'Alexa, how can we increase trust in you?': An Investigation of Trust in Smart Home Voice Assistants

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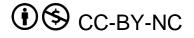
ABSTRACT,

With the emerging rise of Voice Assistants that are transforming households into Smart Homes, these IoT devices become relevant marketing tools that represent crucial opportunities for business intelligence. However, user adoption is limited due to consumers' mistrust. In this paper, trust in Smart Home Voice Assistants is investigated using an identification of trusting bases provided by literature and testing their effect on Willingness to pay (WTP) using a consumer survey. Moreover, the role of demographic variables in this context is researched. The findings display the cognitive trusting base, particularly the trust factor Perceived Usefulness to have a positive effect on WTP. No significant variables moderating the PU – WTP effect have been found. However, differences in trust among age group, education and gender could be observed. When evaluating priorities set by the respondents, Data Security is a major concern. Thus, in order to exploit the potential this technology has for marketing, it is necessary to tackle the Perceived Usefulness by adding new, innovative application fields and productive features to Smart Home Voice Assistants that provide transparency in terms of Data Security.

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Keywords Voice Assistants, Trust, Smart Home, Willingness-to-pay, Internet of Things, Artificial Intelligence

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1. INTRODUCTION

Nowadays, consumer preferences shift towards a more digitalized world, including the usage of voice as an interaction channel (Buvat et al., 2018). Especially voice assistants or so called Intelligent Virtual Assistants (IVA) as a smart home device are gaining rapid popularity among consumers and are increasingly integrated into their daily lives (Purington et al., 2017). Globally, the IVA market size estimation in 2016 rounded up to USD 1005.2 million with a growth rate of 36.7% (Grand View Research, 2018). The connected devices become progressively more available and affordable (Zheng et al., 2018). By 2021, "customer adoption of voice assistants globally is expected to reach 1.83 billion" (Buvat et al., 2018, p. 3).

The underlying functioning of these virtual assistants is enabled by Artificial Intelligence, the Internet of Things, Machine Learning and Natural Language Processing techniques (Liao et al., 2019) that facilitate Human-Computer interaction (Hoy, 2018). This way, devices such as Amazon Alexa, Google Assistant or Microsoft Cortana can perform numerous functions, including entertainment features and assistant functions like playing music, managing appointments, playing games and providing information about a variety of topics, e.g. weather and traffic (Purington et al., 2017).

However, the use of voice assistants is not yet extensively adopted. Due to a lack of trust in these new devices, less people are using it, resulting in a slow diffusion of this technology (Soronen et al., 2008).

1.1. Academic Relevance

Previous papers provide extensive scientific evidence about the acceptance of Artificial Intelligence (AI) and new (disruptive) technologies in general. An Example of this is the Technology Acceptance Model by Davis (1985), that serves as a basis for the research of technology adoption and examines the factors that are having an influence on technology diffusion among a population. This framework has been applied by various researchers, also in the context of voice assistants. Examples of prior studies in the sector of voice assistant adoption are Chowdhury (2018), who applied the theory of planned behaviour (Ajzen, 1991) into the context of non-human interactions and avoidance attitudes towards voice assistants, as well as Nasirian et al. (2017) who evaluated and tested the interaction and trust perspectives of AIbased voice assistants using a conceptual model.

However, even though the factor trust is considered as relevant for the adoption of Smart Home devices as well as voice assistants in general (Liao et al., 2019), this topic has only limitedly been researched. There is neither a conceptual framework model combining dimensions of trust based on trusting bases that refers to voice assistants, nor evidence about correlations between trust and demographic variables. Moreover, previous research has shown that in various business sectors, trust is positively affecting the WTP (Bakaki & Bernauer, 2016). For the AI sector and the voice assistant field, this effect is not yet investigated. Therefore, by closing this research gap, this paper is of academic relevance.

1.2. Practical Relevance

Current research highlights that customers' use of Artificial Intelligence-based platforms enable a technology that provides value for both, the company and the consumer (Brill, 2018). Next to this, AI based voice assistants also improve operational efficiency while the used Natural Language Processes (NLP) can improve customer service through users' activity tracking (Zheng et al., 2018).

Especially Smart Home Voice Assistants provide useful insights for personalized marketing as well as possibilities such as voice search which enhances Search Engine Optimization for companies (Ramamurty et al., 2017). In terms of Business Intelligence, companies can use the collected data as an innovation tool to identify customer needs (Spoor, 2016) by utilizing "embedded sensors and the Internet to collect and communicate data with each other and their users" (Zheng et al., 2018, p. 200). Saffarizadeh et al. (2017) states that, the higher the amount of information a voice assistant user discloses, the higher the value he receives from using these devices. Nonetheless, consumers disclose less personal information due to mistrust, which hinders (marketing) personalization (Saffarizadeh et al., 2017).

1.3. Research Scope

The first contribution this paper aims at is setting up a framework based on the literature and a resulting trust model. Furthermore, the trusting bases and subdimensions of trust in voice assistants are investigated and operationalized to quantify them. Moreover, using existing trust scales derived from the literature, a questionnaire is created. Finally, the paper aims at providing consumer insights on the impact of the sub dimensions (trusting bases) of trust on the willingness to pay (WTP) for voice assistants (with a focus on smart home voice assistants). Moreover, correlations of these trusting bases with demographic variables such as age or gender are investigated, as well as how to mitigate the most influencing dimensions in order to exploit voice assistants as a marketing tool. It is important to note that this paper will focus on the variable trust in voice assistants and its relation to WTP, as well as correlating demographic variables and implications. Therefore, the WTP itself, further affection on intention to use and optimized marketing are not part of the research scope and will thus not be investigated in this paper.

In the end, this paper will provide valuable insights for companies to address trust issues that hinder them from executing marketing opportunities facilitated by voice assistants. These trust issues can be tackled effectively to increase the WTP and to exploit the potential AI-based Voice Assistants have for business intelligence. Hence, consumers can be addressed using data-driven, personalized marketing that is optimized through their use of voice assistants as Smart Home devices.

1.3.1. Research Question

The resulting research question is:

How can a company increase trust in 'smart home' voice assistants with the objective of increasing the willingness-to-pay (WTP)?

Subquestions:

- 1. What are the trusting bases of trust in voice assistants?
- 2. Which trusting base(s) contribute most towards WTP?
- 3. How do the trusting bases correlate with demographic variables like gender, age ... etc.?

4. How can a company mitigate the mistrust by effectively tackling the most influencing trusting base(s)?

2. CONCEPTUAL FRAMEWORK

The overall framework that serves as a foundation for the assessment of trust dimensions is adapted from the conceptual model of Li et al. (2008). For assessing the components of trust, trust definitions are investigated and trusting bases are elaborated upon. To provide a comprehensive picture, various literature sources will be integrated that lead to the creation of a reliable conceptual model for trust in a Smart Home Voice Assistants context.

2.1. Trust

Trust is defined as "a psychological state comprising the intention to accept vulnerability based upon positive expectations of the intentions or behavior of another" (Rousseau et al., 1998). Another widely adopted definition by Mayer et al. (1995) describes trust as "the willingness of a party to be vulnerable to the actions of another party based on the expectation that the other will perform a particular action important to the trustor, irrespective of the ability to monitor or control that other part" (Mayer et al., 1995, p. 712). By having trust in technology, users face the risk of unfulfilled expectations and responsibilities (McKnight, 2011).

Moreover, it is seen as "one of the currencies that humans use to accept a technology in their everyday lives" (Dağlı, 2018, p. 9). Li et al. (2008) even determines trust as a "primary predictor" of technology use (Li et al., 2008, p. 39). Previous research has shown that trust affects the users' intention to use a new technology and that the technology adoption is related to trust concerns (De Kruijff, 2018).

The basis of the trust model that forms the conceptual framework is adapted from Li et al. (2008). The trust model is based on four trusting bases, namely personality base, cognitive base, calculative base and institutional base. These trusting bases form the external determinants that set up the trusting technology construct (Li et al., 2008). In the paper of Li et al., the trust model is tested in the context of new technology. However, in this paper, the framework will be adapted and shifted towards the examination of trust factors in a voice assistant context. This involves the replacement and adjustments of trusting bases and (sub-)dimensions. For instance, 'institutional base' is excluded as this is only relevant in an organizational context (Li et al., 2008). Moreover, due to the focus on WTP, the calculative base is not considered. A visualization of the model used in this paper can be seen in Fig.1.

2.2. Personality Trusting Base

The variable 'Personal Innovativeness' can be categorized into the personality base (Li et al., 2008). Li et al. explains the trusting base 'Personality Base' by psychological research findings that reveal differences between individuals in their willingness to depend on others "across a broad spectrum of situations ... and objects" that remains stable in its intensity and longevity (Li et al., 2008, p. 43). Nasirian et al. defined 'Personal Innovativeness' as the "willingness to try out any new technology" (Nasirian et al., 2017, p.5). This willingness/innovativeness is influencing the variable trust (Li et al., 2008). More specifically, it enhances users' trust in technology (Nasirian et al., 2018). Moreover, Rosen (2005) explained that 'Personal Innovativeness' predicts the intention to use technology (Rosen, 2005, p.62). The resulting hypothesis is *H1: There is a significant positive effect of Personal Innovativeness on Willingness to pay.*

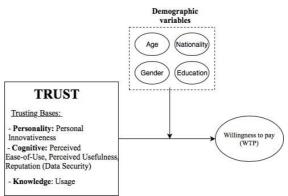


Fig. 1: Trust Model adapted from Li et al. (2008)

2.3. Cognitive Trusting Base

According to Li et al. (2008), the cognitive trusting base is suggested to affect trust in the form of "various cognitive cues and impressions" (Li et al., 2008, p.45). Huang (2017) considered the variables 'Perceived Usefulness' and 'Perceived Ease of Use' to be cognitive factors in predicting continued use of information systems (Huang, 2017). These two components were first defined by Davis (1989) and gained wide popularity as part of the Technology Acceptance Model (TAM). The definition by Davis points out 'Perceived Usefulness' as the extent of the belief that the use of a certain technology leads to job performance enhancements (Davis, 1985). Moreover, the factor 'Perceived Ease-of-Use' is defined as "the degree to which an individual believes that using a particular system would be free of physical and mental effort." (Davis, 1985, p.26). Resulting, these two factors are assumed to have a positive effect on Willingness to pay, i.e. H2: There is a significant positive effect of Perceived Ease-of-Use on Willingness to pay and H3: There is a significant positive effect of Perceived Usefulness on Willingness to pay.

