Gaming Motivation and Problematic Video Gaming:

The Role of Needs Frustration

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Mills, D. J., Milyavskaya, M., Derevensky, J., & Heath, N. L. (in press). Gaming motivation and problematic video gaming: The role of needs frustration. *European Journal of Social Psychology*. Advance online publication. Motivation is often used as a predictor of a problematic style of video game engagement, implying that individuals' gaming undermines optimal functioning. Drawing from recent advances in Self-Determination Theory (SDT), the present study explores the links between gaming motivations, the daily frustration of basic psychological needs, and reports of problematic video gaming (PVG). A sample of 1,029 participants (72.8% male; M = 22.96 years; SD = 4.13 years) completed items regarding their gaming engagement and gaming motivation as well as their experience of needs frustration and PVG symptoms. Results revealed positive associations between gaming motivations and PVG, and between daily needs frustration and PVG. Finally, after comparing several competing models, a mediational model whereby needs frustration explained the association between individuals' gaming motivation and PVG emerged as best fitting the data. The discussion addresses the theoretical and practical implications of these findings in the context of recent research.

Abstract

Keywords: Gaming Disorder; Gaming; Motivation; Basic Needs; Self-Determination Theory

The Role of Needs Frustration

Self-Determination Theory (SDT; Ryan & Deci, 2000, 2017) is a widely used theory of human motivation that emphasizes the quality versus the quantity of motivation in explaining consequences from activity engagement. SDT proposes more adaptive outcomes will occur when activity engagement is freely chosen and based solely on the pleasure and enjoyment of the activity itself (i.e., intrinsic motivation). Studies have shown less intrinsic motivation toward gamblingor exercise-related activities is associated with reports of greater gambling disorder and exercise dependence, respectively (Clarke, 2004; González-Cutre & Sicilia, 2012). Intrinsic motivation is assumed to stem from greater satisfaction of three basic psychological needs (competence, autonomy, and relatedness) during engagement in the activity, which implies that gambling disorder and exercise dependence are associated with lower needs satisfaction during engagement in related activities. However, active impediments to needs satisfaction, or needs frustration, within domains unrelated to the activity may ultimately cultivate a problematic pattern of activity engagement, as individuals become more reliant upon the activity to satisfy these needs (Vansteenkiste & Ryan, 2013). To date, no study has explored the contributions of both activity motivation and the experience of needs frustration in predicting problematic activity engagement. Therefore, within the quickly growing area of video games, the present study explores the unique roles gaming motivation and experiences of needs frustration in explaining problematic video gaming.

Self-Determination Theory (SDT; Ryan & Deci, 2000, 2017) proposes three interrelated types of motivation underlying activity engagement that are easily applied to video games. An *intrinsic motivation* is guided by the pleasure and enjoyment from video games. An *amotivation*

indicates that video games are perceived to add little value to life resulting in a lack of personal intention to engage in them. Existing between an intrinsic motivation and an amotivation lies four subtypes of *extrinsic motivation*. The first, integrated regulation, is guided by the personal expression of self through video game engagement (Ryan & Deci, 2017). The second, identified regulation, is characterized by the alignment of video game engagement with one's values and goals (Ryan, 1995). The third, introjected regulation, indicates experiencing strong internal pressures to engage in video games that are beyond the control of the individual (Ryan & Deci, 2000). Finally, the fourth, external regulation, is driven by the desire to earn rewards through gaming (Ryan, 1995). In line with assumptions of SDT, the abovementioned motivations toward gaming have been found to be positively associated with time spent gaming except for an amotivation toward gaming (Lafrenière, Verner-Filion, & Vallerand, 2012).

Interwoven within SDT's perspective on motivation is the role of three basic psychological needs. The three needs include competence (i.e., perceiving one's abilities are well-matched with the activity), autonomy (i.e., perceiving actions while engaging in the activity are under one's own volition), and relatedness (i.e., perceiving a sense of connectedness with others through engagement in the activity) (Ryan & Deci, 2000, 2017). Reporting greater satisfaction of these needs during activity engagement is expected to concomitantly occur with a stronger intrinsic motivation (Ryan, 1995). Consistent with this assumption, Lafrenière, Verner-Filion, and Vallerand (2012) observed that a stronger intrinsic motivation toward gaming was associated with perceiving higher needs satisfaction during gaming. Other research has demonstrated that higher game enjoyment, stronger intentions to play, and greater time spent gaming are further predicted by greater needs satisfaction during video game engagement (Johnson, Gardner, & Sweetser, 2016; Ryan, Rigby, & Przybylski, 2006; Tamborini, Bowman, Eden, Grizzard, & Organ, 2010).

Therefore, needs satisfaction while gaming appears to explain the appeal of video games (see review by Przybylski, Rigby, & Ryan, 2010), but, as shown within a recent study, does not meaningfully explain problematic video gaming (PVG; Mills, Mettler, & Heath, 2017).

PVG is a pattern of video game engagement that contributes to maladaptive functioning in daily life (King, Haagsma, Delfabbro, Gradisar, & Griffiths, 2013; Petry et al., 2014). PVG is conceptually similar to other behavioral addictions such as gambling disorder and exercise dependence (Hausenblas & Symons Downs, 2002), which have been found to be most strongly associated with an introjected regulation and amotivation. Demographically, PVG is more commonly reported by males than females, and by those spending large amounts of time playing video games (Jeromin, Rief, & Barke, 2016; Lemmens, Valkenburg, & Gentile, 2015). Research has demonstrated that various measures of PVG are associated maladaptive outcomes including depression, impulsivity, conduct disorder, anxiety, and other psychological disorders (Bargeron & Hormes, 2017; Strittmatter et al., 2015; Vadlin, Åslund, Hellström, & Nilsson, 2016). Although beyond the scope of this paper to discuss in detail, it is important to note that a consensus has not been reached regarding the criteria of PVG or the weight that should be given to PVG as a potential disorder (Aarseth et al., 2016; Griffiths et al., 2016; Przybylski, Weinstein, & Murayama, 2017). Nonetheless, PVG represents an area in which to build upon previous applications of SDT to the study of gaming.

Recent developments in SDT suggest *needs frustration*, or the extent to which individuals feel obstructed in their pursuit of satisfying their needs in daily life (Vansteenkiste & Ryan, 2013), may be a more robust predictor of behavioral addictions such as PVG. Studies show the active obstruction component that defines needs frustration is essential to predicting maladaptive outcomes (e.g., depression, interpersonal sensitivity), whereas low needs satisfaction that is void

of any obstruction will weakly predict adaptive outcomes (e.g., life satisfaction, vitality) (Chen et al., 2015; Costa, Ntoumanis, & Bartholomew, 2015; Gunnell, Crocker, Wilson, Mack, & Zumbo, 2013). Previous research has also shown that needs frustration during participation in a mandatory (e.g., school) or loved activity (e.g., sport) is associated with less intrinsic motivation toward the activity as well as maladaptive outcomes (Costa, Coppolino, & Oliva, 2016; Haerens, Aelterman, Vansteenkiste, Soenens, & Van Petegem, 2015). However, in line with Vansteenkiste and Ryan (2013), needs frustration across life domains may cultivate a dependence on one activity for needs satisfaction, alluding to a problematic style of engagement.

The present study has two objectives. The first objective sought to assess the associations among gaming motivations, daily needs frustration, and PVG. It was hypothesized that introjected regulation, external regulation, and amotivation would be positively correlated with competence, autonomy, and relatedness frustration. PVG was hypothesized to be positively associated with all six gaming motivations as well as reports of competence, autonomy, and relatedness frustration.

