

DETERMINANTS OF USERS' INTENTION TO USE SMARTWATCH

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ABSTRACT

The rapid advancement of information and communication technologies in recent years has enabled the design and production of compact mobile devices with robust computing power. These advances have resulted in a plethora of wearable devices, including smartwatches. In particular, smartwatches are highly hyped in the information and communications technology industry for multiple functions that appeal to a broad range of user interests, fitness, health monitoring, and connect to smartphones to extend their functionality with the goal of simplifying users' lives, reducing distractions, and letting them experience the world without fussing with their phones. Despite these benefits and hypes, the smartwatch market has been declining continuously, and current sale estimates are still relatively low. Thus, this study aims to investigate the level of Malaysian individuals' intention to use a smartwatch and to identify and assess their impacts. Therefore, the study will assess the impacts of perceived usefulness, perceived ease of use, cost, privacy, and health risk on user's behavioural intention to use smartwatch. A total of 501 collected data was used to test the study hypotheses using structural equation modelling technique. The results revealed the high level of Malaysian individuals' intention to use a smartwatch. Moreover, perceived usefulness and perceived ease of use have a significant positive effect whereas cost, privacy, health risk did not negatively affect users' intention to use a smartwatch. The extended technology acceptance model explains 34 per cent of the variance in the intention to use a smartwatch. The findings of the study will help future researches and smartwatch developers by providing a guideline for the development of the smartwatch devices.

Keywords: *Wearable DEVICES, Smartwatch, Tam, User Adoption, Intention TO Use*

1. INTRODUCTION

The quick advancement of high-tech information and communication technologies (ICTs) lately has permitted the design and massive production of compacted mobile gadget with strong computing power. Subsequently, the sizes of these devices have reduced significantly so that individual can wear them in form of smart watches, smart glasses, or smart clothing rather than being carried, thus literally leading to wearable computing [1]. Lately, the popularity of wearable computing devices are on the rise and their usage is also increasingly extended. The technological

advancement in the wearable computing devices have yielded in a plethora of other wearable computing devices, including smartwatches. In particular, smartwatches are highly puffed up in information and communications technology industry for the multiple functions that interest to a variety of user's interests, fitness couple with health monitoring. Today's smartwatches connect to smartphones and extend their functionality with the goal of simplifying users' lives, reducing distractions, and letting them experience the world without fussing with their phones.

Interestingly, unlike the smartphone interface, smartwatches enable the wearer to have hands free that allowed unhindered access to information [2]. In spite of their benefits, smartwatches also have many challenges such as design [3, 4], product complexity [5], privacy, security, battery life, price, aesthetics, convenience, accessibility, and social acceptability [6-9]. These challenges need to be tackled to inspire adoption and use of these products. In addition, if smartwatch aesthetics challenges are not addressed, it will continue to face marketing challenges and affect the product's sale [10].

Furthermore, as smartwatches have been widely accepted as the 'next big thing' that will have a significant impact on our daily lives [11, 12]. However, the smartwatch market has been declining continuously. Report from International Data Corporation shows that the year-over-year market growth of smartwatch has declined to 51.6% [13]. Similarly, it was forecasted that 373 million smartwatches are unlikely to be sold before the end of 2020 [14]. Yet, the highest proliferation of wearable devices are currently smartwatches [12, 15-17] and their demand are increasingly estimated to be high [18]. However, the current sale estimates are still relatively low [15]. Furthermore, little is known about what impacts these differences in forecasts and sales and subsequently research is needed to comprehensively understand this gap of technology that is still at the early stages of product lifecycle and commercialisation [19]. In particular, the question of what drives the adoption of smartwatches remain unanswered and only few empirical researches have been conducted on user acceptance and behaviours concerning them [17, 19].