The third sub-dimension of the cognitive base is 'Reputation' (Li et al., 2008). It is investigated that reputation functions as a decision-making tool for trustworthiness (Li et al., 2008, p.45). This is reinforced by Sabater & Sierra (2003), who define reputation as a "trustenforcing, deterrent, and incentive mechanism to avoid cheaters and frauds" (Sabater & Sierra, 2003, p.2). In the context of Smart Home Voice Assistants, Chung et al. (2017) assumed reputation, more specifically in the form of data protection issues, to influence consumers' trust in voice assistants. Lei et al. (2017) unveiled three security vulnerabilities of Smart Home Voice Assistants, with an emphasis on device access gained through (unauthorized) third parties (Lei et al., 2017). Lastly, Zheng et al. (2018) states that there is a significant effect of users' privacy protections perceptions on trust in IoT device manufacturers (Zheng et al., 2018). Data Security Concerns as a trusting factor are therefore assumed to negatively affect the WTP, i.e. H4: There is a negative effect of Reputation in the form of Data Security Concerns on Willingness to pay.

When using innovative devices, users are exposed to risk as well as uncertainty related to the storage and transmission of data (McKnight et al., 2011), more particularly the interconnection with central cloud services that allow easy usage and global access of data collected by IoT devices (Daubert et al., 2015, p.1). This data is derived from the two user-side components, the companion applications and the voice assistant-enabled devices (Chung et al., 2017, p.4).

2.4. Knowledge Trusting Base

This (last) trusting base is also referred to as experiential trust (McKnight et al., 2011). It assumes "a history of trustor-trustee interactions" (McKnight et al., 2011, p. 3). Concluding, the knowledge trusting base is affecting trust in the form of either previous usage or a lack of experience in dealing with voice assistants. Sabater & Sierra indicated direct experiences with virtual agents in the form of 'Usage' to be "the most relevant and reliable information source for a trust model" that is positively contributing towards trust (Sabater & Sierra, 2005, p.4).

This trusting base has been excluded from the research model of Li et al. (2008), as their research was focused on initial trust, e.g. trust that is characterized by no prior interaction of the trustor with the trustee. However, for this paper, this scientifically proven trusting base is considered in the conceptual framework and the study to facilitate comparisons between Users and Non-Users of voice assistants and the role of usage in trust. The resulting hypothesis is *H5: There is a positive effect of Usage on WTP*.

2.5. Willingness to pay (WTP)

Willingness to pay (WTP) is defined as "the maximum price a given consumer accepts to pay for a product or service" (Le Gall-Ely, 2009, p. 3). Li et al. (2008) assumes the price-benefit ratio to be correlated with trust. In case the benefit exceeds the price, the user tends to have more trust and vice versa (Li et al., 2008). However, when critically evaluating this relationship, it becomes apparent that Li et al. proved this relationship to be significant under the condition of initial trusting beliefs, e.g. no prior experience and are driven by self-interest and rationality (Li et al., 2008). Under these circumstances, WTP serves as a rational factor that is crucial in technology purchasing decisions.

There is evidence that, across various business sectors, trust generally has a positive relationship to WTP. This means higher trust results in a higher WTP. Bakaki & Bernauer (2016) argue that their results suggest "low levels of trust in public institutions have a strong negative impact on the public's willingness to pay for forest conservation" (Bakaki & Bernauer, 2016, p.1). Furthermore, Nocella et al. (2010) state that "the role of trust in consumer willingness-to-pay (WTP)...is paramount" (Nocella et al., 2010, p.275). In his research, consumers' trust appears to have a positive influence on WTP (Nocella et al., 2010). In Social Network Sites (SNS) Han & Windsor (2011) also proved user's trust as a variable influencing WTP (Han & Windsor, 2011). In addition, Ba (n.d.) investigated a model that hypothesizes buyers' trust significantly affects their WTP (Ba, n.d.). The results indicate that this is true for expensive products (Ba, n.d.).

3. METHODOLOGY

3.1. Research Design

In order to fulfill the research objective of answering the overall research question and the operationalized subquestions, an adequate research design needs to be chosen. For this, considerations regarding the appropriateness of a method for answering the research questions are essential. In the table below (Fig.2), the methods applied to investigate *trust in voice assistants* can be seen (Fig. 2).

Question	Method
What are the trusting bases	Secondary research:
of trust in voice assistants?	Literature review
Which trusting base(s)	Primary research:
contribute most towards	Survey (SPSS
WTP?	Multiple Linear
	Regression)
How do the trusting bases	Primary
correlate with demographic	research:
variables like gender, age	Survey (SPSS
etc.?	Bivariate Analysis)
How can a company mitigate	Implications based
the mistrust by effectively	on primary
tackling the most influencing trusting base(s)?	research (Survey)

Fig. 2: Methods used in relation to research questions

For this purpose, a combination of a *primary research* model, e.g. utilizing primary data that is "collected directly from researchers for the purposes of their research objectives", enabling me to have full control over the data collection (Chrysochou, 2017, p. 411) and a *secondary research* model, e.g. including "sources and data that are already available" has been used (Chrysochou, 2017, p.411).

The secondary research was done in the 'Conceptual Framework' part of this paper and includes various literature sources. Keywords that were used for the sourcing of adequate literature are Smart Home Voice Assistants, Trust, Artificial Intelligence, Amazon Alexa, Google Assistant and Willingness-to-pay. By implementing this secondary research approach, a trust framework is being adapted. By these means, the first research question of this paper, the identification of trusting bases, can be resolved.

The primary data of this paper is the data that is obtained from the sample using a quantitative data collection approach. A survey is preferred in case "the aim is more to get an overview and not explore in-depth the phenomenon" of consumer behaviour (Chrysochou, 2017. p. 412). Moreover, there is a strive for a generalization of the results. Surveys are commonly used as the provide the possibility of a detailed operationalization and interpretation of data. They allow for effective measurements of pre-set variables (Chryschou, 2017). Using the primary survey data, the trusting base that has the strongest impact on WTP is identified using a Multiple Linear Regression analysis in SPSS that provides a regression equation with coefficients that show the impact of each base on the variable WTP. Thus, the question *Which trusting base(s) contribute most towards WTP?* can be researched. By analyzing the survey results and performing bivariate correlation analyses in SPSS, the third research sub-question target of identifying correlations between trusting bases and demographic variables can be studied. The final research (sub-) question is concerned with the mitigation of trust issues, this can also be tackled by the insights derived through the primary data (survey) analysis. By these means, the main research question of the paper, i.e. *How can a company increase trust in 'smart home' voice assistants with the objective of increasing the willingness-to-pay (WTP)?*, can be answered. The further methodology process needed for conducting a survey can be categorized into the subprocesses Sampling, Designing Questions and Data Collection (Chryschou, 2017).

3.2. Sampling

In the sampling stage of a research, the researcher is determining the sample of participants that will be involved in the search, in this case in the survey. This also includes considerations about approaching possible respondents and defining the population the sample is drawn from. To achieve a high validity, the sample should be large and representative (Chrysochou, 2017).

A total of 421 participants responded to the survey. This constitutes a representative sample size that yields the advantage of an increased validity of findings and the comparison of sub-groups (Hartley, 2013), in this case specifically the comparison between cases in different geographical areas, educational levels, gender and age group. To allow for this sample diversity, apart from an internet connection to fill in the online survey, no exclusion criteria were set.

However, due to incomplete surveys (9%) and not utilizable answers, some responses had to be removed from the data analysis. In particular, all responses that were not completed and are marked with less than two minutes duration were excluded from the analysis. Moreover, careful considerations about the usefulness of incomplete surveys were made on a case-by-case basis, judging their potential insights for the research question. Finally, out of the 421 responses, only 370 responses are considered in the data analysis and are processed further. The sample varies in age groups from 18-54+, the majority (65.7%) being female. Of the 370 valid respondents, 206 (55,67%) are users of voice assistants. Apple's Siri (26.5%) being the most used one. The proportions of Google Assistant and Amazon Echo (Alexa) users are relatively similar with 21.6% of all respondents using Google Assistant and 21.4% using Alexa. Only 3.5% indicated the use of other voice assistant devices, such as Microsoft Cortana or Samsung Bixby. In total, 33 different nationalities from Europe, Asia, America and Africa are included and grouped into continents/geographical areas. Hereby, due to the enormous fraction of 63.2% of all respondents being from Germany, Germany was treated as an own geographical area in the analysis. The age distribution is skewed to the right as 53.1% of the respondents are in age group 18-24. The majority (54,2%) are students, followed by Full-Time employees (28.4%), Part-Time employees (9.9%), Self-Employed participants (4.5%), Retired (2.7%) and respondents with an inability to work (0.3%). This distribution is also reflected in the marital status of the respondents as well as the level of education. 67% indicated that they are single, 19.7% being married and 3.8% are divorced or widowed. When analyzing the level of education of the participants, it becomes apparent that only a small fraction (0.6%) attended High School without a degree. 34.3% of all respondents obtained a High School degree. Besides, 20.6% obtained an Apprenticeship Degree. In the tertiary education field, most respondents have a Bachelor Degree or equivalent, with a total percentage of 25.4% of all respondents. 17.3% completed a Master Degree or equivalent and 1.8% followed and completed a PhD program. A detailed overview of the sample characteristics can be found in the appendix (Appendix, Part 1).

