

Conference Paper

Factors Influencing Customers Acceptance of Using the QR Code Feature in Offline Merchants for Generation Z in Bandung (Extended UTAUT2)

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Abstract

Mobile payment services in Indonesia are shifting to support the cashless society future supported by the many tech-savvy consumers in Indonesia. The leading mobile payment players, namely Go-Pay, OVO, DANA, LinkAja, etc. began to expand their target market network by adding the QR Code feature to perform transactions in offline merchants. With the offering transactions of cashbacks, the consumer will more likely to adopt the QR Code because the promotion can only be obtained by making offline transactions at several affiliated merchants. The purpose of this study is to identify the factors that can influence the behavioral intention and the actual usage for using the QR Code feature in mobile payment to perform transactions in offline merchants. This study will also identify the extent to which the actual usage of the QR Code feature can contribute to the National Non-Cash Movement (GNNT), starting from the scope of generation Z in Bandung. This study will be conducted by using Extended UTAUT2 with nine independent variables of Performance Expectancy (PE), Effort Expectancy (EE), Social Influence (SI), Facilitating Conditions (FC), Hedonic Motivation (HM), Price Value (PV), Habit (HT), Trust (T), and Perceived Risk (PR). These variables will be tested to the Behavioral Intention (BI) and also Actual Use (AU). The result indicates that the Habit is the most significant factors to influence Behavioral Intention. While the others, namely Behavioral Intention, Facilitating Conditions, Hedonic Motivation, and Performance Expectancy influence to Actual Usage and Behavioral Intention.

Keywords: Mobile Payment, QR Code in Offline Merchants, Generation Z, Extended UTAUT2

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1. Introduction

Digital payment often understood as a consumer who uses mobile payment systems to initiate and activate their mobile phone to fulfill the transactions in regards to paying goods and services by using their mobile devices [25]. The completion of the payments and transactions between the buyer and the merchants is considered to be fast and

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straightforward. The payments also can be made anytime and anywhere by holding only a mobile device [30].

Digital payment is also happening in Indonesia, where the shifting of technology to support the cashless society future is also supported by the many tech-savvy consumers in Indonesia [45]. Indonesia's growth regarding digital payment systems is said to be increasingly confident about going all cashless. To that extent, that continuous decline in performing cash transactions occurs to Indonesia of 69%, following the second high in leading the way for cashless payments after Singapore of 76% [46]. In addition, [4] also added, 44.16% of users use smartphones/tablets to access the internet, 4.49% use computers and 39.28% use both. This shows that the adoption of mobile technology is in trend since Indonesia began introducing mobile payments to the public introduced in 2007, when Telkomsel sent its first digital payment, T-Cash. Then, followed by Dompotku in 2007 and XL Tunai in 2012. Having reached its successful performance in 2017, smartphones recovered to reach their highest point compared to bank accounts, debit cards, e-money accounts, and credit cards [1]

Today, mobile payment in Indonesia are beginning to expand their target market network by adding features for offline transactions. This feature of QR Code payment can be found from the leading mobile payment players, which is, Go-Pay, OVO, T-cash (LinkAja), and DANA. According to [14], Go-Pay is the most popular payment provider by 79%, followed by OVO (58.42%), T-cash (LinkAja) (55.52%), and DANA (34.18%). With the offering transactions of cashbacks, shopping discounts from merchants, and discounts when using the mobile payment, the consumer will more likely to adopt the QR Code because the promotion can only be obtained by making offline transactions at several affiliated merchants [29]. In the meantime, QR Code is known as the payment method that has more key drivers than using other payment method. QR Code allows everyone to perform the transactions by simply scanning the QR Code with their mobile phone [12].

In Indonesia, the adoption of the QR Code is considered as new technology. So, this study approaches the generation who was born to exposed to the changes of technology, namely generation Z in Bandung. Generation Z is the largest populations of human category and also the youngest age range in the world today. Generation Z refer to individuals who were born in the mid-1990s until the mid 2000s [49]. This generation is commonly known as the most highly tech-savvy, easy to swallow online products and has more convenient access to electronic devices, and broad information on the internet [24, 49]. In addition, according to the survey conducted by [47], generation Z is leading with their usage of mobile payment to perform transactions of 53% compared to other

generations. They are more likely to seek mobile payment platforms that seen as the best platform to use. To match this study, based from the website of [5], the population in Bandung area based on the age group from 15-19 is 220,024, age group of 20-24 is 258,830, and 25-29 age group is 228,917. The data indicates that the population of generation Z in Bandung is much denser than the other age range.

However, not all community has already adapted to the condition where they are most likely not 100% believe in the concept and mechanism of the digital wallet [42]. In addition, even though there are already many research about the adoption for digital payments, mobile payments, and mobile banking, a study of the use of the QR Code feature on mobile payments in conducting offline merchant transactions has never been done before, especially to evaluate the generation Z in Bandung area. Therefore, this study will focus on looking at the factors that contribute to the behavior to adopt the technology model and customer intention to use the mobile payment by using the Extended UTAUT2 model.

2. Literature Review

2.1. Mobile Payment With QR Code Feature

Mobile payment defines as a payment system to transferring amounts of cash to pay for products or service with the assistance of portable electronic devices such as tablets or mobile phone [16, 18, 23]. According to [13], mobile payment can be used in offline merchants (physical stores) or online merchants through the mobile phone. There are types of mobile payment features provided, one example is the QR Code feature. QR Code payments are now available to use at purchasing through merchants around the world and regarded as the foundation for the future of cashless payments in the digital age [15]. QR Code provides both consumers and merchants with an affordable and convenient payment service with no compelling reasons to issue physical cards, tokens or Electronic Data Capture (EDC) machines. QR Code also offers more cost-effective payment system to be implemented in F&B and retail outlets merchants compare to NFC that still requires unique stickers that can be identified by the EDC machine camera [15].

