Mgale and Yunxuan, 2020). Of the three channels, miller and local collector were the two most commonly used channels in all three regions. Bigger farmers were more likely to use millers and smaller farms more likely to use local collectors. Cooperatives were used more by those with higher level of education. Those findings are largely consistent with observations in other countries (Pham *et al.,* 2019; Mgale and Yunxuan, 2020). Aside from the socio-demographic characteristics of channel users, the hypothesized causal paths to the farmers' past channel use (i.e. past behavior) and intentional use were largely confirmed despite some nuances found amongst the users of the three channels.

The role of past behavior was an unresolved issue for TRA/TPB (Fishbein and Ajzen, 2010). Our data supported the hypothesized causal paths that past behavior was influenced by subjective norm and transaction condition across all three channel user groups. Past behavior was also influenced by economic goals for those selling to millers. As for the relative importance of the influencing factors, transaction condition and subjective contributed more to the past channel choice than economic goal. The proposed influence of past behavior on attitude either directly or indirectly via trust was also supported by the data. This enriches existing understanding of the role of past behavior in TRA/TPB framework and is worthy of application in other sectors and behavioral contexts.

Attitude was shown to have statistically significant, albeit weak, influence on intention of using millers and cooperatives, but not local collectors. Attitude was formed directly through past experience of the channel use (past behavior) and indirectly via trust through the construct of operational honesty in grading and weighing particularly for miller and local collector users. This is similar to the results of Mgale and Yan (2020) who found that farmers' trust in the

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402   channel affects their choice of millers and large-scale traders. It also partly concurs with

403   Schipmann and Qaim, (2011) who suggested that lack of trust in grading processes was one of

404   the reasons for farmers to withdraw from a given marketing channel. However, the trust-attitude

405   path was not supported by those who sold to cooperatives.

406   Subjective norm in the form of family/friends and neighbors was found to have

407   consistently influenced past behavior in all situations and influenced intention to use local

408   collectors and millers, but not cooperatives. This partly corroborates the findings of van Dijk *et*

409   *al.* (2016) that farmers tend to be influenced by their immediate social referents. Pham *et al.*

410   (2019) and Mgale and Yan (2020) both found that access to marketing information affects rice

411   farmers' channel choice in Vietnam and Tanzania. Friends and neighbors could be important

412   sources of market information for Thai rice farmers.

413   Farmers' expected utility maximization in the forms of economic goal and transaction

414   condition is of varied influence in this study. Economic goal to intention path was not supported

415   by miller and local collector users. This partially supports the findings of some previous studies

416   (Howley *et al.,* 2015; Abebe *et al.,* 2016) that famers' decision may not always follow a purely

417   economic rationale and that in developing countries smallholder farmers tend to trade via

418   middlemen even if the profit margin is low especially if there is a personal relationship with the

419   collector (Pham *et al.,* 2019). The level of activity required from the farmer in engaging with

420   these different marketing channels has also been shown to be of influence e.g. if the rice is

421   collected from the farm gate or if the farmer has to take the rice to the mill (Kakati and

422   Chakraborty, 2017), mediated in part by whether the farmers have access to personal transport

423   (Pham *et al.,* 2019). Mgale and Yunxian (2020) also echo this finding that most farmers in their

424   study still sell through local collectors. They cite distance to market and also inability to act

425   outside the farm gate. Also critical is the direct contrast with cooperative users who were perhaps

426   more empowered through collective action to consider their economic goals.

427   Transaction condition in the forms of cash payment, confidence of receiving payment,

428   accessibility and buying any quantity affects the decision of farmers to sell to rice millers and

429   local collectors. Our study reinforces that speed and mode of payment (cash payment in this

430   case) and buying capacity are generally big concerns for small-scale farmers (Blandon *et al.,*

431   2010; Barrett *et al.,* 2012). However, farmers who sold to cooperatives were not statistically

432   significantly influenced by transaction condition, perhaps because there were other factors of

433   more influence. Preliminary interviews suggested that this might have been due to the fact that

434   some farmers did not want to be tied up to cooperatives.