3.3. Designing Questions

The questions that were used in the survey questionnaire are based on the operationalization (see Fig. 3 below) and were adapted from existing frameworks. The questionnaire used can be found in the appendix (Appendix, Part 2).

Trusting Base	Variable	Source	Questions
Personality	Personal	Rosen	9 Likert Scale
·	Innovative-	et al.	Questions
	ness	(2005)	
Cognitive	Perceived	Davis	4 Likert Scale
-	Ease-of-Use	(1989)	Questions
Cognitive	Perceived	/	Existing
-	Usefulness		features
			+ General
Cognitive	Reputation	Liao	5 Likert Scale
	(Data	et al.	Questions
	Security)	(2019)	
Knowledge	Experiences	/	Usage
WTP	Price	Le-Gall-	Max. Price
		Ely	
		(2009)	
Potential	Age, Gender,	/	Demographic
Correlations	Nationality,		Data
	Education		
	etc.		
Weighting	All trusting	/	Ranking;
Trusting	bases		Multiple
Bases'	variables		Linear
Impact on			Regression
WTP			
Willingness	Max. Price	/	Max. Price
to pay (WTP)			WTP in €

Fig. 3: Operationalization of trusting bases, WTP and weighting of trusting bases

In the beginning of the questionnaire, the respondent was asked to mention the use of a voice assistance and if applicable, the frequency of use of such a device. This also involves voice assistants that, in its pure function, are not meant to be Smart Home devices, such as Siri and Google Assistant, as these assistants are subject to be integrated into Smart Homes in the future (Budzinski et al., 2018). Subsequently, the respondent should indicate to which extent he agrees/does not agree to statements regarding the personal innovativeness (Personality trusting base), perceived ease-of-use of voice assistants as well as perceived usefulness (Cognitive trusting base). The statements relating to the Ease-of-Use were designed in two paths, letting the VA users responding to the original statements and the Non-Users to statements that included terms such as 'seems to be' and 'It seems'. To assess the usefulness not only in a general, bigger picture but related to their broad functionalities, different features were presented and differences in usefulness perceptions were investigated. Moreover, an option for presenting useful ideas for functions was added.

In the second half of the survey, the respondent was asked to choose for the most trustworthy company offering voice assistants and indicate their level of concern regarding data security, storage and access through third parties in the process of voice assistant usage (Reputation, Cognitive Base). Furthermore, the WTP was elaborated upon asking for the maximum price respondents are willing to pay, either for a device providing all functions, a device providing all functions that were marked as useful in the previous step, and a device only offering the basic functions (music and information function). These results were combined into a single variable "WTP" to get a more comprehensive impression and measure relationships between WTP and trust. Lastly, the respondents were requested to prioritise the factors Price, Ease-of-Use, Usefulness based on functions, Data Security and Novelty. This way, next to (possible) extreme choices in the Likert Scales and the statistical analyses using Multiple Linear Regression, findings about the importance and weight of the different trusting bases can be analysed.

To keep the respondents' attention, the demographic data such as gender, age group, nationality and highest level of education concluded the survey.

To simplify the data for the quantitative analysis, open questions were avoided. Furthermore, as a measurement tool for the response to the statement-related questions, a Five-Point Likert Scale was used. This scale included options for the respondents to choose between the items 'Strongly Agree - Agree - Neither Agree nor Disagree -Disagree - Strongly Disagree'. The Likert Scale is described as "one of the most fundamental and frequently used psychometric tools" in research (Joshi et al., 2015, p. 396). In terms of neutrality, the option of a symmetric scale to provide the respondent the possibility of choosing "any response in a balanced and symmetric way in either directions" (Joshi et al., 2015, p. 397) which should prevent distortion of results. Nevertheless, in the context of Data Security, the neutral option has not been offered due to the nature of the question ('Are you concerned ...?' with the options Not at all concerned - Not concerned -Concerned – Very concerned) which implies a binary variable with distinct intensity.

In total, the questionnaire encompasses 17 questions (16 for Non-Users), each representing either a specific question or a question block of similar questions, whereas question blocks were exclusively used in the context of Likert-Scales. The time used to complete the survey is approximately five minutes and can be done online at any moment in the period of three weeks, using any display device that has an internet connection (Computer, Tablet, Smartphone etc.).

3.4. Data Collection

An online survey is considered to be the most appropriate survey type as the survey is about trust in Smart Home Voice Assistants which are internet-based devices. Thus, the target group of users and potential users are mostly current users of the internet. Moreover, in order to investigate potential moderator and/or correlating variables relating to differences in nationality, an online survey was useful to facilitate time-efficient and effective participation around the globe. The focus on an international sample was also supported by the original instruction language, English. As a second option, the survey could also be conducted in German to enable a diversified sample in terms of age groups and facilitate participation for the age groups 45+ that possibly do not speak English fluently (anymore).

Distribution channels used were the test pool platform SONA of the Behavioural, Management and Social Sciences (BMS) Faculty of the University of Twente, as well as Social Media (Facebook) and a circle of friends and acquaintances. However, the survey was conducted anonymously to ensure confidentiality. All respondents in the survey participated on a voluntary basis with the possibility to interrupt and continue at a later moment in time.

4. ANALYSIS

The questionnaire is being evaluated using SPSS. The Likert-Scale based responses of each question will be analysed to identify overall impressions of the personality, cognitive and knowledge trusting bases. The difference in relevance and the differing impact of the trusting bases on WTP will be researched 1. using the prioritization ranking to obtain an impression about what consumers perceive to be the most crucial trusting base and 2. Performing a Multiple Linear Regression to detect statistical significance for the influence of the trusting bases on WTP. Moreover, associations (correlations) between the demographic variables Age Group, Gender, Education, Geographical Areas and the trusting bases are investigated as well as potential moderating effects on the trust-WTP relationship. A detailed summary of the Likert-Scale analysis can be found in the Appendix (Appendix, Part 3).

4.1. Analysis of the trusting bases

4.1.1. Personality Trusting Base

In the analysis, it becomes apparent that the sample values the ability to find any information only whenever they want, 93% either agreed or strongly agreed to this statement. Another crucial factor is the ability to access the internet anytime and anywhere they would like to (90.4%). For both statements, 'Strongly agree' is the most chosen option (56.9% respectively 56.1%). When looking at their attitude towards the persuasion of the newest technology trends, its potential to provide far-reaching solutions and endless possibilities, the respondents on average have a rather neutral opinion with a median of 3 (=Neither agree nor disagree).

However, 'Agree' was the most chosen option when reflecting about the positive influence of technology on their personal accomplishments (49%). Another 15.2% strongly agreed, leading to 64.2% agreement to this statement.

The downsides of technology that encompasses people wasting too much time on technology (54.3% agreed/strongly agreed) and the arousal of isolation (56.9%) was also identified by the respondents, with most of them indicating 'Agree' which functions as the median. The view that technology makes our life more complicated, was not shared but also not rejected (neutral with 'Neither Agree nor Disagree' as the median).

4.1.2. Cognitive Trusting Base

Both Users and Non-Users of Smart Home Voice Assistants agreed that voice assistants have an extensive perceived ease-of-use with a median of 2 (='Agree') for each question. These agreements were represented in all aspects of ease-of-use. The general impression is that the usage is easy to learn, clear and understandable and that dealing with a voice assistant is easy (or seems easy for Non-Users) to become skillful at.

In the second sub-dimension of the cognitive trusting base, perceived usefulness, the general usefulness as well as the usefulness of different functions provided by voice assistants in a Smart Home context were investigated.

In general, Smart Home Voice Assistants such as Amazon Echo (Alexa) or Google Home are perceived as useful or very useful by 72.3% of the respondents. According to the results, the most useful Smart Home Voice Assistants functions for the sample are the calendar function (considered to be useful or very useful by 89%), which includes setting up and reminding about appointments, and the function to provide information about several topics, such as traffic information, weather forecasts and/or news (90.1%). Other features relevant for the respondents when using a Smart Home device are the music function (86.5%) which consists of the usage of music streaming services such as Apple Music or Spotify and radio streaming, as well as the call function (80%) to communicate in telephone calls via the Voice Assistant. Functions verified as 'Useful' but not of major relevance are the recipe function (68.7%) and the connection with thermostats (65.4%), lights (75.5%), robot vacuum cleaners (52.1%), other Smart Home Speakers (64.4%) and the use of these devices as a dictionary and/or a calculator (74%).