The second objective sought to assess *how* gaming motivations, daily needs frustration, and PVG were related. At present, theory and previous research offer at least three potential models explaining how these constructs are related. The first model (Model A in Figure 1) suggests gaming motivations and daily needs frustration are best viewed as separate predictors of PVG. This model is supported by the theoretical and empirical evidence reviewed above. Alternatively, gaming motivations may explain the link between needs frustration and PVG. The second model (Model B in Figure 1) draws upon recent results from Lalande and colleagues (2017) who provided evidence that deficits in needs satisfaction are associated with lower life satisfaction through a stronger extrinsic motivation toward a loved activity. The deficits in needs satisfaction, per Lalande and colleagues (2017), bring about an overreliance toward one activity to satisfy basic

needs, which over time undermines the potential of an intrinsic motivation toward an activity. However, it is possible that experiences of needs frustration explain the link between gaming motivations and PVG. This final model (Model C in Figure 1) suggests gaming motivations contribute to reports of PVG through increased experiences of needs frustration. Although this model does not have as much empirical support as Model A or Model B, recent research in a tangentially related area of study demonstrated that perfectionism contributes to the presence of eating disorders through an increase in daily needs frustration (Boone, Vansteenkiste, Soenens, Van der Kaap-Deeder, & Verstuyf, 2014). Conceptualizations of perfectionism do parallel some of the broad qualities of introjected regulation including an inability to control internal pressures. As such, perfectionist individuals demonstrate a unique interaction with their environment that may cultivate a greater susceptibility for experiencing needs frustration. Therefore, with regard to gaming motivation, Model C suggests that the internalization of a strong extrinsic motivation may also contribute to the experience of daily needs frustration because it represents how individuals might interact with their environment similar to reports of perfectionism.

Other models beyond the three outlined above were considered, but were not found to have justification in existing theory or available evidence resulting in their exclusion. Given the previously mentioned links to PVG, gender and time spent gaming were included as covariates.

Methods

Participants

Ethical approval from McGill University was given prior to recruitment, and all participants provided their informed consent before to beginning the online questionnaire. In total, 1,802 participants were recruited through online social networks (e.g., *Facebook, Reddit*) and research forums (e.g., *Psychological Research on the Net*), email invitations, and flyers. The

dataset was reviewed for duplicate IP addresses in which the earliest response was retained with the latter responses excluded (n = 59). Further, participants less than 18 years (n = 43) or over 35 years (n = 61) were excluded, as this range corresponded with significant portion of video game users (Entertainment Software Association, 2016). Additionally, 251 participants did not complete the online survey and were thus excluded. Finally, 359 participants were excluded for responding to an attention item incorrectly, or responding "No" to the question, "Do you play video games most days of the week", suggesting they are not frequent video game users. Comparisons of gender, age, and time spent gaming were conducted between those included and those excluded, which did not reveal any significant differences (p > .05).

The final sample included 1,029 participants (72.8% male; M = 22.96 years; SD = 4.13 years). On average, participants spend 19.70 hours (SD = 15.34 hours) per week gaming. Participants were from the United States (46.4%), Canada (33.1%), as well as various European (12.2%) and Asian (3.4%) countries. Most participants (64%) reported they were presently enrolled at a post-secondary institution.

Measures

Problematic Video Gaming. The 9-item Internet Gaming Disorder Scale (IDGS; Lemmens et al., 2015) was used to assess PVG. Participants rated the frequency they experienced each item over the last year using an altered 6-point scale ranging from *almost never* (1) to *almost always* (6). Internal consistency was acceptable ($\alpha = .85$).

Needs Frustration. Bartholomew, Ntoumanis, Ryan, and Thøgersen-Ntoumani's (2011) Psychological Need Thwarting Scale (PNTS) was adapted to assess needs frustration in daily life. The original scale consists of three 4-item subscales assessing the frustration of each psychological need. An item from the relatedness subscale (REL2) was not included in the questionnaire due to a previous study finding it was confusing and ultimately did not load highly onto its respective factor (Mills, Milyavskaya, Mettler, & Heath, 2017). Participants used the same 6-point scale as the IDGS to rate how frequently they experience each item in their life. Internal consistency was acceptable for the competence ($\alpha = .82$), autonomy ($\alpha = .88$), and relatedness ($\alpha = .89$) frustration subscales.

Gaming Motivation Scale. Lafrenière and colleagues (2012) developed the 18-item Gaming Motivation Scale (GAMS) to assess users' motivation toward gaming. The scale includes 6 three-item subscales that assess intrinsic motivation, amotivation, and each of the four subtypes of extrinsic motivation toward gaming (i.e., integrated regulation, identified regulation, introjected regulation, external regulation). Each item was rated on a 7-point scale ranging from *not agree at all* (1) to *very strongly agree* (7). Internal consistencies were all above .70 for each subscale, except for the intrinsic motivation subscale ($\alpha = .60$).

Analytical Strategy

Statistical analyses were conducted using MPlus version 7.4 with missing values estimated by full information maximum likelihood (Muthén & Muthén, 2015). Bivariate correlations with a Bonferroni modified alpha (p < .001) were conducted to assess the associations among gaming motivations, daily needs frustration, and PVG (Objective 1). Fit indices were calculated to assess the fit of the present data to each of the proposed models within Figure 1 (Objective 2). Goodnessof-fit was determined by values of .08 or lower for root mean square error of approximation (RMSEA), near or above .95 for both comparative fit index (CFI) and Tucker-Lewis Index (TLI), and less than .06 for standardized root mean squared residual (SRMR) (Hu & Bentler, 1999; Kline, 2016). Additionally, Akaike's information criterion (AIC) and Bayesian information criterion (BIC) were used to compare models with lower values suggesting greater parsimony (Kline, 2016; Wagenmakers, 2007). Prior to conducting these analyses, the underlying factor structure of each measure was confirmed with the present data. The input and output of the preliminary and primary analyses as well as the individual items for each of the measures are provided as online supplemental material.

Results

Preliminary Analyses

A complete description of the steps taken to confirm each of the measures is provided in the online supplemental material. Although only minor modifications were made to IGDS to confirm a single factor solution, both the PNTS and the GAMS required further steps. First, an item from the competence subscale of the PNTS was excluded due to loading poorly onto its respective factor. Following its removal, the present data were found to fit the expected 3-factor solution. The revised 3-item competence subscale was found to have adequate internal consistency $(\alpha = .89)$. Second, following several steps that resulted in not confirming the 6-factor solution for the GAMS, the entire intrinsic motivation subscale was excluded. Several steps were taken including an exploratory factor analysis with Promax rotation as well as confirmatory factor analysis with the remaining 15 items. In the end, four unique factors were identified with adequate internal consistency: Integrated-Identified Regulation ($\alpha = .88$), Introjected Regulation ($\alpha = .74$), External Regulation ($\alpha = .70$), and Amotivation ($\alpha = .87$). The interested reader is encouraged to review the online supplemental material for further explanation of these steps. Not surprisingly, participants recruited from a gaming-specific Reddit streams reported greater time spent gaming, stronger motivations toward gaming, and higher PVG than participants recruited using other methods (p < .05). Table 1 presents the means and standard deviations of the included instruments.

Primary Analyses

The first objective was to assess the associations among gaming motivations, daily needs frustration, and PVG. Bivariate correlations (see Table 1) largely supported expected associations. Specifically, PVG was positively associated with each of the four gaming motivations as well as competence, autonomy, and relatedness frustration. Furthermore, competence, autonomy, and relatedness frustration with introjected regulation and amotivation, however, external regulation was not associated with needs frustration.

The second objective assessed how gaming motivations, daily needs frustration, and PVG were related. Average scores for the four gaming motivations, daily frustration of each basic need, and PVG were included as observed variables within the planned path analyses. Time gaming and gender were included as covariates, however several paths were excluded given the non-significant associations observed in Table 1. First, the pathways from time gaming to competence, autonomy, and relatedness frustration were excluded from the proposed models. Second, the pathways from gender to PVG, introjected regulation, external regulation, and amotivation were excluded from the proposed models.

An acceptable fit was found for each model (see Table 2), however, Model C was found to have the lowest AIC, which suggested it is the preferred model (Kline, 2016). Moreover, using an equation the provided by Wagenmakers (2007), the differences in BIC revealed "strong" evidence that Model C is a more parsimonious model compared to Model A and Model B. In fact, according to Wagenmakers' equation, Model C is 30.4 times more likely than Model A and 28.2 times more likely than Model B. As shown in Figure 2, the results of this model show all four gaming motivations are directly associated with PVG. In addition to these direct effects, positive indirect effects for introjected regulation and amotivation were found through autonomy frustration

(Introjected Regulation: $\beta = .03$, p = .01, 95CI[.01, .05]; Amotivation: $\beta = .02$, p = .01, 95CI[.01, .04]). All other effects are included in the online supplemental material.

Discussion

The present study investigated the roles of both gaming motivations and daily needs frustration in explaining PVG. As expected, gaming motivations and daily needs frustration are positively associated with PVG. These associations are consistent with work in other areas of study (Boone et al., 2014; Clarke, 2004; Costa et al., 2016; González-Cutre & Sicilia, 2012; Symons Downs, Savage, & DiNallo, 2013). However, in building upon previous research, the present findings are the first to show that the experience of daily autonomy frustration partially mediates the association between both introjected regulation of and amotivation toward gaming and PVG. The discussion focuses on these two specific motivations to help explain PVG.

