

Factor Analysis of Continuance Intention to Use QR Code Mobile Payment Services: An Extended Expectation-Confirmation Model (ECM)

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ABSTRACT

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QR code mobile payment is a payment method that is quite popular in Indonesia where users only need to open or display a QR code on the m-payment application when making transactions. Users can make payments easily, anywhere and anytime. Apart from the benefits of QR codes on m-payments, there are still obstacles regarding the intention to continue using them. Some users stopped using the QR code service on the m-payment application due to the potential risks involved. The purpose of this study is to find out what factors can affect continued intention to use QR code m-payment. The research model used is the Extended Expectation-Confirmation Model (ECM) by combining ECM and UTAUT and adding trust and perceived risk variables. The number of samples in this study was 313 participants who were users who had used QR code m-payment OVO, GoPay, or ShopeePay with a minimum age of 17 years. The sampling technique used is purposive sampling. This study uses quantitative methods and data analysis with the PLS-SEM approach using SmartPLS version 3. The results of this study are three rejected hypotheses and nine accepted hypotheses. Based on the accepted hypotheses, it shows that social influence, trust, and satisfaction affect continuance intention to use QR code m-payment. Social influence is the biggest factor affecting continuance intention to use QR code m-payment service. These results can be considered for developers and companies such as OVO, GoPay, and ShopeePay.

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1 Introduction

The increasingly developed technology has made progress with the presence of a transaction method using mobile payment or m-payment as a digital payment. M-payment has now dominated the market in both developed and developing countries (Franque *et al.*, 2021). Several innovations have been implemented in the m-payment application service, one of which is the Quick Response (QR) code technology is a code that has a black and white grid arrangement and is usually used to store information that can be read by a cell phone camera (Gao *et al.*, 2018). QR code m-payment is used by opening the m-payment application to display or scan QR code. According to a Kadence (2021) survey, it shows that OVO, GoPay, and ShopeePay are the three most widely used m-payment applications in Indonesia with a percentage of OVO 31%, GoPay 25%, ShopeePay 20%, DANA 19%, and LinkAja 4%. The three applications are OVO, GoPay, and ShopeePay. They also have a QR code service for transaction processing. M-payment applications are very convenient payment method and can be integrated with QR codes (Silalahi *et al.*, 2022).

The main advantage of using QR code m-payment is that users can make payments easily, anywhere and anytime. However, there are still obstacles related to the intention to continue using

them. Some users stop using QR code m-payment because of some potential risks associated with it. For example smartphones can be infected with viruses via QR codes and money stolen from m-payment applications (Gao *et al.*, 2018). So it is important to know the factors that affect the continued intention to use QR code m-payments. Continuing use of a system is very important because the success of a system ultimately depends on its continued use, not its initial use (Lee & Kwon, 2011).

Several previous studies have used different theoretical models to find out continuance intention on m-payment. Shao *et al.* (2019) use the theory of diffusion of innovations (DOI), then there will be an expansion of the IT continuance (ITC) model (Chen & Li, 2017). Several studies use the ECM as the theoretical basis for researching intentions to continue using m-payments (Franque *et al.*, 2021; Humbani & Wiese, 2019; Lu *et al.*, 2017). UTAUT is also used to identify continuance intention in using a technology (Franque *et al.*, 2021; Humbani & Wiese, 2019; Lu *et al.*, 2017).

Although the trend of m-payment is increasing, research related to continued intention to use QR code m-payment is still rarely done. Several previous studies have only focused on continuance intentions towards the use of m-payments and did not focus on continuance intentions for certain service features or specific services contained in m-payments (Cao *et al.*, 2018; Chen & Li, 2017; Franque *et al.*, 2021; Shao *et al.*, 2019). Several previous studies have examined the continuance intention factor using the extended ECM model by combining the ECM and UTAUT models (Singh, 2020; Wu & Wu, 2019). However, there are still limited studies that use the combined ECM and UTAUT models in the context of the QR code m-payment service. In addition, combining the ECM and UTAUT models can explain continuance intention more strongly (Singh, 2020). In addition to the original construction of ECM, the author combines it with the UTAUT model, which uses the effort expectancy and social influence variables. Trust is an important factor on continuance intention because it is connected to transaction of financial (Cao *et al.*, 2018). In addition, perceived risk is also a major concern for m-payment service users. Some users stop using the QR code m-payment service because of the potential associated risks (Gao *et al.*, 2018). So, this research uses the extended ECM model by combining the ECM model (Bhattacharjee, 2001) with UTAUT (Venkatesh *et al.*, 2003) and adding two variables, namely perceived risk (Yuan *et al.*, 2016) and trust (Cao *et al.*, 2018) to identify continued intention in using QR code m-payments.

This study explores the factors that influence continuance intention to use QR code m-payment OVO, GoPay, and ShopeePay. Two research questions are posed:

RQ1. What factors influence continuance intention to use the QR code m-payment service?

RQ2. How is the influence of perceived usefulness, satisfaction, effort expectancy, social influence, perceived risk, and trust on continuance intention to use the QR code m-payment service?

2 Research Model and Hypothesis

2.1 Confirmation

Confirmation refers to a user's assessment of a product, service, or technology (Franque *et al.*, 2021). The user will get a positive confirmation when the expectations are met, and when it is not as expected, the user will get a negative confirmation (Oghuma *et al.*, 2016). So that QR code m-payment users can try to adjust their perceived usefulness to be more in line with reality. Previous research explains that confirmation determines positively on perceived usefulness (Franque *et al.*, 2021; Mouakket, 2015; Yuan *et al.*, 2016). Confirmation of the suitability of a product or service to meet user expectations can trigger satisfaction. Confirmation will determine user satisfaction and perceived usefulness of m-payment services (Susanto *et al.*, 2016).