435   The findings of this study have some interesting managerial implications for farmers and

436   marketing channels. For rice farmers, only the cooperative users' intention was motivated by

437   economic goals, whilst for miller and local collector users, farmers were more motivated by

438   services and accessibility of the marketing channels (transaction condition) i.e. being paid in

439   cash. This seemed to suggest either that miller and local collector users lacked power to

440   negotiate and had to sacrifice higher economic return for market accessibility or that being paid

441   in cash had an advantage for them as individuals that took precedence over any negative aspects

442   of the transaction. It can be argued that to enhance their own economic status and profitability,

443   farmers need to take more collective actions as shown by cooperative users in this study.

444   Although considerations of perceived personal and collective economic benefit versus the

445   perceived loss of personal autonomy were not part of this study, Pham *et al.* (2019) highlighted

446   in their study in Vietnam that the farmers who were involved in more formal networks e.g.

447   farmers group had achieved better price for their paddy rice. Joining cooperatives may also help

448   smallholders to reduce transportation costs through collective action. For rice marketing

449   channels, the study shows the importance of past behavior in influencing farmers' trust and

450   attitudes, which then influence their future intention. The findings highlight the importance of

451   providing farmers with good services in an honest and transparent way. Mode and speed of

452   payment are just as important as flexibility of purchase quantity in keeping suppliers.

453   The findings have also important implications for policy makers in developing

454   interventions to safeguard rice farmers' welfare in selling their produce to markets. Firstly,

455   whilst some farmers have engaged in post-farm gate marketing activity, others simply wish to

456   sell at the farm gate to a local collector regardless of own economic goals. Coupled with the high

457   percentage of single marketing channel dependence, this suggests a need to empower rice

458   farmers and improve farmers' access to market. Measures may include investing in rural

459   infrastructure and enhancing access to marketing information as identified by Pham *et al.* (2019)

460   and Mgale and Yan (2020). The fact that majority of farmers only had primary school education

461   and farmers with high school or above education were more likely to use collective action

462   channel (i.e. cooperative) suggests that rural education system is an area for improvement. The

463   finding that trust and transaction conditions played a substantive role in channel choice

464   demonstrates the farmers' concern about integrity and services offered by the channels. Policy

465   makers may address this issue by developing standards for rice purchasing from farmers,

466   especially ensuring the reliability of the weighing and grading process. Without this assurance,

467   farmers may simply take use the marketing channel of lower economic return but less transaction

468   risk.

469   There are several limitations to this research. The scope of this study is inevitably limited

470   by time, sector and country contexts. As mentioned in the introduction, the use of marketing

471   channels in Thailand has changed over time. Whilst four marketing channels have been reported

472   in literature, this study found that central paddy market was only used by less than 5% of the

473   respondent rice farmers, much lower than the previously found 12.6% (Srisompun, 2014). It is

474   possible this might reflect the impact of the end of the latest RPS in 2014. Due to the small

475   number of users, it was statistically inappropriate to model the determinant factors for the use of

476   central paddy market. Secondly, a theoretical limitation is that this study only looked at past

477   behavior and intentional behavior whilst the original TRA/TPB suggest that intention is a

478   predictor of actual behavior which requires a longitudinal study with multiple data collection

479 points, not an element of the methodology described herein. Future studies can look at measuring

480 all three elements of self-reported or indeed observed behaviors: past behavior, behavioral

481 intention and actual behavior. The covariance between the error terms of past behavior and

482 intention was fairly high, indicating a strong correlation between the two behavioral variables.

483 Whilst we recognize the merit of single indicator for the two behavioral constructs (Hayduk and

484 Littvay, 2012), it is possible that intention and past behavior could be measured with multiple

485 meaningful indicators. Future studies into other influencing variables, and in other contexts will

486 also help to develop a more holistic understanding of marketing channel choice by farmers.