Introjected regulation implies that the drive to engage in video games is facilitated by uncontrollable internal pressures, which resemble withdrawal-like symptoms such as anxiety or irritability when unable to play (Ryan, 1995). Per SDT, these pressures stem from a contingent self-worth that is strongly attached to the activity indicating engagement in the activity is an attempt to prove oneself. A recent study demonstrated that although each of the four types of extrinsic motivation as well as amotivation were positively correlated with a strong attachment of self-esteem to video game engagement, introjected regulation of gaming engagement correlated highly (r > .70) with the pursuit of individual validation through gaming (Beard & Wickham, 2016). As such, introjected regulated video game users are compelled to play video games problematically due to the role gaming has on their perception of self, implying a larger issue surrounding problematic gaming cognitions (see work by King & Delfabbro, 2014, 2016).

On the other hand, amotivation represents a perception that gaming does not provide value to one's life culminating in a lack of personal intention in gaming engagement (Ryan, 1995). Although further research is needed, helplessness within the present context likely refers to an awareness of the negative consequences of gaming engagement, but an inability to stop or manage gaming engagement. As such, the link between amotivation toward gaming and PVG parallels the link between feeling "addicted" to gaming and PVG (Männikkö, Billieux, Nordström, Koivisto, & Kääriäinen, 2017; Rehbein, Kliem, Baier, Mößle, & Petry, 2015). Further, the strong correlation between introjected regulation and amotivation suggests the helplessness in effectively managing gaming engagement may stem from strong internal pressures to engage in gaming, which are coming into conflict with a heightened awareness that gaming no longer adds value to other life domains and may even be undermining quality of life (Ryan & Deci, 2017).

Importantly, both introjected regulation and amotivation predicted increased frustration of all three basic needs; however, current theory does not explicitly address *how* these motivations might facilitate greater needs frustration. Only two studies have shown a mediational role of needs frustration in explaining the presence of a problematic behavior, and both used reports of perfectionism as a predictor of needs frustration within a sport-specific context (Costa et al., 2016) or in daily life (Boone et al., 2014). Results demonstrated that perfectionism contributed to a problematic behavior (e.g., binge eating, exercise dependence) through an increase in needs frustration. Although the strong uncontrolled internal pressures to engage in video games parallels some of broad elements of perfectionism, the overlap with amotivation is less clear. It is possible that perfectionistic individuals feel "addicted" to behaving a specific way and helpless to change their behavior. This overlap between introjected regulation and amotivation with perfectionism does not explain *how* needs frustration is facilitated.

One possible way in which introjected regulation and amotivation facilitate greater daily needs frustration is through a strong disengagement in environments unrelated to gaming resulting in an active avoidance of such environments. PVG is strongly related to the perception that the online gaming community is the only place in which one is valued and appreciated (Liu & Peng. 2009). Indeed, these perceptions speak to a key component embedded within the introjected regulation of gaming engagement as well as the control gaming has over amotivated individuals. Strong beliefs that one is only valued within a gaming community will result in disengagement as well as greater aggression in both gaming as well as non-gaming environments. This increased aggression may bring about an exclusionary response from non-gaming environments. Future research is needed to assess for a recursive association, whereby perceptions of being valued only online stem from experiences of needs frustration. An important implication of Liu and Peng's (2009) finding is that enthusiastic, but otherwise healthy users likely feel valued and appreciated in other areas of their life beyond gaming (e.g., work, school, friendships) limiting the control gaming has on their life. As shown in another study using adolescents, it is likely this stems from a lack of daily needs frustration experienced within a key domain (Yu, Li, & Zhang, 2015).

An interesting finding within the present study was that only autonomy frustration mediated the links introjected regulation and amotivation toward gaming and PVG. Autonomy frustration is critical in facilitating the development of a contingent or fragile self-esteem (Deci & Ryan, 1995). Caution should be taken in interpreting the present results as suggesting that competence and relatedness frustration do not contribute to PVG. Rather, competence and relatedness frustration were found to contribute to PVG through a decline in subjective well-being (Mills et al., 2017). As such, these findings highlight the unique role motivations toward gaming

may play undermining daily autonomy satisfaction, which contributes to a strong attachment of self-esteem onto gaming engagement resulting in greater PVG.

Finally, it should be noted that both integrated-identified regulation and external regulation predicted PVG, but were not associated with needs frustration. This suggests that enthusiastic, but otherwise healthy video game users who perceive gaming (1) provides personal expression (i.e., integrated regulation) and (2) is in alignment with values and goals (i.e., identified regulation) are likely to report some symptoms of PVG (Charlton & Danforth, 2007), which is similar to the conceptualization of a harmonious passion (Vallerand, 2010). On the other hand, external regulation, which relates to being motivated to collect various rewards (e.g., levelling up, in-game awards), is not surprisingly related to PVG, as these elements will often relate to directly to a surface-level appeal of gaming. However, these characteristics by themselves will likely not facilitate a sustained PVG.

Several limitations should be noted. First, the use of self-report data assumes participants correctly interpreted each item and responded truthfully; however, careful steps were taken in reviewing the measures as well as participants' responses to each item to ensure accuracy and validity. Second, because the present data did not confirm the original six-factor structure of the GAMS, the present findings are based upon a slightly incomplete picture of gaming motivation from the perspective of SDT. Future research is needed to revalidate the scale and its ability to adequately measure all six gaming motivations. Finally, due to the crossectional design, the present study is not able to infer causality. However, the data-driven approach taken is a strength of the present study as it compared a set of theoretically-derived models, which ultimately revealed strong evidence of the likely relation among gaming motivations, needs frustration, and PVG.

Notwithstanding these limitations, the present study offers original contributions to both the SDT and PVG literature. Specifically, introjected regulation and amotivation play a role in explaining PVG through in part their effect on the experience of needs frustration. Although speculative, it is possible that clinicians might see changes in individuals' gaming patterns by directly addressing their motivations toward gaming through motivational interviewing (Miller & Rollnick, 2013). Further, a recent study has shown a small but significant reduction of the association between needs frustration and ill-being for those reporting higher dispositional mindfulness (Schultz, Ryan, Niemiec, Legate, & Williams, 2015), a quality of consciousness that may be increased through targeted training during interventions (Brown & Ryan, 2003; Carmody & Baer, 2008). As these are speculative suggestions, research is needed to provide evidence of their merits within a clinical study. Furthermore, research should explore whether the link between needs frustration and PVG depends on the type of video games one primarily plays.

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1 Candar (Malac - 1: Famalac - 2)	4	•	,	-	,		-	`		
1. UCHUCI (IVIAICS -1 , I'CIIIAICS -2)	•									
2. Hours Gaming	26***	·								
3. Problematic Video Gaming	04	.24***	ı							
4. Integrated-Identified	10***	$.36^{***}$.38***	ı						
5. Introjected	02	.24***	.65***	.47**	ı					
6. External	07	$.19^{***}$.36***	.38	.42**	ı				
7. Amotivation	03	.12***	.48***	.04***	.50***	.13	·			
8. Competence Frustration	.20***	.03	.36***	.12***	.35***	60.	.29***	ı		
9. Autonomy Frustration	.15***	.04	.37***	.11**	.33***	.10	.31***	.73***	·	
10. Relatedness Frustration	.20***	.01	.28***	.08	.28***	.03	.27***	.76***	.67***	ı
Mean	I	19.71	2.19	3.15	1.93	3.52	1.94	2.71	2.63	2.48
Standard Deviation	·	15.33	0.84	1.44	1.14	1.51	1.33	1.28	1.62	1.32
*** $p < 001$ (Bonferroni adinsted alpha)										

 Table 1.