2.2. Generation Z

Being adopted from [49], the term Generation Z refers to a group of individuals who were born between the mid-1990s until the mid-2000s. These generations born after

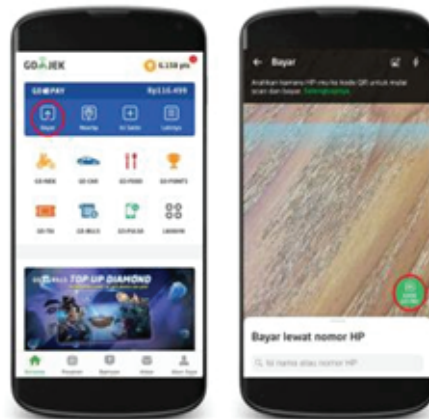


Figure 1: Example of Payment Using QR Code

Generation Y and the grandchildren of the Baby Boomers and raised by Generation X [24]. These generation are known to have common traits and characteristics; one example is that they are commonly dependent on technology [23]. This generation is commonly known as the most highly tech-savvy, easy to swallow online products and has more convenient access to electronic devices, and broad information on the internet [24, 49]. This happens because Generation Z is the first generation growing up with technology and is known to be more inclined to go fully digital [33].

2.3. Extended Unified Theory of Acceptance and Use of Technology 2 (Extended UTAUT2)

UTAUT is a theory that was proposed by [43] and was later refined to be UTAUT2 by [44]. The objective of UTAUT2 to produce the extensions of the prior theory, UTAUT, to primarily focus more to the consumer use context rather than the organizational context to identify the adoption intentions of an IT/IS. This suggests that UTAUT2 are essential to measuring the validity of UTAUT in consumer context compared to the original one [43].

UTAUT2 is a complex model where [43] has added more constructs compared from the previous UTAUT, which are, Hedonic Motivation (HM), Price Value (PV), and Habit (H) while the prior model already carried four factors which are Performance Expectancy (PE), Effort Expectancy (EE), Social Influence (SI) and Facilitating Conditions (FC). This shows that the UTAUT2 model can be able to explain even more comprehensively to the Behavioral Intention (BI) and the Actual Use (AU) by consumers. However, the model in this study will incorporate from the proposed model from [35] in his prior study as entitled Exploring Mobile Wallet Adoption in Indonesia Using UTAUT2 and [40] in his

prior study as entitled Extending UTAUT2 To Explore Consumer Adoption Of Mobile Payments. The study stated that the possible extensions of UTAUT2 model that can explain the mobile payment context are Perceived Risk (PR) and Trust (T) and thus affecting the behavioral intention of the usage of mobile payment. Also, this study will examine until the variable of actual use.

2.3.1. Performance Expectancy towards Behavioral Intention

Performance expectancy defined as the degree where the individuals believe that using technology will contribute benefits in regards to perform the activities. In the context of mobile payment, individuals perceive that using technology to make payments will enhance their performance in doing such purchasing transactions [43].

H₁: Performance Expectancy (PE) positively influence consumer's behavioral intention (BI) to use QR Code in mobile payment to made transactions in offline merchants

2.3.2. Effort Expectancy towards Behavioral Intention

Effort expectancy defined as the degree of ease related to the use of specific technology. In the context of mobile payment, this factor explains where individuals are free from effort and easy to understand the use of certain types of mobile payment [44].

H₂: Effort Expectancy (EE) positively influence consumer's behavioral intention (BI) to use QR Code in mobile payment to made transactions in offline merchants

2.3.3. Social Influence towards Behavioral Intention

Social influence defined as the degree which an individual valued the opinions of their significant others such as families, relatives, or friends to use a particular technology. This individual can be influenced by their significant others to use a specific type of mobile payment [43]

H₃: Social Influence (SI) positively influence consumer's behavioral intention (BI) to use QR Code in mobile payment to made transactions in offline merchants

2.3.4. Facilitating Conditions towards Behavioral Intention and Actual Use

Facilitating conditions defined as the degree to which an individual's perception to have adequate resources or access and supporting environment in order to acquire effective adoption of a technology [43]. In consumer use context in UTAUT2, facilitation is seen to be available and spread widely to each consumer and can have a different level of access between each technology available in the market. If an individual has more access to one facilitation, it is more likely to have a higher intention to use one technology. Thus, UTAUT2 decided to link facilitating conditions to both the behavioral intention and the actual use. Facilitating conditions also have a link between the moderating variables such as age, gender and experience [44].

H_{4a} : Facilitating Conditions (FC) positively influence consumer's behavioral intention (BI) to use QR Code in mobile payment to made transactions in offline merchants

H_{4b} : Facilitating Conditions (FC) positively influence consumer's adoption (AU) to use QR Code in mobile payment to made transactions in offline merchants

2.3.5. Hedonic Motivation towards Behavioral Intention

Hedonic Motivation is defined as the feeling of enjoyment, fun, or pleasure when an individual use a technology [44]. The findings from [11] shows that if an individual uses a technology for its own sake, it is more likely to be engaged with the essence of enjoyment and fun. This individual also pays more attention to a novelty in a technology, where the novelty itself contributes to the hedonic motivation effect. In the consumer use context, this factor is found to be important in measuring technology acceptance and actual use.

H_5 : Hedonic Motivation (HM) positively influence consumer's behavioral intention (BI) to use QR Code in mobile payment to made transactions in offline merchants

2.3.6. Price Value towards Behavioral Intention

Price value explains where the individual sees the technology based on its price and usefulness that further have a significant impact on technology use. In the context of using mobile payment, the price value will be positive if the individual finds that using a particular technology at a specific price, will give them benefits as much as the prices

applied to its technology. Adding this construct will act as a predictor to behavioral intention in using a technology [44].