H1: Confirmation (C) has a positive effect on Perceived Usefulness (PU)

H2: Confirmation (C) has a positive effect on Satisfaction (S)

2.2 Perceived Usefulness

Perceived usefulness in this study is defined as the user's opinion of the expected advantages of using the QR code m-payment. Sustainable relationships will increase if the benefits of the service are felt by users (Franque *et al.*, 2021). Perceived usefulness explains satisfaction positively on m-payment (Franque *et al.*, 2021; Lim *et al.*, 2019). Perceived usefulness determines continuance intention on m-payment (Franque *et al.*, 2021; Liébana-Cabanillas *et al.*, 2021).

H3: Perceived Usefulness (PU) has a positive effect on Satisfaction (S)

H4: Perceived Usefulness (PU) has a positive effect on Continuance Intention (CI)

2.3 Trust

Trust is the user's level of belief in the QR code m-payment service. Lack of security will also increase the risk and uncertainty of QR code m-payment after adoption of usage, so that it can affect the intention of continued use. Several studies have shown the effect of trust on continuance intention in various contexts, such as m-payment (Cao *et al.*, 2018), *m-wallet* (Kumar *et al.*, 2018), and *m-banking* (Susanto *et al.*, 2016). When users feel a service is reliable, it will lead to a sense of satisfaction. If the user does not trust it, then there is a possibility of dissatisfaction with the service, resulting in a negative evaluation (Cao *et al.*, 2018). Trust has been shown to have a significant positive effect on satisfaction (Cao *et al.*, 2018; Poromatikul *et al.*, 2019; Susanto *et al.*, 2016).

H5: Trust (T) has a positive effect on Satisfaction (S)

H6: Trust (T) has a positive effect on Continuance Intention (CI)

2.4 Effort Expectancy

Effort expectancy in this study is the user's perception of the ease of using the QR code m-payment. Effort expectancy in UTAUT can be used to investigate continuance intention (Venkatesh *et al.*, 2011). Marinković *et al.* (2020) used the UTAUT and stated that effort expectancy determines satisfaction. Furthermore, effort expectancy affects continuance intention in using m-payment (Singh, 2020). The research of Gao *et al.* (2018) also shows that there is a positive effect of effort expectancy with continuance intention in the use of QR codes in m-payments.

H7: Effort Expectancy (EE) has a positive effect on Satisfaction (S)

H8: Effort Expectancy (EE) has a positive effect on Continuance Intention (CI)

2.5 Perceived Risk

Perceived risk is the user's response to uncertainty and adverse consequences when using a system or service (Yuan *et al.*, 2016). QR code m-payment services are often associated with risks related to privacy, loss of personal data, and transactions, which are of greatest concern to consumers (Gao *et al.*, 2018). Some users may be concerned if the service provider discloses their personal data to other companies or people. Perceived risk has a significant negative effect on satisfaction in the use of m-payment (Chen & Li, 2017; Yuan *et al.*, 2016). There is a negative effect between perceived risk and continuance intention of using m-payment services (Chen & Li, 2017; Rouibah *et al.*, 2016; Shao *et al.*, 2019).

H9: Perceived Risk (PR) has a negative effect on Satisfaction (S)

H10: Perceived Risk (PR) has a negative effect on Continuance Intention (CI)

2.6 Social Influence

Social influence, in the context of this study, is specified as the influence of the family or friends environment. Since people observe the actions of others when using m-payment services, they are influenced by those around them (Gao *et al.*, 2018). Previous research discovered that social influence explains positively on continuance intention to use m-payment (Gao *et al.*, 2018; Lu *et al.*, 2017).

H11: Social Influence (SI) has a positive effect on Continuance Intention (CI)

2.7 Satisfaction

Satisfaction is defined as a user's feeling of satisfaction in using the QR code m-payment service. However, when users are dissatisfied with the product, they can stop using it (Kumar *et al.*, 2018). In accordance with several previous studies, continuance intention is affected by satisfaction (Joo *et al.*, 2017; Kumar *et al.*, 2018; Yuan *et al.*, 2016).

H12: Satisfaction (S) has a positive effect on Continuance Intention (CI)

The research model used is shown in Figure 1.

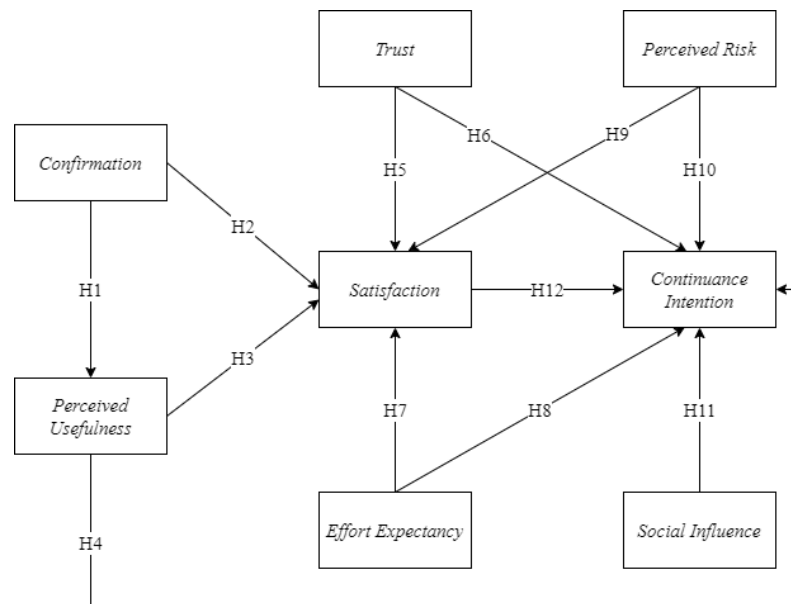


Figure 1. Research model

3 Method

This study uses quantitative methods in testing the extended ECM by combining ECM and UTAUT and adding trust and perceived risk to explain continuance intention to use QR code m-payment services. Quantitative methods are used through surveys using questionnaires. The questionnaire consists of 34 items. Questionnaires were distributed to users of QR code m-payments such as OVO, GoPay, or ShopeePay via WhatsApp, Instagram, and Twitter.