In the survey, the gaming function was considered to be of least importance, least marked with the options 'Useful' or 'Very Useful' (24%). Other functions that does not provide usefulness for the respondents are the order function with only 34,7% indicating it as either useful or very useful (purchasing products that are linked to Google or Amazon by voice ordering), operating the coffee machine (46.7%) through Smart Home Voice Assistants and the connection of cameras with these devices (47.6%). Regarding the last sub-dimension of the cognitive trusting base, namely 'Reputation', as described in the literature review, the data security issue arises. The participants indicated that they are concerned about the security of their data and privacy. In particular, significant concerns about the continuous listening of the device and unauthorized purchases through third parties, that make use of the ordering function, arise. Another consideration is the encrypting of data collected by the device and transferred to the cloud that evokes users' concerns. Moreover, third parties accessing the data is also a major concern. Lastly, the microphone activation without direct interaction is feared. Across the analysis of this dimension, lower means in answers have been identified. This represents higher significance for the relevance of this aspect influencing trust. The sample is to a high extent concerned about their data being collected, processed and stored by the multinational companies such as Google, Microsoft or Amazon. This is further displayed in the answers when asking for the most trustworthy company. There is not a single company offering Smart Home Speakers that seems to be very trustworthy for the participants, all results (30% Google, 22.1% Amazon, 24.1% Microsoft and 6.4%

Apple, which was mentioned upon request) are very close by and represent a similar level of trustworthiness.

A fraction of 14% of all respondents indicated that they do not trust any of these companies with some stating that this is due to the fact that all of the mentioned companies have their servers located in the United States of America (USA).

4.1.3. Knowledge Trusting Base

Of the 370 respondents, 206 (55.67%) indicated that they are current users of voice assistants. The voice assistant with the highest number of users is Apple's Siri (36.29%), followed by Google Assistant (29.62%) and Amazon's Alexa (29.26%). Microsoft Cortana is a less used IoT device (3.33%). Lastly, other voice assistants used (1.48%) are Bixby and Homey. Most users of voice assistants use their device daily (35%) or 2-3 times a week (25.7%). 8.3% described their usage frequency as 'once a week', whereas 8.7% use a VA every two weeks. The percentage of total users that use a voice assistant only once a month or less is 22.3%.

It becomes apparent that Amazon Echo (Alexa) is the device that has the highest usage frequency. 62.18% of the Alexa users mentioned that they are using the assistant on a daily basis. The voice assistant that is used the less (in terms of frequency) is Apple's Siri (23.7%). Thus, Apple's Siri is the device with the highest numbers of users but the lowest usage frequency.

4.1.4. Self-perceived Relevance of Trusting Bases

In the ranking of purchasing factors, the respondents were requested to prioritise multiple factors, each being related to a certain sub-dimension of a trusting base. The analysis supports the idea of data security being a relevant actor in a purchasing decision. The ranking indicates that Data Security (46%) is the number 1 priority, followed by Usefulness (24.7%). In the table (Fig.4), ranking results according to the scores given in each trusting base dimension are seen (e.g. Personality Base: Personal Innovativeness, Cognitive Base: Ease-of-Use, Perceived Usefulness, Data Security Concerns, Knowledge Base: Users, Non-Users).

From these results, it is apparent that Data Security as a purchasing factor is highly valued by low innovative respondents (68.6% of this group), respondents with a medium perceived ease-of-use (75.5%) and respondents with a low perceived usefulness of these devices (33.7%). The majority of Non-Users (60.8%) placed this factor as a priority.

Usefulness is prioritised by respondents that are highly concerned about Data Security (56.3%), respondents that perceive the usefulness to be high (36.0%), the Ease-of-use to be low (37.5%) and are highly innovative (30.8%). Users generally place usefulness higher than Non-Users (29.8%).

In the context of WTP, the price is of major importance. It is identified as a main factor for respondents that are highly innovative (28.2%), perceive the Ease-of-Use to be low (37.5%), the Perceived Usefulness to be medium (22.8%) and are medium concerned about their data (29.6%). Users indicate price as a major factor more often (27.6%).

Ease-of-Use is emphasized as the key purchasing factor for medium innovative respondents (7.8% of this group placed it as their priority), perceive the Ease-of-Use to be low (12.5%), the Perceived Usefulness to be low (15.8%) and are highly concerned about their data (12.5%). Users highlighted Ease-of-Use to be a priority more often than Non-Users, 7.2% of the Users chose Ease-of-Use as the most relevant factor for a purchasing decision.

The factor Novelty has been identified to play a minor role in purchasing decisions as most groups did not prioritise this factor, leading to a maximum value of 2.8% of the data medium concerned respondents.

Trusting	Meaning	Data	Useful-	Price	EoU	Novelty	N
Base Scores	for Trust	S.	ness				(Subgroup)
Personality	Low	68.6%	17.1%	14.3%	0.0%	0.0%	35
High	Innovative						
Personality	Medium	47.6%	23.8%	19.9%	7.8%	1.0%	206
Medium	Innovative						
Personality	Highly	33.3%	30.8%	28.2%	7.7%	0.0%	78
Low	Innovative						
Cognitive:	Low	12.5%	37.5%	37.5%	12.5%	0.0%	8
EoU High	Ease-of-						
	Use						
Cognitive:	Medium	75.7%	8.1%	8.1%	8.1%	0.0%	37
EoU	Ease-of-						
Medium	Use						
Cognitive:	High	58.8%	20.6%	15.5%	5.2%	0.0%	97
EoU Low	Ease-of-						
	Use						
Cognitive:	Low	63.2%	5.3%	15.8%	15.8%	0.0%	19
PU High	PU						
Cognitive:	Medium	52.4%	20.6%	22.8%	3.7%	0.5%	189
PU	PU						
Medium							
Cognitive:	High	33.7%	36.0%	20.2%	9.0%	1.1%	89
PU Low	PU						
Cognitive:	Less	62.9%	20.6%	14.3%	2.3%	2.3%	175
DS High	concerned						
Cognitive:	Medium	21.1%	35.2%	29.6%	11.3%	2.8%	71
DS	concerned						
Medium							
Cognitive:	Highly	6.3%	56.3%	25.0%	12.5%	0.0%	16
DS Low	concerned						
Knowledge:	1	34.3%	29.8%	27.6%	7.2%	1.1%	181
User							
Knowledge:	1	60.8%	18.2%	14.7%	6.3%	0.0%	143
Non-User							

Fig. 4: Number 1 priority purchasing factors within subgroups of trusting bases

4.2. Willingness to Pay (WTP)

In order to measure willingness to pay for a Smart Home Voice Assistant in Euro (\mathcal{E}) , the respondents were asked to indicate maximum prices that they would pay for a standalone Smart Home device that (1) provides all functions mentioned in the 'Usefulness' part (2) provides all functions you consider to be useful and (3) provides only very basic function, e.g. music streaming service/radio function and information function.

The means of the max. price are (1) $\notin 139.12$ with a standard deviation (SD) of $\notin 96.757$, (2) $\notin 118.86$ with a SD of $\notin 66.86$ and (3) $\notin 78.6$ with a SD of $\notin 73.734$. It can be seen that there is a very high price range of $\notin 244$ for a level 1 device, $\notin 247$ for a level 2 device and $\notin 242$ for a level 3 device, with the lowest WTP price being $\notin 0$ and the highest WTP price being $\notin 500$.

4.3. Correlations between Demographic Variables and Trusting Bases

To assess the impact of the demographic variables on the model, correlations between the demographics and the trusting bases will be identified. This allows for a deeper insight of trust and helps to set implications for the objective of increasing trust. A comprehensive overview about the analysis including all correlation coefficients can be found in the appendix (Appendix, Part 5).

4.3.1. Age Group and Trust in Voice Assistants The SPSS bivariate correlation analysis (Spearman's Rho correlation) made apparent that age and perceived ease-ofuse are correlated. The responses to three of the four Likert-Scale questions suggest a significant correlation with the variable age group. Younger people in this sample assume the ease-of-use to be higher than the perceived ease-of-use from older respondents. Another weak to moderate correlation has been found between age and usefulness for the functions calendar, call and lights. Younger age groups show lower means, reflecting higher perceived usefulness of these functions than older age group perceive them to be (see Appendix, Fig. 5.4 for a comparison of means between age groups and trust factors across significant correlations).

4.3.2. Gender and Trust in Voice Assistants

It appears that Gender affects the personality trusting base (personal innovativeness dimension) in six out of nine statements. The agreement of the desire of internet access, technology trends, technology as an efficiency tool to achieve more accomplishments as well as the solutions are more present in responses of men. Regarding the technology drawbacks, time waste and isolation issues are perceived to be stronger by women. Concluding, women of this sample do have a more negative attitude towards technology and are less personal innovative. Concerning the effect of Gender on the cognitive trusting base, correlations between Gender and the dimensions Perceived Usefulness and Reputation (Data Security) could be verified. The general usefulness of Smart Home Voice Assistants is perceived to be higher by men. A comparison of the mean values of male and female can be found in the Appendix (Fig.5.5). Interestingly, the trusting (sub-)dimension with the most correlation is Data Security. All 4 statement show correlations with the variable Gender. It can be seen that men are generally more concerned about data and privacy issues relating to technology than women. This can be proven by the higher means of females, which in this case represent less concerns.

4.3.3. Education and Trust in Voice Assistants

When looking at the correlation between education and the trusting bases, education seems to affect the perceived ease-of-use (cognitive trusting base), as well as the perceived usefulness in terms of the call and camera functions.