Furthermore, most studies on smartwatches have focused on overview descriptions, concepts, challenges, and technical aspects [20] and how they can be utilized for various purposes: healthcare, fitness tracking, biometric sensors, etc. [12, 21]. This implies that empirical studies on smartwatches have been more 'technology driven' instead of 'audience-centred'. Meanwhile, scholars have argued that attempts to understand the actual audience's perceptions and intentions should be pivotal, especially when IT innovation is in the early stages coupled with obscure knowledge of the needs of consumers [11, 12, 22]. Likewise, Wu, Wu, and

Chang (2016) add that it is essential to explore both technological and potential adopters' perceptions of smartwatches [20]. Therefore, this study intends to fill the aforementioned research gaps, to identify important factors affecting users' intention to use smartwatches (ITU SW), since scholars [10, 12] assert that the adoption of new products remains an important issue for researchers and practitioners. In doing so, the study attempts to answer the following research questions: What is the level of users' intention to use smartwatch? Do innovation characteristics affect users' intention to use smartwatch?.

Therefore, this study is expected to contribute to existing body of literatures in many aspects. First, the study intends to provide insights on users' intention to use smartwatches. Therefore, this study will reveal the influence of the three additional innovation characteristics on users' intention to use smartwatch. Thus, findings from the additional construct can be used by smartwatch developers to identify individuals' needful needs and could be considered in future designs of smartwatch devices so as to increase product sales. Secondly, the study will also link the factors of perceived usefulness (PU), perceived ease of use (PEOU), cost (CT), privacy (PRV), and health risk (HR), to the TAM theories. The above research question will be answered through the lens of the famous Technology adoption model with important additional constructs reviewed from the literature.

The remainder of the paper is structured as follows. Section 2 describe the past and current related work, section 3 discusses the theoretical background with the hypotheses and research model while section 5 presents the research methods. Finally, the paper ends with the discussion of findings, followed by limitation and future work suggestions.

2.1 Overview of Wearable devices

The advances in technology have yielded smaller components with better performance, which today makes it possible to fit computers into devices that could easily be worn on the body. Wearable Technology, wearable user interfaces or simply wearables could be seen as a subset of Internet of Things and is both a way to collect more data but also a way to provide the user with information faster and more convenient. The main purpose of wearable

devices is to simplify people's daily life and free up the arms [23]. Previously, wearable devices was mainly used in the field of military [24]. Nowadays, combined with aesthetics and fashion design, demand of wearable devices continues to grow in various fields [24]. They are used in many areas such as finance, education, gaming, enterprise, and healthcare and medicine.

2.1. Overview of Smartwatch

After several decades of technological development from the initial conception of strapping a computing device to one's wrist, the smartwatch has finally become a viable device that extends the functions of smartphones to a more intimate level [12, 25]. In the mobile device industry, smartwatches, have been widely accepted as the 'next big thing' that will have significant impact on our daily lives [11]. These devices can measure our heartbeat, energy consumed, steps taken and distance travelled [26]. Thus, this study defined smartwatch based on the definition of Cecchinato, Cox, and Bird as a wrist-worn device with computational power, that can connect to other devices via short range wireless connectivity; provides alert notifications; collects personal data through a range of sensors and stores them; and has an integrated clock [11, p.2136].

2.2 Individual Adoption Theories

Individual adoption and use of information technologies is one of the most mature streams of information system researches [27]. As today's technological products such as wearable devices and its related technologies are developing at a fast rate. Thus, a great effort has been made to develop a more reliable and comprehensive model that can explain and predict users' acceptance and use of information technology. The simplicity of TAM and the recent empirical support it has garnered has made TAM widely popular with information system researchers [12, 28]. Thus, this study chooses to use TAM as the basic theoretical model for investigating the potential user acceptance of the smartwatch product for a number of reasons. The first being that TAM has a simple and solid structure that can explain an individual's key considerations of technological product acceptance [20, 29]. As such, it has been proven to be quite reliable in the prediction of user acceptance of several new technology products and services. Additionally, researchers have indicated

that the main constructs employed in TAM (i.e. PU and PEOU) have been proven to be accepted as the most powerful variable for predicting technology usage intention and acceptance [27, 30] and they are fundamentally a subset of perceived innovation characteristics [20, 28].

