Short version of the Smartphone Addiction Scale in Chinese adults: Psychometric properties, sociodemographic, and health behavioral correlates

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Background and aims: Problematic smartphone use (PSU) is an emerging but understudied public health issue. Little is known about the epidemiology of PSU at the population level. We evaluated the psychometric properties of the Smartphone Addiction Scale – Short Version (SAS-SV) and examined its associated sociodemographic factors and health behaviors in Chinese adults in Hong Kong. **Methods:** A random sample of 3,211 adults aged ≥ 18 years (mean $\pm$ SD: 43.3 ± 15.7 , 45.3% men) participated in a population-based telephone survey in Hong Kong and completed the Chinese SAS-SV. Multivariable linear regressions examined the associations of sociodemographic factors, health behaviors, and chronic disease status with SAS-SV score. Data were weighted by age, sex, and education attainment distributions of the Hong Kong general population. **Results:** The Chinese SAS-SV is internally consistent (Cronbach's $\alpha = .844$) and stable over 1 week (intraclass correlation coefficient = .76, $p < .001$). Confirmatory factor analysis supported a unidimensional structure established by previous studies. The weighted prevalence of PSU was 38.5% (95% confidence interval: 36.9%, 40.2%). Female sex, younger age, being married/cohabitated or divorced/separated (vs. unmarried), and lower education level were associated with a higher SAS-SV score (all $ps < .05$). Current smoking, weekly to daily alcohol drinking, and physical inactivity predict greater PSU after controlling for sociodemographic factors and mutual adjustment. **Discussion and conclusions:** The Chinese SAS-SV was found valid and reliable for assessing PSU in Hong Kong adults. Several sociodemographic and health behavioral factors were associated with PSU at the population level, which may have implication for prevention of PSU and future research.

Keywords: problematic mobile phone use, smartphone dependence, smartphone ownership, epidemiology, population studies, validation studies

INTRODUCTION

Problematic use of mobile information and communication technologies (ICTs) that directly lead to significant and persistent functional impairments and distress may constitute behavioral addiction (Kardefelt-Winther et al., 2017). Emerging research has found that problematic use of mobile phone is associated with several health hazards, ranging from psychosocial disturbances like anxiety and depression (Elhai, Dvorak, Levine, & Hall, 2017) to potentially fatal injuries from traffic accidents wherein mobile phone use was a contributing factor (Cazzulino, Burke, Muller, Arbogast, & Upperman, 2014; Kim, Min, Kim, & Min, 2017), and social problems like poor family and peer relationship (Hawi & Samaha, 2017; Wang et al., 2017). However, partly due to the diverse psychometric scales used to assess mobile phone use dependency, very little is known about its prevalence and

course, and its conceptualization is much debated (Billieux, Maurage, Lopez-Fernandez, Kuss, & Griffiths, 2015; Panova & Carbonell, 2018).

The Smartphone Addiction Scale (SAS), originally developed in Korean but published in English, is a contemporary scale developed to assess problematic smartphone use (PSU; Kwon, Lee, et al., 2013). Alternative measures of PSU include Smartphone Addiction Inventory (Lin et al., 2014), Smartphone Addiction Proneness Scale (Kim, Lee, Lee, Nam, & Chung, 2014), Problematic Mobile Phone Use Questionnaire – Revised (Kuss, Harkin, Kanjo, & Billieux, 2018), and Implicit Association Test measures for

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Smartphone and Internet Addiction (Roh et al., 2018). The short version of SAS (SAS-SV; Kwon, Kim, Cho, & Yang, 2013) is among the most widely used instruments with validated translations in Turkish (Noyan, Darçın, Nürmedov, Yılmaz, & Dilbaz, 2015), Italian (De Pasquale, Sciacca, & Hichy, 2017), Spanish and French (Lopez-Fernandez, 2017), and Arabic (Sfendla et al., 2018), making it a useful instrument for cross-cultural comparisons and further research. However, the psychometric properties of the Chinese translation of SAS-SV are unknown.