3.1 Participants

The sampling technique used is purposive sampling. The specified criteria are users who have used QR code m-payment OVO, GoPay, or ShopeePay and are at least 17 years old. Using a minimum sample of 250, because in PLS-SEM the sample size of more than 250 can increase accuracy and consistency (Sholihin & Ratmono, 2021). This study obtained 313 valid participants. Based on gender, 183 (58.5%) of the 313 participants were female and 130 (41.5%) were male. Most of the participants were 17-25 years old 240 (76.7%), and most of them had a high school education or equivalent 197 (62.9%). Students dominated 207 (66.1%) of the total 313 participants. The most participants were ShopeePay users, with 284 (90.7%). Furthermore, most of the participants used the QR code m-payment more than once a week 159 (50.8%). Table 1 shows the profile of participants.

Table 1. Demographic information of the participants

Category	Number	Percentage (%)
<i>Gender</i>		
Male	130	41.5
Female	183	58.5
<i>Age</i>		
17-25 years old	240	76.7
26-35 years old	60	19.2
36-45 years old	13	4.2
46-50 years old	0	0.0
<i>Last education</i>		
Junior High School/Equivalent	0	0.0
Hmkijuhg igh School/Equivalent	197	62.9
Diploma (D1/D2/D3/D4)	14	4.5
Bachelor degree (S1)	101	32.3

Master (S2)	1	0.3
Doctorate (S3)	0	0.0
<i>Jobs</i>		
Student		
Private employees	207	66.1
Government employees	48	15.3
Self-employed	15	4.8
Housewife	33	10.5
Freelancer	4	1.3
Doesn't work	2	0.6
Teacher	2	0.6
Extracurricular Coach	1	0.3
Private employees	1	0.3
<i>Mobile payment used</i>		
OVO	120	38.3
GoPay	200	63.9
ShopeePay	284	90.7
<i>Frequency of using QR code mobile payment OVO, GoPay, or ShopeePay</i>		
At least 1 time a day	8	2.6
>1 time a week	159	50.8
1 time a week	35	11.2
<1 time a week	111	35.5

3.2 Measures

The data was obtained from a survey using a questionnaire. The questionnaire consists of eight constructs and 34 items. They are perceived usefulness (PU) (four items), confirmation (C) (four items), satisfaction (S) (five items), effort expectancy (EE) (four items), social influence (SI) (five items), perceived risk (PR) (five items), trust (T) (four items), and continuance intention (CI) (three items). The research questionnaire consists of demographic information and questions of construct. The questionnaire uses 7 choices on a Likert scale with 1 = strongly disagree, 2 = disagree, 3 = slightly disagree, 4 = neutral, 5 = slightly agree, 6 = agree, and 7 = strongly agree (Jogiyanto, 2008). The items used in this study were adapted from several sources (Alalwan *et al.*, 2016; Bhattacharjee, 2001; Gao *et al.*, 2018; Park *et al.*, 2019; Susanto *et al.*, 2016; Venkatesh *et al.*, 2003; Venkatesh *et al.*, 2012; Yuan *et al.*, 2016).

3.3 Data Analysis

Data analysis is a process after collecting participant data or other data sources (Sugiyono, 2013). Then the data analysis was carried out by dividing it into two steps are demographic analysis and statistical analysis. Demographic analysis by processing and analyzing demographic data from the results of distributing questionnaires. The next stage is statistical analysis to test the model and test the proposed hypothesis using the Partial Least Square-Structural Equation Modeling (PLS-SEM) approach with SmartPLS version 3.

4 Results and Discussion

4.1 Evaluation of the Measurement Model (Outer Model)

4.1.1 Individual Item Reliability

Individual item reliability is done by checking the outer loading value of each variable. The indicator or item is declared valid if the outer loading value is above 0.7 (Hair *et al.*, 2017). Based on Table 2, the outer loading is more than 0.7, so it can be said to be valid and can be continued on the test afterwards.

Table 2. Outer loading value

	C	CI	EE	PR	PU	S	SI	T
C1	0.863							
C2	0.878							
C3	0.886							
C4	0.846							
CI1		0.922						
CI2		0.934						
CI3		0.934						
EE1			0.878					
EE2			0.908					
EE3			0.900					
EE4			0.874					
PR1				0.898				
PR2				0.930				
PR3				0.940				
PR4				0.911				
PR5				0.877				
PU1					0.865			
PU2					0.888			
PU3					0.914			
PU4					0.859			
S1						0.901		
S2						0.915		
S3						0.890		
S4						0.889		
S5						0.872		
SI1							0.857	
SI2							0.884	
SI3							0.835	
SI4							0.864	
SI5							0.719	
T1								0.790
T2								0.894
T3								0.927
T4								0.879

4.1.2 Internal Consistency Reliability

The internal consistency reliability test is based on the value of composite reliability and Cronbach's alpha values with the condition that the value is above 0.7 (Hair *et al.*, 2017). Based on Table 3, composite reliability and Cronbach's alpha have values above 0.7, so that the variable is in accordance with the requirements.

