A latent profile approach for the study of internet gaming disorder, social media addiction, and psychopathology in a normative sample of adolescents

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Abstract

Introduction

Over the past couple of decades, advancements in digital technologies have brought about positive applications in health, education, and global connectivity. However, for a small minority of individuals, the overuse of these technologies has been associated with negative consequences, ranging from subjective distress to psychopathological symptoms.⁴ Among the various theoretical models of technology-based addictions, Griffiths’ symptom-centered model⁵ has been widely applied to the conceptualization of many technology-based addictions including internet...
gaming disorder (IGD; included in Section III - Emerging Measures and Models - of the most recent fifth edition of the Diagnostic and Statistical Manual of Mental Disorders (DSM-5), and social media addiction (SMA)) which has no status in the DSM-5.

Griffiths’ (2005) model comprises six symptoms: salience refers to when an addictive activity dominates a person’s thinking, feelings, and behavior; mood modification refers to when people engage in specific activities to help change their mood states; tolerance refers to the need to increase the amounts of engagement in the addictive behavior to achieve the former effects; withdrawal refers to the unpleasant feeling states occurring when individuals decrease or suddenly reduce their addictive activities; conflict refers to the intrapsychic and interpersonal problems arising as a consequence of addictive activities; and relapse refers to the unsuccessful efforts to stop engaging in the addictive behavior if the individual is trying to cease.

With regard to the risk factors related to IGD and SMA, previous studies have found relationships between these two technology-based addictions and comorbid disorders such as attention-deficit/hyperactivity disorder, obsessive-compulsive disorder, depressive symptoms, mood and anxiety symptoms. In addition to psychopathological symptoms, impulsivity has been found to be one of the most predictive personality factors of IGD and SMA, especially in adolescent and emerging adult populations.

These studies have been based on variable-centered approaches providing specific information on the importance of each factor to the outcome variable but are imprecise when assuming the homogeneity of the sampled individuals. In light of these limitations, person-centered approaches are useful in examining “similarities and differences among individuals with respect to how variables relate to each other.” The advantages of this approach is that they: (i) can assess whether distinct groups of individuals can be identified via their naturalistic groupings of factors; (ii) offer complex combinations among all possible factors at all possible levels of each factor; and (iii) are appropriate for clinical practice because decisions concerning assessment and treatments are often focused on the individual rather than on the variable or factor.

Given the ambiguous associations between psychopathology and IGD and SMA, as well as the need to apply a research method able to pinpoint the heterogeneity of the technology-based addictions in adolescent populations, the present study used the person-centered approach of latent profile analysis (LPA) to identify groups of adolescents who had similar profiles for multiple dimensions of psychopathology and online addictions. As this statistical method defines unobserved subgroups based on observed indicators without specifying the number of profiles in advance, it is considered a more appropriate method to address research questions that are exploratory in nature and to understand the diversity and complexity in multiple risk factor exposures in adolescent psychopathology.

The present study intended to identify profiles of adolescents characterized by unique patterns of psychopathological risks (somatization, obsessive-compulsive, interpersonal sensitivity, depression, anxiety, hostility, phobic anxiety, paranoid ideation, psychoticism), impulsivity, and two technology-based addictions (ie, IGD and SMA). Gender was also included as suggested by previous literature. Given that no previous studies have reported a specific number of profiles, at least two classes were expected, one with low levels on all risk indicators and the other with high levels on all risk indicators. This study adds to previous literature in this field in that it focused on early-, mid- and late-adolescents. To our best knowledge, no study has done so; rather, other authors have concentrated on youths and emerging adults, prevalently assessing those samples as homogeneous groups.

**Method**

**Participants and procedure**

A sample of 643 participants (312 males; \(M_{age} = 16.02 \text{ years, SD} = 1.43\)), was recruited from high schools in the regions of Central Italy through a convenience sampling, and divided into three age groups: early adolescence (14–15 years; \(n=259, 40.3\%\) of the entire sample), mid-adolescence (16–17 years; \(n=252, 39.2\%\) of the entire sample), and late adolescence (18–19 years; \(n=132, 20.5\%\) of the entire sample). Participants were invited to complete an online self-report questionnaire which took approximately 50 mins to complete. Data collection took place during March to May 2017 (data are available upon request to the authors at: http://dx.doi.org/10.17632/n8ksj69mtt.2).

**Measures**

The Internet Gaming Disorder Scale–Short Form (IGDS9-SF; Italian translation and validation) is a nine-item, single-factor instrument based on DSM-5 IGD core...
It was devised to assess the severity and concomitant detrimental effects of IGD by examining both online and/or offline gaming activities over a 12-month period. The items are answered on a 5-point Likert scale ranging from 1 (Never) to 5 (Very often). Examples of items are: “Do you feel the need to spend increasing amount of time engaged in gaming in order to achieve satisfaction or pleasure?” and “Have you continued your gaming activity despite knowing it was causing problems between you and other people?”. Higher scores indicate a higher degree of gaming disorder. In the present study, the instrument exhibited very good reliability (Cronbach’s α = 0.88).