Moreover, the personality base, more specifically personal innovativeness correlates with education. Technology trends, the attitude towards new possibilities Technology provides and the belief in new technology to increase personal accomplishments are more agreed to by lower educational groups. Moreover, the higher the educational group, the lower the perceived ease-of-use (see Appendix, Fig. 5.6).

4.3.4. Geographical Area (Nationality) and Trust in Voice Assistants

Regarding potential cross-cultural differences that could have been observed in the geographical area demographic variables, no evidence for correlations with the trusting bases has been found.

4.4. Identifying the most relevant trusting base(s): Multiple Linear Regression

The identification of the most crucial trusting base(s) affecting the willingness to pay (WTP) for a voice assistant was done using a Multiple Linear Regression in SPSS. Hereby, the scores of the questions related to each trusting base sub-dimension (e.g. Personality Base: Personal Innovativeness, Cognitive Base: Perceived Ease-of-Use, Perceived Usefulness and Reputation/Data Security) were thought to be independent variables affecting the dependent variable WTP. When running the analysis, the knowledge base factor Usage has been excluded from SPSS, resulting in no support for hypothesis *H5*.

The resulting regression equation equals to:

WTP = 581.862-3.374*D+3.493*PE+8.586*PU+3.013*PI

Whereas:

D = Data Security Concerns (Cognitive trusting base) PE = Perceived Ease-of-use (Cognitive trusting base) PU = Perceived Usefulness (Cognitive trusting base) PI = Personal Innovativeness (Personality trusting base)

The model is statistically significant with p<0.01, thus p<Alpha (see Appendix, Part 6, Fig. 6.1), this means that, taking into account the positive coefficients, there is a significant positive relationship between trust and WTP. However, only the variable Perceived Usefulness is independently affecting WTP with a significance level being lower than Alpha = 0.05 (0.001), supporting H1 to be true.

The findings suggest that some variables are not affecting the WTP independently, however they might be significant in combination with other variables, therefore it is necessary to check for multicollinearity. As can be seen in the appendix (Appendix, Part 6, Fig. 6.2), multicollinearity can be excluded as all variance inflation factors (VIF) are ranged between 1.040 and 1.389. This means that it is not possible to predict any of the trusting base variables from the others.

Trusting Base	Coefficient	Coefficients Std. Error	Sig.	Hypothesis supported
Personality Base: Personal Innovativeness	8.586	2.386	0.001	yes (H1)
Cognitive Base: Perceived Ease- of-Use	8.013	3.659	0.412	no (H2)
Cognitive Base: Perceived Usefulness	3.493	3.209	0.279	no (H3)
Cognitive Base: Reputation	-3.374	7.753	0.664	no (H4)

Fig. 5: Multiple Linear Regression predicting WTP

Moderators / Trusting Base			Group Gender Nationality GRESSION) (REGRESSION) (PROCESS)		Education (REGRESSION)	Usage (PROCESS)
Personality	Base:	C 0.02	<i>C</i> -0.74	C -8.59	<i>C</i> -0.11	<i>C</i> 14.83
Personal Innovativenes	~	SE 0.54	SE 1.29	SE 3.73	SE 0.53	SE 6.68
mnovativenes	8	Sig. 0.97	Sig. 0.56	Sig. 0.02	Sig. 0.98	Sig. 0.03
Cognitive	Base:	C 0.45	<i>C</i> -2.20	C -3.53	<i>C</i> -0.38	<i>C</i> -0.32
Perceived Ease-of-	se-of-	SE 0.83	SE 2.05	SE 3.23	SE 0.85	SE 3.02
Use		Sig. 0.59	Sig. 0.29	Sig. 0.28	Sig. 0.66	Sig. 0.92
Cognitive	Base:	C 0.082	C 0.96	<i>C</i> 1.67	C 0.01	C -1.33
Perceived		SE 0.30	SE 0.70	SE 1.62	<i>SE</i> 0.29	SE 3.30
Usefulness		Sig. 0.79	Sig. 0.18	Sig. 0.30	Sig. 0.97	Sig. 0.69
Cognitive	Base:	C 0.10	C -3.25	C -3.85	C 0.16	C 0.72
1	(Data	SE 1.03	SE 2.47	<i>SE</i> 3.49	SE 1.05	SE 8.71
Security Conc	Concerns)	Sig. 0.92	Sig. 0.19	Sig. 0.27	Sig. 0.88	Sig. 0.93

Fig. 6: Identification of Moderators

4.5. Demographic Variables moderating the Trust - WTP Effect

When identifying demographic variables potentially moderating the effect between the trusting bases and WTP, multiple linear regressions with interaction variables as well as the Hayes' PROCESS function in SPSS has been used. Whereas no significant moderator has been identified for the significant Perceived Usefulness effect on WTP, two findings should be noted. It becomes apparent that Nationality, allocated along various geographical areas, moderates the Personal Innovativeness – WTP relationship. Particularly, the European Area Nationalities (excluding Germany) significantly moderate this effect (<0.02).

Moreover, another significant moderator in the Personal Innovativeness – WTP relationship is Usage (p<0.03). When analyzing other potential moderators, no significance emerged (see Fig. 6 for coefficients, standard error of the estimate and significance levels).

5. DISCUSSION AND IMPLICATIONS

In the first part of the discussion, findings derived from the analysis are evaluated with the objective of identifying the most relevant trusting base. Subsequently, implications are drawn based on the analysis, resulting in responding to the main research question of this paper, namely how to increase trust in voice assistants for increasing consumers' willingness-to-pay (WTP).

5.1. Findings based on the Correlations and Multiple Linear Regression

The findings resulting out of the evaluation of the Likert-Scales, the bivariate correlation analysis and the Multiple Linear Regression suggest that the trusting model is basically valid. All trusting bases were thought to be relevant by the respondents and play a major role in the trust of voice assistants. The most useful functions are the calendar and information function, followed by music and call functions. Game Apps on Smart Home Voice Assistants are generally perceived to be of least importance.

Correlations found suggest perceived ease-of-use to be higher for younger age groups and indicate a higher perceived usefulness for these ages in terms of call, light and camera function. Men generally perceive voice assistants to have a higher usefulness but simultaneously have more doubt when it comes to data security issues. The findings provide evidence that women have less personal innovativeness. Relating to education, less educated groups are identified to have a more positive attitude towards technology.

Concerning the most relevant trusting base for WTP, the cognitive trusting base appears to have a major relevance for the respondents. Not only in the Likert-Scale more extreme answers (resulting in lower/higher means) emerged, but also the fact that two sub-dimensions of the cognitive trusting base (Perceived Usefulness and Data Security Concerns) are listed in the top 3 prioritization ranking. To verify this subjective judgement, a significant (positive) effect of Perceived Usefulness on WTP has been found in the Multiple Linear Regression, thus a positive relationship between Perceived Usefulness and WTP can be assumed, which in turn supports hypothesis **H3** to be true. Therefore, the cognitive trusting base, more precisely the variable Perceived Usefulness can be proven to have the most relevance for the Trust-WTP relation.

5.2. Implications: How to increase Trust in Smart Home Voice Assistants

Resulting from the findings obtained in the analysis, some general implications can be drawn. The main focus for voice assistant providers should be on the perceived usefulness trust dimension, as this is identified to be the most crucial one having an effect on consumers' willingness-to-pay (WTP).

Answering the main research question, the usefulness can be increased by adding various new application fields and providing new functions that highlight productivity rather than entertainment. These include e.g. security features, connection to TVs and the use of voice assistants in cars. Moreover, elderly care might also be an innovative application field. In order to embrace this opportunity, more detailed explanations are necessary to advice also less educated people. This field is especially interesting with regard to the findings that older people do not necessarily have less personal innovativeness and Non-Users generally perceive Smart Home Voice Assistants to have a high ease-of-use and are easy to learn.

To exploit the potential voice assistants (especially Smart Home devices) have for a company's marketing, potential future users should be targeted using the data security topic. As the results suggest major safety concerns, more transparency should be provided and also advertised in the pre-purchase marketing, in particular related to men as they have higher degree of doubts.

7. CONCLUSION

In this paper, the topic trust in Smart Home Voice Assistants was dealt with. More specifically, an answer of the research question of how to increase trust in these devices with the objective of increasing the willingness to pay was aimed at. Next to the literature review that provided components of trust (trusting bases), a quantitative approach was used. In the analysis of the survey with a sample size of 370 respondents, major findings suggested the relevance of the cognitive trusting base, including perceived usefulness, to be the most relevant trusting base/dimension for WTP as this has been the only trusting dimension marked as statistically significant. No moderators impacting the Perceived Usefulness - WTP effect have been found. However, the findings yield correlations between the trusting bases and the demographic variables Age, Gender and Education. Concluding, the perceived usefulness trusting dimension should be tackled by focusing on the development of innovative, productive features for Smart Home Voice Assistants that will increase customer value and their trust in these IoT-devices.

6. LIMITATIONS AND FUTURE RESEARCH

Due to a restriction in resources, the study conducted in this paper has some limitations. First of all, most respondents in the survey are students with an age ranging from 18-34 and a High School or Bachelor Degree which might decrease the representation of this sample as a whole population. Moreover, cross-cultural differences could not be observed due to a limited number of participants from other countries than Germany. This provides an interesting starting point for further research in the field of trust in voice assistants. Moreover, as identified in this paper, usefulness and new application fields play a major role to promote the trust in Smart Home Devices. Options for elderly care and their consequences for trust could be investigated especially in regard to the lack of caregivers. Lastly, data security focused marketing is a subject to look further into.