Table 3. Composite reliability and cronbach's alpha value

Variable	Composite reliability	Cronbach's alpha
Confirmation (C)	0.925	0.891
Continuance Intention (CI)	0.950	0.922
Effort Expectancy (EE)	0.939	0.913
Perceived Risk (PR)	0.961	0.949
Perceived Usefulness (PU)	0.933	0.904
Satisfaction (S)	0.952	0.937
Social Influence (SI)	0.919	0.889
Trust (T)	0.928	0.896

4.1.3 Convergent Validity

The convergent validity test is carried out based on the value of the average variance extracted (AVE) and accepted if it is more than 0.5 (Hair *et al.*, 2017). Table 4 shows that The AVE for every variable is more than 0.5.

Table 4. AVE value

Variable	AVE
Confirmation (C)	0.754
Continuance Intention (CI)	0.865
Effort Expectancy (EE)	0.792
Perceived Risk (PR)	0.831
Perceived Usefulness (PU)	0.777
Satisfaction (S)	0.798
Social Influence (SI)	0.696
Trust (T)	0.764

4.1.4 Discriminant Validity

Discriminant validity is based on the fornell-larcker and cross loading values (Hair *et al.*, 2017; Yamin & Kurniawan, 2011). Fornell-larcker based on the AVE root value on the correlation between variables, provided that each variable's value must be larger than the correlation with other variables and the requirement for the measurement of cross loading is that the value of the indicator on the variable itself must be greater than the value of the indicator on other variables. (Hair *et al.*, 2017). Based on Table 5 and 6, the cross loading and fornell-larcker values are appropriate so that it can be continued in the next test.

Table 5. Cross loading

	C	CI	EE	PR	PU	S	SI	T
C1	0.863	0.476	0.555	-0.104	0.590	0.675	0.428	0.566
C2	0.878	0.548	0.477	-0.208	0.578	0.670	0.455	0.522
C3	0.886	0.494	0.532	-0.154	0.670	0.686	0.400	0.577
C4	0.846	0.490	0.604	-0.042	0.686	0.704	0.423	0.617
CI1	0.545	0.922	0.498	-0.226	0.537	0.665	0.640	0.607
CI2	0.494	0.934	0.468	-0.235	0.488	0.638	0.637	0.581
CI3	0.571	0.934	0.505	-0.200	0.514	0.687	0.672	0.626
EE1	0.516	0.431	0.878	-0.121	0.520	0.575	0.276	0.425
EE2	0.565	0.453	0.908	-0.183	0.571	0.613	0.310	0.470
EE3	0.551	0.480	0.900	-0.075	0.584	0.676	0.390	0.521
EE4	0.591	0.507	0.874	-0.075	0.552	0.677	0.412	0.531
PR1	-0.161	-0.170	-0.148	0.898	-0.103	-0.248	-0.119	-0.179
PR2	-0.114	-0.191	-0.097	0.930	-0.025	-0.198	-0.210	-0.136
PR3	-0.185	-0.272	-0.169	0.940	-0.051	-0.243	-0.223	-0.204
PR4	-0.072	-0.187	-0.045	0.911	0.021	-0.169	-0.216	-0.098
PR5	-0.104	-0.239	-0.087	0.877	0.008	-0.167	-0.233	-0.108
PU1	0.634	0.503	0.547	-0.033	0.865	0.627	0.373	0.505
PU2	0.651	0.469	0.580	-0.064	0.888	0.649	0.336	0.568
PU3	0.646	0.507	0.559	-0.038	0.914	0.675	0.362	0.546
PU4	0.642	0.466	0.523	0.009	0.859	0.631	0.321	0.544
S1	0.737	0.686	0.661	-0.223	0.643	0.901	0.585	0.681
S2	0.729	0.611	0.672	-0.166	0.700	0.915	0.525	0.629
S3	0.686	0.633	0.640	-0.191	0.668	0.890	0.532	0.630
S4	0.667	0.674	0.641	-0.204	0.623	0.889	0.553	0.614
S5	0.700	0.580	0.584	-0.236	0.639	0.872	0.477	0.595
SI1	0.393	0.577	0.312	-0.199	0.271	0.472	0.857	0.386
SI2	0.373	0.559	0.255	-0.194	0.268	0.463	0.884	0.349
SI3	0.416	0.578	0.374	-0.095	0.387	0.514	0.835	0.430
SI4	0.476	0.659	0.399	-0.254	0.407	0.588	0.864	0.492
SI5	0.376	0.527	0.286	-0.165	0.301	0.444	0.719	0.405
T1	0.531	0.435	0.416	-0.101	0.457	0.532	0.310	0.790
T2	0.555	0.568	0.480	-0.168	0.501	0.613	0.430	0.894
T3	0.605	0.609	0.495	-0.170	0.535	0.647	0.477	0.927
T4	0.608	0.637	0.522	-0.127	0.635	0.663	0.497	0.879

Table 6. Fornell-larcker

	C	CI	EE	PR	PU	S	SI	T
C	0.868							
CI	0.578	0.930						
EE	0.626	0.527	0.890					
PR	-0.144	-0.236	-0.125	0.911				
PU	0.729	0.552	0.626	-0.036	0.882			
S	0.788	0.714	0.717	-0.228	0.732	0.893		
SI	0.491	0.699	0.394	-0.220	0.395	0.599	0.834	
T	0.659	0.651	0.550	-0.163	0.613	0.706	0.498	0.874

4.2 Evaluation of the Structural Model (Inner Model)

4.2.1 Path Coefficient

Path coefficient which shows the magnitude of the influence of each independent variable (Widarjono, 2020). If the path coefficient value is more than 0.1, then it will affect the model. Table 7 indicates that perceived usefulness, effort expectancy, and perceived risk on continuance intention have path coefficient value below 0.1, indicating that they have no effect. While the other eight variables have a significant positive effect and perceived risk on continuance intention has a significant negative effect.

