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Computers in Human Behavior

Computers in Human Behavior xxx (2007) xxx-xxx

www.elsevier.com/locate/comphumbeh

Online flow experiences, problematic Internet use and Internet procrastination

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Abstract

This study explores the theoretical and practical overlap between online procrastination, prob-10 lematic Internet use, and flow on the Internet. At the theoretical level there is a great deal of inter-11 relatedness between these three concepts (for example, all three concepts deal with issue of a lack of 12 13 control over time spent online and acknowledge the distracting and entertaining properties of the Internet); yet, one can also argue that the concepts are theoretically distinct (for example, flow is 14 15 a total absorption in the work at hand, whereas procrastination is the avoidance of the work at 16 hand). All three concepts have been used to describe either desirable (flow) or undesirable (procras-17 tination and problematic Internet use) states when online. In this study a sample of 1399 Internet 18 users was obtained from a survey placed on a South African online information technology magazine. Using the problematic Internet use questionnaire (PIUQ), the distraction subscale of the online 19 cognition scale (OCS), and a modified version of the Flow scale it was found that there were strong 20 21 positive relationships between all three variables (the strongest relationship being between problematic Internet use and online procrastination). The results also suggested that procrastination may be 22 23 a connector between PIU and flow; also that PIU is a connector between procrastination and flow, but that flow is independent of the relationship between PIU and procrastination. These results are 24 discussed in relation to previous studies on problematic Internet use and in particular, whether these 25 26 relationships are unique to respondents involved in the information technology sector. 27 © 2007 Published by Elsevier Ltd.

Keywords: Procrastination; Flow theory; Flow experiences; Problematic Internet use; Information technology workers

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31 **1. Introduction**

The use of the Internet (i.e. the World-Wide Web, email, and other associated online 32 applications) for personal purposes while at work is said to cost organisations large 33 amounts of time and money (Mahatanankoon, Anandarajan, & Igbaria, 2004). Most 34 researchers have labelled the personal use of the Internet at work as deviant, using terms 35 such as Internet abuse in the workplace (Anandarajan, 2002), problematic Internet use 36 (Davis, Flett, & Besser, 2002), cyberloafing (Lim, 2002), and even Internet addiction (Grif-37 fiths, 2003a; Young & Case, 2004). This has resulted in organisations trying to control the 38 amount of time that employees spend using company infrastructure, equipment and time. 39 Interventions have included online monitoring, training, rehabilitative action (e.g. disci-40plinary action or counselling), or policy control, each with varying degrees of success 41 (Whitty & Carr, 2006; Young & Case, 2004). On the other hand, Belanger and Van Slyke 42 (2002) have argued that a certain level of play and experimentation (i.e. non-work use of 43 the Internet) is desirable and can lead to learning that is of significant value to an organi-44 sation. Of course, this raises questions of how much time an employee should be allowed 45 to spend online during work times for personal use and also what types of activities might 46 lead to play and experimentation that would ultimately be beneficial to the organisation 47 (Mahatanankoon et al., 2004). This view of online procrastination however, lacks empir-48 ical support. Empirical work suggests rather, that employees perform non-work activities 49 out of habit (Lee, Lim, & Wong, 2005). There is some evidence that procrastination activ-50 ities provide relaxation and escape opportunities which people perceive as positive (Woods 51 & Griffiths, in press). 52

There is now a large, and growing, body of work that has examined the causes, corre-53 lates, predictors, and outcomes of problematic Internet use (PIU). In addition, Davis 54 et al. (2002) has conceptualised procrastination as another aspect of PIU. There is also 55 a small, but growing, interest in examining optimal online experiences or online flow 56 (Chen, 2006; Chen, Wigand, & Nilan, 1999; Hoffman & Novak, 1996; Pace, 2004; Wan 57 & Chiou, 2006). As is shown in the following sections, superficially, there is a great deal 58 of overlap in the definitions and practical conceptualisations of these three concepts; PIU, 59 online flow experiences, and procrastination online. This study explores the theoretical 60 and practical overlaps between these three concepts within a population of working 61 adults. 62

63 1.1. Problematic Internet use (PIU)

In recent years there have been a wide variety of investigations examining the "addic-64 tive" use of the Internet and the predictors of Internet addiction in studies from around 65 the world (Campbell, Cumming, & Hughes, 2006; Nalwa & Anand, 2003; Ng & Wie-66 mer-Hastings, 2005; Palesh, Saltzman, & Koopman, 2004; Simkova & Cincera, 2004; 67 Suhail & Barges, 2006; Whang, Lee, & Chang, 2003; Yang & Tung, 2006). A great deal 68 of the debate has centred on whether the Internet is indeed "addictive" given that there 69 is no psychiatric classification for "addictions" in the DSM-IV (American Psychiatric 70 Association, 2000). Instead, there has been a proliferation of terms to describe the "over-71 use" or "abuse" of the Internet. Terms used have included "unregulated Internet use" 72 (LaRose, Lin, & Eastin, 2003), "compulsive Internet use" (Meerkerk, Van Den Eijnden, 73 & Garretsen, 2006), "Internet dependence" (Yuen & Lavin, 2004), "excessive Internet 74

use" (Suhail & Barges, 2006), "pathological Internet use" (Davis, 2001; Niemz, Griffiths, 75 & Banyard, 2005), and "problematic Internet use" (Caplan, 2002; Davis et al., 2002). 76 Alternatively, some researchers argue that a person's overuse or abuse of the Internet is 77 a behavioural manifestation of other things that may be problematic in their lives (Caplan, 78 2002). For example, online gambling may rather be a sign of pathological gambling as 79 opposed to Internet addiction, or spending excessive amounts of time searching for online 80 pornography may be indicative of some form of sexual "addiction". Another contention is 81 that spending an excessive amount of time online may be a way of avoiding marital or 82 work problems or from escaping other social confrontations (Caplan, 2002). 83