H₆: Price Value (PV) positively influence consumer's behavioral intention (BI) to use QR Code in mobile payment to made transactions in offline merchants

2.3.7. Habit towards Behavioral Intention and Actual Use

Habit explains where the individual performs a behavior towards technology, and they use it automatically. It is also linked to experience, where experience in the passage of chronological time can form different levels of habit or in another word; the experience can reflect the results of habit. When an individual tends to perform actions towards technology, and they have a specific period, it will produce one individual habit in performing those actions [32, 44]. Thus, habit links to two variables, which are,

H_{7a}: Habit (HT) positively influence consumer's behavioral intention (BI) to use QR Code in mobile payment to made transactions in offline merchants

H_{7b}: Habit (HT) positively influence consumer's adoption (AU) to use QR Code in mobile payment to made transactions in offline merchants

2.3.8. Trust towards Behavioral Intention

Trust is a subjective belief or an expectation of an individual where other parties will meet the expectation and do not perform opportunistically in a situation, mainly in financial transactions area [19]. In the context of mobile payment, trust is considered to be highly relevant to its context because it involves transactions of money using electronic devices. Other reasons include where mobile payment systems are related to its novelty, tangled environment, and involvement of its mobile network operators [39]. As trust is found to be significant to behavioral intention [40], this study will extend to the model of UTAUT2 and put trust as the variables.

H₈: Trust (T) positively influence consumer's behavioral intention (BI) to use QR Code in mobile payment to made transactions in offline merchants

2.3.9. Perceived Risk towards Behavioral Intention

Perceived Risk comes from an individual's feeling of uncertainty towards specific behavior due to the fear of something unexpected and undesirable [40]. In the context of mobile payment, it indicates from the prior study from [27] that the consumer will

experience the vulnerability to security violations regarding the use of such wireless communications infrastructure. If the perceived risk is high, the individual will more likely choose not to adopt the mobile payment and switch it back to the traditional mode of purchasing transactions [37].

H₉: Perceived Risk (PR) negatively influence consumer's behavioral intention (BI) to use QR Code in mobile payment to made transactions in offline merchants

2.3.10. Behavioral Intention towards Actual Use

Behavioral intention is where the individual's behavior affects the actual use behavior of one's technology [43]. Several findings from [3, 26, 44] found that the behavioral intention could significantly affect the actual use behavior of technology. Thus, according to the prior study, this study will formulate the following hypothesis for behavioral intention,

H₁₀: Behavioral Intention (BI) positively influence consumer's adoption (AU) to use QR Code in mobile payment to made transactions in offline merchants

2.3.11. Moderating Variables

In the UTAUT2 model, generated by [44], the method posits three moderating variables, which are, age, gender, and experience that can help the factors to influence the behavioral intention or actual usage behavior. However, this study will not apply any of these moderating variables to see the direct effect of ones variable to another without partially supported or influenced by the moderating variables. In regards to the findings from Yuan & Kanthawala (2015), found that the moderating variables of gender, age and, experience were not significant to support the variables to influence the behavioral intention to use.

2.4. Conceptual Framework

Based on the several works of literature reviewed, this study will propose a conceptual framework drawn below in Figure 2,

3. Methodology

This research will use a quantitative method, where the researcher spreads a questionnaire containing the socio-demography variable and also the level of acceptance

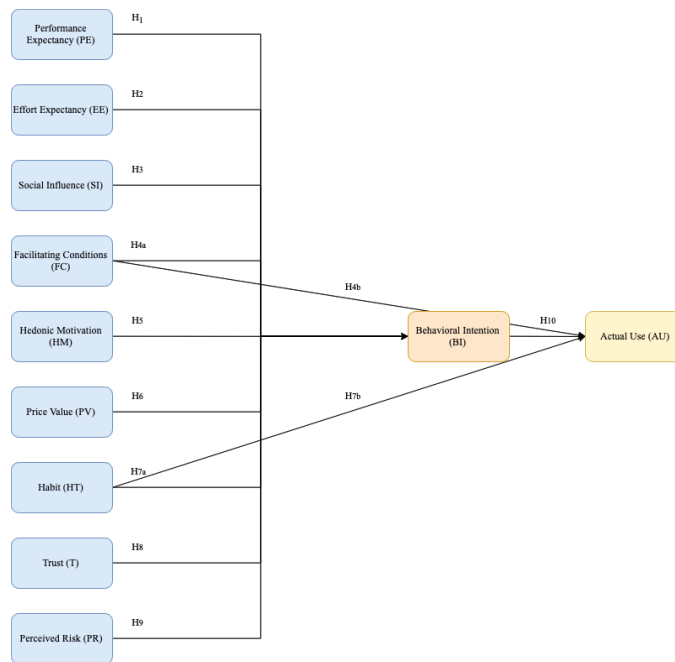


Figure 2: Conceptual Framework

variable with 32 questions. The questionnaires use 7-Point Likert Scale ranging from 1: “Strongly Disagree” and 7: “Strongly Agree” to get the responses along a continuum of possible responses from the participants 34. As for question number 32, for Actual Use variable, the answer will be using 5-point interval scales which are never use, once in three months (rarely), one or two times a month (occasionally), one or two times in 2-3 days (often), many times per day (very often). The collected data will be processed with PLS-SEM method using Smart PLS 3.0. in [48] suggest that the appropriate sample size for using PLS-SEM starts from 100 to 200. So, the sample size for this study will be 200 respondents.

4. Discussion

The data generated from the set of questionnaires were 200 respondents spread around Bandung area with the age group between 17 and 24 years. The respondents have used the QR Code feature that is used to perform transaction on offline merchants. The number of respondents obtained are qualified with the specified minimum sample requirements that covers the generation Z in Bandung area.