Table 7. Path coefficient

Hypothesis	Path	Path coefficient
H1	C → PU	0.729
H2	C → S	0.329
H3	PU → S	0.200
H4	PU → CI	0.063
H5	T → S	0.207
H6	T → CI	0.218
H7	EE → S	0.259
H8	EE → CI	0.033
H9	PR → S	-0.107
H10	PR → CI	-0.052
H11	SI → CI	0.397
H12	S → CI	0.240

4.2.2 Coefficient of Determination (R^2)

If the R^2 value is around 0.75, it is said to be strong, around 0.50 is said to be moderate, and around 0.25 is said to be weak (Hair *et al.*, 2017). According to the results in Table 8, the dependent variables are continuance intention and perceived usefulness, which can be said to be moderate. Meanwhile, satisfaction can be said to be strong. It can be interpreted that perceived usefulness, perceived risk, trust, effort expectancy, social influence, and satisfaction explain continuance intention moderately. In addition, confirmation moderates the perceived usefulness. While the confirmation, perceived usefulness, trust, effort expectancy, and perceived risk explain satisfaction strongly.

Table 8. Coefficient of determination

Variable	R^2	Description
Continuance Intention (CI)	0.655	Moderate
Perceived Usefulness (PU)	0.532	Moderate
Satisfaction (S)	0.761	Strong

4.2.3 T-Test

Table 9 presents the results of hypothesis test on the structural model. The hypotheses were tested by a two-tailed test using the bootstrapping method with a significance level of 5%. If the

t -value is greater than 1.96 then the hypothesis can be accepted. The results show that there are nine accepted hypotheses because the t -value is greater than 1.96. The other three were rejected because they had a t -value of less than 1.96.

Table 9. T-test

Hypothesis	Path	T-test
H1	C → PU	23.669
H2	C → S	5.323
H3	PU → S	3.842
H4	PU → CI	0.925
H5	T → S	3.145
H6	T → CI	3.347
H7	EE → S	5.214
H8	EE → CI	0.571
H9	PR → S	3.243
H10	PR → CI	1.602
H11	SI → CI	6.419
H12	S → CI	2.744

4.2.4 Effect Size (f^2)

Based on Table 10 show that confirmation of perceived usefulness has a large effect. Then, the three pathways that have a moderate effect are confirmation on satisfaction, effort expectancy on satisfaction, and social influence on continuance intention. The three pathways that have no effect are perceived usefulness, effort expectancy, continuance intention, and perceived risk on continuance intention. While the other five variable relationships have a small effect.

Table 10. Effect size

Hypothesis		f^2			Influence
No	Path	$R^2_{included}$	$R^2_{excluded}$	Σf^2	f^2
H1	C → PU	0.532	0.000	1.137	Large
H2	C → S	0.761	0.721	0.168	Medium
H3	PU → S	0.761	0.745	0.066	Small
H4	PU → CI	0.655	0.654	0.005	Not supported
H5	T → S	0.761	0.740	0.091	Small
H6	T → CI	0.655	0.633	0.065	Small
H7	EE → S	0.761	0.726	0.148	Medium
H8	EE → CI	0.655	0.655	0.001	Not supported
H9	PR → S	0.761	0.750	0.046	Small
H10	PR → CI	0.655	0.653	0.007	Not supported
H11	SI → CI	0.655	0.558	0.282	Medium
H12	S → CI	0.655	0.642	0.041	Small

4.2.5 Predictive Relevance (Q^2)

The predictive relevance test must have a value larger than zero so that the variable has a predictive relationship with other variables (Hair *et al.*, 2017). In Table 11, the overall value of the dependent variable, which is above zero, has a predictive relationship with other variables.

Table 11. Predictive relevance

Variable	Q^2
Continuance Intention (CI)	0.556
Perceived Usefulness (PU)	0.410
Satisfaction (S)	0.596

4.2.6 Relative Impact (q^2)

If the relative impact value is about 0.02 then it has a small effect, has a medium effect if the relative impact value is around 0.15 and has a large effect if the relative impact value is 0.35 (Hair *et al.*, 2017). Based on Table 12, the relationship between confirmation and perceived usefulness which has a large effect, and social influence on continuance intention has a medium effect, perceived usefulness, effort expectancy, continuance intention, and perceived risk on continuance intention have no effect. While the other ten hypotheses have a small effect.

Table 12. Relative impact

Hypothesis		q^2			Influence
No	Path	$Q^2_{included}$	$Q^2_{excluded}$	Σq^2	q^2
H1	C → PU	0.410	0.000	0.695	Large
H2	C → S	0.596	0.565	0.077	Small
H3	PU → S	0.596	0.586	0.025	Small
H4	PU → CI	0.556	0.556	0.000	Not Supported
H5	T → S	0.596	0.580	0.040	Small
H6	T → CI	0.556	0.538	0.041	Small
H7	EE → S	0.596	0.569	0.067	Small
H8	EE → CI	0.556	0.557	-0.002	Not Supported
H9	PR → S	0.596	0.588	0.020	Small
H10	PR → CI	0.556	0.555	0.002	Not Supported
H11	SI → CI	0.556	0.473	0.187	Medium
H12	S → CI	0.556	0.544	0.027	Small

4.2.7 Model Fit

Model fit aims to assess how well the hypothesized model structure fits the empirical data to help identify model specification errors (Hair *et al.*, 2017). The fit models used in this study were standardized root mean square residual (SRMR), chi-square statistic (χ^2), and normal fit index (NFI). The results show that the SRMR is 0.051, which means that the suitability of this research model is said to be good (good fit). The chi-square value is 1609.276, which means that this research model has a good fit. While the NFI value, which is 0.845, which means the suitability of this research model is still acceptable (marginal fit). Values ranging from $0.80 < \text{NFI} < 0.90$ are still acceptable (Khairi *et al.*, 2021). The results obtained for the fit model are shown in Table 13.