The term "problematic Internet use" is preferred in this study because it specifically 84 avoids emotive qualities that are attached to terms such as "addiction" and "pathology" 85 (Caplan, 2002). Instead, we adopt LaRose et al. (2003) approach which is based on Ban-86 dura (1991) social cognitive theory of self-regulation. This approach takes the view that 87 excessive behaviour (in one's self-monitoring, judgments, and self-reactions) is something 88 that may periodically arise and that may, over time, be self-remedied. Within this theory 89 an Internet user is not simply diagnosed as an Internet "addict" (also referred to as a 90 "dependent" in some studies) or as "normal". Rather, a user may experience periods of 91 deficient self-regulation (that manifests as overuse of the Internet) followed by periods 92 of spontaneous self-regulated behaviour (or "normal" (Davis, 2001) use of the Internet). 93 Deficient self-regulation occurs when a person believes they do not have the ability to be 94 able to stop their behaviour. In the absence of efficient self-regulatory mechanisms individ-95 uals are unmotivated, bored, and remain uncertain about their own abilities. Bandura 96 (1991) noted that in order to relieve these feelings that an individual would resort to 97 self-anaesthetisation (e.g. taking drugs or other substances) or other means to escape from 98 reality (i.e. spending excessive amounts of time avoiding an activity in which they lack self-99 100 efficacy).

Of course, the term "problematic Internet use" is rather ambiguous as it implies that all 101 forms of online criminal, negligent, inappropriate, or excessive behaviour might be encom-102 passed by the term. Here, we use Beard and Wolf (2001) definition of PIU as the "use of 103 the Internet that creates psychological, social, school, and/or work difficulties in a person's 104 life" (p. 378). Beard and Wolf (2001) specifically exclude Internet-related crimes (down-105 loading or distributing illegal material, hacking, and online sexual harassment) from this 106 definition. While these behaviours are certainly problematic they are also recognised in 107 many countries as crimes, whereas the behaviours we wish to describe by PIU are socially 108 or psychologically problematic for the individual engaging in these activities without being 109 inherently "crimes". The definition of PIU might also include specific PIU such as exces-110 sive online gambling or spending excessive amounts of time viewing and downloading por-111 nography (Griffiths, 2003a). 112

Within the "addiction" model of PIU, symptoms include some or all of the following: 113 needing to spend increasing amounts of time online (Niemz et al., 2005), loss of control in 114 the amount of time spent online (LaRose et al., 2003), unsuccessful attempts to reduce the 115 amount of time spent online (Beard & Wolf, 2001), preoccupation with online activities 116 (Beard, 2005), withdrawal symptoms (Niemz et al., 2005), changes in mood or anxiety lev-117 els (Campbell et al., 2006; Yang & Tung, 2007), denial or deception regarding the amount 118 of time spent online (Caplan, 2002; Yang & Tung, 2007), and negative social, physical, 119 financial or emotional consequences as a result of spending time online (Widyanto & 120 McMurran, 2004). Within the self-regulation model the same symptoms are recognised 121

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as indicators of deficient self-regulation although this model recognises a continuum of deficient self-regulatory behaviours rather than the absolute presence/absence of a certain proportion of these symptoms as symptomatic of PIU (LaRose et al., 2003).

125 *1.2. Internet procrastination*

Procrastination refers to the inability of an individual to convert an intention to act into 126 the actual performance of an act (Blunt & Pychyl, 2005). Lavoie and Pychyl (2001) noted 127 that procrastination refers not to the complete avoidance of a task, but rather to delaying 128 the commencement or completion of a task. There is still little consensus over whether pro-129 crastination is a personality trait or a consequence of a person's situational circumstances. 130 The personality trait approach to procrastination shows mixed evidence for relationships 131 between procrastination behaviour and numerous personality traits such as extraversion, 132 low levels of conscientiousness, and external locus of control (Senécal, Lavoie, & Koest-133 ner, 1997; Steel, Brothen, & Wambach, 2001). The situational approach to procrastination 134 notes that individuals tend to procrastinate more when the focus tasks are perceived as less 135 enjoyable or intrinsically unpleasant (Blunt & Pychyl, 2005). Within this approach, if a 136 task is made too difficult or boring then it would be perceived by individuals as unpleasant 137 and would be avoided (Senécal et al., 1997). Similarly, procrastination is most likely to 138 occur when a person is feeling anxious as a result of being evaluated or where they feel 139 they lack control in completing a task (Senécal et al., 1997). It must be noted that procras-140 tination is only remotely related to problems of time management as procrastinators often 141 know exactly what they should be doing, but still fail to do it. Most, but not all, evidence 142 suggests that procrastination is associated with poorer performance and negative work 143 outcomes (Steel et al., 2001). 144

Procrastination using the Internet, also referred to as cyberslacking (Lavoie & Pychyl. 145 2001) or cyberloafing (Lim, 2002), usually refers to people using the Internet (at work) for 146 non-work (personal interest) purposes in order to avoid doing work-tasks because these 147 are perceived as boring, unpleasant, or too challenging. Lim (2002) argues that these types 148 of online behaviours are part of production deviance in the workplace, resulting in lower 149 employee outputs. Lavoie and Pychyl (2001) found that 51% of respondents in their survey 150 reported frequent Internet procrastination. Also, Greenfield and Davis (2002) reported 151 that employees in their sample spent more than 3 h/week online conducting non-work-152 related activities. The Internet provides people with a wealth of entertaining distractors 153 such as emailing or chatting to friends and/or family online, as well as access to sports 154 scores, entertainment news, and hobbies. The Internet is frequently perceived as entertain-155 ing and positively related to perceived stress relief (Lavoie & Pychyl, 2001). Lim (2002) 156 suggested that using the Internet to procrastinate is relatively easy for most employees 157 (especially those employees who are expected to use computers) to disguise, since "cyber-158 loafers" have the appearance of being at work while actually engaging in non-work tasks. 159