4.1. Descriptive Analysis

The descriptive analysis will cover the explanation of socio-demography of the respondents. It presents in Table 1 below,

TABLE 1: Socio-Demography Information

Demographic		Frequency	Percent (%)
Gender	Male	86	43%
	Female	114	57%
Age	17	18	9%
	18	31	16%
	19	30	15%
	20	37	19%
	21	56	28%
	22	16	8%
	23	8	4%
	24	4	2%
Current Education Level	Senior High School	31	16%
	Diploma	11	6%
	Bachelor Degree	146	73%
	Master Degree	2	1%
	Employee	10	5%
Used the QR Code Feature in Mobile Payment to Perform Transactions in Offline Merchants	Yes	200	100%
	No	-	-

4.2. Partial Least Square Structural Equation Modelling (PLS-SEM)

4.2.1. Descriptive Statistics

From the table given above, the highest mean can be found in the indicators of EE3 with 6.440 that represents the respondents’ understanding of “Easiness in using the technology”. The high mean results indicates that most of the respondents do not feel difficult to use Go-Pay with the QR Code feature. While the lowest mean can be found in the indicators of AU with 3.465 that represents the respondents’ understanding of

TABLE 2: Descriptive Statistics

Variables	Indicators	Mean	Std. Deviation	Min	Max
PE	PE1	5.880	1.066	2.000	7.000
	PE2	5.990	1.145	2.000	7.000
	PE3	5.335	1.301	1.000	7.000
EE	EE1	6.355	0.883	4.000	7.000
	EE2	6.305	0.856	3.000	7.000
	EE3	6.440	0.766	4.000	7.000
	EE4	6.260	0.901	3.000	7.000
SI	SI1	4.400	1.446	1.000	7.000
	SI2	4.505	1.556	1.000	7.000
	SI3	4.495	1.449	1.000	7.000
FC	FC1	5.970	1.034	1.000	7.000
	FC2	6.195	0.898	2.000	7.000
	FC3	6.120	0.952	2.000	7.000
HM	HM1	5.835	1.135	2.000	7.000
	HM2	5.885	1.163	1.000	7.000
	HM3	5.400	1.360	1.000	7.000
PV	PV1	6.145	1.065	1.000	7.000
	PV2	5.900	1.015	2.000	7.000
	PV3	5.870	0.966	2.000	7.000
HT	HT1	4.695	1.487	1.000	7.000
	HT2	3.635	1.809	1.000	7.000
	HT3	3.720	1.764	1.000	7.000
T	T1	5.915	0.999	2.000	7.000
	T2	5.780	1.092	1.000	7.000
	T3	5.765	1.053	2.000	7.000
PR	PR1	3.770	1.660	1.000	7.000
	PR2	3.645	1.624	1.000	7.000
	PR3	3.705	1.720	1.000	7.000
BI	BI1	5.490	1.058	2.000	7.000
	BI2	5.140	1.233	1.000	7.000
	BI3	5.070	1.373	1.000	7.000
AU	AU	3.465	0.565	3.000	5.000

“Usage frequency in using the technology”. The low mean results indicates that the frequency in using the QR Code feature in mobile payment is still in low adoption.

The maximum and minimum values in Table 2 above show the highest and lowest values that respondents choose from each question indicator. From the table, the highest value for the variables PE, EE, SI, FC, HM, PV, HT, T and PR is 7, where the value of 7 is the highest value of the 7-point Likert Scale in the questionnaire set. Then, the highest value for the AU variable is 5, where the value of 5 is also the highest value of the 5-point Interval Scale in the questionnaire set. While the minimum value in each indicators show to be different from each indicators.

4.2.2. Outer Model (Measurement Model)

Indicator Reliability

Indicator reliability can be measured by seeing the outer loadings for each indicators. The minimum value of the indicator reliability should be 0.70 or higher [48] to be acceptable, or between 0.40 to 0.70 are said to be adequate [22]. The results of the indicator reliability can be obtained through performing the PLS Algorithm in Smart PLS 3.0 by looking at the “Outer Loadings” of the model. Based on the result shown in Table 3, all of the indicators are reliable because it has fulfilled the minimum value of indicator reliability.

Internal Consistency Reliability

The internal consistency reliability were conducted to evaluate the constructs measures are reliable or not [20]. To determine the reliability, Cronbach's alpha and composite reliability were both considered because the Cronbach's alpha represent the lower bound of the reliability and the composite reliability represent the upper bound of the reliability. The requirement for composite reliability values should be 0.70 or higher, but 0.6 or higher for exploratory research to be acceptable [6]. From Table 4, the result shows that all of the variables has exceed the minimum value of 0.70. It means that each constructs has high level of internal consistency reliability and said to be adequate to be assessed further.

Convergent Validity

To measure the convergent validity, it needs to be underlined that the common rule of thumb is the constructs to have minimum value of outer loadings of 0.708. Based

TABLE 3: Indicator Reliability

Variables	Indicator	Outer Loadings	Indicator Reliability	Conclusion
PE	PE1	0.859	0.737881	Reliable
	PE2	0.791	0.625681	Reliable
	PE3	0.855	0.731025	Reliable
EE	EE1	0.876	0.767376	Reliable
	EE2	0.865	0.748225	Reliable
	EE3	0.907	0.822649	Reliable
	EE4	0.858	0.736164	Reliable
SI	SI1	0.871	0.758641	Reliable
	SI2	0.891	0.793881	Reliable
	SI3	0.918	0.842724	Reliable
FC	FC1	0.858	0.736164	Reliable
	FC2	0.883	0.779689	Reliable
	FC3	0.869	0.755161	Reliable
HM	HM1	0.922	0.850084	Reliable
	HM2	0.918	0.842724	Reliable
	HM3	0.890	0.7921	Reliable
PV	PV1	0.869	0.755161	Reliable
	PV2	0.842	0.708964	Reliable
	PV3	0.883	0.779689	Reliable
HT	HT1	0.873	0.762129	Reliable
	HT2	0.933	0.870489	Reliable
	HT3	0.909	0.826281	Reliable
T	T1	0.890	0.7921	Reliable
	T2	0.906	0.820836	Reliable
	T3	0.875	0.765625	Reliable
PR	PR1	0.869	0.755161	Reliable
	PR2	0.863	0.744769	Reliable
	PR3	0.867	0.751689	Reliable
BI	BI1	0.892	0.795664	Reliable
	BI2	0.916	0.839056	Reliable
	BI3	0.928	0.861184	Reliable
AU	AU	1.000	2.000	Reliable

on Table 5, all of the variables has already acquired the requirement. Further, the convergent validity is measured by seeing the Average Variance Extracted (AVE) to