Although not without criticism, researchers have indicated that the constructs of TAM alone cannot fully explain user's technology acceptance, thus many scholars have suggested that TAM needs to be extended with other constructs to provide a stronger theoretical model [31, 32]. As an example, the study of [12] and [15] extended TAM to study the adoption of smartwatch devices. In this research, we take the postulate that TAM needs to be completed to gather more efficiently the factors of smartwatch adoption and combine it with other constructs from the literature. Furthermore, this will provides a complete account of the causal mechanisms underlying the relationships as well as unique insights that cannot be obtained with just using the original TAM's constructs [19]. Finally, consumer acceptance of a new technology is a complicated phenomenon that needs an extended robust model [20, 33].

2.2.1 Empirical Studies Related Wearable Technologies

The popularity and significance of wearable technologies necessitated for the need to investigate the factors affecting wearable technology adoption. This section present some of the reviewed empirical studies related to the context of smartwatch adoption. Therefore, it is vital to extend this streams of research by acquiring the empirical result of users' intention to use smartwatch. [34] conducted a study in Malaysia from university students perspective to identify the influencing features of smartwatch acceptance by examining the consumer recognition towards the technology. They infer that these technologies have the convenient for online ticketing services and thus allows users to effortlessly acquire the information they need. Equally, [35] investigated smart product adoption from the perspective of consumer resistance studies. The authors identified PU, perceived novelty, perceived price, intrusiveness, privacy, dependence, and self-efficacy as key factors of why student resist smart product. They added that the slow pace of

consumer adoption of this innovation is a major disappointment for the smartwatch industry.

[36] combine three models of TTF, DOI, and new product adoption model to investigate the factors that affect user's intention to adopt smartwatch. They argued that smart device manufacturers have a strong interest to understand those factors and to explore different ways to increase users' intention. To examine the factors affecting the intention of non-users of smartwatches, [15] integrate TAM with visibility construct. A total of 226 responses were collected from Malaysian university student through self-administered questionnaire. The author asserts that currently, little is known about what impacts the relatively low sales of smartwatches. In another study on a smartwatch, [16] apply TAM to determine the influential psychological constructs of smartwatch users'.

Also, [12] carried out research to determine factors affecting consumers intention to accept smartwatches. They suggest that users' studies on smartwatches by academician is now at a very nascent stage and most of the studies did not use specific smart wearable devices as the subject of the study and concentrated on the specific function of the devices. Equally, [20] integrate three theories of UTAUT, TAM, and DOI to explore antecedents of consumer intention to accept smartwatch. The authors opined that certain factors of willingness, necessity, and social factors had not been explored adequately by past smartwatch studies. In a quantitative approach based on purposive sampling, [37] investigate how individual's innovativeness affects consumer smartwatch purchase. The author argued that there's still exists little scientific understanding on consumer's perceptions to accept or purchase smartwatch devices. Moreover, [19] determined the antecedents of wearable devices and its perceived value with regards to consumers'. They mention that only few studies were carried out on factors affecting user acceptance and behaviours' of smartwatches.

[10] examines the effect of user innovativeness and their relationship among innovative attributes and investigates their effects on user purchase intention of wearable devices. The author mentioned that little is known on why certain individuals are more likely to accept wearable devices. Likewise, [38] investigated Apple

smartwatch behavioural intention from the perspectives of users in Germany. They infer that smartwatch manufacturers need to emphasise the usefulness and enjoyment of their devices and tackle or reduce users' perceived risk of the device. [3] conducted a study to compare US and Indian consumers perceptions, preferences, attitudes, and behavioural intentions toward wearable trackers. The author mentioned that little research has been conducted on the relationship among the factors that affects consumers intention to use wearable devices. To find the factor that affect users purchase intention, [36] develops a rich framework that consists of design aesthetic, software, and hardware attributes. Results indicate that design aesthetics significantly influence smartwatch purchase intention. Equally, [39] investigated the vital factors affecting smartwatch adoption behaviours. They mentioned that in order to expand the smartwatch industry, more studies are needed to comprehend the factors that affect users intention to use smartwatch.