160 1.3. Flow on the Internet

Flow describes a state of intrinsic enjoyment in the task at hand that occurs when a person's perceived skills closely match the perceived demands or challenges of that task (Csikszentmihalyi, 1997). Csikszentmihalyi (1997) listed eight components of flow; a clear goal, challenges that match an individual's skills, control over the task, immediate and effi-

cient feedback, concentration and focus, loss of self-consciousness, loss of a sense of time. 165 and an activity that becomes autotelic (i.e. a task is perceived worthy for its own sake). 166 Flow is therefore a state of consciousness (usually characterised by a loss of a sense of time 167 passing) that is sometimes experienced by individuals who are deeply involved in an activ-168 ity they are enjoying. Flow is most often experienced by individuals as pleasure, satisfac-169 tion and as playfulness (Webster, Trevino, & Ryan 1993). Unfortunately, most studies 170 inconsistently assess these eight factors. For example, Webster et al. (1993) only assessed 171 the control, concentration and focus, and autotelic components. For employers, the state 172 of flow is desirable if it refers to relevant work tasks. However, where employees need to 173 attend to many different tasks (especially if the control element of flow is absent), or where 174 the state of flow is on a non-work task, employers would consider flow to be non-desir-175 able. Flow might have a negative impact on work performance if it not directed towards 176 the person's required work task. 177

Numerous authors have applied flow to human-computer interactions (Sharafi, Hed-178 man, & Montgomery, 2006; Webster et al., 1993) and to Internet usage (Chen, 2006; Chen 179 180 et al., 1999; Hoffman & Novak, 1996; Pace, 2004; Wan & Chiou, 2006). Hoffman and Novak (1996) and Chen (2006) have argued that qualities of the World-Wide Web such 181 as control, ease of use, immediate feedback, interactivity, and access to entertainment 182 would inherently make the World-Wide Web open to flow experiences. Chen (2006) found 183 that most World-Wide Web users therefore experience positive moods while online. In 184 addition, Wan and Chiou (2006) have argued that the interactive and engaging properties 185 of online games offer users the opportunity for optimal (flow) experiences. Chen et al. 186 (1999) found that some of the most flow-inducing activities were information retrieval, 187 reading and writing emails, creating webpages, playing online games, and chatting online. 188 Pace (2004) also recognised that certain aspects (such as poor interface usability, interest-189 ing content, and offline distractors) of the online environment act as distractors to optimal 190 flow experiences in online information-seeking, suggesting that there are also qualities of 191 the Internet that act as potential procrastination distractors as well as qualities that 192 encourage flow. 193

194 *1.4. Overlaps between PIU, and Internet procrastination and flow on the Internet*

Davis et al. (2002) consider the relationship between Internet procrastination and PIU 195 to be so close that they include Internet procrastination in their Online Cognition Scale as 196 part of the measure of PIU. The definition of procrastination involves escaping from work 197 that is unpleasant, boring or too difficult. As the Internet is often described as entertaining 198 and interesting, it is naturally assumed to be a good distractor to enable procrastination 199 (Davis et al., 2002; Lavoie & Pychyl, 2001). Within the deficient self-regulatory approach 200 to PIU (LaRose et al., 2003) the excessive use of the Internet may be a sign that someone is 201 bored, unmotivated, or uncertain about their abilities in completing their work tasks and 202 may therefore escape to non-work tasks on the Internet. In addition, some of the symp-203 204 toms of PIU include issues of time management (i.e. unsuccessful attempts to reduce using the Internet and being deceitful about how much time is spent online). While time man-205 agement is only loosely related to procrastination, problematic Internet users are often 206 acutely aware of the excessive amount of time they spend online (as is someone who pro-207 crastinates). Furthermore, using the Internet to procrastinate at work could be an expla-208 nation for why problematic Internet users report productivity losses at work. 209

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Hypothesis 1: We expect a strong positive correlation between Internet procrastination and PIU.

There are quite a number of the characteristics of flow that are strongly reminiscent of 213 PIU. The most obvious example of flow overlapping with PIU is time distortion (i.e. due 214 215 to absorption in the task at hand). One of the symptoms of PIU is the loss of control over the amount of time spent online. Someone who is in a state of flow is characterised as 216 being so deeply involved in the task at hand that they are unaware of time passing 217 218 (Csikszentmihalyi, 1997). However, it is interesting to note that flow is most often charac-219 terised as a positive state, whereas PIU is characterised as a negative state. This is because 220 flow usually refers to tasks relevant to the job being performed, whereas PIU implies 221 engaging in non-relevant job tasks. However, if one is in a "state of flow" when engaging 222 in a non-job-related task one might easily categorise this as resembling aspects of PIU. It is 223 also possible for someone to be in a state of flow for work-related tasks and yet still exhibit 224 symptoms of PIU. For example, if someone becomes so absorbed in their online work that 225 they lose track of time it is possible that they may miss other social and family commit-226 ments. While this might be desirable for an organisation (at least in the short term) it may result in negative family or social consequences (one of the symptoms of PIU) over 227 228 extended periods of time. However, there are also some characteristics of flow that are 229 missing from PIU. In particular, problematic Internet users are not characterised as hav-230 ing a clear goal or a sense of control over the task at hand. Indeed, Wan and Chiou (2006) found that experiences of flow in online gaming were negatively correlated with an addic-231 tive inclination. However, Wan and Chiou (2006) found that "addicted" online gamers did 232 233 not experience high flow states (in fact "addicts" flow states were significantly lower than 234 "non-addicts"). This contradictory finding suggests that while a flow state might lead to 235 more potentially problematic behaviours (like spending more time engaging in an activity), 236 there are other factors that are better predictors of actual addiction.

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Hypothesis 2: We would expect a weak negative relationship between PIU and flow.