TABLE 4: Internal Consistency Reliability

Variable	Cronbach's Alpha (α)	Composite Reliability (CR)	Conclusion
PE	0.784	0.874	Reliable
EE	0.899	0.930	Reliable
FC	0.876	0.924	Reliable
SI	0.840	0.903	Reliable
HM	0.896	0.935	Reliable
PV	0.833	0.899	Reliable
HT	0.890	0.932	Reliable
T	0.869	0.920	Reliable
PR	0.834	0.900	Reliable
BI	0.899	0.937	Reliable
AU	1.000	1.000	Reliable

the extent of how much the construct can explain more than half of the variance of its indicators. According to [6], the minimum AVE value should 0.50 or above. The result shows that all of the variables are above 0.50, as stated below in Table 5,

TABLE 5: Convergent Validity

Variables	Average Variance Extracted (AVE)	Conclusion
PE	0.698	Valid
EE	0.769	Valid
FC	0.802	Valid
SI	0.757	Valid
HM	0.828	Valid
PV	0.748	Valid
HT	0.820	Valid
T	0.793	Valid
PR	0.750	Valid
BI	0.832	Valid
AU	1.000	Valid

Discriminant Validity

Discriminant validity explains the extent to which a construct can be completely different from other constructs. By identifying discriminant validity, it can be ascertained whether

a construct can explain something different that is not represented by another construct in a model. The method used for assess the discriminant validity is by using the Fornell-Larcker criterion [17] and the cross-loading factors. As seen from both Table 6 and Table 7, all of the square root of AVE are greater than the correlation among all latent variables. Also, the outer loadings of each indicators are greater than any of its cross-loadings on other constructs. Thus, it can be concluded that all of the variables has passed the discriminant validity test.

TABLE 6: Discriminant Validity - Fornell-Larcker Criterion

	PE	EE	SI	FC	HM	PV	HT	T	PR	BI	AU
PE	0.836										
EE	0.529	0.877									
SI	0.387	0.168	0.896								
FC	0.460	0.568	0.147	0.870							
HM	0.584	0.414	0.378	0.409	0.910						
PV	0.440	0.359	0.275	0.310	0.410	0.865					
HT	0.458	0.177	0.551	0.198	0.379	0.249	0.905				
T	0.409	0.481	0.318	0.379	0.520	0.511	0.341	0.890			
PR	0.166	-0.009	0.294	0.059	0.175	0.028	0.398	0.034	0.866		
BI	0.602	0.404	0.471	0.438	0.571	0.428	0.685	0.526	0.208	0.912	
AU	0.238	0.187	0.172	0.152	0.165	0.211	0.223	0.235	-0.016	0.324	1.000

4.2.3. Inner Model (Structural Model)

To test the hypotheses that were built before in the conceptual framework, the research will evaluate the model quality such as the path coefficients, path significant (p-value or t-statistics) and coefficient of determination (R2). The t-value of path significance should be larger than 1.96 for it to be significantly affect the endogenous variable. To determine the R2, the acceptable value are 0.75, 0.50 and 0.25 described as substantial, moderate and weak, respectively.

From Table 8, it presents that there are five variables that show a significant value to its dependent variables because the t-statistics has exceed the minimum value of 1.96. These variables are mentioned from the most significant to the least significant, which are HT BI, BI AU, FC BI, HM BI, and PE BI, respectively. Habit shows to have the highest path coefficient (β) as an independent variable towards Behavioral Intention with the value of 0.479. It means that every increase in Habit in 1 point will increase behavioral intention by 0.479. The value of 0.479 also shows that the variable of Habit