The Bergen Social Media Addiction Scale (BSMAS; Italian translation and validation) evaluates experiences in the use of social media within a 12-month period. It comprises six items rated on a 5-point Likert scale (from 1 = Very rarely to 5 = Very often) and related to core addiction elements, ie, salience, mood modification, tolerance, withdrawal, conflict, and relapse. Examples of items include: “How often during the last year have you used social media so much that it has had a negative impact on your job/studies?” and “How often during the last year have you felt an urge to use social media more and more?”. In the present study, the internal consistency of the scale was good (Cronbach’s α = 0.78).

The Barratt Impulsiveness Scale for Adolescents (BIS-11-A; Italian translation and validation) is the most widely used 30-item self-report instrument assessing the trait of impulsiveness. Each item is rated on a 4-point Likert scale (from 1 = Never to 4 = Very often). Representative items include: “I do things without thinking” and “I say things without thinking”. A total score is calculated with higher scores indicating higher levels of impulsiveness. In the present study, the overall impulsiveness score was calculated following Fossati et al.’s suggestions. The internal consistency of the scale was good (Cronbach’s α = 0.78).

The Symptom Checklist-90-R (SCL-90-R31 Italian translation and validation) is a 90-item self-report inventory that assesses the extent to which respondents have experienced the nine primary symptoms of psychopathology in the past seven days, namely, Somatization (SOM), Obsessive-Compulsive (O-C), Interpersonal Sensitivity (I-S), Depression (DEP), Anxiety (ANX), Hostility (HOS), Phobic Anxiety (PHOB), Paranoid Ideation (PAR), and Psychoticism (PSY). Each item is rated on a 5-point scale (from 1 = No problem to 5 = Very serious). In the present study, the subscales of the SCL-90-R showed good to excellent internal consistencies (Cronbach’s α = 0.93 for Somatization, 0.92 for Obsessive-Compulsive, 0.81 for Interpersonal Sensitivity, 0.89 for Depression, 0.90 for Anxiety, 0.74 for Hostility, 0.85 for Phobic Anxiety, 0.88 for Paranoid Ideation, and 0.89 for Psychoticism).

Statistics analysis

Data analyses included descriptive statistics (means and standard deviations) and Latent Profile Analyses (LPAs) for each age group to identify classes of adolescents with similar patterns across the individual risk factors (ie, gender, IGD, SMA, impulsiveness and the nine dimensions of psychopathology). The number of latent profiles was determined using three methods: (i) information-theoretic method, (ii) likelihood ratio statistical test method, and (iii) entropy-based criterion. The first method comprised the Akaike Information Criterion (AIC), the Bayesian Information Criteria (BIC), and the Sample-Size Adjusted BIC (SSA-BIC) with lower values indicating more parsimonious models. The second method comprised the Lo-Mendell-Rubin Adjusted Likelihood Ratio Test (LMRT) with a significant p-value (<0.05). The final method comprised last criterion entropy values ranging from 0 to 1 with higher values indicating a better differentiation between profiles. Analyses were conducted using Mplus 8.

Ethics

The research study complied with the general ethical principles of the Declaration of Helsinki and was approved by the Institutional Review Board of International Telematic University Uninettuno (n.8/7/17). Permission was required from school heads to conduct the research. Written informed consent was obtained from students aged over 18 years and from parents or legal guardians for students aged under 18 years.

Results

Means and standard deviations of the study variables are shown in Table 1. This sample does not exceed the clinical cut-offs indicated in previous literature in any of the considered variables (for norms and cut-off points, please see).

On the basis of the aforementioned individual risk factors, a series of latent profile models including two-to-four classes were estimated for the three age groups. The fit indices for each LPA are shown in Table 2.
Although the values of AIC, BIC and SSA-BIC were lower for the three- and four-class solutions, the two-class model was retained for 14–15 year age group due to the significant LMRT value \( (p=0.007) \) and the highest Entropy value. Class 1 comprised 212 participants (81.85% of the group) characterized by low levels of psychopathological symptoms and higher (non-clinical) levels of IGD, SMA, and impulsiveness, whereas Class 2 comprised 47 participants (15.15% of the group) with higher (non-clinical) levels of IGD, SMA, impulsiveness and psychopathological symptoms.
symptoms. Both classes reported extremely high levels of impulsiveness, followed by high levels of SMA, and moderate levels of IGD. Although scores of psychopathological symptoms in Class 2 were not clinical, Depression and Anxiety levels were higher than other scores in youths aged 14–15 (Figure 1).