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1.5. PIU, Internet procrastination, and flow on the Internet for different Internet activities

A recent review would suggest that PIU for specific types of online activities (such as 240 online gambling, cybersex, or online gaming) received more support than the contention 241 that Internet addiction requires its own psychiatric classification (Yellowlees & Marks, 242 2007). Davis (2001) would refer to this as "specific pathological Internet use" (p. 188) 243 in his cognitive-behavioural model. Evidence from a number of studies have shown that 244 interactive activities (e.g. erotica, online chatting, online games, and online shopping) tend 245 246 to result in higher levels of PIU than activities such as email and information-seeking 247 Q1 (Leung, 2004; Li & Chung, 2006; Meerkerk et al., 2006; Whang et al., 2003). On the other 248 hand, Widyanto and McMurran (2004) found no relationship between PIU and the interactive activities of the Internet. Other researchers have supported these findings by focus-249 sing on populations of Internet users who engage in "higher risk" activities such as online 250 gambling (Griffiths, 2003b) or online gaming (Ng & Wiemer-Hastings, 2005; Wan & 251 Chiou, 2006). Evidence from the flow literature would also suggest that some activities 252 253 are more flow-inducing than others (Ghani & Deshpande, 1994; Wan & Chiou, 2006).

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Griffiths (2003b) argues that the online environment changes the situational and struc-254 tural characteristics of an activity. For example, issues such as ease of access, affordability 255 and anonymity might make online pornography and other online sexual pursuits more 256 attractive for some people. The characteristics of an activity that make it more absorbing 257 and susceptible to excessive use include emotional or mental escape, immersion, and inter-258 activity (Griffiths, 2003b). Woods, Griffiths, Chappell, and Davies (2004) add skill devel-259 opment to these characteristics. Woods et al. (2004) imply therefore that tasks with high 260 flow characteristics are positively related to PIU. 261

Hypothesis 3: Online activities that are more immersive, interactive, and which foster
 mental or emotional escape, or skill development will have higher levels of PIU and
 flow on the Internet.

266 **2. Methods**

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267 *2.1. Sample*

While access to the Internet has grown by approximately 113% in South Africa in the 268 last 6 years, the Internet penetration rate is still rather modest compared to developed 269 countries at approximately 10% of the total population (Internet World Stats, 2006). 270 For example, the Internet penetration rates in Sweden (75%), United Kingdom 271 (63%), Japan (67%), and the United States (69%) are significantly higher (Internet 272 World Stats, 2006). However, in certain sectors in South Africa the Internet penetration 273 rate is much higher, particularly in the employed, skilled, urban population (Wolma-274 275 rans, 2004). The majority of studies that have identified groups "at risk" of problematic Internet use have focused on students (Davis et al., 2002; LaRose et al., 2003; Niemz 276 et al., 2005; Simkova & Cincera, 2004; Suhail & Barges, 2006) or high school chil-277 dren (Nalwa & Anand, 2003; Tsai & Lin, 2001; Wan & Chiou, 2006; Yang & Tung, 278 2007). A smaller proportion of more recent studies have used a general working 279 population (Meerkerk et al., 2006; Whang et al., 2003; Widyanto & McMurran, 280 2004). Within the context of low Internet penetration rates, people with access to com-281 puter technology and online facilities are more vulnerable to problematic Internet use 282 than those with little access. It is for this reason that this study investigates a techno-283 logically literate working population accessed from an online information technology 284 magazine. 285

A total of 1399 usable responses were obtained. This sample included 1065 males and 286 334 females, most likely in the age groups 24–29 years old (N = 378) and 30–35 years old 287 (N = 314), with a Diploma (N = 403) or Bachelor's degree (N = 259). The majority of 288 the respondents had access to the Internet from work (N = 1311) and/or home 289 (N = 1039), usually with access from both locations. Most respondents had been using 290 291 the Internet for longer than 5 years (N = 1196) and usually connected to the Internet in sessions of fewer than 2 h (N = 703) or between 2 and 5 h (N = 393) at a time per 292 day. Respondents accessed the Internet 7 days a week (N = 566) with a large number 293 of respondents only accessing the Internet 5 days a week, every work day (N = 495). 294 Nearly all respondents used the Internet for email (N = 1325) although a large propor-295 296 tion of respondents also used other online communication facilities such as online

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297 chatting (N = 239), Instant Messaging (N = 593), and Online telephony (N = 597). A total of 633 respondents identified themselves as working in the information technology 298 sector and 737 respondents worked in other sectors such as healthcare, retail, marketing, 299 and banking or finance. t-Test comparisons between respondents in the information tech-300 nology section and respondents working in other sectors produced no statistically signif-301 icant differences for the three primary variables of PIU (t = .01, NS), Internet 302 procrastination (t = 1.17, NS), and flow on the Internet (t = 1.21, NS). The two informa-303 tion sectors and other sectors were therefore treated as a single group for the analyses. 304 305 This lack of significance between sectors might be an artefact of the sample. The sample 306 was obtained from an information technology magazine website. It is therefore highly likely that all respondents have (at least) some interest in information technology issues 307 even if they were not formally employed in that sector. A respondent might be employed 308 in another sector, but may be involved in information technology or may even be 309 310 involved with work on the Internet (e.g. a Web developer for an insurance company). 311 Alternatively, just because a respondent is employed in the information technology 312 sector, does not necessarily mean that they use the Internet as an integral part of their 313 iob.