 Correlation Matrix Presenting the Associations Among Gender, Hours Spent Gaming, Problematic Video Gaming, Gaming

p < .001 (Bonterroni adjusted alpha)

Fit Statistics Motivations	for the Pi and the Fi	ropose. rustrat	d Models ion of Ba	: Presen	ted in Figu ds in Expla	Fit Statistics for the Proposed Models Presented in Figure 1 Explaining the Roles of Gaming Motivations and the Frustration of Basic Needs in Explaining Problematic Gaming	ıg the Roles oj 1atic Gaming	f Gaming
	χ^{2}	df	CFI	TLI	χ^2 df CFI TLI SRMR	RMSEA [90CI]	AIC	BIC
Model A	42.51	11	42.51 11 .99 .96	96.	.04	.05 [.04, .07]	32569.42	32835.98
Model B	40.13	11	66.	96.	.04	.05 [.03, .07]	32567.79	32834.35
Model $C^{*\wedge \#}$ 20.72 11 1.00 .99	20.72	11	1.00	66.	.01	.03 [.01, .05]	32547.07	32813.63
* Indicates the lowest AIC of all the models	lowest AIC	of all th	ie models					

Ц • Ľ. . . 1 Ď Jola A Mai ĥ 41+ 4 . Table 2. 4:0 Fit Stat

 $^{\scriptscriptstyle \wedge}$ Indicates the lowest BIC of all the models

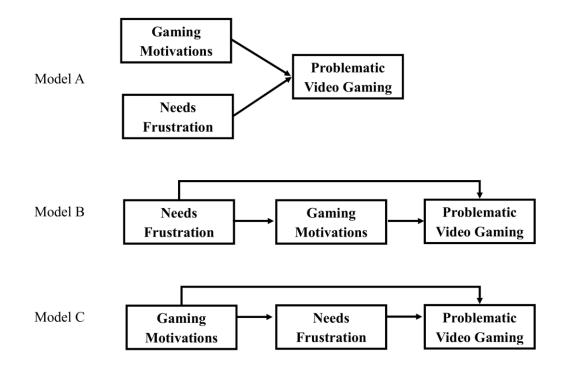


Figure 1. Proposed models depicting the associations among gaming motivations, needs frustration, and problematic gaming.

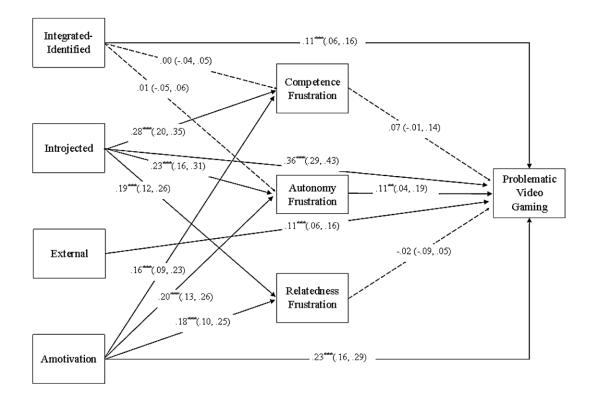


Figure 2. Path estimates for Model C. Correlations among the four gaming motivations as well as the frustration of each basic psychological need were not included for simplicity. Dash lines represent nonsignificant estimates. Significant estimates are indicated by a solid line and an (*). Confidence intervals (95%) are provided in parentheses.

* *p* < .05; ** *p* < .01; *** *p* < .001

Supplemental Material for the Article Titled:

Gaming motivation and problematic gaming: The role of needs frustration

Online Publication Only

Notes: Below are codes for each latent variable. The code followed by a number indicates individual items, which are presented on the following page.

Needs Frustration Coding:

COMP = Competence Frustration Subscale

AUT = Autonomy Frustration Subscale

REL = Relatedness Frustration Subscale

Gaming Motivation Coding:

INTG_ID = Integrated & Identified Regulation Subscales

INTRO = Introjected Regulation Subscale

EXT = External Regulation Subscale

AM = Amotivation Subscale

Problematic Gaming Coding:

PVGU = Problematic Gaming

Items from the measures that were included in the present study

- AUT1 I feel forced to follow decisions made for me
- AUT2 I feel pushed to behave in certain ways
- AUT3 I feel prevented from making choices
- AUT4 I feel under pressure to follow another's plan for me
- COMP1 There are situations where I am made to feel inadequate
- COMP2 There are times when I am told things that make me feel incompetent
- COMP3 Situations occur in which I am made to feel incapable
- *COMP4 I feel inadequate because I am not given opportunities to fulfill my potential
- REL1 I feel others can be dismissive of me
- **REL2 I feel that other people are envious when I achieve success
- REL3 I feel other people dislike me
- REL4 I feel I am rejected by those around me
- PVGU1 During the last year, how often have you been feeling miserable when you were unable to play a game?
- PVGU2 During the last year, how often have there been periods when all you could think of was the moment that you could play a game?
- PVGU3 During the last year, how often have you felt unsatisfied because you wanted to play more?
- PVGU4 During the last year, how often have you lost interest in hobbies or other activities because gaming is all you wanted to do?
- PVGU5 During the last year, how often have you experienced serious conflicts with family, friends or partner because of gaming?
- PVGU6 During the last year, how often have you hidden the time you spend on games from others?
- PVGU7 During the last year, how often have you had arguments with others about the consequences of your gaming behavior?
- PVGU8 During the last year, how often were you unable to reduce your time playing games, after others had repeatedly told you to play less?
- PVGU9 During the last year, how often have you played games so that you would not have to think about annoying things?
- *IM1 Because it is stimulating to play.

- *IM2 For the pleasure of trying/experiencing new game options (e.g., classes, characters, teams, races, equipment)
- *IM3 For the feeling of efficacy I experience when I play
- INTG1 Because it is an extension of Me
- INTG2 Because it is an integral part of my life
- INTG3 Because it is aligned with my personal values
- ID1 Because it is a good way to develop important aspects of myself
- ID2 Because it is a good way to develop social and intellectual abilities that are useful to me
- ID3 Because it has personal significance to me
- INTR1 Because I feel that I must play regularly
- INTR2 Because I must play to feel good about myself
- INTR3 Because otherwise I would feel bad about myself

To acquire powerful and rare items (e.g., armors, weapons) and virtual currency

EXT1 (e.g., gold pieces, gems) or to unlock hidden/restricted elements of the game (e.g., new characters, equipment, maps)

- EXT2 For the prestige of being a good player
- EXT3 To gain in-game awards and trophies or character/avatar's levels and experiences points
- AM1 It is not clear anymore; I sometimes ask myself if it is good for me
- AM2 I used to have good reasons, but now I am asking myself if I should continue
- AM3 Honestly, I don't know; I have the impression that I'm wasting my time

* Included in the study, but excluded following a confirmation of the factor structure ** An original item in the PNTS that was not included in present study.

Initial model fit of Internet Gaming Disorder Scale:

 $\chi^{2}(82) = 417.91, p = <.001$ RMSEA = .12, 90CI [.11, .13] CFI = .84 TLI = .78 SRMR = .06

Internet Gaming Disorder Scale: Confirmatory Factor Analysis Following Modifications

Notes: The inclusion of the highlighted (in yellow) correlations was suggested based on the provided modification indices. Each of these items hint at the affect their gaming engagement has had upon their relationships, specifically confrontations (e.g., arguments friends, family, or partners; being told to cut back in gaming). The input and output of the final model is provided below.

MODEL INPUT:

Analysis:

ESTIMATOR = MLR

Model:

PVGU BY PVGU6 PVGU2 PVGU3 PVGU4 PVGU5 PVGU1 PVGU7 PVGU8 PVGU9;PVGU8WITH PVGU5;PVGU8WITH PVGU7;PVGU7WITH PVGU5;

MODEL FIT INFORMATION: FOLLOWING APPLICATION OF MODIFICATIONS

Chi-Square Test of Model Fit

Value	113.741*
Degrees of Freedom	24
P-Value	0.0000
Scaling Correction Fac	tor 1.2887
for MLR	

* The chi-square value for MLM, MLMV, MLR, ULSMV, WLSM and WLSMV cannot be used for chi-square difference testing in the regular way. MLM, MLR and WLSM chi-square difference testing is described on the Mplus website. MLMV, WLSMV, and ULSMV difference testing is done using the DIFFTEST option.