Table 13. Results of model fit

	Limit	Value	Description
SRMR	< 0.08	0.051	Good fit
Chi-square	> 0.05	1609.276	Good fit
NFI	≥ 0.90	0.845	Marginal fit

4.3 Discussion

Based on the findings, there are three hypotheses supported from 12 hypotheses. Perceived usefulness, effort expectancy, and perceived risk have no effect on continuance intention. Trust, social influence, and satisfaction have an effect on continuance intention. The three factors that have a large effect on continuance intention are social influence, followed by satisfaction, and trust.

The results indicate that perceived usefulness is positively influenced by confirmation. Previous literature has also confirmed this relationship (Bhattacharjee, 2001; Franque *et al.*, 2021; Mouakket, 2015; Yuan *et al.*, 2016). Based on the results that have been obtained, it shows that users feel that their initial expectations are appropriate, either the suitability of the service level or service function, and that the user feels that the QR code service really has benefits, so that it affects the perceived usefulness factor. Furthermore, users feel that if their initial expectations are appropriate and fulfilled, it tends to increase satisfaction. Developers and service providers such as OVO, GoPay, and ShopeePay must provide quality services to exceed user expectations and make users feel satisfied.

Perceived usefulness determines satisfaction positively, but does not determine on continuance intention. Previous literature has also confirmed this relationship (Bhattacharjee, 2001; Lai *et al.*, 2016; Susanto *et al.*, 2016). This finding contradicts previous research on perceived risk and continuance intention (Bhattacharjee, 2001; Franque *et al.*, 2021; Liébana-Cabanillas *et al.*, 2021). Based on the results that have been obtained, it shows that users will feel satisfied if they feel useful or beneficial when using the QR code m-payment service, such as making transactions faster because scanning or showing the code is easier, and can be done anywhere and anytime. However, the perceived usefulness of users such as ease, speed, and effectiveness in transactions, does not affect continuance intention to use QR code m-payment service. When users feel the benefits obtained from the service, it is not necessarily the case that they will use it again. Developers and service providers further improve performance to be able to increase the effectiveness of user performance, including in terms of data security and transaction reader speed.

Trust has a significant positive effect on satisfaction and continuance intention. Previous literature has also confirmed this relationship (Cao *et al.*, 2018; Kumar *et al.*, 2018; Poromatikul *et al.*, 2019; Susanto *et al.*, 2016). Based on the results obtained, it shows that when users feel that the QR code m-payment service is trustworthy, secure, and reliable, it will foster a sense of satisfaction and increase continuance intention. Developers and service providers must continue to improve QR code services with more attention to convenience and can further improve the security system strictly to maintain user trust when transacting.

Effort expectancy has a significant positive effect on satisfaction, but has no effect on continuance intention. Previous literature has also confirmed this relationship (Marinković *et al.*, 2020). However, these results are in line with previous research between effort expectations and continuance intention (Chopdar & Sivakumar, 2019; Zhao & Bacao, 2020). Therefore, the smaller the effort, the more user satisfaction there will be. However, effort expectancy does not affect continuance intention to use QR code m-payment. All of the items show that QR code m-payment service is easy to learn and use, but this convenience is not what affects on continuance intention. This can be due to direct transactions or cash, which can also be done easily, so that convenience is not their consideration. The recommendations that can be given are that service providers must update the ease of service, for example, by simplifying the payment process or transactions and being able to provide digital literacy to the public.

Perceived risk has a significant negative effect on satisfaction, but has no impact on continuance intention. Previous literature has also confirmed this relationship (Chen & Li, 2017; Yuan *et al.*, 2016) and Liébana-Cabanillas *et al.* (2015) found that perceived risk has no impact on continuance intention. Based on the results that have been obtained, it shows that the perceived risk of users of the QR code m-payment service will affect user satisfaction. However, users are not worried about the risks and it will not affect the user's intention to continue using the service. This can happen if the majority of respondents come from Gen Z. Gen Z is not too concerned with risk because most of them already know how it works and want to do things quickly, easily, and instantly. So, the recommendations that can be given are that service providers must pay more attention to and improve security and convenience to avoid scams, or fraud that can occur in QR code services. In addition, the company can also provide warnings and directions to users regarding the potential for crime.

Social influence has a significant positive effect on continuance intention. Previous literature has also confirmed this relationship (Gao *et al.*, 2018; Lu *et al.*, 2017). It shows that continuance intention is influenced by the social environment, such as family, friends, relatives, or co-workers who have used the QR code m-payment service and recommend it. Social influence has the biggest effect on continuance intention so that social influence is the main factor affecting continuance intention to use QR code m-payment services. Recommendations that can be given are more socialization or dissemination of information to the public to influence the public to use the QR code m-payment service and also collaborate with various merchants to better reach users.

Continuance intention is positively influenced by satisfaction. Previous literature has also confirmed this relationship (Joo *et al.*, 2017; Kumar *et al.*, 2018; Yuan *et al.*, 2016). The author assumes that satisfaction in using the QR code m-payment service is one of them due to users feel happy and interested to continue using it. So, service providers must continue to

ensure that the QR code service can function properly. Provide promos in the form of discounts or cashback for users when making transactions to provide user satisfaction.

This study has several limitations, based on the demographic results of participants, users of the QR code m-payment service are dominated by ShopeePay users, so this study explains more about ShopeePay users than OVO and GoPay. The results are also more dominated by students. The results of data collection also found 17 invalid data that had to be deleted.

Future research can use different research subjects on m-payment applications that have a QR code feature for transactions, such as the type of mobile banking application. The research model can be further developed, such as combining the ECM model with other models. Perceived usefulness, effort expectancy, and perceived risk have no effect on continuance intention, so that further researchers can review each of the indicators on these variables.