The original SAS contains 33 items assessing “daily life disturbance,” “positive anticipation,” “withdrawal,” “cyberspace-oriented relationship,” “overuse,” and “tolerance” symptoms of PSU (Kwon, Lee, et al., 2013). A content validation process was conducted with three items from the “daily life disturbance,” four from the “withdrawal,” and one each from the “cyberspace-oriented relationship,” “overuse,” and “tolerance” subscales selected to form the 10-item SAS-SV (Kwon, Kim, et al., 2013). Although the scale developers did not report its factor structure (Kwon, Kim, et al., 2013), validation studies have suggested a single factor structure for the SAS-SV (De Pasquale et al., 2017; Hawi & Samaha, 2017; Lopez-Fernandez, 2017; Noyan et al., 2015; Sfendla et al., 2018). Confirmatory factor analysis (CFA) is warranted to validate the unidimensional structure of the instrument in Chinese.

Identifying sociodemographic and health behavioral risk factors associated with PSU can help identify at-risk population for further research and guide targeted interventions. Most related studies to date were conducted in relatively small, convenience samples of adolescents or young adults (typically college students) with uncertain generalizability (De-Sola Gutiérrez, de Fonseca, & Rubio, 2016). As smartphone use continues to rise among adults and older people, population studies are needed for a broader understanding of the PSU phenomenon.

Hong Kong is a technologically advanced city of China and ranked 6th out of 175 countries in the ICT Development Index in 2016 (International Telecommunication Union, 2016). The proportion of smartphone users surged rapidly from 54.0% in 2012 to 85.6% in 2016 (Census and Statistics Department, 2017). Therefore, capitalizing on the extensive smartphone penetration in Hong Kong, this study aims to (a) evaluate the psychometric properties of the Chinese translation of SAS-SV and (b) examine the sociodemographic and health behavior correlates of PSU in the general adult population in Hong Kong.

METHODS

Participants and procedure

The Hong Kong Family and Health Information Trends Survey (hereafter referred to as “the Survey”), under the project “FAMILY: A Jockey Club Initiative for a Harmonious Society,” is a cross-sectional, population-based telephone study. The Survey aimed to examine the use of ICTs in relation to family and individual health and behavior in the Hong Kong general public. Details of the methods have been reported in previous phases of the Survey

(Shen et al., 2017; Wang, Viswanath, Lam, Wang, & Chan, 2013). The present phase, conducted from February to May 2017, included the measure of PSU.

The Public Opinion Programme of the University of Hong Kong, a reputable local survey agency, was commissioned to implement the Survey. Cantonese-speaking Hong Kong residents aged ≥ 18 years were selected by a two-stage, probability-based sampling procedure. In the first stage, landline telephone numbers were randomly generated using known dialing codes retrieved from the official numbering plan for telecommunication services, which covers nearly all residents in Hong Kong. After excluding invalid numbers according to computer and manual dialing records, the generated numbers were randomly listed for telephone contact. In the second stage, upon successful contact with a household, an eligible household member whose coming birthday was closest to the interview date was invited to participate in the anonymous survey. Participation is completely voluntary with no incentive. Each interview was administered in Cantonese by a trained interviewer using a web-based computer-assisted telephone interview system. A total of 4,054 out of 5,573 eligible subjects completed the Survey (response rate = 70.2%), with 80.0% (3,243/4,054) reporting owning a smartphone. Smartphone ownership was associated with younger age, unmarried, employed, higher education level, and higher monthly household income (all $ps < .001$), but not with sex ($p = .63$; Table 1).

Measures

The SAS-SV contains 10 items, each scores on a Likert scale of 1 (*strongly disagree*) to 6 (*strongly agree*). The sum of these items gives an overall SAS-SV score (range: 10–60) with higher score indicating PSU. A native Chinese fluent in English translated the English SAS-SV into Chinese with local adaptations. We modified the 8th item of the original SAS-SV “Constantly checking my smartphone so as not to miss conversations between other people on Twitter or Facebook” by replacing “Twitter or Facebook” with “WhatsApp, Facebook, or WeChat,” which were more relevant to this study population. Backward translation was conducted by a native English speaker fluent in Chinese. Discrepancies were resolved by discussion and further modification before the instrument was finalized for use in this study.

Respondents reported their smoking status (categorized into “never smoker,” “former smoker,” and “current smoker”), alcohol drinking habit [“never drinker,” “former drinker,” “occasional drinker (<1 day/month),” “monthly drinker (1–3 days/month),” and “weekly to daily drinker (≥ 1 day/week)”], and number of days in the past week they performed at least 10 min of moderate physical exercise (recoded as “none,” “1–4 days,” and “5–7 days”). History of doctor-diagnosed chronic diseases was also reported (dichotomized as “yes” or “no”).