RMSEA (Root Mean Square Error Of Approximation)

Estimate	0.060	
90 Percent C.I.	0.049	0.072
Probability RMSEA <=	.05	0.059

CFI/TLI

CFI	0.963
TLI	0.944

Chi-Square Test of Model Fit for the Baseline Model

Value	2450.575
Degrees of Freedom	36
P-Value	0.0000

SRMR (Standardized Root Mean Square Residual)

Value 0.036

STANDARDIZED MODEL RESULTS: STDYX Standardization

	Two-Ta	iled
Estimate	S.E. Est./S.E.	P-Value

PVGU	BY				
PVGU6		0.690	0.020	33.890	0.000
PVGU2		0.738	0.021	34.476	0.000
PVGU3		0.514	0.028	18.557	0.000
PVGU4		0.521	0.031	16.883	0.000
PVGU5		0.621	0.029	21.760	0.000
PVGU1		0.689	0.022	30.692	0.000
PVGU7		0.509	0.032	15.811	0.000
PVGU8		0.575	0.029	19.636	0.000
PVGU9		0.671	0.023	28.885	0.000

PVGU8	WITH				
PVGU5		0.580	0.037	15.569	0.000
PVGU7		0.370	0.038	9.630	0.000

PVGU7	WITH			
PVGU5	0.256	0.040	6.474	0.000
Intercepts				
PVGU1	2.043	0.039	52.548	0.000
PVGU2	1.791	0.033	53.936	0.000
PVGU3	2.417	0.054	44.948	0.000
PVGU4	1.523	0.024	62.929	0.000
PVGU5	1.564	0.032	49.152	0.000
PVGU6	1.967	0.036	54.776	0.000
PVGU7	1.368	0.021	64.161	0.000
PVGU8	1.491	0.029	50.727	0.000
PVGU9	1.599	0.025	63.929	0.000
Variances				
Variances PVGU	1.000	0.000	999.000	999.000
	1.000	0.000	999.000	999.000
		0.000	999.000	999.000
PVGU		0.000	999.000 17.006	999.000 0.000
PVGU Residual V	Variances			
PVGU Residual V PVGU1	variances 0.526	0.031	17.006	0.000
PVGU Residual V PVGU1 PVGU2	Variances 0.526 0.456	0.031 0.032	17.006 14.450	0.000 0.000
PVGU Residual V PVGU1 PVGU2 PVGU3	Variances 0.526 0.456 0.736	0.031 0.032 0.028	17.006 14.450 25.830	0.000 0.000 0.000
PVGU Residual V PVGU1 PVGU2 PVGU3 PVGU4	Variances 0.526 0.456 0.736 0.729	0.031 0.032 0.028 0.032	17.006 14.450 25.830 22.667	0.000 0.000 0.000 0.000
PVGU Residual V PVGU1 PVGU2 PVGU3 PVGU4 PVGU5	Variances 0.526 0.456 0.736 0.729 0.615	0.031 0.032 0.028 0.032 0.035	17.006 14.450 25.830 22.667 17.348	0.000 0.000 0.000 0.000 0.000
PVGU Residual V PVGU1 PVGU2 PVGU3 PVGU4 PVGU5 PVGU6	Variances 0.526 0.456 0.736 0.729 0.615 0.524	0.031 0.032 0.028 0.032 0.035 0.028	17.006 14.450 25.830 22.667 17.348 18.628	$\begin{array}{c} 0.000\\ 0.000\\ 0.000\\ 0.000\\ 0.000\\ 0.000\\ 0.000\end{array}$
PVGU Residual V PVGU1 PVGU2 PVGU3 PVGU4 PVGU5 PVGU6 PVGU7	Variances 0.526 0.456 0.736 0.729 0.615 0.524 0.741	0.031 0.032 0.028 0.032 0.035 0.028 0.033	17.006 14.450 25.830 22.667 17.348 18.628 22.622	0.000 0.000 0.000 0.000 0.000 0.000 0.000

Initial model fit of Psychological Need Thwarting Scale:

 $\chi^{2}(82) = 319.21, p = <.001$ RMSEA = .08, 90CI [.07, .09] CFI = .95 TLI = .93 SRMR = .04

Psychological Need Thwarting Scale: Confirmatory Factor Analysis Following Modifications

Notes: Although initially adequate, the RMSEA was concerning requiring a review of the individual loadings. COMP4 was found to load poorly (.11) in the initial CFA alluding to its exclusion (as shown in red). The inclusion of the highlighted (in yellow) correlation was suggested based on provided modification indices. The input and output of the final model is provided below.

MODEL INPUT:

Analysis:

ESTIMATOR = MLR

Model:

COMP BY COMP1 COMP2 COMP3 COMP4; AUT BY AUT1 AUT4 AUT2 AUT3; REL BY REL1 REL2 REL3; AUT4 WITH AUT1;

MODEL FIT INFORMATION: FOLLOWING APPLICATION OF MODIFICATIONS

Chi-Square Test of Model Fit

Value214.283*Degrees of Freedom40P-Value0.0000Scaling Correction Factor1.3415for MLR

* The chi-square value for MLM, MLMV, MLR, ULSMV, WLSM and WLSMV cannot be used for chi-square difference testing in the regular way. MLM, MLR and WLSM chi-square difference testing is described on the Mplus website. MLMV, WLSMV, and ULSMV difference testing is done using the DIFFTEST option.

RMSEA (Root Mean Square Error Of Approximation)

Estimate	0.066	
90 Percent C.I.	0.058	0.075
Probability RMSEA <=	.05	0.001

CFI/TLI

CFI	0.968
TLI	0.956

Chi-Square Test of Model Fit for the Baseline Model

Value	5551.988	
Degrees of Freedom	55	
P-Value	0.0000	

SRMR (Standardized Root Mean Square Residual)

Value 0.030

STANDARDIZED MODEL RESULTS: STDYX Standardization

	Estimate	Two-Tailed imate S.E. Est./S.E. P-Value			
	Listimute	0.L. L3t./	5.L. I	uiue	
COMP E	Υ				
COMP1	0.895	0.015	56.897	0.000	
COMP2	0.828	0.013	61.869	0.000	
COMP3	0.831	0.010	88.570	0.000	
AUT BY	<i>I</i>				
AUT1	0.843	0.020	36.472	0.000	
AUT2	0.754	0.020	37.276	0.000	
AUT3	0.743	0.014	59.489	0.000	
AUT4	0.790	0.016	48.042	0.000	
REL BY	r				
REL1	0.866	0.012	72.632	0.000	
REL3	0.840	0.016	52.877	0.000	
REL4	0.853	0.014	62.309	0.000	

AUT WITH COMP	0.850	0.018	46.567	0.000
REL WITH				
COMP	0.861	0.018	47.704	0.000
AUT	0.794	0.022	36.178	0.000
AUT4 WITH				
AUT1	0.429	0.043	9.948	0.000
Intercepts				
COMP1	1.978	0.037	53.211	0.000
COMP2	1.927	0.036	53.877	0.000
COMP3	1.833	0.033	55.289	0.000
AUT1	1.777	0.031	56.625	0.000
AUT2	1.817	0.032	56.702	0.000
AUT3	1.761	0.031	56.163	0.000
AUT4	1.953	0.037	52.569	0.000
REL1	1.643	0.028	58.923	0.000
REL3	1.792	0.032	55.362	0.000
REL4	1.904	0.036	53.045	0.000
Variances				
COMP	1.000	0.000	999.000	999.000
AUT	1.000	0.000	999.000	999.000
REL	1.000	0.000	999.000	999.000
Residual Varian	ces			
COMP1	0.312	0.024	12.920	0.000
COMP2	0.310	0.022	13.878	0.000
COMP3	0.195	0.018	10.711	0.000
AUT1	0.448	0.030	14.806	0.000
AUT2	0.431	0.031	14.143	0.000
AUT3	0.290	0.024	12.147	0.000
AUT4	0.376	0.026	14.450	0.000
REL1	0.250	0.021	12.097	0.000
REL3	0.294	0.027	11.029	0.000
REL4	0.272	0.023	11.661	0.000

Initial model fit of Gaming Motivation Scale:

```
\chi^{2}(82) = 81781, p = <.001
RMSEA = .08, 90CI [.07, .08]
CFI = .90
TLI = .87
SRMR = .07
```

Gaming Motivation Scale: Confirmatory Factor Analysis Following Modifications

Notes: Several modifications were made to the initial model, however, the fit indices did not ever reach a satisfactory level. Moreover, the inclusion of these modifications resulted in issues with the model becoming not positive definite. Therefore, given the low Cronbach alpha observed for the intrinsic motivation subscale ($\alpha = .60$), it was possible that the items were not loading on to their correct factors. Thus, the data was included in an exploratory factor analysis (EFA) with a Promax rotation in order to assess factor loadings.