6. Conclusions

488 This study looked at marketing channels used by Thai rice farmers and the data showed some

489 statistically significant differences in channel use by region, gender, education and farm size. It

490 then examined how economic, relational, and psychological factors driving Thai rice farmers'

491 decision-making toward their intentional choice of marketing channel. The theoretical model

492 was proposed and tested and suggests that farmers' channel use intention is influenced directly

493 by attitude, subjective norm, economic goal and transaction condition, and indirectly by past

494 behavior and trust via attitude. Subjective norm, economic goal and transaction conditions also

495 influence past behavior directly. We believe we have made some substantive contributions to the

496 study of this subject with these findings. Further research can test the nuanced interaction of

497 these factors in influencing self-reported attitudinal and behavioral intention and the actual

498 behavior exhibited in practice. Past behavior in all situations was found to have significantly

499 influenced attitude, which then consistently influenced intention albeit weakly especially in the

500 negative intentional group. This finding is of interest for wider exploration in wider industrial

501 and behavioral contexts.

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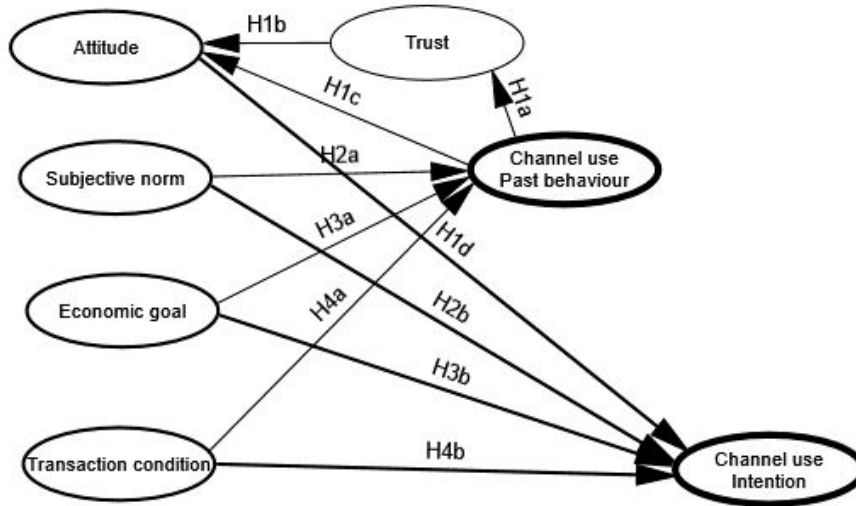
Figure 1. Conceptual model of factors influencing marketing channel use behavior
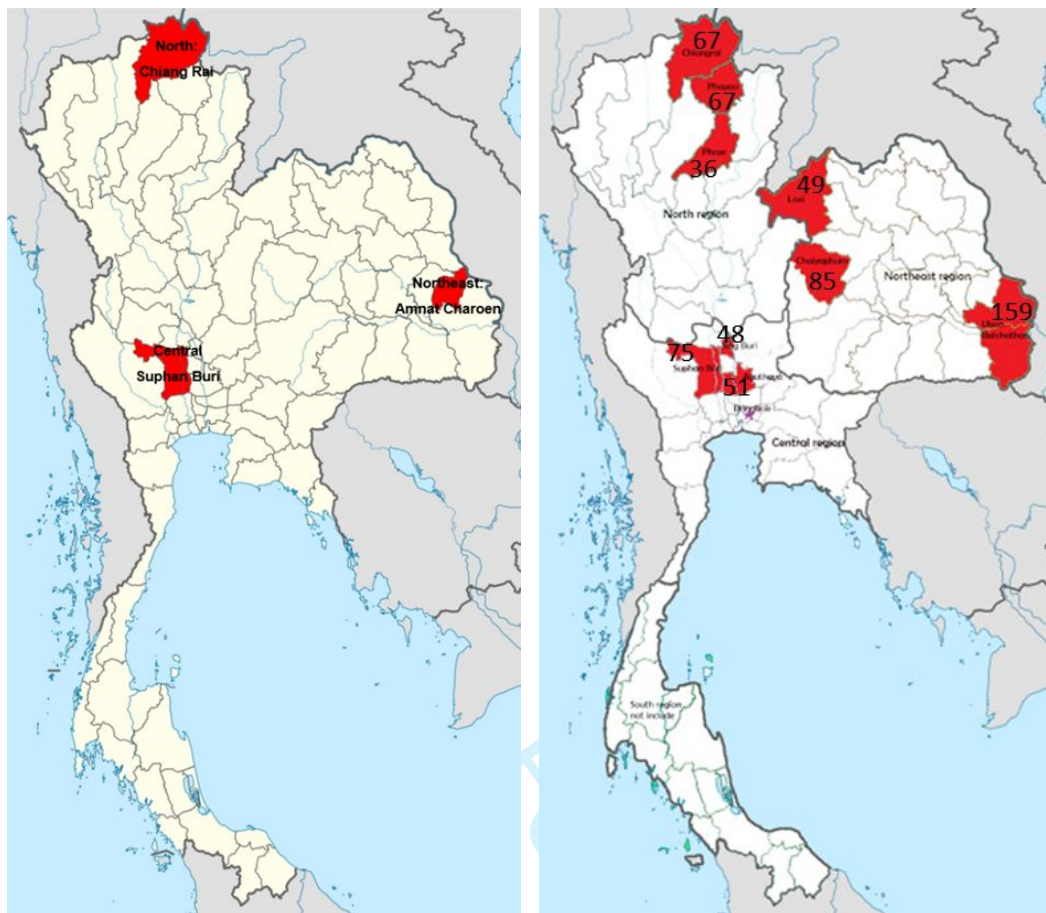
2a. Provinces for phase-one interviews   2b. Provinces for phase-two surveys
(with number of responses)