314 2.2. Questionnaire

The questionnaire consisted of five different sections. The first section consisted of nine 315 biographical questions, including gender, age, education, marital status, and job sector. 316 The second section consisted of six questions about participants' use of the Internet, 317 including questions around where participants' connect to the Internet, the amount of 318 time spent online, and the types of activities that they engage with online. The third section 319 320 consisted of Thatcher and Goolam's (2005) Problematic Internet Use Questionnaire 321 (PIUQ). The PIUQ consists of 20 items, based on Young's (1996) criteria for Internet addiction, on a five-point Likert-type scale from Rarely or Not applicable (1) to Always 322 (5). The PIUQ was shown to have sound internal reliability ($\alpha = 90$) and validity in the 323 324 validation study (Thatcher & Goolam, 2005). The fourth section consisted of the distrac-325 tion subscale of Davis et al. (2002) Online Cognition Scale, used to assess procrastination. 326 The distraction subscale consists of seven items on a seven-point Likert-type scale from 327 Strongly Disagree (1) to Strongly Agree (7) with a high score indicating a strong propensity to be distracted (procrastinate) while using the Internet. Davis et al. (2002) found good 328 internal reliability ($\alpha = 81$) for the distraction subscale. The final section consisted of an 329 adaptation to Webster's et al, (1993) Flow scale. The Flow scale consists of 12 items 330 adapted from Webster et al.'s (1993) original scale by replacing the words "Lotus 1-2-331 3" with "the Internet". The Flow scale is measured on a seven-point Likert-type scale 332 from Strongly Disagree (1) to Strongly Agree (7) with a high score indicating strong flow 333 334 experiences. Webster et al. (1993) reported adequate internal reliability for the original Flow scale ($\alpha = .72$). The PIUQ ($\alpha = .92$), distraction subscale ($\alpha = .89$), and the adapted 335 336 Flow scale ($\alpha = .74$) each demonstrated satisfactory internal reliability and appropriate factorial validity for this sample. While the internal reliability for the Flow scale was 337 not high it fell within the acceptable limits of .70 for survey research (Anastasi, 1988; Mur-338 phy & Davidshofer, 2005) and the factor analysis suggested a single factor solution (of 339 flow) rather than a solution that supported Webster et al. (1993) individual flow 340 dimensions. 341

2.3. Procedure 342

The questionnaire was placed on the website of a prominent South African information 343 technology magazine, with two different links from the homepage (one link in the "Sur-344 veys" menu and another link from the lead story). While the website had a particular 345 South African information technology focus it is not possible to state categorically that 346 only people resident in South Africa responded to the questionnaire, given the nature of 347 the World-Wide Wide. A summary of the lead story was also included in the weekly email 348 sent to all subscribers to the online magazine. The links directed potential respondents to a 349 Q2 brief explanation of the research (see Appendix). Completion of the survey was voluntary 350 and respondents were able to complete the questionnaire at their own pace. Once respon-351 dents had completed the survey they were directed to a webpage that provided information 352 on the symptoms of problematic Internet use. Contact details of a free counselling service 353 were provided on this webpage for follow-up consultations if a respondent felt they 354 required additional assistance. The questionnaire was made available on the website for 355 12 days. Thereafter, the links were removed from the website, the data was converted into 356 a spreadsheet and emailed to the researchers after removing any potentially identifying 357 indicators. 358

359 *2.4. Analyses*

As an initial first step, principal components factor analyses were run on each of the 360 three scales; PIUQ scale, Distraction subscale, and the Flow scale. Using Kaiser (1974) 361 criterion for unity and the slope of the scree plot, each of the scales were indicative of 362 single-factor solutions with the first factor explaining 42% (PIUO scale), 61% (Distraction 363 subscale), and 35% (Flow scale) of the variance, respectively. There are essentially three 364 options in calculating scale totals. Firstly, one can sum each of the items that make up 365 the scales as designed. However, since two of the scales have been amended from the ori-366 ginal scales there may be problems with respect to reliability and validity in this sample 367 and many of the items may not really contribute equally to the scale total. Secondly, 368 one could use the mean scores for each item to get a scale total, but this again would 369 assume that each item contributes equally to the scale total. Finally, one could use 370 weighted scores (based on the orthogonal rotation factor analysis). This option answers 371 the weaknesses inherent in the first two options as the weightings reflect the relative 372 "importance" of each item in the scale. The scale totals used in these analyses were based 373 on the weighted contribution from the factor analysis loadings of the first factor after vari-374 max rotation. The relationships between the key variables and potential scale overlaps 375 were examined using Pearson's correlation coefficients and partial correlation coefficients. 376 The partial correlations enable one to determine any possible moderating influences 377 between the three key variables in the study. The impact of different types of uses of 378 the Internet was examined in two ways. First, t-test comparisons were made between 379 the use and non-use of a particular Internet activity for each of the three key study vari-380 ables. Second, Extended Automatic Interaction Detection (XAID) procedure (Hawkins & 381 Kass, 1982) was used to determine which Internet activities were the best predictors of 382 each of the three key study variables. XAID is one of a family of regression-type data anal-383 ysis methods where the predictors are all categorical variables and the single dependent 384 385 variable is interval in nature. The XAID is best represented as a tree structure with the best

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predictor "splits" occurring higher in the tree. In order to further examine the relative contributions of flow on the Internet and Internet procrastination towards predicting PIU, a forwards stepwise multiple linear regression was computed. Due to the large sample size, the .01 and .001 levels of significance were chosen for the analyses (except for the XAID analyses where the .01 and .05 levels of significance were chosen due to the rigour of the XAID analysis).

392 **3. Results**

393 3.1. Means and standard deviations of the weighted scales

An examination of the means and standard deviations of the PIUQ and the Internet distraction scales indicates that they are skewed towards the lower end of the scale. This implies that the majority of the sample scored low on PIU and Internet procrastination. In contrast, an examination of the mean and standard deviation of the Flow on the Internet scale indicates that this scale is slightly skewed towards higher scores implying that experiences of flow on the Internet are relatively common Table 1.