Two issues were identified immediately within the results of the EFA. First, only three factors provided eigenvalues greater than 1. Second, the intrinsic motivation items loaded poorly onto a common factor with one item (IM2) loading more highly onto a factor with items representing external regulation. Therefore, the intrinsic motivation subscale was excluded completely, and the EFA was conducted again with the remaining 15 items.

Results from the second EFA once again suggested only three factors provided eigenvalues greater than 1 within items from the integrated regulation and identified regulation loading on a common factor and external regulation and amotivation loading on unique factors. Items from the introjected regulation subscale weakly crossloaded onto factors representing integrated-identified regulation and amotivation. Therefore, it was not possible to confirm this model.

Although the eigenvalue was under 1, including the fourth factor allowed for introjected regulation to load onto its own factor. Moreover, it significantly reduced item loading residuals.

As such, the four-factor solution was confirmed using a CFA. At first, the data did not fit the model well ($\chi 2(82) = 626.59$, p = < .001; RMSEA = .08, 90CI [.07, .09]; CFI = .90; TLI = .88; SRMR = .08), however, the inclusion of the highlighted (in yellow) correlations, as suggested based on modification indices, adequate fit was obtained. The input and output of the final model is provided below.

MODEL INPUT:

Analysis:

ESTIMATOR = MLR

Model:

INTG_ID BY INTG1 INTG2 INTG3 ID1 ID2 ID3; INTROJ BY INTR1 INTR2 INTR3; EXT BY EXT2 EXT1 EXT3; AM BY AM1 AM2 AM3; ID2 WITH ID1; EXT3 WITH EXT1; INTR3 WITH INTR2;

MODEL FIT INFORMATION: FOLLOWING APPLICATION OF MODIFICATIONS

Chi-Square Test of Model Fit

Value	307.072*
Degrees of Freedom	81
P-Value	0.0000
Scaling Correction Fac	tor 1.2152
for MLR	

* The chi-square value for MLM, MLMV, MLR, ULSMV, WLSM and WLSMV cannot be used for chi-square difference testing in the regular way. MLM, MLR and WLSM chi-square difference testing is described on the Mplus website. MLMV, WLSMV, and ULSMV difference testing is done using the DIFFTEST option.

RMSEA (Root Mean Square Error Of Approximation)

Estimate	0.052	
90 Percent C.I.	0.046	0.058
Probability RMSEA <= .	05	0.280

CFI/TLI

CFI	0.959
TLI	0.947

Chi-Square Test of Model Fit for the Baseline Model

Value	5647.377
Degrees of Freedom	105
P-Value	0.0000

SRMR (Standardized Root Mean Square Residual)

Value

STANDARDIZED MODEL RESULTS: STDYX Standardization

0.048

Two-Tailed			
stimate S.E. Est./S.E. P-Value			
819 0.0	015 53.47	0.000	
826 0.0	015 54.88	36 0.000	
787 0.0	017 45.67	0.000	
62 0.02	3 29.286	0.000	
6 0.02	7 20.917	0.000	
60 0.01	7 44.160	0.000	
661 0.0	026 25.05	0.000	
583 0.0	034 17.05	0.000	
	27 17 07	3 0.000	
476 0.0	12.69	5 0.000	
.00872 0.0	17 50.97	2 0.000	
360 0.0	20 42.10	1 0.000	
765 0.0	25 30.26	3 0.000	
0.648 0	0.032 20.3	306 0.000	
0.629 0	0.041 154	433 0.000	
	819 0.0 826 0.0 787 0.0 52 0.02 56 0.02 50 0.01 725 0.0 661 0.0 583 0.0 568 0.0 476 0.0 372 0.0 360 0.0 765 0.0 0.648 0 0.629 0	e S.E. Est./S.E. 819 0.015 53.47 826 0.015 54.88 787 0.017 45.67 52 0.023 29.286 56 0.027 20.917 50 0.017 44.160 725 0.025 28.59 661 0.026 25.05 583 0.034 17.05 568 0.037 12.29 476 0.037 12.69 872 0.017 50.97 860 0.020 42.10 765 0.025 30.26 0.648 0.032 20.3 0.629 0.041 15.4	

AM WITH INTG_ID INTROJ EXT	0.112 0.653		7 3.025) 16.421 4.115 (
ID2 WITH ID1	0.422	0.034	12.509	0.000
EXT3 WIT EXT1	H 0.606	0.027	22.212	0.000
INTR3 WIT INTR2	H 0.321	0.049	6.490	0.000
Intercepts INTG1 INTG2 INTG3 ID1 ID2 ID3 INTR1 INTR2 INTR3 EXT1 EXT2 EXT3 AM1 AM2	$\begin{array}{c} 1.650\\ 1.713\\ 1.628\\ 1.741\\ 1.773\\ 1.871\\ 1.493\\ 1.299\\ 1.340\\ 1.892\\ 1.774\\ 1.845\\ 1.303\\ 1.279\end{array}$	$\begin{array}{c} 0.032\\ 0.029\\ 0.031\\ 0.033\\ 0.038\\ 0.026\\ 0.020\\ 0.035\\ 0.037\\ 0.035\\ 0.037\\ 0.021\\ 0.024\\ \end{array}$	55.292 53.284 49.379 56.996 64.053 38.624 50.661 50.567 50.146 61.180 52.830	0.000 0.000 0.000 0.000 0.000 0.000
AM3 Variances INTG_ID INTROJ EXT AM	1.000	0.000 0.000 9) 999.000 999.000
Residual Varia INTG1 INTG2 INTG3 ID1	0.330 0.318 0.381	0.025 0.027	13.140 12.806 14.039 18.818	$0.000 \\ 0.000$

ID2	0.691	0.030	23.401	0.000
ID3	0.422	0.026	16.130	0.000
INTR1	0.474	0.037	12.872	0.000
INTR2	0.563	0.035	16.115	0.000
INTR3	0.660	0.040	16.568	0.000
EXT1	0.795	0.033	23.889	0.000
EXT2	0.554	0.050	11.143	0.000
EXT3	0.773	0.036	21.675	0.000
AM1	0.239	0.030	8.014	0.000
AM2	0.260	0.035	7.381	0.000
AM3	0.415	0.039	10.752	0.000

Model A: Path Analysis

MODEL INPUT:

Analysis:

ESTIMATOR = MLR

Model:

COMP WITH REL AUT INTG_ID INTRO AM; AUT WITH REL INTG_ID INTRO AM; REL WITH INTRO AM; INTG_ID WITH INTRO EXT AM; INTRO WITH EXT AM; EXT WITH AM; GENDER WITH HOURS; COMP ON GENDER; AUT ON GENDER; REL ON GENDER; INTG_ID ON HOURS GENDER; INTG_ID ON HOURS; EXT ON HOURS; PVGU ON COMP REL AUT INTG_ID INTRO EXT AM HOURS;

MODEL FIT INFORMATION:

Number of Free Parameters 54

Loglikelihood

H0 Value-16230.710H0 Scaling Correction Factor1.2168for MLR-16208.904H1 Scaling Correction Factor1.1844for MLR

Information Criteria

 Akaike (AIC)
 32569.419

 Bayesian (BIC)
 32835.982

 Sample-Size Adjusted BIC
 32664.472

 (n* = (n + 2) / 24)

Chi-Square Test of Model Fit

Value 42.513* Degrees of Freedom 11 P-Value 0.0000 Scaling Correction Factor 1.0258 for MLR

* The chi-square value for MLM, MLMV, MLR, ULSMV, WLSM and WLSMV cannot be used for chi-square difference testing in the regular way. MLM, MLR and WLSM chi-square difference testing is described on the Mplus website. MLMV, WLSMV, and ULSMV difference testing is done using the DIFFTEST option.