We also collected data on sex, age, and marital status (“unmarried,” “married/cohabitated,” “divorced/separated,” and “widowed”). Information on indicators of socioeconomic status (SES), including highest education level (“primary or below,” “secondary,” and “tertiary or above”), employment status (“employed,” “unemployed,” “home-makers,” “full-time students,” and “retired”), and monthly

Table 1. Sociodemographic of the sample by smartphone ownership ($N = 4,054$)

	Smartphone owner ($n = 3,243$) n (%) ^a	Smartphone non-owner ($n = 811$) n (%) ^a	χ^2 (p value)
Male	1,468 (45.3)	358 (44.1)	.620
Age (years)			<.001
18–24	360 (11.1)	10 (1.2)	
25–34	675 (20.8)	25 (3.0)	
35–44	701 (21.6)	35 (4.3)	
45–54	683 (21.1)	96 (11.9)	
55–64	538 (16.6)	181 (22.3)	
65+	285 (8.8)	465 (57.3)	
Marital status			<.001
Unmarried	1,060 (32.7)	104 (12.8)	
Married/cohabitated	1,996 (61.6)	481 (59.4)	
Divorced/separated	83 (2.6)	49 (6.0)	
Widowed	104 (3.2)	177 (21.8)	
Employment status			<.001
Employed	1,805 (55.7)	131 (16.1)	
Unemployed	160 (4.9)	61.1 (7.5)	
Homemakers	558 (17.2)	165 (20.3)	
Full-time students	255 (7.9)	2 (0.2)	
Retired	465 (14.4)	453 (55.9)	
Highest education level			<.001
Primary or below	428 (13.2)	531 (65.5)	
Secondary	1,701 (52.5)	249 (30.1)	
Tertiary or above	1,114 (34.4)	31 (3.9)	
Monthly household income			<.001
$\geq \$40,000$	942 (32.2)	45 (6.6)	
\$30,000–\$39,999	438 (15.0)	38 (5.5)	
\$20,000–\$29,999	697 (23.8)	94 (13.7)	
\$10,000–\$19,999	549 (18.8)	147 (21.5)	
$< \$9,999$	298 (10.2)	360 (52.6)	

Note. ^aWeighted by age, sex, and education attainment distribution of the Hong Kong general population.

household income [$\geq \$40,000$,” “\$30,000–\$39,999,” “\$20,000–\$29,999,” “\$10,000–\$19,999,” and “ $\leq \$9,999$ ” (reported in HKD; HKD7.8 $\approx$ USD1)], was also collected.

Statistical analysis

Cronbach’s α was used to assess the internal consistency of the Chinese SAS-SV. Item analyses were conducted by calculating the corrected item-total correlation and Cronbach’s α for items deleted from the scale. One hundred subjects were randomly selected to evaluate 1-week test–retest reliability using intraclass correlation coefficient.

CFA with diagonally weighted least square estimation was used to verify the single factor structure for SAS-SV using LISREL version 9.30 (Li, 2016). All 10 items of SAS-SV were specified as ordinal variables. The asymptotic covariance matrix needed for the estimation was computed using PRELIS included in the LISREL application. Additional covariations among items 1–3 and items 4–7 in the SAS-SV were expected as they were similarly formulated and conceptualized to assess the “daily life disturbances” symptom and “withdrawal” symptom, respectively, in the original SAS (Kwon, Lee, et al., 2013). Residual covariances were thus freely estimated among items 1–3 and items 4–7 a priori to reflect the intended design of SAS-SV (Cole, Ciesla, & Steiger, 2007). No post-hoc modifications

were allowed. Fit index value of <0.10 for root mean square error of approximation, >0.90 for comparative fit index, and >0.90 for Tucker–Lewis index indicate satisfactory model fitness (Hu & Bentler, 1995; MacCallum, Browne, & Sugawara, 1996).