5 Conclusion

The concept of continuance intentions are still not described in detail in this context based on the IS study. To fill this gap, it is proposed to combine ECM and UTAUT to investigate the continuance intention to use QR code m-payment. The results represent the factors that affect continuance intention to use QR code m-payment services are trust, social influence, and satisfaction. Social influence is the factor that has the greatest effect on continuance intention followed by satisfaction, and trust. Trust, perceived risk, social influence, perceived usefulness, effort expectancy and confirmation determine satisfaction. But, perceived usefulness, perceived risk and effort expectancy and have no effect on continuance intention. Based on this, social influences such as family, friends, relatives, or co-workers have the most important role in affecting the continuance intention to use QR code m-payment. The QR code service providers carry out more socialization or disseminate information to the public to influence the public's intention to use the QR code m-payment service and collaborate with various merchants to better reach users and affect the continuance intention to use QR code m-payment.

6 References

- Alalwan, A. A., Dwivedi, Y. K., Rana, N. P., & Williams, M. D. (2016). Consumer adoption of mobile banking in Jordan: Examining the role of usefulness, ease of use, perceived risk and self-efficacy. *Journal of enterprise information management*, 29(1), 118-139. <https://doi.org/10.1108/JEIM-04-2015-0035>
- Bhattacharjee, A. (2001). Understanding information systems continuance: An expectation-confirmation model. *MIS Quarterly*, 25, 351-370. <https://doi.org/10.2307/3250921>
- Cao, X., Yu, L., Liu, Z., Gong, M., & Adeel, L. (2018). Understanding mobile payment users' continuance intention: a trust transfer perspective. *Internet Research*, 28(2), 456-476. <https://doi.org/10.1108/IntR-11-2016-0359>
- Chen, X., & Li, S. (2017). Understanding continuance intention of mobile payment services: an empirical study. *Journal of Computer Information Systems*, 57(4), 287-298. <https://doi.org/10.1080/08874417.2016.1180649>
- Chopdar, P. K., & Sivakumar, V. (2019). Understanding continuance usage of mobile shopping applications in India: the role of espoused cultural values and perceived risk. *Behaviour information technology*, 38(1), 42-64. <https://doi.org/10.1080/0144929X.2018.1513563>
- Franque, F. B., Oliveira, T., & Tam, C. (2021). Understanding the factors of mobile payment continuance intention: empirical test in an African context. *Heliyon*, 7(8), e07807. <https://doi.org/10.1016/j.heliyon.2021.e07807>
- Gao, S., Yang, X., Guo, H., & Jing, J. (2018). An empirical study on users' continuous usage intention of QR code mobile payment services in China. *International Journal of E-Adoption*, 10(1), 18-33. <https://doi.org/10.4018/IJEA.2018010102>
- Hair, J. F., Hult, G. T. M., Ringle, C. M., & Sarstedt, M. (2017). *A primer on partial least squares structural equation modeling (PLS-SEM) 2nd edition*. Sage Publications.

- Humbani, M., & Wiese, M. (2019). An integrated framework for the adoption and continuance intention to use mobile payment apps. *International Journal of Bank Marketing*, 37(2), 646-664. <https://doi.org/10.1108/IJBM-03-2018-0072>
- Jogiyanto. (2008). *Metodologi penelitian sistem informasi : Pedoman dan contoh melakukan penelitian di bidang sistem teknologi informasi*. CV Andi Offset.
- Joo, Y. J., Park, S., & Shin, E. K. (2017). Students' expectation, satisfaction, and continuance intention to use digital textbooks. *Computers in Human Behavior*, 69, 83-90. <https://doi.org/10.1016/j.chb.2016.12.025>
- Kadence. (2021). *Digital payment and financial services usage and behavior in indonesia*. Retrieved 10 Maret from <https://kadence.com/digital-payments-and-e-wallet-usage-and-behavior-in-indonesia/>
- Khairi, M. I., Susanti, D., & Sukono, S. (2021). Study on Structural Equation Modeling for Analyzing Data. *International Journal of Ethno-Sciences Education Research*, 1(3), 52-60. <https://doi.org/10.46336/ijeer.v1i3.295>
- Kumar, A., Adlakaha, A., & Mukherjee, K. (2018). The effect of perceived security and grievance redressal on continuance intention to use M-wallets in a developing country. *International Journal of Bank Marketing*, 36(7), 1170-1189. <https://doi.org/10.1108/IJBM-04-2017-0077>
- Lai, H.-M., Chen, C.-P., & Chang, Y.-F. (2016). Expectation-confirmation model of information system continuance: a meta-analysis. *International Journal of Educational Pedagogical Sciences*, 10(7), 2325-2330. <https://doi.org/10.5281/zenodo.1125535>
- Lee, Y., & Kwon, O. (2011). Intimacy, familiarity and continuance intention: An extended expectation–confirmation model in web-based services. *Electronic Commerce Research and Applications*, 10(3), 342-357. <https://doi.org/10.1016/j.eleap.2010.11.005>
- Liébana-Cabanillas, F., Ramos de Luna, I., & Montoro-Ríos, F. (2015). User behaviour in QR mobile payment system: the QR Payment Acceptance Model. *Technology Analysis Strategic Management*, 27(9), 1031-1049. <https://doi.org/10.1080/09537325.2015.1047757>
- Liébana-Cabanillas, F., Singh, N., Kalinic, Z., & Carvajal-Trujillo, E. (2021). Examining the determinants of continuance intention to use and the moderating effect of the gender and age of users of NFC mobile payments: a multi-analytical approach. *Information Technology Management*, 22, 1-29. <https://doi.org/10.1007/s10799-021-00328-6>
- Lim, S. H., Kim, D. J., Hur, Y., & Park, K. (2019). An empirical study of the impacts of perceived security and knowledge on continuous intention to use mobile fintech payment services. *International Journal of Human–Computer Interaction*, 35(10), 886-898. <https://doi.org/10.1080/10447318.2018.1507132>
- Lu, J., Wei, J., Yu, C.-S., & Liu, C. (2017). How do post-usage factors and espoused cultural values impact mobile payment continuation? *Behaviour information technology*, 36(2), 140-164. <https://doi.org/10.1080/0144929X.2016.1208773>
- Marinković, V., Đorđević, A., & Kalinić, Z. (2020). The moderating effects of gender on customer satisfaction and continuance intention in mobile commerce: a UTAUT-based perspective. *Technology Analysis Strategic Management*, 32(3), 306-318. <https://doi.org/10.1080/09537325.2019.1655537>
- Mouakket, S. (2015). Factors influencing continuance intention to use social network sites: The Facebook case. *Computers in Human Behavior*, 53, 102-110. <https://doi.org/10.1016/j.chb.2015.06.045>
- Oghuma, A. P., Libaque-Saenz, C. F., Wong, S. F., & Chang, Y. (2016). An expectation-confirmation model of continuance intention to use mobile instant messaging. *Telematics and Informatics*, 33(1), 34-47. <https://doi.org/10.1016/j.tele.2015.05.006>
- Park, J., Amendah, E., Lee, Y., & Hyun, H. (2019). M-payment service: Interplay of perceived risk, benefit, and trust in service adoption. *Human Factors Ergonomics in Manufacturing Service Industries*, 29(1), 31-43. <https://doi.org/10.1002/hfm.20750>
- Poromatikul, C., De Maeyer, P., Leelapanyalert, K., & Zaby, S. (2019). Drivers of continuance intention with mobile banking apps. *International Journal of Bank Marketing*, 38(1), 242-262. <https://doi.org/10.1108/IJBM-08-2018-0224>
- Rouibah, K., Lowry, P. B., & Hwang, Y. (2016). The effects of perceived enjoyment and perceived risks on trust formation and intentions to use online payment systems: New perspectives