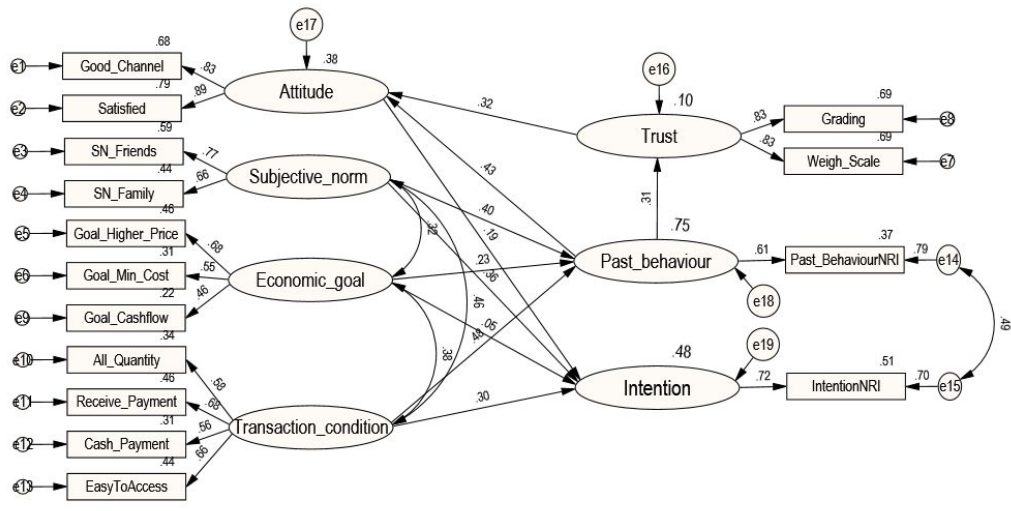**Figure 2. Study area with provinces marked in red**