TABLE 7: Discriminant Validity - Cross Loadings

	PE	EE	SI	FC	HM	PV	HT	T	PR	BI	AU
PE1	0.859	0,492	0,251	0,434	0,496	0,385	0,363	0,365	0,064	0,527	0,168
PE2	0.791	0,484	0,280	0,384	0,475	0,378	0,288	0,343	0,138	0,454	0,201
PE3	0.855	0,357	0,436	0,337	0,494	0,343	0,486	0,319	0,215	0,523	0,230
EE1	0,464	0.876	0,137	0,513	0,353	0,305	0,152	0,421	0,010	0,372	0,180
EE2	0,479	0.865	0,120	0,478	0,304	0,263	0,206	0,363	0,028	0,329	0,183
EE3	0,450	0.907	0,181	0,458	0,401	0,349	0,109	0,454	-0,057	0,363	0,186
EE4	0,462	0.858	0,149	0,542	0,391	0,336	0,158	0,446	-0,012	0,350	0,106
SI1	0,362	0,186	0.871	0,098	0,291	0,282	0,444	0,319	0,195	0,420	0,140
SI2	0,306	0,119	0.897	0,125	0,339	0,213	0,506	0,271	0,309	0,396	0,137
SI3	0,368	0,145	0.918	0,169	0,384	0,243	0,529	0,265	0,288	0,446	0,183
FC1	0,363	0,419	0,111	0.858	0,286	0,230	0,152	0,274	0,028	0,362	0,144
FC2	0,425	0,606	0,106	0.883	0,350	0,291	0,169	0,376	0,057	0,350	0,147
FC3	0,411	0,464	0,161	0.869	0,423	0,286	0,194	0,338	0,067	0,426	0,110
HM1	0,535	0,385	0,317	0,387	0.922	0,368	0,322	0,450	0,135	0,493	0,135
HM2	0,581	0,467	0,346	0,419	0.918	0,395	0,349	0,522	0,094	0,535	0,173
HM3	0,478	0,278	0,367	0,312	0.890	0,355	0,362	0,444	0,248	0,527	0,142
PV1	0,308	0,230	0,228	0,196	0,249	0.869	0,144	0,394	0,023	0,333	0,137
PV2	0,382	0,259	0,292	0,229	0,317	0.842	0,248	0,393	0,028	0,337	0,203
PV3	0,437	0,414	0,205	0,356	0,467	0.883	0,247	0,520	0,021	0,428	0,203
HT1	0,464	0,263	0,444	0,242	0,357	0,239	0.873	0,320	0,306	0,686	0,258
HT2	0,381	0,084	0,518	0,125	0,308	0,220	0.933	0,277	0,391	0,572	0,176
HT3	0,383	0,108	0,540	0,156	0,358	0,213	0.909	0,324	0,394	0,580	0,156
T1	0,374	0,448	0,231	0,319	0,448	0,425	0,264	0.890	0,008	0,436	0,194
T2	0,328	0,443	0,317	0,349	0,513	0,518	0,312	0.906	0,023	0,492	0,215
T3	0,392	0,396	0,296	0,342	0,425	0,417	0,331	0.875	0,059	0,473	0,217
PR1	0,121	-0,030	0,259	0,022	0,140	0,021	0,390	0,022	0.869	0,185	-0,067
PR2	0,164	0,034	0,241	0,082	0,169	-0,013	0,357	0,028	0.863	0,191	0,033
PR3	0,145	-0,033	0,267	0,046	0,144	0,071	0,280	0,040	0.867	0,161	-0,008
BI1	0,533	0,397	0,378	0,451	0,537	0,368	0,556	0,513	0,136	0.892	0,288
BI2	0,524	0,351	0,430	0,381	0,484	0,410	0,625	0,500	0,165	0.916	0,352
BI3	0,587	0,359	0,476	0,370	0,540	0,393	0,688	0,429	0,263	0.928	0,248
AU	0,238	0,187	0,172	0,152	0,165	0,211	0,223	0,235	-0,016	0,324	1.000

has positive effect to behavioral intention. The second highest path coefficient (β) is Behavioral Intention towards Actual Use with the value of 0.317. It means that every increase in Behavioral Intention in 1 point will increase actual use by 0.317. The value of

TABLE 8: Inner Model (Structural Model)

Structural Path	Path Coefficients (β)	T-Statistics	P Values
Performance Expectancy Behavioral Intention	0.139	2.055	0.040
Effort Expectancy Behavioral Intention	0.013	0.217	0.829
Social Influence Behavioral Intention	0.031	0.491	0.624
Facilitating Conditions Behavioral Intention	0.141	2.613	0.009
Facilitating Conditions Actual Use	0.013	0.191	0.849
Hedonic Motivation Behavioral Intention	0.149	2.490	0.013
Price Value Behavioral Intention	0.067	1.020	0.308
Habit Behavioral Intention	0.479	8.246	0.000
Habit Actual Use	0.003	0.032	0.975
Trust Behavioral Intention	0.127	1.770	0.077
Perceived Risk Behavioral Intention	-0.057	1.403	0.161
Behavioral Intention Actual Use	0.317	3.292	0.001

0.317 also shows that the variable of Behavioral Intention has positive effect to actual use. Facilitating Conditions has a path coefficient (β) towards Behavioral Intention of 0.141. It means that every increase in Facilitating Conditions in 1 point will increase the behavioral intention by 0.141. The value of 0.141 also shows that the variable of Facilitating Conditions has positive effect to actual use. Hedonic Motivation has a path coefficient (β) towards Behavioral Intention of 0.149. It means that every increase in Hedonic Motivation in 1 point will increase the behavioral intention by 0.149. The value of 0.149 also shows that the variable of Hedonic Motivation has positive effect on behavioral intention. Performance Expectancy has a path coefficient (β) towards behavioral intention of 0.139. It means that every increase in Performance Expectancy in 1 point will increase the behavioral intention by 0.139. The value of 0.139 also shows that the variable of Performance Expectancy has positive effect on behavioral intention.

However, the remaining variables of Effort Expectancy, Social Influence, Facilitating Conditions, Price Value, Trust, and Perceived Risk towards behavioral intention has found to be not significant towards Behavioral Intention. Also, the variables of Habit

also found to be insignificant towards Actual Use. These all happen because the t-statistics is below 1.96. So, the path coefficients (β) in these variables are not applicable to represent the sample of QR Code users in Bandung.

Besides path coefficient (β), the R^2 value was also calculated to explain the variance of each endogenous variables. R^2 value shows how much variance can be explained by exogenous variables (independent variables) to endogenous variables to endogenous variables (dependent variables). The minimum standards for the measurement, according to [48], are 0.75, 0.50 and 0.25 described as substantial, moderate and weak, respectively. Below are present the results of the R^2 square from the model testing in Table 4.8,

TABLE 9: R^2 Square

Variables	R2 Square
Behavioral Intention (BI)	0.662
Actual Use (AU)	0.106

The result of the procedure shows that the R^2 square of Behavioral Intention (BI) is 0.662 or 66.2% and considered as moderate. It means that the amount of 66.2% is the ability of the independent variable (PE, EE, FC, SI, HM, PV, HT, T, and PR) to explain the dependent variable moderately. For the variable of Actual Use (AU), the R^2 square shows the amount of 0.106 or 10.6% and considered as weak. It means that the amount of 12% is the ability of the independent variable (FC, HT, and BI) to explain the dependent variable weakly.