Several international recognised research firm conducted surveys to find the perception of consumers on smartwatch. For example, the Acquity Group survey 28000 consumers from 28 countries to investigated the purchase intent of five IoT products [40]. The study revealed that to ignite growth from user's adoption perspective within the next five years, the smartwatch manufacturers must offer a compelling customer value proposition, ensure a superior customer experience. Another investigation by Gartner shows that only 12%, 9% and 7% from the 9,592 online respondents of U.S., U.K., and Australia currently use smartwatch [41]. From the Malaysian region, the 2016 survey by the Digital Nations' revealed that only 7% of individuals currently use smartwatch from the 1018 surveyed responders [42].

From the literature reviewed, this study is distinct from many of the researches in a number of ways. This study differs from other studies as it utilised an extended model of TAM. By extending the model to include the influences of cost, privacy, and health risk, the model is designed to give a comprehensive perspective on adoption of information system studies. The model includes many of the antecedents used in the studies mentioned above in a more comprehensive perspective, making it possible to show the relative influences of the antecedents related to user

perceptions and personality traits. Also, these studies failed to include a more vivid product characteristics that can inform the smartwatch vendors on how they can increase the product sales so as to increase the diffusion of these devices among the individuals.

Although, there are few smartwatch studies that uses students' as their population samples', however these studies tend to only focus on student sample from only a particular faculty or departments, which tends to limits the generality of the findings. For example, the study of Chuah et al. (2016) uses students from only business department of Malaysian university and that of Liew et al. (2016) also collected data from only one Malaysian university. Still, a more holistic picture for the findings from these studies (i.e. Chuah et al., 2016 and; Liew et al., 2016) could not be obtained since they used fewer respondents of less than 250. Likewise, the reviews above have shown that most of the studies have not included product characteristics such as health risk, that could inform the smartwatch vendors on how they can further expand their market by knowing the risk associated with the usage of the devices. Thus, the examination of these unique smartwatch characteristic would enable the comprehension of the factors that may possibly influence the acceptance decisions.

A more closer look at the reviewed literature of smartwatch adoption studies shows that the mainstream of academic researchers have employed the TAM model for their study e.g. [12, 15, 16, 34, 43]. However, many of these researchers have deemed it necessary to extend TAM model by incorporating external variables such as perceived enjoyment, trust, and availability in order to improve the explanatory power of the model. In particular, smartwatch devices have different characteristics that can influence user adoption behaviour; therefore, it is important to identify appropriate external variables that can fully explain user's decision in adopting these products [44]. Prominent authors have inferred that incorporating context-specific constructs to the TAM model will make it more applicable to diverse technological contexts [45, 46]. As such, this study has identify those vital construct that could fully explain how to understand and increase the smartwatch diffusion among the potential consumers.

In spite of the fact that [15] and [34] studies do contribute to the some understanding of smartwatch adoption in Malaysian context, there is still a need to conduct more studies to cover important aspects related to individuals' cost, risk, and privacy, to see how these dimensions could shape the Malay individuals' behavioural intention to use smartwatch. Hence, this study will serve to narrow the research gap found by examining the determinants of smartwatch adoption from the extended TAM model. However, past studies failed to incorporate few of the most important innovation characteristics in a more holistic model. Therefore, the current study integrates these variables into one model and investigate their influence on the intention to use smartwatch.

3. Research Model and Hypotheses Development

Based on the literature reviewed, this study has identified various variables and dimension for the main constructs of the study, which are the relationship between innovation characteristics, and the dependent variable as shown in Figure 1.

3.1 Relationship Between Perceived Usefulness and Intention to Use

All the three theories stated that PU is positively related to user adoption behaviour. Various smartwatch studies had found the relationship between PU and intention to use [e.g. 10, 15, 32, 34, 43]. For instance, [34], found PU to influence Malaysian student acceptance of smartwatch, but [15] studies did not. Applied to wearable device context, PU relates to the benefits the consumer thinks he or she will take from the future use of a new product in terms of time saving, convenience, access to additional information [35] and improve efficiency and performance [43]. For example, users can use their smartwatch devices to easily upload their physical workout data, photos, and videos to smartphone. [19] infer that the productivity of users could be improved through various types of activities like communicating with their friend anytime, anywhere, organised tasks and meetings, among others. Smartwatches are more portable than smartphones, which enables users' to complete tasks more quickly and efficiently, which is, consequently, likely to increase consumers' intention. Hence, the higher the added value of the smartwatch over smartphones, the more the

consumer will be inclined to adopt it. Hence this study hypothesis that:

H1: Perceived usefulness will have a positive impact on the user's behavioural intention to use smartwatch.