RMSEA (Root Mean Square Error Of Approximation)

Estimate	0.053	
90 Percent C.I.	0.037	0.070
Probability RMSEA <= .	.05	0.363

CFI/TLI

CFI	0.990
TLI	0.958

Chi-Square Test of Model Fit for the Baseline Model

Value	3051.935
Degrees of Freedom	44
P-Value	0.0000

SRMR (Standardized Root Mean Square Residual)

Value 0.042

STANDARDIZED MODEL RESULTS: STDYX Standardization

E	stimate		vo-Tailed S.E. P-	
COMP ON GENDER	0.195	5 0.031	6.338	0.000
AUT ON GENDER	0.145	5 0.030	4.791	0.000
REL ON GENDER	0.189	0.031	6.105	0.000
INTG_ID ON HOURS GENDER	0.355	0.028 7 0.026		0.000 2 0.293
INTRO ON HOURS		0.032	6.741	0.000
EXT ON HOURS	0.193	0.032	6.079	0.000
AM ON HOURS	0.088	0.032	2.710	0.007
PVGU ON COMP REL AUT INTG_ID INTRO EXT AM HOURS	-0.019 0.117 0.117 0.362 0.113 0.232	0.038 0.028 0.035 0.026 0.034	-0.546 3.040 4.199 10.247 4.297 6.771	$\begin{array}{c} 0.585 \\ 0.002 \\ 0.000 \\ 0.000 \\ 0.000 \\ 0.000 \\ 0.000 \end{array}$
COMP WI REL AUT INTG_ID INTRO AM	0.748 0.723	0.019 0.022 0.029	1.889 10.746	0.000 0.059 0.000

AUT WITH REL INTG_ID INTRO AM	0.664 0.035 0.286	0.025 0.029	1.381 9.896	0.167 0.000
REL WITH INTRO AM				
INTG_ID WIT INTRO EXT AM	0.406 0.334	0.030	11.141	0.000
INTRO WITH EXT AM	0.366			
EXT WITH AM		0.031	2.485	0.013
GENDER WI HOURS		0.027	-9.492	0.000
Means GENDER HOURS				
Intercepts PVGU COMP REL AUT INTG_ID INTRO EXT AM	0.739 1.561 1.337 1.726 1.814 1.453 2.081 1.346	0.085 0.096 0.095 0.095 0.102 0.054 0.064 0.052	8.649 16.251 14.052 18.098 17.751 26.785 32.586 25.889	$\begin{array}{c} 0.000\\ 0.000\\ 0.000\\ 0.000\\ 0.000\\ 0.000\\ 0.000\\ 0.000\\ 0.000\\ 0.000\end{array}$

Variances	1.000			
GENDER	1.000	0.00	0 999.00	0 999.000
HOURS	1.000	0.000	999.000	999.000
Residual Varia	nces			
PVGU	0.520	0.027	19.172	0.000
COMP	0.962	0.012	80.310	0.000
REL	0.964	0.012	82.240	0.000
AUT	0.979	0.009	111.261	0.000
INTG_ID	0.868	0.020	43.519	0.000
INTRO	0.954	0.014	69.610	0.000
EXT	0.963	0.012	78.357	0.000
AM	0.992	0.006	174.941	0.000

R-SQUARE

Observed		Two-Tailed			
Variable	Estimate	S.E. I	Est./S.E.	P-Value	
PVGU	0.480	0.027	17.719	0.000	
COMP	0.038	0.012	3.169	0.002	
REL	0.036	0.012	3.053	0.002	
AUT	0.021	0.009	2.395	0.017	
INTG_ID	0.132	0.020	6.594	0.000	
INTRO	0.046	0.014	3.371	0.001	
EXT	0.037	0.012	3.039	0.002	
AM	0.008	0.006	1.355	0.175	

Model B: Path Analysis with Indirect Effects

MODEL INPUT:

Analysis:

ESTIMATOR = MLR

Model:

COMP WITH REL AUT; AUT WITH REL; GENDER WITH HOURS; COMP ON GENDER; AUT ON GENDER; REL ON GENDER; INTG_ID ON COMP AUT HOURS GENDER; INTRO ON COMP REL AUT HOURS; EXT ON HOURS; AM ON COMP REL AUT HOURS; INTG_ID WITH INTRO EXT AM; INTRO WITH EXT AM; EXT WITH AM; PVGU ON COMP REL AUT INTG ID INTRO EXT AM HOURS;

MODEL INDIRECT:

PVGU IND COMP; PVGU IND AUT; PVGU IND REL;

MODEL FIT INFORMATION

Number of Free Parameters 54

Loglikelihood

H0 Value -16229.892 H0 Scaling Correction Factor 1.2126 for MLR H1 Value -16208.904 H1 Scaling Correction Factor 1.1844 for MLR Information Criteria

 Akaike (AIC)
 32567.785

 Bayesian (BIC)
 32834.347

 Sample-Size Adjusted BIC
 32662.837

 (n* = (n + 2) / 24)

Chi-Square Test of Model Fit

Value40.125*Degrees of Freedom11P-Value0.0000Scaling Correction Factor1.0461for MLR

* The chi-square value for MLM, MLMV, MLR, ULSMV, WLSM and WLSMV cannot be used for chi-square difference testing in the regular way. MLM, MLR and WLSM chi-square difference testing is described on the Mplus website. MLMV, WLSMV, and ULSMV difference testing is done using the DIFFTEST option.

RMSEA (Root Mean Square Error Of Approximation)

Estimate	0.051	
90 Percent C.I.	0.034	0.068
Probability RMSEA <= .	.05	0.440

CFI/TLI

CFI	0.990
TLI	0.961

Chi-Square Test of Model Fit for the Baseline Model

Value	3051.935
Degrees of Freedom	44
P-Value	0.0000

SRMR (Standardized Root Mean Square Residual)

Value 0.037

STANDARDIZED MODEL RESULTS: STDYX Standardization

Est	imate		vo-Tailed /S.E. P-	
COMP ON GENDER	0.201	0.032	2 6.300) 0.000
AUT ON GENDER	0.149	0.032	2 4.673	3 0.000
REL ON GENDER	0.192	2 0.032	2 5.963	1 0.000
INTG_ID ON COMP AUT HOURS GENDER	0.024 0.352	0.043 0.029		
INTRO ON COMP REL 4 AUT HOURS	0.127	0.038 0.040	0.476 3.202	0.634
EXT ON HOURS	0.193	0.032	6.080	0.000
	0.099 0.057 0.182 0.105	0.052 0.047	1.101 3.881	0.271 0.000
PVGU ON COMP REL AUT INTG_ID INTRO EXT AM	-0.019 0.116 0.115	0.035 0.038 0.028 0.035	3.037 4.179 10.264 4.303	0.583 0.002 0.000

HOURS	0.056	0.027	2.056	0.040
COMP WITH REL AUT	0.749	0.018 0.018	41.269 39.148	
AUT WITH REL	0.665	0.022	30.355	0.000
GENDER WI HOURS		0.027	-9.514	0.000
EXT		0.030	13.804 11.266 -1.162	0.000
INTRO WITH EXT AM	H 0.389 0.422	0.027 0.036		
EXT WITH AM	0.085	0.033	2.559	0.010
Means GENDER HOURS	2.852 1.285		6 178.48 26.258	1 0.000 0.000
Intercepts PVGU COMP REL AUT INTG_ID INTRO EXT AM	0.738 1.537 1.328 1.710 1.637 0.678 2.081 0.620	$\begin{array}{c} 0.085\\ 0.101\\ 0.100\\ 0.100\\ 0.112\\ 0.075\\ 0.064\\ 0.084\\ \end{array}$		$\begin{array}{c} 0.000\\ 0.000\\ 0.000\\ 0.000\\ 0.000\\ 0.000\\ 0.000\\ 0.000\\ 0.000\\ 0.000\\ \end{array}$
Variances GENDER HOURS	1.000 1.000) 0.000 0.000		

Residual Variances

PVGU	0.516	0.027	18.997	0.000
COMP	0.960	0.013	75.039	0.000
REL	0.963	0.012	77.845	0.000
AUT	0.978	0.010	102.896	0.000
INTG_ID	0.866	0.020	43.695	0.000
INTRO	0.846	0.022	38.750	0.000
EXT	0.963	0.012	78.353	0.000
AM	0.897	0.019	47.245	0.000

R-SQUARE

Observed		Two-Tailed			
Variable	Estimate	S.E. I	Est./S.E.	P-Value	
PVGU	0.484	0.027	17.791	0.000	
COMP	0.040	0.013	3.150	0.002	
REL	0.037	0.012	2.981	0.003	
AUT	0.022	0.010	2.337	0.019	
INTG_ID	0.134	0.020	6.780	0.000	
INTRO	0.154	0.022	7.043	0.000	
EXT	0.037	0.012	3.040	0.002	
AM	0.103	0.019	5.437	0.000	