Given the large sample size ($n > 300$), we examined the normality of SAS-SV score by checking its skewness and kurtosis, with skewness of ≤ 2 and kurtosis of ≤ 7 indicating a normal distribution (Kim, 2013). Floor and ceiling effects were evaluated by measuring the proportions of subjects scoring the lowest and highest SAS-SV scores, with $\leq 15\%$ indicating satisfactory floor and ceiling effects (Terwee et al., 2007).

To enhance the representativeness of the results of this study, all data were weighted by the sex, age, and education attainment distributions of the Hong Kong general population using random iterative method (Izrael, Hoaglin, & Battaglia, 2004). Sociodemographic profiles of the participants with or without a smartphone were compared using χ^2 tests. To determine the prevalence of PSU in the Hong Kong population for cross-cultural comparison, the widely used SAS-SV cut-off scores of ≥ 31 for males and ≥ 33 for females were used as proposed by the scale developers (Kwon, Kim, et al., 2013). Multivariable linear regressions were used to compute regression coefficients (b) of SAS-SV score in relation to all sociodemographic factors.

The associations of health behaviors and chronic disease status with SAS-SV scores were also examined with adjustment of the sociodemographic factors. Given the ordinal responses, age, highest education level, monthly household income, and physical activity were analyzed as continuous variables in separate models to calculate the p values for linear trend (p for trend). All statistical analyses (except CFA) were conducted using Stata/IC version 13.1 (Stata-Corp LLC, College Station, TX, USA).

Ethics

The study was conducted in accordance with the Declaration of Helsinki. Ethical approval was granted by the institutional review board of the University of Hong Kong/Hospital Authority West Cluster. All participants were informed about the study and provided verbal informed consent. The telephone interviews were tape-recorded for quality checking with participants' consent. The records were then erased 6 months after the completion of the Survey.

RESULTS

Sample characteristics

After excluding those with missing data in any item of SAS-SV ($n = 32$), 3,211 subjects were included in the regression analyses. The mean (SD , range) age was 43.3 (15.7, 18–96) years and 45.3% were male.

Psychometric properties of the Chinese SAS-SV

Table 2 shows that the 10 items had corrected item-total correlations of .41–.66, exceeding the acceptable cut-off of .40. The Cronbach's α of the scale was .844. Removal of any item slightly reduced the internal consistency, which ranged from .819 to .843. Intraclass correlation for 1-week test-retest reliability was .76 ($p < .001$; $n = 100$). CFA showed that the single factor structure for the Chinese

translation of SAS-SV exhibited acceptable fit (root mean square error of approximation = 0.084; comparative fit index = 0.98; Tucker–Lewis index = 0.96).

The mean (SD) SAS-SV score was 28.9 (10.1) and generally followed a normal distribution (skewness = 0.28; kurtosis = 2.55). Of all subjects, 1.9% achieved the lowest possible score and 0.2% achieved the highest possible score, suggesting minimal floor and ceiling effects ($\leq 15\%$).

Sociodemographic correlates of SAS-SV score

The mean [95% confidence interval (CI)] SAS-SV scores were 27.9 (27.4, 28.4) in males and 29.7 (29.3, 30.2) in females ($p < .001$). The corresponding prevalence (95% CI) of PSU was 37.7% (35.3%, 40.3%) and 39.2% (36.9%, 41.5%), respectively ($p = .56$).

Both bivariable and multivariable analyses showed that being female ($p < .001$) and younger (p for trend $< .001$) were associated with higher scores in SAS-SV compared to male and older respondents (Table 3). After adjusting for other sociodemographic factors, higher SAS-SV scores were observed in married/cohabitating ($b = 2.72$; 95% CI = 1.60, 3.84) and divorced/separated ($b = 3.31$; 95% CI = 0.79, 5.82) respondents relative to those who were unmarried. Education level was inversely associated with SAS-SV score (p for trend = .008). Compared to those with monthly household income of “ $\geq \$40,000$,” a higher SAS-SV score was observed for respondents in the category of “\$10,000–\$19,999” ($b = 2.32$; 95% CI = 1.16, 3.47) but not the other income groups. Homemakers had lower SAS-SV score ($b = -1.35$; 95% CI = -2.51, -0.19) than employed respondents.

