Determinants of Online Social Business Network Usage Behavior—Applying the Technology Acceptance Model and Its Extensions

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Usage of online social business networks like LinkedIn and XING have become commonplace in today’s workplace. This research addresses the question of what factors drive the intention to use online social business networks. Theoretical frame of the study is the Technology Acceptance Model (TAM) and its extensions, most importantly the TAM2 model. Data has been collected via a Web Survey among users of LinkedIn and XING from January to April 2010. Of 541 initial responders 321 finished the questionnaire. Operationalization was tested using confirmatory factor analyses and causal hypotheses were evaluated by means of structural equation modeling. Core result is that the TAM2 model generally holds in the case of online social business network usage behavior, explaining 73% of the observed usage intention. This intention is most importantly driven by perceived usefulness, attitude towards usage and social norm, with the latter effecting both directly and indirectly over perceived usefulness. However, perceived ease of use has—contrary to hypothesis—no direct effect on the attitude towards usage of online social business networks. Social norm has a strong indirect influence via perceived usefulness on attitude and intention, creating a network effect for peer users. The results of this research provide implications for online social business network design and marketing. Customers seem to evaluate ease of use as an integral part of the usefulness of such a service which leads to a situation where it cannot be dealt with separately by a service provider. Furthermore, the strong direct impact of social norm implies application of viral and peer-to-peer marketing techniques while it’s also strong indirect effect implies the presence of a network effect which stabilizes the ecosystem of online social business service vendors.

Keywords: Technology Acceptance Model (TAM); Online Social Business Networks; Usage Intention; Structural Equation Modeling; Network Effect

Introduction

Online social business networks have been growing rapidly in recent years, with LinkedIn and XING being among the most established (Boyd & Ellison, 2007). So far, only a small number of studies have used psychological methods to investigate factors which are relevant for typical usage behavior. Especially noteworthy theoretical approaches in this field are the Technology Acceptance Model (TAM) formulated by Davis (1986) and its extensions (Venkatesh & Davis, 2000), which are aiming on explanation of usage intention.

Despite minor criticisms (e.g. Straub & Burton-Jones, 2007; Benbasat & Barki, 2007), the Technology Acceptance Model is widely accepted and empirically tested, including several meta-analyses (Yousaiz, 2007; Schepers & Wetzels, 2007; Ma & Liu, 2004; King & He, 2006). Recently a series of applications of the Technology Acceptance Model in other Internet and Web 2.0 technologies (e.g. internet shopping malls (Gentry & Calantone, 2002), online banking (Pikkarainen et al., 2004), mobile internet user acceptance (Venkatesh et al., 2003) and internet acceptance in general (Kim et al., 2007) have been published.

Given the impression of the repeated successful application of this type of model in related fields of online services, it seemed promising to us to employ them for our focus on Online Social Business Networks. This promise is also underlined by the fact that the family of TAM-models also focuses on the effects of peer influences, which should be a key driver for the online social networking services in focus.

Models and Research Questions

The original Technology Acceptance Model developed by Davis (1989) is based on the works of Fishbein & Ajzen (1975) and Ajzen & Fishbein (1980) who developed the Theory of Reasoned Action (TORA) which will be outlined as a starting point. Within the framework of TORA, two determinants explain the intention to perform a specific behavior. The first de-
terminant is attitude towards the behavior, which is defined as the degree to which performance of the behavior is positively or negatively valued (Ajzen & Fishbein, 1980). The second determinant is subjective norm which is the perceived social pressure to engage or not to engage in a behavior (Ajzen & Fishbein, 1980). The intention to perform a specific behavior finally affects whether people perform the action.

Building on this structure of a psychological attitude model, the Technology Acceptance Model was developed by Davis et al. (1989) to explain the use of information systems and computer adoption. TAM (Davis, 1989) consists of five constructs which channel the influence of external variables specific to the application. The model is depicted graphically in Figure 1. Its schema can be described as follows: the external variables influence perceived usefulness (PU) and perceived ease of use (PEU) of an information system, which are both influencing a person’s attitude towards use (ATT), in turn influences behavioral intention to use (INT), which finally influences actual usage behavior. A major difference to TORA is that the original TAM does not include effects of subjective norm. Taylor & Todd (1995) provide a more detailed comparison of the two models.

It should be further noted that Davis later recommended to drop attitude (ATT) out of the model formulation (Davis et al., 1989), while it has been reintroduced in the newer TAM2 extension of the model (Venkatesh & Davis, 2000).

The variables are specified as follows: Perceived usefulness (PU) is defined as the extent to which a person believes that using a specific technology/information system will enhance her performance (Lee et al., 2003). The meta-analysis of Scheepers & Wetzels (2006) reports PU being one of the strongest direct and indirect influences on usage intention. Perceived ease of use (PEU) is defined as the extent to which a person believes that using the specific technology/information system will be free of effort (Lee et al., 2003). Scheepers & Wetzels (2006) report only substantial indirect effects of PEU on ATT over PU, direct effects of PEU on ATT and INT are reported as being often weak.