Figure 3. Modified model of channel choice behavior for Thai rice farmers (Model 2)

(Model fit indices are in Table 3)

Table 1. Socio-demographic characteristics of the respondents, pooled number of responses by channel use, and Chi-Square test for independence of channel used by key attributes

Attribute	Region	N	Min	Max	Mean	SD	Median
Age	North	170 (27%)	25	71	50.21	9.03	52
(year)	Northeast	293 (46%)	18	84	54.1	10.80	54
	Central	174 (27%)	27	82	53.76	11.00	55
	Total	637(100%)	18	84	52.97	10.53	54
Education	North	170	0	16	7.5	4.03	6
(year)	Northeast	293	0	16	5.89	2.97	4
	Central	174	2	18	6.74	3.77	4
	Total	637	0	18	6.55	3.56	6
Household size	North	170	1	7	4.24	1.22	4
(person)	Northeast	293	1	12	4.82	1.85	5
	Central	174	1	10	4.06	1.64	4
	Total	637	1	12	4.46	1.68	4
Rice sold	North	149	2%	100%	59.2%	21.0%	52.9%
(percentage)	Northeast	208	4%	100%	54.3%	20.3%	52.2%
	Central	77	89%	100%	99.8%	1.2%	100%
	Total	434	2%	100%	64.1%	25.1%	58.8%
Rice farm size							
(Rai)	North	170	2	90	10.21	8.77	8
(1 Rai = 0.16 ha)	Northeast	293	2	54	15.28	9.76	13
	Central	174	2	240	28.06	26.22	23
	Total	637	2	240	17.42	17.27	12

		Channel Used N					
Unique number of respondents	Region	Miller	Miller & Co-op	Co-op	Local collector	Miller & Local collector	All three channels
170	North	56	11	43	57	3	0
293	Northeast	170	1	20	93	8	1
174	Central	92	21	22	33	5	1
637	**Total**	**318**	**33**	**85**	**183**	**16**	**2**

Pooled number of responses	Region *	Miller (pooled n)	% of 637	Co-op (pooled n)	% of 637	Local collector (pooled n)	% of 637
184	North	70	11.0%	54	8.5%	60	9.4%
304	Northeast	180	28.3%	22	3.5%	102	16.0%
202	Central	119	18.7%	44	6.9%	39	6.1%
690	Total	369	57.9%	120	18.8%	201	31.6%

Chi-Square test for independence (channel used by gender, education and farm size)*							
		Miller		Co-op		Local collector	
		n	% by row	n	% by row	n	% by row
Gender	Male	148	54.0%	58	21.20%	68	24.8%
(X^2=6.65, .036)	Female	221	53.1%	62	14.90%	133	**32.0%**
Education	≤ 7 years	265	55.0%	62	12.90%	155	32.2%
(X^2=29.06, <.001)	> 7 years	89	50.3%	53	**29.90%**	35	19.8%
Farm size	≤ 12 Rais	150	45.3%	60	18.10%	121	**36.6%**
(X^2=20.16, <.001)	> 12 Rais	219	**61.0%**	60	16.70%	80	22.3%

*Chi-Square test of channel used by region: X^2=56.065, p <.001

Table 2. Descriptives, reliability and validity test results for key latent variables and comparison by channel used

		Intention[a]	Past behavior[a]	Attitude	Trust	Subjective Norm	Economic Goal	Transaction condition
Pooled number of responses = 690	Mean	0.81	0.89	5.92	5	5.27	4.04	4.61
	Min	0.14	0.17	1.82	1.39	2.36	1.51	2.67
	Max	1	1	6.89	6.48	6.06	4.51	5.06
	SD	0.314	0.24	1.09	1.21	0.83	0.44	0.51
By channel used	N	Mean	Mean					
Miller	369	0.81	0.89	5.84	4.79	5.26	4.05	4.58
Local Collector	201	0.85	0.93	6.05	5.33	5.33	4.04	4.75
Co-op	120	0.76	0.79	5.95	5.09	5.18	4.01	4.48
(ANOVA test p)		**	***	*	***	(.26)	(.71)	***