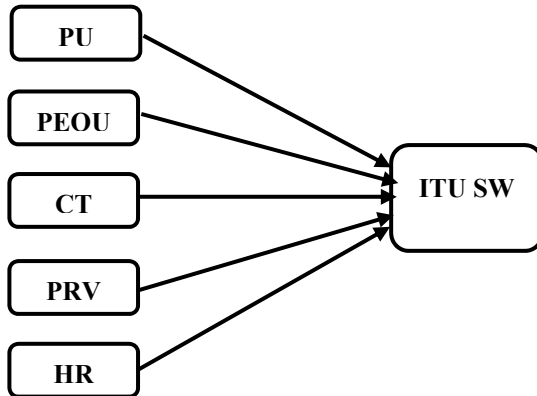


Figure 1

3.2 Relationship Between Perceived Ease of Use and Intention to Use

Many of the previous studies have state that PEOU is a significant determinant of behavioural intentions to use [47, 48]. TAM and UTAUT stated that both PEOU and EE were positively related to intention to use [27, 48]. This relationship was also reported from the studies of [15] on smartwatch acceptance. Prior researchers have found that individuals need more time to comprehend and use product that are more complex [49]. The complexity of smartwatch is reduced with easy to understand graphics that would enable the user to simply tap the graphics icon to trigger an action. For example, a user can simply tap a 'music' icon to directly listen to music that will enable the user to spend less time to understand and familiarise themselves with the operational procedures of the devices. Moreover, the operations of smartwatch devices are generally more complicated, since users are required to continuously wear them and use with other smart devices like smartphone at the same time. Thus, the easier a technology product is to use, the more that the user form a positive intention toward using that technology [48]. Therefore, the study hypothesise that:

H2: Perceived ease of use will have a positive impact on user's intention to use smartwatch.

3.3 Relationship Between Cost and Intention to Use

Cost is one of the inhibiting factors that hinders the acceptance of technology product [50, 51]. Before users decide to purchase a product, they usually compares the benefits and the cost of that product or service. If the perceived fee exceeds the perceived benefits, it is then viewed as expensive, and users will be less likely to adopt it [1]. As a result, the majority of the scholars suggested a negative association between the cost and intention to use while only few scholars hypothesises cost to positively affect users' intention behaviour. For example, the study of [16] on smartwatch, [1] on healthcare wearable devices and on m-commerce all hypothesis cost to negatively affect user's intention whereas those of [52] on media tablets and that of [35] on smartwatch hypothesises positively.

Subsequently, other studies found that users do not consider the cost of the product as long as the products bring benefits and value to the user [53, 54]. For example, the studies on wearable devices' of [10], on multimedia message service of ([55], and on smart meter by [53] all found cost not to affect users intention. Thus, in view of these contradicting findings, this study hypothesis that:

H3: Cost will have a negative impact on users intention to use smartwatch.

3.4 Relationship Between Privacy and Intention to Use

Privacy concern is one of the biggest challenges that hinder the used of new technology product especially those that involve the use and sharing of data. Users are generally concern about how their data were gathered secretly and constantly without their permission [56]. The fear over the misuse of personal data by the consumers have become one of the biggest challenges facing online businesses [57]. Researchers believe that these concerns will increase with the development of smart products [35, 58] and smart services [59, 60]. As wrist worn wearable devices interact daily with the end user and they encompasses ubiquity, invisibility, pervasiveness, and invasiveness as their characteristic [35, 58].