The lack of inclusion of the Subjective Norms (SN) which was deemed relevant for system usage led to development of the already mentioned extension of the Technology Acceptance Model by Venkatesh & Davis (2000) called TAM2. The model is depicted graphically in Figure 2.

TAM2 is of course of special interest for us, given our focus on Social Business Networks and served as theoretical backbone of this study. In our application it has been further extended by inclusion of external moderators which will be discussed in detail in the operationalization section. According to the theory they can be classified into the following groups: organizational factors, technology factors such as technology complexity (operationalized by PEU) and individual factors such as age, gender or experience (Sun & Zhang, 2006). Given that Social Business Networking is an individual and non-corporate behavior, we did not include organizational factors in our study.

If our approach was to be summarized in a single sentence, it would be: “What are the psychological mechanisms associated with Social Business Network usage, seen from a perspective of attitude research in terms of the TAM2 model?”

Of course, such a statement needs some decomposition in order to be tested systematically and empirically. In particular, we focused on examining several research questions of different granularity, which arose before the background of the TAM2 model.

The first and most obvious question is whether TAM2 is an appropriate approach to model Social Business Network usage. Given that TAM2 can be considered appropriate, the next question regards to the pattern of effects as postulated by the model. What are the strongest influences and what are the orders of magnitude of direct and indirect effects? Given the case of Social Networking Services we expect the influence of Social Norm to be substantive. However, the relative roles of Perceived Usefulness and Perceived Ease of Use of the technology are not clear at this point and will be subject to closer examination.

Hypotheses

H1 (ATT > INT): The more positive the attitude (ATT), the stronger the intention (INT) to use online social business networks in the next four weeks.

H2 (SN > INT): The more positive the attitude of the subjective norms (SN), the stronger the intention (INT) to use online social business networks in the next four weeks.

H3 (PU > ATT): The more positive the perceived usefulness (PU), the more positive the attitude (ATT) towards the use of online social business networks in the next four weeks.

H4 (PEU > PU): The more positive the perceived ease of use (PEU), the more positive the perceived usefulness (PU) towards the use of online social business networks in the next four weeks.

H5 (PEU > ATT): The more positive the perceived ease of use (PEU), the more positive the attitude (ATT) towards the use of online social business networks in the next four weeks.

H6 (SN > PU): The more positive the attitude of the subjective norms (SN), the more positive the perceived usefulness (PU) towards the use of online social business networks in the next four weeks.
Sample

Data were collected via a Web survey of social business network users over four months, from January 2010 to April 2010. Social business network users of the two major social business networks, LinkedIn and XING, were invited to participate in the survey. The questionnaire was hosted on a university server using the open-source survey tool LimeSurvey. A total of 541 respondents started the questionnaire with 321 finishing it.

Thirty-three percent of the respondents were female, and sixty-seven percent were male. Over seventy percent of the respondents were between thirty one and fifty years old. About half of the respondents were self-employed. The empirical distribution of gender is very close to those communicated by XING (XING Mediendaten, 2010) and shows a lesser proportion of women than communicated by LinkedIn (LinkedIn Audience, 2013). The distribution of ages deviates slightly from those communicated both by XING and LinkedIn, showing a lower proportion in the age class from 27 to 30 years and higher proportion in the age class from 41 to 50 years. Table 1 summarizes the samples demographics. Taken together, our sample shows small deviations from what can be considered the population’s marginal distribution. However, we can see no evidence for a systematic sampling or self-selection bias.

Operationalization

Before formulating the questionnaire items, qualitative interviews of about 30 minutes in length had been conducted with eight persons to gather deeper insight into the behavior under investigation and its relevant determinants, following recommendations by Fishbein & Ajzen (2010).

After analyzing the collected qualitative interviews, questions had been formulated in concordance to recommendations of Davis et al. (1989) and Fishbein & Ajzen (2010). All constructs have been operationalized using multi-item scales to avoid problems regarding reliability (Ajzen, 2005). After formulation of the first draft of the questionnaire, cognitive pre-tests had been conducted with several persons. This resulted in shortening the quite lengthy questionnaire and a reformulation of selected items.

Respondents were asked to score on a 7 point Likert scale in the questionnaire with different endpoints specific to the question asked. The final version was coded in the online questionnaire tool and tested again. Several items have been excluded after the pretests and a smaller amount has been excluded after data collection and conducting some descriptive analyses. All items were formulated in German, so we will only present English translations here, which can be found in Table 2. The original questionnaire can be requested from the authors.

Data Analysis

Confirmatory Factor Analysis

Confirmatory factor analysis was used to examine the convergent validity of each multi-item scale, through specifying a

Table 1.
Demographic of the survey respondents.