Health behavior correlates of SAS-SV score

Table 4 shows that current smokers and weekly to daily drinkers had higher SAS-SV scores than never smokers ($b = 1.48$; 95% CI = 0.19, 2.78) and never drinkers ($b = -1.32$; 95% CI = -2.44, -0.20), respectively. Lower SAS-SV scores were observed in respondents who

Table 2. The Chinese Smartphone Addiction Scale – Short Version (reported in English)

Based on your current situation, to what extent do you agree with the following statements? ^a		Mean (SD) score ^b	Corrected item-total correlation	Cronbach's α if item deleted ^c
1	Missing planned work due to smartphone use	2.4 (1.4)	.43	.839
2	Having a hard time concentrating in class, while doing assignments, or while working due to smartphone use	2.7 (1.6)	.44	.839
3	Feeling pain in the wrists or at the back of the neck while using a smartphone	3.0 (1.7)	.41	.843
4	Won't be able to stand not having a smartphone	3.5 (1.7)	.54	.830
5	Feeling impatient and fretful when I am not holding my smartphone	2.6 (1.5)	.64	.820
6	Having my smartphone in my mind even when I am not using it	2.6 (1.5)	.65	.820
7	I will never give up using my smartphone even when my daily life is already greatly affected by it	3.4 (1.7)	.55	.829
8	Constantly checking my smartphone so as not to miss conversations between other people on WhatsApp, Facebook, or WeChat	3.6 (1.5)	.56	.828
9	Using my smartphone longer than I had intended	2.9 (1.5)	.66	.819
10	The people around me tell me that I use my smartphone too much	2.1 (1.3)	.57	.828

Note. SD : standard deviation.

^aEach item scores on a scale of 1 (*strongly disagree*) to 6 (*strongly agree*). ^bWeighted by age, sex, and education attainment distribution of the Hong Kong general population. ^cOverall Cronbach's $\alpha = .844$.

Table 3. Associations of sociodemographic characteristics with SAS-SV score ($n = 3,211$)

	Mean (SD) SAS-SV score ^a	<i>b</i> [95% CI]	
		Crude association	Adjusted association ^b
Sex			
Male	27.9 (9.7)	Ref.	Ref.
Female	29.7 (10.4)	1.87 [1.17, 2.58]***	1.67 [0.86, 2.48]***
Age (years)			
18–24	29.9 (8.8)	Ref.	Ref.
25–34	30.5 (9.6)	0.69 [–0.60, 1.98]	0.07 [–1.92, 2.06]
35–44	29.4 (10.6)	–0.50 [–1.78, –0.78]	–2.61 [–4.72, –0.50]*
45–54	27.6 (10.1)	–2.28 [–3.57, –0.99]**	–4.50 [–6.68, –2.32]***
55–64	28.3 (10.4)	–4.18 [–2.86, –0.16]**	–4.18 [–6.51, –1.85]***
65+	26.7 (10.5)	–3.17 [–4.74, –1.59]***	–6.80 [–9.58, –4.01]***
<i>p</i> for trend		<.001	<.001
Marital status			
Unmarried	28.8 (9.4)	Ref.	Ref.
Married/cohabitated	29.0 (10.5)	0.26 [–0.49, 1.02]	2.72 [1.60, 3.84]***
Divorced/separated	28.9 (11.1)	0.10 [–2.17, 2.37]	3.31 [0.79, 5.82]*
Widowed	27.6 (10.0)	–1.12 [–3.18, 0.94]	1.57 [–0.86, 4.00]
Highest education level			
Primary or below	29.2 (11.4)	Ref.	Ref.
Secondary	28.9 (10.2)	–0.29 [–1.37, 0.80]	–1.11 [–2.39, 0.17]
Tertiary or above	28.7 (9.5)	–0.55 [–1.69, 0.59]	–1.76 [–3.26, –0.26]*
<i>p</i> for trend		.320	.008
Employment status			
Employed	29.0 (10.1)	Ref.	Ref.
Unemployed	28.0 (9.7)	–1.00 [–2.65, 0.65]	–1.23 [–2.99, 0.53]
Homemakers	28.8 (10.4)	–0.18 [–1.14, 0.79]	–1.35 [–2.51, –0.19]*
Full-time students	30.2 (8.4)	1.22 [–0.11, 2.55]	0.67 [–1.53, 2.86]
Retired	28.2 (11.0)	–0.84 [–1.88, 0.20]	0.80 [–0.73, 2.33]
Monthly household income			
≥\$40,000	28.2 (9.8)	Ref.	Ref.
\$30,000–\$39,999	29.2 (10.1)	1.00 [–0.15, 2.15]	0.92 [–0.24, 2.08]
\$20,000–\$29,999	29.2 (10.3)	1.00 [0.01, 2.00]*	0.78 [–0.26, 1.82]
\$10,000–\$19,999	30.7 (10.5)	2.49 [1.42, 3.55]***	2.32 [1.16, 3.47]***
<\$9,999	27.2 (10.0)	–1.00 [–2.32, 0.32]	0.09 [–1.39, 1.58]
<i>p</i> for trend		.110	.040