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APPENDIX

Part 1: Demographic data about respondents

Part 2: Survey

Part 3: Likert-Scale Question Analysis

Part 4: Self-perceived ranking of VA purchasing factors

Part 5: Correlations of trusting bases with demographic variables (5.1. Age Group, 5.2. Gender, 5.3. Education)

Part 6: Assessing the relevance of trusting bases for WTP and moderating effects: Multiple Linear Regression, Univariate Analysis of Variance and Multicollinearity checks

PART 1: DEMOGRAPHIC DATA ABOUT RESPONDENTS

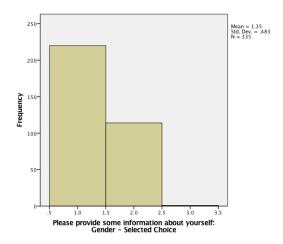


Fig. 1.1. : Gender of respondents (1=Female, 2=Male)

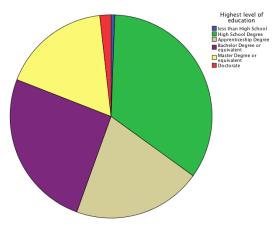


Fig. 1.2: Education levels of respondents

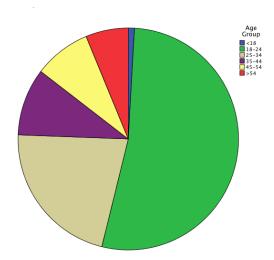


Fig. 1.3: Age Groups of respondents

Country	# Respondents	Country2	Respondents
Austria	3	Lithuania	1
Brazil	2	Mexico	2
Bulgaria	2	Netherlands	16
Canada	3	Pakistan	1
China	5	Poland	4
Congo	1	Portugal	2
Denmark	2	Russia	3
Egypt	2	South Africa	2
France	3	Switzerland	2
Germany	234	Syria	1
Greece	2	Taiwan	1
Hungary	3	Turkey	3
India	3	UK	9
Indonesia	1	USA	6
Iran	2	Vietnam	1
Italy	3	Iran	2

Latvia

1

Fig. 1.4: Nationalities of respondents



Fig 1.5: Geographical Visuliations of respondents' nationalities, created via Excel Bing Maps

<u>PART 2: SURVEY</u> Trust in Smart Home Voice Assistants Start of Block: Block 5 Q22 Dear participant,

This survey is part of my Bachelor thesis in the study programme International Business Administration (University of Twente). In my thesis, I aim at investigating different factors that are important for the trust in Smart Home Voice Assistants like Amazon's Alexa, Google Home etc and how this correlates with age group, gender, nationality and other variables. More specifically, I aim to assess possible factors that are the main reason of people having trust/not having trust in voice assistants.

Explanation Smart Home Voice Assistants/Smart Speakers: Internet-based devices that allow you to interact with, using a codeword (e.g. "Okay, Google" or "Alexa") followed by a command. This way, the devices provide different functions, including entertainment and productivity features.

The survey will take approximately 5 minutes and is anonymous. All data is treated confidentially.

If you have any questions, please do not hesitate to contact me.

Thank you!

Rabea Adams r.j.adams@student.utwente.nl End of Block: Block 5 Start of Block: Default Question Block Q1 What type of voice assistant(s) do you use?

 Apple Siri (1)

 Google Assistant (2)

 Amazon Echo (Alexa) (3)

 Microsoft Cortana (4)

 Other, Please specify (5)

 I don't use a voice assistant (6)

 Display This Question:

 If What type of voice assistant(s) do you use? != I don't use a voice assistant

 Q2 How often do you use a voice assistant?

 •Daily (1)

 •2-3 times a week (2)

 •Once a week (3)

 •Every two weeks (4)

 •Once a month (5)

•Less than once a month (6)

End of Block: Default Question Block

Start of Block: Block 1

Q4 Please indicate to what extent you agree/disagree to the following statements:

	Strongly agree (1)	Agree (2)	Neither agree nor disagree (3)	Disagree (4)	Strongly disagree (5)
I feel it is important to be able to find any information online whenever I want (1)	0	0	0	0	0
I feel it is important to be able to access the internet any time I want (2)	0	0	O	0	0

I think it is important to keep up with the latest trends in technology (3)	0	0	0	0	0
Technology will provide solutions to most of our problems (4)	0	0	o	0	0
With technology, everything is possible (5)	0	0	0	0	0
I feel that I get more accomplished because of technology (6)	0	0	0	0	0
New technology makes people waste too much time (7)	0	0	o	0	0
New technology makes people more isolated (8)	0	0	0	0	0
New technology makes life more complicated (9)	0	0	o	0	0

Display This Question:

If What type of voice assistant(s) do you use? != I don't use a voice assistant Q5 Please indicate to what extent you agree/disagree to the following statements:

	Strongly agree (1)	Agree (2)	Neither agree nor disagree (3)	Disagree (4)	Strongly disagree (5)
Interacting with voice assistants is easy to learn (1)	0	0	0	0	0
Interacting with voice assistants is clear and understandable (2)	o	0	o	o	0
I feel that it is easy to become skilful in using voice assistants (3)	0	0	0	0	0
I think that voice assistants are easy to use (4)	0	0	0	0	0

Display This Question:

If What type of voice assistant(s) do you use? = I don't use a voice assistantQ6 Please indicate to what extent you agree/disagree to the following statements:

	Strongly agree (1)	Agree (2)	Neither agree nor disagree (3)	Disagree (4)	Strongly disagree (5)
Interacting with voice assistants seems to be easy to learn (1)	0	0	0	0	0

Interacting with voice assistants seems to be clear and understandable (2)	0	0	0	0	0
It seems that it is easy to become skilful in using voice assistants (3)	0	0	0	0	0
It seems that voice assistants are easy to use (4)	0	0	0	0	0

Q7 How useful are the following Smart Home Voice Assistant functions in your opinion? (If you do not use a voice assistant, please indicate how useful you would find these functions if you had one)

	Very useful (1)	Useful (2)	Useless (3)	Very useless (4)
Smart Home Voice Assistants in general (1)	0	0	0	0
Playing music (radio+streaming function) (2)	0	0	0	0
Playing games (game function) (3)	0	0	0	0
Setting up and reminding about appointments, alarm or timer (calendar+alarm function) (4)	0	0	o	0
Connect with phone (call function) (5)	0	0	0	0

Connect with Google/Amazon for purchasing products (order function) (6)	0	0	0	0
Provide information for cooking (recipe function) (7)	o	0	0	0
Provide information about weather, traffic, news or sports (info function) (8)	0	0	0	0
Connect with thermostat (9)	0	0	0	0
Connect with lights (10)	o	0	0	0
Connect with coffee machine (11)	0	0	0	0
Connect with vacuum cleaner robot (12)	0	0	0	0
Connect with other Smart Home Speakers (e.g. additional Amazon Echo Dot) (13)	0	0	0	0
Connect with cameras (14)	0	0	0	0
Function as dictionary/calculator (15)	0	0	0	0
Q8 What other functions would you consider to be useful?				

End of Block: Block 1

Start of Block: Block 2

Q10 In your opinion, what is the most trustworthy company offering smart home voice assistants?

•Google (1)

oAmazon (2)

oMicrosoft (3)

•Other, Please Specify: (4) _____

Q11 Please indicate how concerned you are about the following, based on the most trustworthy company you chose above:

	Very concerned (1)	Concerned (2)	Not concerned (3)	Not concerned at all (4)
I am concerned that the device is always listening (1)	0	0	0	0
I am concerned that other people might activate/access the device and trigger unauthorized purchases (2)	0	0	0	0
I am concerned that information communicated between the device and the service provider is not always encrypted (3)	0	0	0	0
I am concerned that microphones on these devices are activated without a user's direct action (4)	0	0	0	0
I am concerned that third parties gain access to my data collected by the device (5)	0	0	0	0
End of Block: Block 2				

Start of Block: Block 3

Q12 What is the maximum price (in €) you are willing to pay for a Smart Home Voice Assistant?

Level 1 Device: Stand-alone Smart Speaker with the ability of providing all functions mentioned

Level 2 Device: Stand-alone Smart Speaker with the ability of providing all functions you marked as useful or very useful

Level 3 Device: Stand-alone Smart Speaker with the ability of providing the basic functions (music + information functions)

(Note: The price of currently launched Smart Home voice assistants ranges from 59€ up to 350€)

•Level 1 Device (1)_____

•Level 2 Device (2) _____

•Level 3 Device (3)

Q13 Please prioritise the following purchasing decision factors:

(Please click on the items and pull them to bring them into the right order)

_____ Price (1)

_____ Ease-of-use (2)

Usefulness based on functions (3)

_____ Data Security (4)

_____ Novelty (5)

End of Block: Block 3

Start of Block: Block 4

Q14 Please provide some information about yourself:

Gender

oFemale (1)

oMale (2)

Other, Please specify: (3)

Q15 Age Group

o<18 (1)

o**18-24** (2)

025-34 (3)

o35-44 (4)

o45-54 (5)

o>54 (6)

Q16 Nationality

Q17 Highest level of education

oless than High School (1)

•High School Degree (2)

•Apprenticeship Degree (3)

•Bachelor Degree or equivalent (4)

•Master Degree or equivalent (5)

ODoctorate (6)