Furthermore, as smartwatches collect and store sensitive data, users' privacy could be

endangered since users do not have any control over what the service providers or firms might do with the data [38, 61]. Moreover, third parties and or hackers could intercept the data while users are syncing the data to other connected devices such as smartphone. Such concerns represent an increase in non-monetary effort, and will thus negatively impact the users' intention [62]. Firms should inform users about how collected information will be used prior to collection. Such information will allow consumers to assess better the risks of disclosing personal data [59] and might reduce the difficulties of users experienced in knowing where and how their information is stored and who is authorised to access and use them [35]. Thus, consumers who are concerned about their privacy might deter to use smartwatch. Previous studies have empirically shown that privacy has a negative impact on user adoption behaviour [53, 57, 63, 64]. Thus, this study hypothesis that:

H4: Privacy will have a negative impact on user intention to use smartwatch.

3.5 Relationship Between Health Risk and Intention to Use

Health risk is a physical risk that users might associate with the adoption of a new product [65] that could deter users intention to the use of the wearable devices. In the context of smart products, the use of smartwatch devices may increase such perceived risk. In a recent study conducted by McKinsey Global Institute, leading industry experts noted that security breaches in the wearable devices exposes consumers to significant health threats [66]. As the use of smartwatch devices requires the user to always attached it to ones' wrist and the continuous exchange of data between the smartwatch and connected devices, many specialists have pointed out that this could give rise to health hazards to the wearer [67]. As such, wearable devices' are suspected of producing radiation harmful to some part of the human body [68], and recently, the World Health Organisation cautioned that this kind of radiation could even cause cancer. Users are particularly concerned about exposure to electromagnetic radiation due to prolonged use of smartwatch devices [69-71].

Moreover, wearers are increasingly concerned about the accuracy and reliability of the data gathered from the smartwatch sensors [70, 72].

Such concerns about the hazards of smartwatch devices might lead users to associate them with a health risk related to their health and can delay potential user's intention [70, 73]. Prior studies found the relationship between users' health concern and intention to be negative. For instance, the work of [35] found health risk to affect consumers use of smart services. Thus, this study hypothesises that:

H5: Health risk will have a negative impact on user intention to use smartwatch

4.1 Research Methods and Design

A self-administered questionnaire was administered to the students of the designated five Malaysian research Universities. The study target student because they are more open to the adoption of new technology products [10, 74, 75] and represent the largest groups of modern technology users' [47, 76]. Furthermore, the use of students would provide a vivid description of the characteristics of the general population in Malaysia, thus would allow for the findings of the study to be generalisable [10]. The students are only selected for the study if they satisfied all the three inclusion criteria as written in the instruction part of the questionnaire. The study collected a total of 501 data through quota sampling, which is a non-probability sampling technique. All the measurement items were measured on a 7-Point Likert Scale with a multi scale items adjusted specifically to fit smartwatch context as shown in Table 1.

4.1.1 Common Method Bias

This study collected all its data from a single source for both the endogenous and exogenous variable, thus might prompt for the occurrence of common method bias [77]. Harman's Single Factor was used to check the response bias. The percentage of variance explained in this study is 24.436%, which is below the threshold value of 50%, indicating that there is no significant common method bias [78, 79].

4.1.2 Demographic Profiles of the Respondents

From the data, more than two third of the respondents were less than 30 years old, thus this shows that the majority of the respondents are in their young age. The gender analysis revealed that 70% of the respondents were female while the remaining were male. Also, for the ethnicity, more

than half of the populace are Malay, followed by Chinese with 23% while only 44 respondents are Indians with 7.9% representation. This finding revealed almost the true representative of the ethnicity of the Malaysian citizens.

4.2 Research Model Analysis

The study uses partial least square method to test the research model and analysed the data using SmartPLS version 3.2.8. The study follows a two stage process by assessing the measurement model and the analysis of the structural model. Measurement model extracts information on the validity and reliability of the items of the indicators whereas, structural model measures the relationship between the latent constructs besides testing the hypotheses of the study.