4.3. Hypothesis Testing

The results of this study indicate that Performance Expectancy has an effect where the more users feel that using the QR Code feature in mobile payments is useful and increases their productivity, the more users have the intention to adopt the feature in their daily life. Users also feel that by adopting the technology, they get benefits by doing the transaction. This findings is consistent with the prior study conducted by [36] about mobile payment adoption, [2] about mobile banking adoption, [26] about online banking adoption, and [3] about internet banking adoption.

Effort Expectancy is explained as the degree where users find it easy to use QR Code technology. In this finding, Effort Expectancy was found to be insignificant in this research towards behavioral intention to adopt the QR Code feature, which is contrary to prior findings, namely [35], [36], [2], and [3]. However, this finding is supported by [9]

TABLE 10: Hypothesis Testing

Hypotheses	Structural Path	Path Coefficients (β)	T-Statistics	Information	
				Results	Conclusion
H ₁	PE BI	0.139	2.055	Positively Significant	Accepted
H ₂	EE BI	0.013	0.217	Not Significant	Rejected
H ₃	SI BI	0.031	0.491	Not Significant	Rejected
H _{4a}	FC BI	0.141	2.613	Positively Significant	Accepted
H _{4b}	FC AU	0.013	0.191	Not Significant	Rejected
H ₅	HM BI	0.149	2.490	Positively Significant	Accepted
H ₆	PV BI	0.067	1.020	Not Significant	Rejected
H _{7a}	HT BI	0.479	8.246	Positively Significant	Accepted
H _{7b}	HT AU	0.003	0.032	Not Significant	Rejected
H ₈	T BI	0.127	1.770	Not Significant	Rejected
H ₉	PR BI	-0.057	1.403	Not Significant	Rejected
H ₁₀	BI AU	0.317	3.292	Positively Significant	Accepted

where it stated that the high level of mobile phone usage making the user too familiar with the rapid development of a technology. Although this finding shows insignificance relationship, the user feels that using the QR Code feature for mobile payment is easy to use, easy to learn, clear and understandable and easy to be mastered. This may occur when viewed from the respondents of this research, which is generation Z with the age range 17-24, that generation Z are commonly dependent on technology. They were born when technology was growing rapidly making them known as tech-savvy. They already feel familiar with the use of existing technologies. Thus, the level of difficulty to use the technology, namely Effort Expectancy, is not a consideration for the behavioral intention to use the QR Code feature in mobile payment.

The results of this study indicate that Social Influence does not make the users have the intention to adopt the QR Code feature in mobile payment. This even might happen because the user does not value opinions from the people closest to them to

use technology. The users feel that they could still have the intention to use QR Code without their significant others being really use the technology. This finding is in line from [2] about the adoption of mobile banking, where their environment seems less interested in recommending a technology to others.

The results of this research indicate that Facilitating Conditions can draw users to have the intention to adopt the technology. Users feel that they can easily have the access to knowledge to use the QR Code feature. Users also feel that the QR Code feature is compatible with their devices, making it easy for them to use it. Even though the access to resources to use the QR Code feature on mobile payment is not too high, compared to the other two factors above, it does not mean that the users do not acquire the intention to adopt the QR Code feature in mobile payment. This finding is in line with the prior study conducted by [26].

Facilitating Conditions to the actual adoption to use the QR Code features has found to be not significant. This might be happen because the Facilitating Conditions can only affect to actual use indirectly through behavioral intention and not directly to the actual usage. These users may still possessed the resources and knowledges needed, but it does not enough to make them to actually use the QR Code feature in mobile payment. This finding is in line with [9].

Hedonic Motivation is described as an encouragement in the form of enjoyment, fun, and feeling entertained when individuals use the technology. In this study, Hedonic Motivation affect the intention of users to use the QR Code feature in mobile payment. Users may feel that with the use of the QR Code feature in mobile payment, they get a novelty feeling due to performing their activities with the technology. This finding is in line with the prior study from [50] and [35]. The findings stated that the users intention is also triggered not only from the utilitarian motivation or its functionality, but also they pay attention to the user experience they get in performing the technology.

Price Value explains that when individuals using the QR Code in mobile payment at a specific price, it will give them benefits as much as the prices applied to its technology. The results from this study indicates that the Price Value do not affect the users' intention to adopt the QR Code feature in mobile payment. The users may feel that the technology give them a reasonable price, valuableness of the money spent and valuableness of the technology itself. On the contrary, the advantages given from using the mobile payment application, such as, cashbacks or promotions do not necessarily affect their interest to adopt the behavioral intention of the QR Code payment. This might be happen because the users may feel that the cashbacks or promotions are not something they really need. Also, this is in contrast with this finding where the users tend to use the technology for

their enjoyment and personal satisfaction. This is in line with the prior study from [35] and [36].

Habit is the extent to which the users use the QR Code feature gradually which then turns into a habit. In this study, habit is the most significant variable to influence the behavioral intention. This finding may occurred because the users of this study are generation Z, where they are known to be accustomed to being exposed to mobile payments and other technologies, making it their habit to use these technologies. This finding is consistent with the prior study conducted by [35] and [26] that also stated that habit was significantly influenced the behavioral intention to use the application.

In this study, habit has found to be not significant to the actual adoption to the use of QR Code feature. This can happen because Habit can only affect to actual use indirectly through behavioral intention and not directly to the actual usage. This is in line with the understanding of habit, where habit can be developed from the activities that performed gradually and produce a habit. However, the questionnaire also shows that the users have not really made the use of the QR Code feature in mobile payment as their habit. This might happen because the adoption of cashless users in Indonesia are still in the inception stage, where this stage shows how low the adoption of users to become cashless [41]. Moreover, the use of a mobile payment with a QR Code was actually implemented in 2018, when the industry of mobile payment application began to emerge and offered cashback or promotion to the public. This finding is not in line with the prior study conducted by [3] that stated otherwise.