End of Block: Block 4

PART 3: LIKERT-SCALE QUESTIONS ANALYSIS

Ability1Strongly AgreeAccess1Strongly AgreeTrends3NeitherSolutions3NeitherPossibilities endless3NeitherAccomplishments2AgreeTime Waste2AgreeIsolation2AgreeComplicated3NeitherEase-of-Use UsersMedianMeaningEasy to learn2AgreeSkillful2AgreeSkillful2AgreeEasy to use2Agree	Likert-Scale Questions	Median	Meaning
Trends3NeitherSolutions3NeitherPossibilities endless3NeitherAccomplishments2AgreeTime Waste2AgreeIsolation2AgreeComplicated3NeitherEase-of-Use UsersMedianMeaningEasy to learn2AgreeClear and understandable2AgreeSkillful2Agree	Ability	1	Strongly Agree
Solutions 3 Neither Possibilities endless 3 Neither Accomplishments 2 Agree Time Waste 2 Agree Isolation 2 Agree Complicated 3 Neither Ease-of-Use Users Median Meaning Easy to learn 2 Agree Skillful 2 Agree	Access	1	Strongly Agree
Possibilities endless 3 Neither Accomplishments 2 Agree Time Waste 2 Agree Isolation 2 Agree Complicated 3 Neither Ease-of-Use Users Median Meaning Easy to learn 2 Agree Clear and understandable 2 Agree Skillful 2 Agree	Trends	3	Neither
Accomplishments 2 Agree Time Waste 2 Agree Isolation 2 Agree Complicated 3 Neither Ease-of-Use Users Median Meaning Easy to learn 2 Agree Clear and understandable 2 Agree Skillful 2 Agree	Solutions	3	Neither
Time Waste 2 Agree Isolation 2 Agree Complicated 3 Neither Ease-of-Use Users Median Meaning Easy to learn 2 Agree Clear and understandable 2 Agree Skillful 2 Agree	Possibilities endless	3	Neither
Isolation 2 Agree Complicated 3 Neither Ease-of-Use Users Median Meaning Easy to learn 2 Agree Clear and understandable 2 Agree Skillful 2 Agree	Accomplishments	2	Agree
Complicated 3 Neither Ease-of-Use Users Median Meaning Easy to learn 2 Agree Clear and understandable 2 Agree Skillful 2 Agree	Time Waste	2	Agree
Ease-of-Use Users Median Meaning Easy to learn 2 Agree Clear and understandable 2 Agree Skillful 2 Agree	Isolation	2	Agree
Easy to learn 2 Agree Clear and understandable 2 Agree Skillful 2 Agree	Complicated	3	Neither
Easy to learn 2 Agree Clear and understandable 2 Agree Skillful 2 Agree			
Clear and understandable 2 Agree Skillful 2 Agree	Ease-of-Use Users	Median	Meaning
Skillful 2 Agree	Easy to learn	2	Agree
	Clear and understandabl	e 2	Agree
Easy to use 2 Agree	Skillful	2	Agree
	Easy to use	2	Agree

Ease-of-Use Non-Users	Median	Meaning
Easy to learn	2	Agree
Clear and understandable	2	Agree
Skillful	2	Agree
Easy to use	2	Agree

Usefulness	Median	Meaning
Usefulness		
General	2	Agree
Music	2	Agree
Games	3	Neither
Calendar	2	Agree
Call	2	Agree
Order	3	Neither
Recipe	2	Agree
Info	2	Agree
Thermostat	2	Agree
Lights	2	Agree
Coffee machine	3	Neither
Vacuum cleaner	2	Agree
Smart Home Speaker	2	Agree
Cameras	3	Neither
Dictionary/Calculator	2	Agree

Knowledge	Median	Meaning
Users	0,724324324	4 0,7243
Non-Users	0,27567567	6 0,2757

Reputation	Median	Meaning
Listening	2	Agree
Unathorized purchases	2	Agree
Encryptency	2	Agree

Mic Activation	2	Agree
3rd Party Data Access	2	Agree

PART 4: RANKING SELF-PERCEIVED FACTORS FOR PURCHASING A VA

Factor	1st
Data Security	7 46.0%
Usefulness	24.7%
Price	21.9%
Ease-of-Use	6.8%
Novelty	0.6%

Fig. 4.1: Ranking Results Self-Perceived Importance of Purchasing Factors

Please prioritise the following purchasing decision factors: (Please click on the items and pull them to bring them into the right order) – Usefulness based on functions

		Frequency	Percent	Valid Percent	Cumulative Percent
Valid	1	80	21.6	24.7	24.7
	2	95	25.7	29.3	54.0
	3	79	21.4	24.4	78.4
	4	61	16.5	18.8	97.2
	5	9	2.4	2.8	100.0
	Total	324	87.6	100.0	
Missing	System	46	12.4		
Total		370	100.0		

Please prioritise the following purchasing decision factors: (Please click on the items and pull them to bring them into the right order) – Data Security

		Frequency	Percent	Valid Percent	Percent
Valid	1	149	40.3	46.0	46.0
	2	66	17.8	20.4	66.4
	3	46	12.4	14.2	80.6
	4	44	11.9	13.6	94.1
	5	19	5.1	5.9	100.0
	Total	324	87.6	100.0	
Missing	System	46	12.4		
Total		370	100.0		

Please prioritise the following purchasing decision factors: (Please click on the items and pull them to bring them into the right order) – Novelty

		Frequency	Percent	Valid Percent	Cumulative Percent
Valid	1	2	.5	.6	.6
	2	8	2.2	2.5	3.1
	3	16	4.3	4.9	8.0
	4	32	8.6	9.9	17.9
	5	266	71.9	82.1	100.0
	Total	324	87.6	100.0	
Missing	System	46	12.4		
Total		370	100.0		

Fig 4.2.. Ranking Prioritization Purchasing Factors

Please prioritise the following purchasing decision factors: (Please click on the items and pull them to bring them into the right order) – Price

		Frequency	Percent	Valid Percent	Cumulative Percent
Valid	1	71	19.2	21.9	21.9
	2	79	21.4	24.4	46.3
	3	71	19.2	21.9	68.2
	4	89	24.1	27.5	95.7
	5	14	3.8	4.3	100.0
	Total	324	87.6	100.0	
Missing	System	46	12.4		
Total		370	100.0		

Please prioritise the following purchasing decision factors: (Please click on the items and pull them to bring them into the right order) – Ease-of-use

		Frequency	Percent	Valid Percent	Cumulative Percent
Valid	1	22	5.9	6.8	6.8
	2	76	20.5	23.5	30.2
	3	112	30.3	34.6	64.8
	4	98	26.5	30.2	95.1
	5	16	4.3	4.9	100.0
	Total	324	87.6	100.0	
Missing	System	46	12.4		
Total		370	100.0		

		-					
Trusting	Meaning	Data					
Base	for	Secu-	Useful-	Price	EoU	Novelty	N (Curle and Curle)
Scores	Trust	rity	ness	Price	EOU	Noverty	(Subgroup)
Personality	Low	60 GM	17.10/		0.00	0.01/	35
High	Innovative	68.6%	17.1%	14.3%	0.0%	0.0%	35
Personality	Medium						
Medium	Innovative	47.6%	23.8%	19.9%	7.8%	1.0%	206
Personality	Highly						
Low	Innovative	33.3%	30.8%	28.2%	7.7%	0.0%	78
Completions	Low Ease-of-						
Cognitive: EoU High	Lase-or-	12.5%	37.5%	37.5%	12.5%	0.0%	8
Cognitive:	Medium	12.376	57.376	57.5%	12.376	0.0%	•
EoU	Ease-of-						
Medium	Use	75.7%	8.1%	8.1%	8.1%	0.0%	37
	High						
Cognitive:	Ease-of-						
EoU Low	Use	58.8%	20.6%	15.5%	5.2%	0.0%	97
Cognitive:	Low						
PU High	PU	63.2%	5.3%	15.8%	15.8%	0.0%	19
Cognitive:							
PU	Medium						
Medium	PU	52.4%	20.6%	22.8%	3.7%	0.5%	189
Cognitive:	High						
PU Low	PU	33.7%	36.0%	20.2%	9.0%	1.1%	89
Cognitive:	Less						
DS High	concerned	62.9%	20.6%	14.3%	2.3%	2.3%	175
Cognitive:							
DS	Medium	-	25.20	20.000		2.024	
Medium	concerned	21.1%	35.2%	29.6%	11.3%	2.8%	71
Cognitive:	Highly	6.20		25.00	13.50	0.01	
DS Low	concerned	6.3%	56.3%	25.0%	12.5%	0.0%	16
Knowledge:	,			-	-		
User	/	34.3%	29.8%	27.6%	7.2%	1.1%	181
Knowledge:	,	60 00V			c	0.004	
Non-User	/	60.8%	18.2%	14.7%	6.3%	0.0%	143

Fig. 4.3: Priority Ranking of Subgroups

Trusting Dimension Scores	s Min	Max	Range	Range/3	(Low)	(Med)	(High)
Ease-of-Use	8	40	32	10,66666667	18,66666667	29,33333333	40
Usefulness	-60	-15	45	15	-45	-30	-15
Data Security Concern	5	20	15	5	10	15	20
Personality Base	10	30	20	6,666666667	16,66666667	23,333333333	30

Fig. 4.4: Assessing Subgroups for Comparison of each trusting base ranking prioritizations (e.g. dividing scores into three (High/Medium/Low) groups)

5) CORRELATIONS OF TRUSTING BASES WITH DEMOGRAPHIC VARIABLES (BIVARIATE ANALYSIS)

5.1. AGE GROUP

Personality	Correlation	Sig
Ability	-0,032	0,554
Access	-0,028	0,613
Trends	-0,03	0,584
Solution	0,078	0,154
Possibilities	0,031	0,571

Accomplish	0,027	0,626	
Time Waste	0,016	0,765	
Isolation	-0,027	0,616	
Complicated	0,029	0,591	
			NOT RELEVANT!