400 3.2. Results of the overlaps between PIU, Internet procrastination, and flow on the Internet

The results of the correlations between the three constructs for the total sample are pre-401 sented in Table 2. Each of the correlations was positively statistically significant at p < .001402 with medium (between .40 and .60) to large (>.60) effect sizes. The correlation between 403 PIU and Internet procrastination was equivalent at .58 after partialling-out (all partial 404 correlations are also shown in Table 2) the effect of flow on the Internet. This would sug-405 406 gest that the strong positive relationship between PIU and procrastination is unaffected by their relationships with flow. The correlation between PIU and flow on the Internet was 407 reduced to .28 after partialling-out the effect of Internet procrastination. This would sug-408 gest that procrastination may be a connector between PIU and flow. Finally, the correla-409 tion between flow on the Internet and Internet procrastination was reduced to .18 after 410 411 partialling-out the effect of PIU. This would suggest that PIU is a connector between procrastination and flow. 412

413 *3.3. Multiple regression results with PIU as the dependent variable*

Assuming that PIU is the dependent variable in these relationships (i.e. PIU is a result of Internet procrastination and flow on the Internet) a multiple stepwise linear regression was computed. The multiple stepwise linear regression indicated that four variables explained a significant proportion of the variance (see Table 3). The variables of Internet

	Mana	Standard daviat
Table 1 Means and stand	dard deviations of the weighted s	scales

	Mean	Standard deviation	Max.	Min.
PIUQ	24.61	8.63	63.88	12.78
Internet distraction	16.41	7.59	37.95	5.42
Flow on the Internet	26.75	6.50	41.37	5.27

Table 2

Correlations and partial correlations (after partialling-out the third variable in parentheses) between PIU, Internet procrastination and flow on the Internet

	Internet procrastination	Flo	w on the Internet	
PIU	.67** (.58)	.48	.48** (.28)	
Internet procrastination	_	.44** (.18)		
** Significant at $p < .001$.				
Table 3				
Multiple stepwise linear regression on	to PIU			
	R^2	ΔR^2	F	
Internet procrastination	.45	-	849.40**	
Time spent online	.50	.05	529.99**	
Flow on the Internet	.54	.04	406.04**	
Number of days per week online	.55	.01	318.81**	
**				

** Significant at p < .001.

procrastination ($R^2 = .45$, p < .001), length of time online in a session ($\Delta R^2 = .05$, 418 p < .001), flow on the Internet ($\Delta R^2 = .04$, p < .001), and number of days per week online 419 $(\Delta R^2 = .01, p < .001)$ were found to be the most important variables in explaining the 420 greatest proportion of the variance in problematic Internet use within this sample. The 421 correlation between days per week online and length of time online in a session was sta-422 tistically significant, but weak (r = .16) suggesting that possibility of multicollinearity 423 was low. Also, these two variables are theoretically distinct (i.e. one person may spend 424 a moderate amount of time online every day of the week while another person might spend 425 a large amount of time online one day a week and yet both may experience problematic 426 Internet use) which would suggest the inclusion of both variables in the model. While 427 the regression model was statistically significant (p < .05) with the addition of 10 other 428 variables, these variables explained less than 1% additional variance each which was 429 deemed a negligible amount. It was likely that statistical significance in a number of these 430 instances might be an artefact of the large sample size. 431

432 3.4. Comparisons of PIU, Internet procrastination, and flow on the Internet by Internet 433 activity

For the PIU variable the *t*-tests showed that with the exception of using the Internet for 434 email and general web browsing, the use of all other activities was significant at the 435 p < .001 level. For the flow on the Internet variable, the *t*-tests showed that the use of 436 all 11 Internet activities produced a significantly higher level of flow at the p < .001 level. 437 The *t*-test results for the Internet procrastination variable showed that using the Internet 438 for email, general web browsing, file transfer protocols, and telnet were non-significant. 439 The use of news websites was significant at p < .01 and all other activities were significant 440 at $p \le .001$ (online chatting, instant messaging, online games, peer-to-peer file sharing, and 441 blogging. Due to the more nuanced results for the Internet procrastination variable, only 442 these means and *t*-test results are shown in Table 4. 443

Table 4

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Internet procrastination	Non-use mean	Use mean	t-Statistic
File transfer protocols (FTP)	16.40	16.43	08
Email	15.97	16.44	51
General web browsing	13.79	16.48	-2.07
Telnet	16.25	17.59	-2.17
News websites	16.05	17.32	-2.83^{*}
Online telephony	15.96	17.65	-3.67^{**}
Peer-to-Peer file transfers (P2P)	15.95	17.90	-4.08^{**}
Instant messaging	15.46	17.71	-5.48^{**}
Blogging	15.86	19.71	-6.66^{**}
Online games	15.83	20.45	-7.66^{**}
Online chat	15.57	20.47	-9.27^{**}

T-tests for Internet procrastination and Internet activities

Significant at p < .01.

** Significant at p < .001.

The XAID analyses produced some interesting patterns of association with the dependent variables. For the flow on the Internet variable, the best predictor was the use of online chat. If a respondent did not use online chat the next best predictor was using blogging. For the PIU variable a similar patter emerged except that two more variables were considered significant in the model. If a respondent did not use blogging then the use of online gaming was the next best predictor. If a respondent used online chat, the next best predictor was using online telephony. For the Internet procrastination variable the predictive abilities of the blogging and online gaming activities were reversed, with online gaming

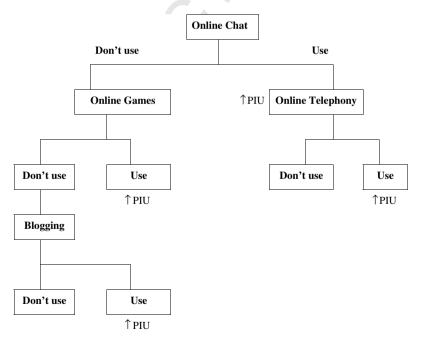


Fig. 1. XAID tree diagram for online activity predictors of PIU.

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being a better predictor of Internet procrastination if online chat was not used. In each case, only online chat was significant at p < .01. Other variables in the model were significant at p < .05. The XAID tree model for PIU is given in Fig. 1.