- from an Arab country. *Electronic Commerce Research Applications*, 19, 33-43. <https://doi.org/10.1016/j.elerap.2016.07.001>
- Shao, Z., Zhang, L., Li, X., & Guo, Y. (2019). Antecedents of trust and continuance intention in mobile payment platforms: The moderating effect of gender. *Electronic Commerce Research and Applications*, 33, 100823. <https://doi.org/https://doi.org/10.1016/j.elerap.2018.100823>
- Sholihin, M., & Ratmono, D. (2021). *Analisis SEM-PLS dengan WarpPLS 7.0 untuk hubungan nonlinier dalam penelitian sosial dan bisnis*. Penerbit Andi.
- Silalahi, P. R., Tambunan, K., & Batubara, T. R. (2022). Dampak penggunaan QRIS terhadap kepuasan konsumen sebagai alat transaksi. *ULIL ALBAB: Jurnal Ilmiah Multidisiplin*, 1(2), 122-128.
- Singh, S. (2020). An integrated model combining ECM and UTAUT to explain users' post-adoption behaviour towards mobile payment systems. *Australasian Journal of Information Systems*, 24. <https://doi.org/10.3127/ajis.v24i0.2695>
- Sugiyono. (2013). *Metodologi penelitian kuantitatif, kualitatif dan R&D*. Alfabeta.
- Susanto, A., Chang, Y., & Ha, Y. (2016). Determinants of continuance intention to use the smartphone banking services: An extension to the expectation-confirmation model. *Industrial Management Data Systems*, 116(3), 508-525. <https://doi.org/10.1108/IMDS-05-2015-0195>
- Venkatesh, V., Morris, M. G., Davis, G. B., & Davis, F. D. (2003). User acceptance of information technology: Toward a unified view. *MIS Quarterly*, 27(3), 425-478. <https://doi.org/10.2307/30036540>
- Venkatesh, V., Thong, J. Y., Chan, F. K., Hu, P. J. H., & Brown, S. A. (2011). Extending the two-stage information systems continuance model: Incorporating UTAUT predictors and the role of context. *Information Systems Journal*, 21(6), 527-555. <https://doi.org/10.1111/j.1365-2575.2011.00373.x>
- Venkatesh, V., Thong, J. Y., & Xu, X. (2012). Consumer acceptance and use of information technology: extending the unified theory of acceptance and use of technology. *MIS Quarterly*, 36(1), 157-178. <https://doi.org/10.2307/41410412>
- Widarjono, A. (2020). *Analisis multivariat terapan dengan program SPSS, AMOS, dan SMARTPLS*. UPP STIM YKPN.
- Wu, C.-G., & Wu, P.-Y. (2019). Investigating user continuance intention toward library self-service technology: The case of self-issue and return systems in the public context. *Library Hi Tech*, 37(3), 401-417. <https://doi.org/10.1108/LHT-02-2018-0025>
- Yamin, S., & Kurniawan, H. (2011). *Generasi baru mengolah data penelitian dengan partial least square path modeling*. Salemba Infotek.
- Yuan, S., Liu, Y., Yao, R., & Liu, J. (2016). An investigation of users' continuance intention towards mobile banking in China. *Information Development*, 32(1), 20-34. <https://doi.org/10.1177/0266666914522140>
- Zhao, Y., & Bacao, F. (2020). What factors determining customer continuingly using food delivery apps during 2019 novel coronavirus pandemic period? *International journal of hospitality management*, 91, 102683. <https://doi.org/10.1016/j.ijhm.2020.102683>