<table>
<thead>
<tr>
<th>Measure</th>
<th>Frequency</th>
<th>Percent</th>
<th>Valid N</th>
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<tbody>
<tr>
<td>Gender</td>
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<td></td>
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<tr>
<td>Female</td>
<td>105</td>
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<td></td>
</tr>
<tr>
<td>Male</td>
<td>212</td>
<td>66.9</td>
<td></td>
</tr>
<tr>
<td>Age</td>
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<td></td>
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<tr>
<td>Under 26</td>
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<tr>
<td>27 - 30</td>
<td>30</td>
<td>9.6</td>
<td></td>
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<tr>
<td>31 - 40</td>
<td>107</td>
<td>34.2</td>
<td></td>
</tr>
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<td>41 - 50</td>
<td>121</td>
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</tr>
<tr>
<td>Self employed</td>
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<td>51.8</td>
<td></td>
</tr>
</tbody>
</table>

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separate single factor model for each of the five theoretical constructs. After testing every single factor, model all factor models were included simultaneously in a final model in order to investigate joint reliability and validity of measurement items.

We started with the measurement setup of the classical Technology Acceptance Model (TAM) developed by Davis (1989), but excluded the Attitude construct (ATT), as later recommended by Davis et al. (1989). Standardized factor loadings of the measurement items and correlations between latent constructs are depicted in Figure 3.

The chi-square statistic was found to be non-significant, indicating a close fit between model and data. All other indicators surpassed the recommended levels as well (compare Hu & Bentler, 1999), which indicates that the operationalization is statistically appropriate.

In the next step, we extended the core model by relevant constructs of Technology Acceptance Model 2 (Venkatesh & Davis, 2000). Again a setup of several tests had been performed. Figure 4 shows results for the extended Technology Acceptance Model (TAM2).

Here the chi-square statistic was found to be significant, indicating presence of a misfit between model and observations. However, it should be noted that the chi-square statistic is assumed to be generally very sensitive to sample size. If it is medium or large, as in our case where n equals 321, the statistic tends to be significant even though differences between the data generated and the proposed model are low. All other indicators surpass the recommended levels (compare Hu & Bentler, 1999), which leads us to the conclusion that the operationalization of the TAM2 is statistically appropriate.

### Structural Equation Modeling

Both the original Technology Acceptance Model (TAM) and its extended version, the Technology Acceptance Model 2 (TAM2) have been tested in a path model in order to answer our list of research questions. Both models show an acceptable overall fit to the data as can be seen from Figures 5 and 6. Again, TAM2 shows a significant misfit between model and observations, while its non-inferential fit-measures are looking well. As in the case of test of operationalization we conclude that this significant chi-square statistic is due to our relatively large sample size and does not indicate any serious misfit of the model. The proportion of explained variance of constructs is considerable with a maximum of 72.7% for Behavioral Intention (INT) in the case of TAM2.

Examining the pattern of effects of both models, the most striking observation is the both negligible and insignificant direct effect of Perceived Ease of Use (PEU) on Intention (INT) in both models (and on Attitude (ATT) in TAM2). These are contrasted by substantial indirect effects mediated over Perceived Usefulness (PU). In the big picture, this pattern can be interpreted as follows: It seems that across all users, social business networks are evaluated as easy to use and users utilize

![Figure 3](image1.png)

Simultaneous confirmatory factor analyses of the constructs included in the classic TAM.

![Figure 4](image2.png)

Simultaneous factor analyses of the constructs included in the extended TAM2.

![Figure 5](image3.png)

Path model for the classic Technology Acceptance Model.

![Figure 6](image4.png)

Path model for the extended Technology Acceptance Model 2.
the platforms rather than satisfy their social needs (as measured by Perceived Usefulness (PU)) instead of being attracted by the technical innovation.

As expected, the influences of Social Norm (SN) on Attitude (ATT) and Behavioral Intention (INT) are considerable, with even stronger indirect effects being mediated over Perceived Usefulness (PU). This can be interpreted as evidence for the presence of a “network-effect” regarding usage-related utility which exceed direct influences of peers in magnitude.

**Discussion**

The above results have immediate consequences for design and marketing of Social Business Networks. First, and contrary to hypothesis of both original and modified Technology Acceptance Model, there is no evidence for a direct effect of Perceived Ease of Use (PEU) which can be considered in separation from Perceived Usefulness (PU).

Second, there is a considerable direct effect of Social Norm (SN) which implies potential for peer-to-peer and recommendation marketing techniques. This effect is exceeded in magnitude by an indirect effect mediated by Perceived Usefulness (PU) which implies a substantial “network effect”. Given such a mechanism, growth of social networking services seems to be autocatalytic to some extent, leading to a situation where market potential is absorbed by larger or more mature players.

Above findings also have implications for theory since they raise the question whether there are significant moderators which could eventually explain the vanishing direct effect of Ease of Use (PEU). Castaneda et al. (2007) examines the moderating effect of user experience on TAM’s structure of coefficients, which appears promising for our present situation.

**REFERENCES**