4.2.1 Measurement Model

The measured model was assessed through four criteria's. As the first stage of PLS analysis, measurement model was evaluated regarding their reliability (individual item reliability determined by the indicator's outer loadings and internal consistency reliability which is determined by composite reliability) and validity.

Individual item reliability is determined by indicator's outer loadings. Indicators with outer loadings below 0.708 are deleted [80, 81]. After running the PLS algorithm, all items values were above the threshold value and ranges from 0.710 to 0.983, thus none were deleted. Next, the internal consistency of the indicators items were assessed through convergent validity. Based on the recommended criteria as suggested by [80, 81], the composite reliability values of all constructs as shown in Table 1 are well above 0.70 which demonstrate high level of internal consistency reliability. Additionally, the convergent validity was evaluated through average variance extracted. An AVE value of 0.50 or higher indicates that the construct explains on average or more than half of the variance of its indicators as recommended by [80, 81]. As shown in Table 1, all of the AVE values were well above the required minimum level of 0.50. Thus, the measures of all the construct of the path model have high levels of convergent validity.

Finally, the discriminant validity was assessed through heterotrait-monotrait ratio of

correlations (HTMT) as the dominant past approached like Fornell-Larcker criterion (Fornell & Larcker, 1981) and cross loading (Chin, 1998) were inadequately sensitive to detect discriminant validity in the variance-based PLS-SEM as argued by [80]. The threshold values suggested in the literature are 0.85 [82, 83] whereas others suggest 0.90 [84, 85]. As shown in Table 1 all computations yielded values of below the threshold value of 0.85. In addition, bootstrapping procedure was run to check whether the HTMT inference criteria has been met. The result in the confidence intervals indicate that the values in the upper confidence interval limit were all below the values of one. Thus, the HTMT inference criteria indicates that all HTMT values are significantly difference from one. Henceforth, this provided support for the measures' discriminant validity [80, 81]. Overall, all measurement criteria had been met and provided support for the measures' reliability and validity.

4.2.2 Assessment of the Structural Model

After the measurement model were confirmed as reliable and valid, the analysis proceed to assess the structural model. Before testing the relationship between the latent variables, collinearity issues in the structural model need to be addressed to test the bias of the path coefficient in case the estimation involves significant level of collinearity among the predictor constructs. All the predictor constructs yielded the values of 1.067 to 1.363, thus satisfied the variance inflation factor (VIF) value of less than 5, which indicates that they is no collinearity problems among the constructs of the study.

Next, determine the significance of path coefficients, a nonparametric bootstrapping technique with 5000 re-sample [80] was applied to obtained t-value with the critical values for one tail with significance level of 5%. The results of the structural model estimation are summarised in Table 1. From the results, three out of the six hypotheses are supported.

From Table 1, the relationship of PU -> ITU, PEOU -> ITU, and were all significant. Hence, H1, and H2 were supported. However, the relationship between PC -> ITU, PP -> ITU, HR -> ITU, were all not significant. Next, the structural model's predictive accuracy was assessed through R² values. After the PLS algorithm was run, an R²

value of 0.336 was obtained, which indicate that 34% of variance can be explained in the intention to use smartwatch by the constructs of the study.

users’ intention to use smartwatch. This implies that higher PU will increase users’ intention to use smartwatch. This result is consistent with the TAM theory and existing literature [39, 47, 86], even though the result contrasted with some studies on smartwatch adoption. For example, the study of [15] on student smartwatch adoption and that of [20] on Taiwanese smartwatch acceptance. This results implies that students in Malaysia perceived smartwatch as a task-oriented smart device used for utilitarian usefulness that they can use to increase work productivity and

quickly accomplish certain tasks. Also, PEOU showed a positive impact on intention to use smartwatch and this implies PEOU is an essential factor of users’ intention to use smartwatch. This finding suggests that enhancing and improving individuals’ intention to use smartwatch is positively associated with the improvement of PEOU of smartwatch devices and is consistent with the TAM [47] and other past studies on wearable devices [e.g. 36].