Ease-of-Use Users			
Easy to learn	0,077	0,289	
Clear and understandable	0,089	0,224	
Skillful	0,06	0,412	
Easy to use	0,09	0,218	NOT RELEVANT!

Ease-of-Use Non-Users			
Easy to learn	0,23	0,005	
Clear and understandable	0,284	0,001	
Skillful	0,245	0,003	
Easy to use	0,161	0,054	RELEVANT

Usefulness	0,006	0,917
Music	0,022	0,691
Games	-0,049	0,369
Calendar	0,146	0,007
Call	0,109	0,046
Order	-0,106	0,052

D:	0.007	0.112
Recipe	0,087	0,112
Info	0,018	0,738
Thermostat	0,103	0,06
Lights	0,167	0,002
Coffee machine	0,071	0,193
Vacuum cleaner	0,09	0,102
Smart Home Speaker	0,63	0,253
Cameras	-0,46	0,403
Dictionary/Calculator	0,064	0,246
USAGE	-0,283	0
Data Security		
Listening	0,01	0,849
Purchases	-0,085	0,119
Encryptancy	-0,067	0,224
Mic Activation	0,007	0,903
wite Activation	0,007	0,903
3rd Party Access	-0,064	0,244

5.2. GENDER

Personality	Correlation	Sig
Ability	-0,068	0,217
Access	-0,13	0,018

Trends	-0,207	0
Solution	-0,202	0
Possibilities	-0,096	0,079
Accomplish	-0,211	0
Time Waste	0,253	0
Isolation	0,211	0
Complicated	0,161	0,03



Ease-of-Use Users

Easy to learn	-0,011	0,884
Clear and understandable	0,093	0,204
Skillful	0,058	0,425
Easy to use	0,082	0,262

Ease-of-Use Non-Users

Easy to learn	-0,007	0,929
Clear and understandable	0,025	0,769
Skillful	-0,005	0,95
Easy to use	0,044	0,6

NOT RELEVANT!

NOT RELEVANT!

Music	0,033	0,542
Usefulness	-0,146	0,008

Games	0,043	0,438	
Calendar	-0,044	0,426	
Call	0,005	0,928	
Order	0,059	0,284	
Recipe	0,049	0,375	
Info	-0,088	0,11	
Thermostat	-0,026	0,639	
Lights	-0,074	0,179	
Coffee machine	0,047	0,398	
Vacuum cleaner	0,042	0,44	
Smart Home Speaker	-0,107	0,05	
Cameras	-0,074	0,177	
Dictionary/Calculator	0,097	0,075	

RELEVANT

Usage

-0,211

0,004

Data Security		
Listening	0,126	0,021
Purchases	0,176	0,001
Encryptancy	0,164	0,003
Mic Activation	0,148	0,007
3rd Party Access	0,152	0,006

RELEVANT

5.3. EDUCATION

Personality	Correlation	Sig
Ability	-0,001	0,988
Access	0,014	0,8
Trends	-0,171	0,002
Solution	-0,059	0,286
Possibilities	-0,128	0,019
Accomplish	-0,113	0,04
Time Waste	-0,045	0,414
Isolation	0,005	0,933
Complicated	-0,031	0,573

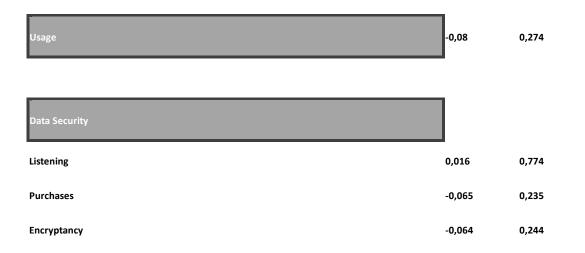
RELEVANT

Ease-of-Use Users		

Easy to learn	0,14	0,054
Clear and understandable	0,069	0,345
Skillful	0,063	0,392
Easy to use	0,073	NOT 0,315 RELEVANT

Ease-of-Use Non-Users		
Easy to learn	0,171	<mark>0,0</mark> 4
Clear and understandable	0,137	0,1
Skillful	0,094	0,26

Easy to use	0,078	0,35	RELEVANT
Usefulness	0,01	0,849	
Music	0,046	0,406	
Games	-0,068	0,213	
Calendar	-0,026	0,64	7
Call	-0,108	0,049	
Order	-0,106	0,053	
Recipe	0,066	0,225	
Info	0,04	0,466	
Thermostat	0,033	0,542	
Lights	0,047	0,394	
Coffee machine	-0,011	0,84	
Vacuum cleaner	0,052	0,346	
Smart Home Speaker	0,028	0,607	٦
Cameras	-0,12	0,028	
Dictionary/Calculator	0,054	0,323	RELEVANT



Mic Activation	-0,041	0,451	
3rd Party Access	-0,064	0,242	NOT RELEVANT

Statements/Age	μ (18-24)	μ (25- 34)	μ (35-44)	μ (45- 55)	μ (>54)
Easy of Use	16.2	20.2	16.3	21.8	19.6
Usefulness Call	1.9	2.1	2.03	2.36	2.38
Usefulness Lights	1.87	1.96	2.24	1.93	2.14
Usefulness Cameras	1.62	1.95	1.79	1.79	1.95

Fig. 5.4.: Comparison of means across trust factors significantly correlating with age group

Statements/Gender	µ Male	µ Female
Technology Access	1.48	1.64
Trends	2.43	2.95
Tech as a Solution	2.39	2.8
Accomplishments	2.1	2.52
Time Waste	2.93	2.33
Isolation	2.79	2.31
General Usefulness	2.06	2.3
Data: Listening	1.73	1.99
Data: Purchases	1.87	2.21
Data: Encrypting	1.79	2.06
Data: Mic. Access	1.6	1.87
Data: Access	1.65	1.94

Fig. 5.5: Comparison of means across trust factors significantly correlating with gender

Statements/ Education	U	U	<u> </u>		MA	PhD
Tech Trends	2.5	3.03	2.78	2.62	2.41	3.17
Possibilities	3.5	3.35	3.16	2.85	3.03	4.0
Accomplish- ments	3.0	2.49	2.46	2.19	2.29	2.33
Easy to learn	1.0	1.91	1.96	2.06	2.24	3.0
Usefulness cameras	2.5	2.77	2.71	2.46	2.57	2.17
Usefulness	2.0	1.97	2.13	1.85	1.83	1.83

Call

Fig. 5.6: Comparison of means across trust factors significantly correlating with gender

6) ASSESSING THE RELEVANCE OF TRUSTING BASES FOR WTP AND MODERATING EFFECTS: MULTIPLE LINEAR REGRESSION & UNIVARIATE ANALYSIS OF VARIANCE + MULTICOLLINEARITY CHECKS

Model	R	R Square	Adjusted R Square	Std. Error of the Estimate		
1	.401 ^a	.161	.123	166.68468		
Pe	rceivedUsefu	ilness, EaseOf	UseTotal, NEW			
Model		Sum of Squares	df	Mean Square	F	Sig.
Model 1	Regression			Mean Square 118411.933	F 4.262	
Model	Regression Residual	Squares	12 4			Sig. .003

Coefficients^a

		Unstandardize	d Coefficients	Standardized Coefficients		
Model		В	Std. Error	Beta	t	Sig.
1	(Constant)	581.862	100.894		5.767	.000
	PerceivedUsefulness	8.586	2.386	.386	3.599	.001
	NEWPersonality	3.013	3.659	.092	.824	.412
	EaseOfUseTotal	3.493	3.209	.118	1.089	.279
	DataSecurity	-3.374	7.753	047	435	.664

Fig. 6.1: Multiple Linear Regression to identify most relevant trusting base

Coefficients^a

	Coefficients ^a
Collinearity Statistics	
Tolerance VIF	Collinearit
.899 1.112	Model Tolerance
2	1 WTP .961
_	DataSecurity .798
	EaseOfUseTotal .840
	NEWPersonality .792

a. Dependent Variable: DataSecurity

Coefficients^a

		Collinearity Statistics		
Model		Tolerance	VIF	
1	DataSecurity	.795	1.257	
	EaseOfUseTotal	.796	1.257	
	NEWPersonality	.754	1.327	
	PerceivedUsefulness	.820	1.219	
a. Dependent Variable: WTP				

Coefficients^a

		Collinearity Statistics		
Model		Tolerance	VIF	
1	PerceivedUsefulness	.765	1.307	
	NEWPersonality	.791	1.264	
	WTP	.850	1.176	
	DataSecurity	.909	1.100	
a. Dependent Variable: EaseOfUseTotal				

Coefficients^a

		Collinearity Statistics			
Model		Tolerance	VIF		
1	PerceivedUsefulness	.758	1.319		
	WTP	.846	1.183		
	DataSecurity	.901	1.110		
	EaseOfUseTotal	.830	1.205		
a. Dependent Variable: NEWPersonality					

Fig. 6.2. Multicollinearity Checks using the variance inflation factors (VIF)

Mode	el	Tolerance	VIF
1	WTP	.961	1.040
	DataSecurity	.798	1.253
	EaseOfUseTotal	.840	1.191
	NEWPersonality	.792	1.263
	Dependent Variable		

Dependent Variable: PerceivedUsefulness