Validity test results for key latent variables

	CR	AVE	MSV	1	2	3	4	5
1. Attitude (Satisfaction)	0.842	0.728	0.204	**0.853**				
2. Trust	0.819	0.693	0.197	0.444	**0.833**			
3. Subjective Norm	0.675	0.511	0.209	0.369	0.269	**0.715**		
4. Economic Goal	0.601	0.359	0.198	0.199	-0.052	0.301	**0.600**	
5. Transaction condition	0.775	0.465	0.209	0.364	0.239	0.457	0.391	**0.682**

Note: [a] Variables were negatively skewed and therefore transformed using formula 1/(K-old variable) where K = largest possible value + 1 (Pallant, 2020, p. 98);

* p < .10; ** P < .05; *** P < .01

Table 3. Model fit indices for the proposed model and modified model

Model fit indices [a]	Threshold of good model fit	Model 1 (proposed)	Model 2 (modified[b])
X^2 (chi square)	-	303.336	213.691
df (Degree of freedom)	-	79	78
p	Expect $p < .05$ when N > 250	.000	.000
Normed chi square	X^2/df < 3	3.840	2.740
Goodness of fit index	GFI >.95	.945	.960
Root Mean Square Error of Approximation	RMSEA <.06	.064	.050
p of Close Fit	PCLOSE >.05	.001	.466
Adjusted GFI	AGFI >.92	.916	.939
Tucker-Lewis Index	TLI >.92	.893	.935
Comparative Fit Index	CFI >.92	.920	.951

a.   *summarized from Hair et al. 2013*

b.   *modified by adding a covariance line between the error terms of the past behavior and intention*

Table 4. SEM test results of the modified model and multiple group analysis of users of different channels based on the modified model

Proposed causal paths			Test results interpretation	Model 2 (N=690)		Miller (N=369)		Local collector (N=201)		Cooperative (N=120)		MGA path comparison between the three channels	
				std. β	p	std. β	p	std. β	p	std. β	p	X^2 (df=2)	p
Past behavior	→ Trust	H1a	Supported	.309	***	.187	.025	.252	.032	.321	.022	.75	.689 NS
Trust	→ Attitude	H1b	Partly supported	.324	***	.366	***	.311	***	.079	.351	13.53	.012
Past behavior	→ Attitude	H1c	Supported	.432	***	.341	***	.390	.008	.799	***	11.69	***
Attitude	→ Intention	H1d	Partly supported	.192	***	.231	***	.192	.155	.224	***	6.83	.003
Subjective norm	→ Past behavior	H2a	Supported	.398	***	.368	***	.303	***	.250	.032	1.46	.482 NS
Subjective norm	→ Intention	H2b	Partly supported	.359	***	.294	***	.338	.004	.160	.209	3.26	.196 NS
Economic goal	→ Past behavior	H3a	Partly supported	.226	***	.353	***	.159	.568	.328	.151	.99	.609 NS
Economic goal	→ Intention	H3b	Partly supported	.051	.462	.142	.724	.090	.552	.479	***	9.91	.012
Transaction condition	→ Past behavior	H4a	Supported	.476	***	.398	***	.319	***	.341	.041	.975	.614 NS
Transaction condition	→ Intention	H4b	Partly supported	.298	***	.256	***	.286	.016	.330	.669	4.38	.112 NS
				R^2									
Past behavior				.750		.690		.356		.651		X^2=92.338	
Intention				.482		.414		.486		.748		(df=36) ***	

*** p < .001;   Coefficients are shaded if p > .050;

NS: the path weight is not significantly different across the three channel user groups with the significance probability being higher than .05.