Results from the unverified relationship of the study showed that cost, privacy, and health risk did not negatively affect users intention to use smartwatch. This implies that higher cost would not hinder users’ intention to use smartwatch and that potential users considered the cost of adopting the smartwatch devices as negligible when they compare the anticipated benefits arising from using the devices. Furthermore, result for the privacy entails that the increase of privacy of smartwatch devices will not lead to decrease in the intention to use smartwatch. Equally, the health risk findings implies that users’ perceived health risk perception would not deter their intention to use smartwatch devices.

5.1.1 Theoretical Contributions

This study proposed three additional innovation characteristics on the intention to use smartwatch context. These findings extended the scope of intention to use, especially in the context of smartwatch adoption. Therefore, the study added novelty to the TAM [47] by proposing cost, privacy, health risk as additional variables in the context of studies on smartwatch adoption. Thus, this study extended the scope of TAM research in terms of innovation characteristics by investigating the impact of these variables on the intention to use

Table 1: Hypothesis Result

Hypothesis	Beta	Std Error	t-Value	LL	UP	Decision
H1	0.264	0.049	5.370**	0.14	0.288	Supported
H2	0.224	0.046	4.863**	0.075	0.22	Supported
H3	-0.038	0.050	0.814	-0.116	0.04	Not supported
H4	-0.044	0.038	1.153	-0.096	0.029	Not supported
H5	-0.049	0.046	0.6	-0.105	0.043	Not supported

**p<0.05

5.1 Discussion of Study Findings

This study investigated the level and determinants of students’ intention to use smartwatch. Thus, the study examined the effects of perceived usefulness, perceived ease of use, cost, privacy, and health risk on user’s behavioural intention to use smartwatch. In view of these objectives, a research model was developed from the theory of TAM and existing literature. The study focused on five innovation characteristics (perceived usefulness, perceived ease of use, cost, privacy, and health risk) as determinants of students’ intention to use smartwatch due to their possible influence.

Based on the results, the studied verified that perceived usefulness has positive impact on

smartwatch. This finding contributed to future researches to focus on other factors as a determinant of users' intention when future researches want to study in smartwatch contexts. Finally, the study provided a better understanding of the important innovation characteristics as the determinants of individuals' intention in the context of studies on smartwatch adoption.

5.1.2 Practical Contributions

The finding of this study showed the level of individuals' intention to use smartwatch is very high. It implies that Malaysian individuals' desire to use smartwatch in their daily life. The findings of this study are useful for the following developments: Firstly, the study provides a guideline for smartwatch developers. It revealed the determinants of Malaysian individuals' intention to use smartwatch. The findings also indicated that PU followed by PEOU were the most important determinant of Malaysian individuals' intention to use smartwatch. Therefore, smartwatch developers should ensure that these devices offer more benefits to potential users in their daily life, and they can increase the simplicity of these devices by providing more attractive and consistent screen design regarding the menus, navigations and contents.

On the contrary, the results from the cost implies that smartwatch vendors do not necessarily need to reduce the cost of these devices. However, in order to ensure wider adoption among individuals, especially among those with lower income, there is a need to still reduce their prices. This will not only increase the sales of the product but will also bring more employment and revenue to the relevant stakeholders. Overall, the results of this study provide a useful insight that can be used by smartwatch manufacturers and other practitioners in building and developing better strategies and policies based on individuals' viewpoints.

5.1.3 Limitations and Future Research

The research design for the current study was explicitly developed to achieve its objectives, but this study had some limitation. The first limitation is the sample of the study which consisted of students from only Malaysian research universities. Thus, this sample may not represent the

entire populace and sufficient to generalize to the whole populace. Therefore, future researches can cover a large sample size, which can include all segments of the Malaysian populace. Secondly, this study only investigated smartwatch in general and did not investigate one specific smartwatch brand. Thus, future studies can study a specific brand such as Apple watch and can include other factors in the model such as brand influence and brand loyalty. Third, this study only investigated only a one-product category; smartwatch devices, which potentially limited the generalizability to other wearable devices. Therefore, this study confined its findings to smartwatch products only and the results cannot be used to generalize other wearable devices. Hence, future researches can investigate other types of wearable devices like smart glasses.

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