STANDARDIZED TOTAL, TOTAL INDIRECT, SPECIFIC INDIRECT, AND DIRECT EFFECTS: STDYX Standardization

Two-Tailed Estimate S.E. Est./S.E. P-Value

Effects from COMP to PVGU

Total0.1760.0473.7360.000Total indirect0.1080.0264.1440.000

Specific indirect

PVGU INTG_ID COMP 0.008 0.006 1.475 0.140

PVGU INTRO COMP	0.077	0.017	4.484	0.000
PVGU AM COMP	0.023	0.014	1.629	0.103
Direct PVGU COMP	0.068	0.040	1.687	0.092

Effects from AUT to PVGU

Total	0.207	0.045	4.647	0.000
Total indirect	0.091	0.023	3.984	0.000

Specific indirect

PVGU INTG_ID AUT	0.003	0.005	0.568	0.570
PVGU INTRO				
AUT PVGU AM	0.046	0.015	2.982	0.003
AUT	0.042	0.012	3.449	0.001
Direct PVGU				
AUT	0.116	0.038	3.037	0.002

Effects from REL to PVGU

Total	0.000	0.043	0.007	0.994
Total indirect	0.020	0.022	0.883	0.377

PVGU INTRO REL	0.007	0.014	0.475	0.635
PVGU AM REL	0.013	0.012	1.102	0.271
Direct PVGU REL	-0.019	0.035	-0.550	0.583

Specific indirect

Model C: Path Analysis with Indirect Effects

MODEL INPUT:

Analysis:

ESTIMATOR = MLR

Model:

INTG_ID WITH INTRO EXT AM; INTRO WITH EXT AM; EXT WITH AM; GENDER WITH HOURS; COMP ON INTG_ID INTRO AM GENDER; AUT ON INTG_ID INTRO AM GENDER; REL ON INTRO AM GENDER; COMP WITH REL AUT ; AUT WITH REL AUT ; AUT WITH REL; INTG_ID ON HOURS GENDER; INTRO ON HOURS; EXT ON HOURS; PVGU ON COMP REL AUT INTG ID INTRO EXT AM HOURS;

MODEL INDIRECT: PVGU IND INTG_ID; PVGU IND INTRO; PVGU IND AM;

MODEL FIT INFORMATION:

Number of Free Parameters 54

Loglikelihood

H0 Value -16219.535 H0 Scaling Correction Factor 1.2167 for MLR H1 Value -16208.904 H1 Scaling Correction Factor 1.1844 for MLR Information Criteria

Akaike (AIC)	325	547.069
Bayesian (BIC)	32	813.632
Sample-Size Adjusted	BIC	32642.122
$(n^* = (n+2) / 24)$		

Chi-Square Test of Model Fit

Value	20.716*
Degrees of Freedom	11
P-Value	0.0364
Scaling Correction Fact	tor 1.0263
for MLR	

* The chi-square value for MLM, MLMV, MLR, ULSMV, WLSM and WLSMV cannot be used for chi-square difference testing in the regular way. MLM, MLR and WLSM chi-square difference testing is described on the Mplus website. MLMV, WLSMV, and ULSMV difference testing is done using the DIFFTEST option.

RMSEA (Root Mean Square Error Of Approximation)

Estimate	0.029	
90 Percent C.I.	0.007	0.048
Probability RMSEA <= .	.05	0.963

CFI/TLI

CFI	0.997
TLI	0.987

Chi-Square Test of Model Fit for the Baseline Model

Value	3051.935
Degrees of Freedom	44
P-Value	0.0000

SRMR (Standardized Root Mean Square Residual)

Value 0.014

STANDARDIZED MODEL RESULTS: STDYX Standardization

Two-Tailed Estimate S.E. Est./S.E. P-Value				
COMP ON INTG_ID INTRO AM GENDER	0.275 0.160	0.038 0.037	7.203 4.336	0.000 0.000
AUT ON INTG_ID INTRO AM GENDER	0.234 0.195	0.040 0.036	5.806 5.461	0.000 0.000
REL ON INTRO AM GENDER	0.175	0.037	4.689	0.000
INTG_ID ON HOURS GENDER		0.028 3 0.026		0.000 0 0.276
INTRO ON HOURS	0.239	0.035	6.866	0.000
EXT ON HOURS	0.193	0.032	6.071	0.000
AM ON HOURS	0.117	0.034	3.434	0.001
PVGU ON COMP REL AUT INTG_ID INTRO EXT AM	0.067 -0.019 0.114 0.363 0.110 0.228	0.038 0.027 0.035 0.026	1.680 -0.553 3.040 4.183 10.249 4.263 6.739	$\begin{array}{c} 0.093 \\ 0.580 \\ 0.002 \\ 0.000 \\ 0.000 \\ 0.000 \\ 0.000 \\ 0.000 \end{array}$

HOURS	0.055	0.027	2.052	0.040
INTG ID WIT	Н			
		0.029	14.590	0.000
EXT	0.342	0.030		
AM	0.000		-0.010	
1	0.000	0.000	0.010	0.772
INTRO WITH	ł			
EXT	0.391	0.026	15.006	0.000
AM	0.485	0.032	15.325	0.000
EXT WITH	0.100	0.000		0.001
AM	0.106	0.032	3.273	0.001
GENDER WI	гн			
0	-0.256	0.027	-9.468	0.000
noons	0.250	0.027	7.400	0.000
COMP WITH	ł			
REL	0.716	0.020	35.036	0.000
AUT	0.677	0.022	31.063	0.000
AUT WITH				
REL	0.619	0.024	25.294	0.000
Means				
GENDER	2 951	0.01	7 1710/	5 0.000
HOURS	1.285		26.258	
HOUKS	1.285	0.049	20.238	0.000
Intercepts				
PVGU	0.728	0.085	8.609	0.000
COMP	0.795	0.119	6.669	0.000
REL	0.711	0.112	6.330	0.000
AUT	0.983	0.116	8.498	0.000
INTG ID	1.811	0.101		
INTRO	1.385		24.268	0.000
EXT	2.082	0.064	32.600	0.000
AM	1.302	0.055	23.798	0.000
		-		
Variances				
GENDER	1.000	0.00	0 999.00	0 999.000
HOURS	1.000	0.000	999.000) 999.000

Residual Variances

PVGU	0.500	0.027	18.282	0.000
COMP	0.818	0.024	33.853	0.000
REL	0.863	0.022	39.104	0.000
AUT	0.841	0.022	38.414	0.000
INTG_ID	0.868	0.020	43.572	0.000
INTRO	0.943	0.017	56.513	0.000
EXT	0.963	0.012	78.440	0.000
AM	0.986	0.008	122.820	0.000

R-SQUARE

Observed		Two-Tailed		
Variable	Estimate	S.E. I	Est./S.E.	P-Value
PVGU	0.500	0.027	18.298	0.000
COMP	0.182	0.024	7.557	0.000
REL	0.137	0.022	6.207	0.000
AUT	0.159	0.022	7.243	0.000
INTG_ID	0.132	0.020	6.612	0.000
INTRO	0.057	0.017	3.433	0.001
EXT	0.037	0.012	3.035	0.002
AM	0.014	0.008	1.717	0.086

STANDARDIZED TOTAL, TOTAL INDIRECT, SPECIFIC INDIRECT, AND DIRECT EFFECTS

STDYX Standardization

Two-Tailed Estimate S.E. Est./S.E. P-Value

Effects from INTG_ID to PVGU

Total0.1150.0284.1830.000Total indirect0.0010.0040.1980.843

Specific indirect

PVGU COMP INTG_ID 0.000 0.002 0.178 0.859

PVGU AUT INTG_ID	0.001	0.003	0.170	0.865
Direct PVGU INTG_ID	0.114	0.027	4.183	0.000
Effects from IN	TRO to P	VGU		
Total Total indirect	0.405 0.041		11.338 4.261	
Specific indire	ect			
PVGU COMP INTRO	0.018	0.011	1.617	0.106
PVGU REL INTRO	-0.004	0.007	-0.554	0.580
PVGU AUT INTRO	0.027	0.010	2.723	0.006
Direct PVGU INTRO	0.363	0.035	10.249	0.000
Effects from AN	M to PVG	U		
Total Total indirect	0.258 0.030			
Specific indirect				
PVGU Comp Am	0.011	0.007	1.587	0.113

PVGU REL AM	-0.003	0.006	-0.546	0.585
PVGU AUT AM	0.022	0.009	2.614	0.009
Direct PVGU AM	0.228	0.034	6.739	0.000