With regard to gender, the probability that an individual in the first class was female was 0.47 and in the second class 0.60. In short, females were less likely to be in Class 1 than males, but more likely to be in Class 2 than males. As for the age groups 16–17 years and 18–19 years, the LMRT \( (p<0.05) \) and Entropy values indicated that the three-class solution fitted better than the two- and four-class models. More specifically, in the 16–17 year age group, the lower AIC, BIC, and SSA-BIC values, together with a lower significant value of LMRT, favored a three-class solution. Class 1 (78.96% of the group, \( n=199 \)) comprised individuals with low levels of psychopathological symptoms and higher (non-clinical) levels of impulsiveness, SMA, and IGD; Class 2 (14.28% of the group, \( n=36 \)) comprised individuals characterized by moderate (non-clinical) levels of psychopathological symptoms, with higher (non-clinical) scores in depression, impulsiveness, SMA, and IGD; Class 3 (6.75% of the group, \( n=17 \)) comprised individuals with high psychopathological symptoms, especially somatization, anxiety, phobic anxiety, and psychoticism, and lower levels of impulsivity, SMA and IGD in comparison to Classes 1 and 2. As for class 1 and 2, these subjects did not exceed clinical cutoffs for any of the considered variables. Although psychopathological symptoms in youths aged 16–17 did not reach clinical significance, it must be noted that in Class 2 mid-adolescents scored higher on Depression, whereas in Class 3 (where all scores were higher than Class 2), they showed highest scores on Somatization, Anxiety, Hostility and Psychoticism (Figure 2).

As for gender, the probability that an individual in Class 1 was female was 0.54, in Class 2 was 0.50, and in Class 3 was 0.29. In short, females were more likely to be in Class 1 than males, but less likely to be in Class 3 than males. However, males and females had the same probability of being in Class 2.

When looking at the final age group (18–19 years), the LMRT \( p \)-value was not significant for the two-class solution, indicating that this class solution did not fit the data better than a one-class solution and, consequently, there were no relevant classes in this subgroup.

Indicator means for each class in each age group are shown in Table 3.

**Discussion**

The present study sought to define adolescent profiles characterized by different patterns of psychopathological risks, impulsivity, and two specific technology-based addictions – internet gaming disorder (IGD) and social media addiction (SMA). Overall, results from the Latent Profile Analysis (LPA) provided a nuanced understanding of the relative contribution of each factor to risk exposure across age. For this purpose, the total sample was divided...
into three age groups: early, mid and late adolescence, following Steinberg’s suggestions.35

Findings from the LPAs demonstrated that there are no relevant differences between groups as regards IGD, SMA and impulsiveness, but there are groups with low, medium and high (yet non-clinical) patterns of psychopathological symptoms, independently of their level of IGD, SMA and impulsiveness. More specifically, only two profiles were identified in the 14–15 year age group, whereas three profiles emerged in the other two age groups. The two profiles of early adolescents were similar to technology-based addictions and impulsivity, but distinct in psychopathological risks. Although scores of psychopathological symptoms in Class 2 were not clinical, Depression and Anxiety levels were higher than other scores in youths aged 14–15. We can speculate that adolescents in the first profile might use videogames and social media to increase levels of emotional activation, while youths in the second profile might use technologies wanting to reduce the psychological discomfort.36–39

In mid-adolescence (16–17 years), the three profiles identified were similar in psychopathological risks, but different in technology-based addictions and impulsivity. Indeed, mid-aged adolescents in the third profile were characterized by higher levels of psychopathological risks associated with higher levels of SMA, but lower levels of IGD. Although psychopathological symptoms in youths aged 16–17 did not reach clinical significance, it must be noted that in Class 3 (where all scores were higher than Class 2), they showed highest scores on Somatization, Anxiety, Hostility and Psychoticism. Differently from younger adolescents, these youths show a more complex psychological functioning.

In brief, LPAs showed that no different groups exist in this sample with regards to levels of IGD and SMA, but different classes emerge both in 14–15 and 16–17 youths with regards to psychopathological risk. This was an unexpected result considering previous literature, which suggested correlations between these variables. Nonetheless, a seminal work found strongest associations in the adult population, rather than in adolescents, probably because older subjects experienced the negative consequences of the technology misuse for a longer period, and this has eventually led to comorbid psychopathology.11

With regard to gender, the LPAs showed that in all age groups, females were more likely to belong to less problematic profiles (namely, Class 1) than males, except for their levels of impulsivity. However, it is noteworthy that the levels of IGD and SMA were similar between the profiles in all age groups but the profiles were differentiated principally with respect to the levels of psychopathological symptoms. Here, although boys and girls showed similar scores in IGD and SMA, they belonged to different profiles given the causal relationship that could associate problematic use of social networking and gaming with psychopathology (which was unexplored in this cross-sectional study). For example, problematic gaming and excessive use of social networking sites in males might lead to an increase of psychopathological symptoms, whereas in girls they might reduce the symptoms on the basis of the differences between boys and girls in choosing the type of videogame and social networking site (for males: playing games, shooter games, and Facebook; for females: adventure games, quiz games, and Instagram)40,41 and in the motivations driving them (for males: competition; for females: emotional closeness).42
The results of the present study should be interpreted in light of some limitations. First, the homogeneity of the sample in terms of race and geographical origin does not enable broad generalization of the results to a wider population. Second, although the psychometric tools used in this research were valid and reliable, they were self-report measures and are open to well-known biases (such as those associated with social desirability and memory recall). Third, the cross-sectional nature of the present research limited the possibility to draw meaningful conclusions about the cause-and-effect relationship between the variables examined.

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Author contributions

All authors contributed toward data analysis, drafting and revising the paper, gave final approval of the version to be published and agree to be accountable for all aspects of the work.

Disclosure

The authors declare no conflicts of interest in this work.

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