455 **4. Discussion**

456 *4.1. Overlaps between PIU and Internet procrastination and flow on the Internet*

The strong positive correlation between PIU and Internet procrastination (Hypothesis 457 1) was as expected. Davis et al. (2002) have made a strong argument for including procras-458 tination as a dimension of PIU despite the fact that they are distinct theoretical concepts. 459 One of the possible reasons for the strong correlation was the choice of Internet procras-460tination scale. The distraction subscale of Davis et al. (2002) Online Cognition Scale 461 (which Davis et al. (2002) used to assess PIU) was used to assess Internet procrastination 462 in this study. This may have exaggerated the strength of the correlation between PIU and 463 Internet procrastination. Despite this choice of scale, there is still good reason to believe 464 the relationship between PIU and Internet procrastination is positive and strong. The 465 Internet clearly has many distractive qualities (LaRose et al., 2003; Lavoie & Pychyl, 466 2001) since it is easy to learn to use, and it can provide appropriate challenges to those 467 that require the additional challenge. 468

The moderate positive correlation between PIU and flow on the Internet (Hypothesis 2) 469 was not as expected and is opposite to that obtained by Wan and Chiou (2006) who 470 looked specifically at online gaming. Flow is generally considered to be a positive individ-471 ual experience and PIU is considered to be a negative personal experience, therefore one 472 might expect a negative correlation between flow experiences on the Internet and PIU. The 473 474 result found in this study would imply that extended experiences of flow on the Internet might even result in PIU symptoms. In other words, flow experiences are the reason 475 why someone would spend longer than intended engaged in an online activity. Since most 476 respondents' experiences of flow on the Internet are moderately high and reported PIU is 477 relatively low, this might mean that a marginal increase in flow experiences on the Internet 478 would lead to small (non-diagnosable) increases in PIU symptoms. This would mean that 479 individuals and employers might benefit from additional flow experiences and yet not suf-480 fer too badly from PIU symptoms. However, this correlation also implies that a person 481 who has a very high PIU score is also likely to be in a high flow state and that person 482 might therefore be experiencing time distortion and a loss of control over the amount 483 of time spent on Internet tasks. Griffiths (2003b) provides an alternative explanation. 484 The positive relationship between flow on the Internet and PIU could be because they 485 share many of the same qualities (i.e. they are immersive, provide opportunities for per-486 ceived control, and are intrinsically motivating). These findings parallel the research inves-487 tigating the negative effects of work dependence or "workaholism" (Scott, Moore, & 488 Miceli, 1997; Spence & Robbins, 1992). The ambiguity inherent in the relationship 489 between flow on the Internet and PIU might (in future) be explained by different types 490 of PIU. For example, Scott et al. (1997) identified compulsive-dependent "workaholics" 491 as more at risk to harmful psychological and physical problems as a result of their work 492 behaviour than achievement-oriented "workaholics". 493

The results of the partial correlations also produce an interesting set of results. The relationship between PIU and Internet procrastination is independent of the impact of flow on

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the Internet on this relationship. Therefore, a person who is procrastinating is also likely 496 to be displaying symptoms of PIU (or vice versa). If they are also in a state of flow while 497 engaging in procrastinating activities this does not materially affect PIU. However, the 498 relationship between PIU and flow on the Internet is affected by the removal of Internet 499 procrastination as a variable. In other words, the relationship between PIU and flow on 500 the Internet is significantly stronger when a person is using the Internet to procrastinate. 501 Conversely, if a person is not using the Internet to procrastinate the relationship between 502 flow on the Internet and PIU is weakened. Similarly, the relationship between Internet 503 procrastination and flow on the Internet is affected by the removal of PIU as a variable. 504 505 This means that the relationship between Internet procrastination and flow on the Internet is weaker when PIU symptoms are absent. If a person is in a state of flow while avoiding 506 work-related tasks this is more likely to result in PIU. By implication, if a person is in a 507 state of flow for work-related tasks, the relationship with procrastination would not exist 508 and a person would be less likely to experience PIU. However, this is not to say (theoret-509 510 ically) that a person cannot be in a state of flow for work-related tasks and still experience 511 PIU. After all, if a person spends long hours performing work-related tasks online while neglecting family and other social commitments it is likely that PIU symptoms would 512 emerge over time. 513

These results provide a small degree of support for Belanger and Van Slyke (2002) con-514 tention that a small amount of play is important for preventing boredom (and possibly 515 even encouraging creativity). The results also support the findings from Woods and Grif-516 fiths (in press) who found that some (procrastination) activities are perceived positively as 517 a temporary escape from other life experiences. The results would suggest that too much 518 play or procrastination would result in PIU, since the relationship between flow and play-519 fulness is strongly implied in the definition of flow (Csikszentmihalyi, 1997). The results of 520 521 the partial correlation between Internet procrastination and flow on the Internet would suggest that a small proportion of flow and procrastination is manageable by the individ-522 ual without resulting in well-being detriments and the appearance of PIU symptoms would 523 be dependent on the degree of control that a person has over the amount of time spent 524 engaged in non-work tasks. 525

526 4.2. Regression analysis

The regression analyses provide some support for the correlation and partial correla-527 tion results. Internet procrastination explains the greatest proportion of the variance 528 (45%) of any of the variables entered into the model. This result is entirely in line with 529 the results of the correlations as supported by Davis et al. (2002) and LaRose et al. 530 (2003). The variable explaining the next greatest amount of the variance (5%) entered into 531 the model was the amount of time spent online in a session. This finding is consistent with 532 533 LaRose et al. (2003), Lin and Tsai (2002), Morahan-Martin and Schumacher (2000), and Young (1996) but contradicts Widyanto and McMurran (2004) who found that time spent 534 535 online was unrelated to PIU. It must be noted that Widyanto and McMurran (2004) sample was quite small in relation to the other studies. Flow on the Internet explained a fur-536 ther 4% of the variance in PIU and the number of days that a person spends online per 537 week a further 1%. Given the intercorrelations between Internet procrastination and flow 538 on the Internet, it is likely that much of the variance explained is shared. Likewise, there is 539 540 likely to be some shared variance in the two variables related to time spent online (i.e.

average time spent online per session and the number of days per week spent online). What
 is revealing from the regression model is that procrastination and the amount of time spent
 online are better predictors of PIU than flow experiences.