In this study, Trust has found to be not significant to the behavioral intention to use the QR Code feature. Trust is explained where other parties will not behave opportunistically towards other individuals. It turns out that the trustworthiness of the technology, the commitments that the application offered and the application commitment to keep the customers' interest in using the technology do not necessarily affect the intention to adopt. This might happen because the users are exposed to multiple choices of other mobile payment system. Also, the users may not feel to have the urgency to trust one's particular technology because the users pay attention to novelty of technology. This study is in line with the prior research by [31] conducted in Indonesia that stated Trust do not significant to the behavioural intention.

In this study, Perceived Risk has found to be not significant to behavioral intention to adopt the technology. Perceived Risk is explained as the users' feeling of uncertainty towards something unexpected and undesirable. This finding shows that the users do not feel risky or have doubts about using the technology. This finding is in contrary with the prior study from [38] and [3].

Behavioral Intention is described as the users' intention and preferences to continue to use the QR Code feature. The results of this study indicate that the users are more likely to continue to use the technology in the future, have the willingness to use it in daily usage and to use it frequently. Although their intention is considered as high, but their frequency in actual use is still relatively average. Most of them use it occasionally or often. This finding is in line with [3] and [26] that also stated that behavioral intention affects actual use.

5. Conclusion and Recommendation

Based on the discussion presented in the previous chapter, the researcher has identified the factors that can influence the adoption of the QR Code feature in mobile payment to perform transactions in offline merchants. The findings shows that five out of twelve hypotheses proposed in this study are found to be accepted in regards to the usage the QR Code features in offline merchants. The accepted hypotheses are from the most significant, which are, HT BI, BI AU, FC BI, HM BI, and PE BI, respectively. All of these hypotheses are accepted because it fulfills the minimum requirement of t-statistics of more than 1.96 and has positive value of path coefficients, which is known as significant. The significant' variables represent the population of the object in this study, which is Generation Z in Bandung that already used the QR Code. However, the remaining seven hypotheses are rejected because the t-statistics are less than 1.96 which is not significant. The rejected hypotheses can only represent the respondents of this study, but not the population in this study.

This research adopted the theoretical foundations from [40], which is the new produced conceptual framework of Extended UTAUT2 that affect to behavioral intention and actual usage. This research has found out that seven out of twelve hypotheses were not significant to the behavioral intention and actual usage of the QR Code feature. Thus, a modified conceptual framework is provided as the result from this research as seen in Figure 3,

The modified conceptual framework can be used further as the reference or literature for those who want to do a research in regards the use of QR Code feature in offline merchants among Generation Z in Bandung.

Seeing that users greatly value their habit to use the QR Code feature, generation Z already understands that using technology that often changes is common to them. However, to increase their use of a technology, it is highly recommended for the industry QR Code from mobile-based payment to monitor current user interface provided for this

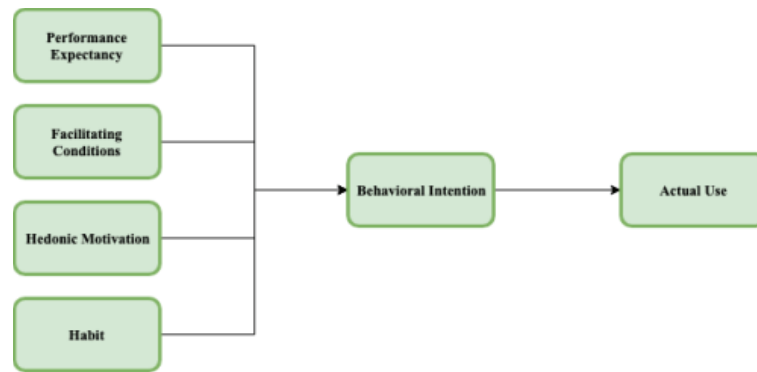


Figure 3: Modified Conceptual Framework

generation. Even though generation Z is adapt to fast-changing environment, but it also needs to be underlined to use the originality of it. This is due to the fact that this adoption accounts for until the Transitioning stage, considered as not so preferable stage, and also to reach more inclusive users that can understand the easiness and the basic of the user interface provided. Also, the researcher suggest the industry to reach even more inclusive merchants, not only in the F&B area, but also in the fashion industry, miscellaneous industry, goods and services industry, etc.. The users will be exposed to the QR Code payment gradually, making them interested to pay with this method.

The researcher also suggest to provide the access of facilitations to a more inclusive audience. The industry can consider making a special user interface for people with disabilities and also promoting knowledge of how to use it to the target market. In addition, access can be added to the users who are still under the age of 17 with age restrictions/parental guidance. Users who are still under the age of 17, can still get full access for all the features, but still under parents' consent so their financial can still be maintained. This is due to the limitations of generation Z that can use this feature. The researcher suggest the industry to provide it so all of the generation Z from 1995 to 2010 born can experience the use of the feature.

More programs and features should be added as the feeling of the experience is necessary in this research. The company should take a look into the target market of the users. Because, the users tend to see the application from their experiences (user-based experiences). This issue can be done by adding new programs in the application frequently. Also, other programs can also be added and implemented based on the user's important days, such as birthdays or marriage days. This can increase the sense of engagement in novelty, where this generation really pay their attention in this factors to use the application.

In regards to the user's productivity, the researcher suggest the industry to improve their system to increase the efficiency and effectivity. One example to implement is that the industry should work with more conventional banks. This action is done so that the balance in the application can be integrated with the bank. By doing this, the user does not need to go back and forth to open two applications just to make a top-up on the mobile application. This will increase their productivity, where they will not feel difficult and become accustomed to using the QR Code in mobile payment.

Since the variance that explains the Actual Use is considered as weak, future research can extend the variables or segmentations to further increasing significant value of the variance. So, this study can explain more than behavioral intention, but in an actual use contextual to actually see the usage of the QR Code frequently in daily life.

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