544 *4.3.* The impact of different online activities

The results give strong support for the contention that the online activities that are most 545 interactive (i.e. online games, online chat, online telephony, and blogging) are the best pre-546 dictors of PIU, flow on the Internet, and Internet procrastination. The results are similar 547 to the majority of previous studies (Chen et al., 1999; Leung, 2004; Li & Chung, 2006; 548 Meerkerk et al., 2006; Ng & Wiemer-Hastings, 2005; Whang et al., 2003). These online 549 activities each have a number of qualities in common including high degrees of social inter-550 action, perceived control, and they are intrinsically motivating, and absorbing. Interactive 551 activities provide opportunities for skill development and therefore also have a high degree 552 of perceived control (Griffiths, 2003b). These qualities are similar to those found in flow 553 554 experiences hence providing theoretical support for the positive relationship between PIU and flow on the Internet. This provides support for Woods et al. (2004) contention 555 that the structural and situational elements of different tasks provided a better explanation 556 of excessive behaviour than trait-based explanations. 557

Since the study was conducted in South Africa, the question arises about the generalis-558 ability of these results to more industrialised countries. The information technology indus-559 try in South Africa is similar in nature to information technology sectors in many countries. 560 In this industry, access to the Internet is ubiquitous, convenient, and fast. People working in 561 this industry in South Africa are technologically-literate and evidence of their access and 562 use of the Internet reported in this study parallels (or in most cases is higher than) that 563 564 reported in studies from Canada (Davis et al., 2002), Taiwan (Lin & Tsai, 2002), The Netherlands (Meerkerk et al., 2006), the United Kingdom (Niemz et al., 2005; Widyanto & 565 McMurran, 2004), and the United States (Morahan-Martin & Schumacher, 2000; Morgan 566 & Cotton, 2003). However, to assess the true generalisability of these results further studies 567 in different countries and different work sectors would be necessary. 568

569 4.4. Limitations of the study

Since it was not possible to determine the country of origin of respondents based on the 570 questionnaire, it is difficult to determine the generalisability of the results. It is assumed 571 that the vast majority of respondents were resident in South Africa given the nature of 572 the data collection point (a South African information technology website), but it is also 573 possible that this assumption is false. The use of Davis et al. (2002) distraction scale as a 574 measure for Internet procrastination has already been acknowledged as potentially prob-575 lematic given Davis et al. (2002) conceptualisation of PIU. Other studies should use a 576 more independent measure of procrastination, although this would involve some modifi-577 578 cation of items since no other scale of Internet procrastination exists. One obvious limitation is the use of an Internet-based sample as this may lead to sampling bias. Since the 579 sample was obtained from an Internet-based information technology magazine, this means 580 that respondents would be biased towards the Internet as a useful technology and may 581 even have been procrastinating by filling out the online survey instead of completing other 582 work-related tasks. 583

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Webster et al. (1993) assessment of flow is incomplete measuring only four of the eight 584 components of flow (i.e. control, concentration and focus, and autotelic components). 585 This is problematic since it does not allow for a full examination of theoretical and prac-586 tical overlaps in all the aspects of flow with PIU and procrastination. Furthermore, Hoff-587 man and Novak (1996) have argued that the state of flow is dynamic. The measure of flow 588 used in this study assessed flow over an indeterminate period and a range of different tasks. 589 It is possible that flow is experienced for some online tasks some of the time, but not for 590 other tasks. The measure of flow used in this study was not sensitive enough to assess these 591 dynamic changes in flow states for different activities. Similar criticisms might be levelled 592 593 at the sensitivity of the procrastination measure to different Internet activities. Eleven different Internet activities were examined in this study. This list of activities is by no means 594 exhaustive (online sexual addiction and online gambling are not included, for example) or 595 highly nuanced in terms of specific tasks (ordering a product online or reading a news 596 report on a sporting event, for example). Future research might consider widening the list 597 598 of Internet activities or limiting the area of investigation to a particular activity in order to 599 adopt a more nuanced approach to different online tasks. Similarly, the measures of PIU, flow and procrastination should focus specifically on identifying Internet activities that are 600 work- and non-work-related. 601

602 *4.5. Conclusions and directions for future research*

The results from this study have demonstrated that there is empirical evidence for an 603 overlap between the concepts of flow on the Internet, Internet procrastination, and 604 PIU. Following Belanger and Van Slyke (2002) the results of this study suggest that stud-605 ies need to investigate the respective proportions of time spent socialising, playing, pro-606 607 crastinating, and working online. This would mean, following the situational approach to procrastination (Blunt and Pychyl, 2005), that one would need to investigate the qual-608 ities of the different online work tasks that lead to boredom or excessive challenge and 609 whether play or procrastination might contribute to improved work performance. The dif-610 ficulty with this line of inquiry from a flow perspective is that the Internet is a "multi-activ-611 ity medium" (Chen et al., 1999, p. 590). This means that people might have matched skills 612 and demands for some activities (e.g. online gambling or online gaming) but not for others 613 614 (e.g. online literature searching). This implies that studies should probably operationalise flow, skill and challenge in terms of specific online activities. Mahatanankoon et al. (2004) 615 Personal Web Usage (PWU) scale might prove useful in this regard. Additionally, there 616 are many different applications of the Internet (Chen et al., 1999), some of which (down-617 loading software or playing online games) may be potentially more conducive to PIU than 618 other activities (Lin & Tsai, 2002; Morahan-Martin & Schumacher, 2000; Yang & Tung, 619 2007). Several studies (including this study) have suggested that PIU is more likely to 620 621 occur when the Internet is used for social functions rather than work-related (instrumental) functions (Li & Chung, 2006; Yang & Tung, 2007). Further research is required to 622 623 unravel this issue to determine whether PIU is a separate concept from excessive work within the information technology worker population. Lin and Tsai (2002) also found that 624 Internet dependents were more likely to be sensation seekers which, in turn, might mean 625 that they become bored with routine work more easily and hence use the Internet to pro-626 crastinate (leading to PIU). The relationship between sensation seeking, PIU and procras-627 628 tination requires further investigation. Finally, there is some evidence to suggest that

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- people should be made aware of the dangers of online activities that have high flowquotients.

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