Note. SD: standard deviation; SAS-SV: Smartphone Addiction Scale – Short Version; CI: confidence interval.

^aWeighted by age, sex, and education attainment distribution of the Hong Kong general population. ^bMutually adjusted for other variables in the table.

* $p < .05$. ** $p < .01$. *** $p < .001$.

performed ≥ 5 days (vs. 0 day) of moderate physical exercise per week ($b = -1.32$; 95% CI = $-2.44, -0.20$). No association between chronic disease status and SAS-SV was observed ($b = -0.50$; 95% CI = $-1.42, 0.43$).

DISCUSSION

This is the first study to (a) evaluate the psychometric properties of the Chinese version of SAS-SV, (b) provide an estimate of PSU prevalence, and (c) examine the socio-demographic and health behavior associated with PSU in a large population-representative sample of adults including older people. The Chinese translation of SAS-SV showed good internal consistency (Cronbach's $\alpha = .844$), comparable to those observed in other languages (.79–.91) (De Pasquale et al., 2017; Haug et al., 2015; Hawi & Samaha, 2017; Lopez-Fernandez, 2017; Noyan et al., 2015; Sfindla et al., 2018), with satisfactory test–retest reliability.

Results from item analyses suggested that all items were measuring the same construct of the scale and non-redundant. CFA supported a unidimensional structure for the Chinese SAS-SV with adequate model fitness, as suggested by previous factor analyses in other languages (De Pasquale et al., 2017; Hawi & Samaha, 2017; Lopez-Fernandez, 2017; Noyan et al., 2015; Sfindla et al., 2018). The SAS-SV score also followed a normal distribution (a desirable property for many statistical tests) with minimal floor and ceiling effects. Overall, the findings of this study supported the Chinese SAS-SV as a useful tool to measure PSU in Hong Kong adults, which can be further adapted for use in other Chinese subpopulations.

The prevalence of PSU observed in our sample (38.5%) was high compared to those observed in western, convenience samples of adults from Spain (12.5%) and Belgium (21.5%) (Lopez-Fernandez, 2017), and students from Switzerland (16.9%) using the SAS-SV (Haug et al., 2015). This discrepancy may be explained in part by the

Table 4. Associations of health behaviors and chronic disease status with SAS-SV score

	<i>n</i> (%) ^a	<i>b</i> [95% CI]	
		Model 1 ^b	Model 2 ^c
Cigarette smoking			
Never smoker	2,572 (80.1)	Ref.	Ref.
Former smoker	328 (10.2)	0.22 [−1.05, 1.50]	−0.64 [−1.38, 1.25]
Current smoker	311 (9.7)	1.81 [0.54–3.09]**	1.48 [0.19, 2.78]*
Alcohol drinking habit			
Never drinking	1,386 (43.2)	Ref.	Ref.
Former drinking	128 (4.0)	0.91 [−1.02, 2.85]	1.14 [−0.86, 3.14]
Occasional drinking	1,111 (34.6)	1.34 [0.49, 2.19]**	1.24 [0.39, 2.10]**
Monthly drinking	266 (8.3)	1.14 [−0.26, 2.55]	1.03 [−0.38, 2.44]
Weekly or more drinking	320 (10.0)	1.91 [0.61, 3.21]**	1.50 [0.18, 2.83]*
Moderate physical exercise			
None	1,908 (59.5)	Ref.	Ref.
1–4 days/week	870 (27.1)	−0.36 [−1.22, 0.50]	−0.45 [−1.32, 0.41]
5–7 days/week	431 (13.4)	−1.32 [−2.44, −0.20]*	−1.32 [−2.44, −0.20]*
<i>p</i> for trend		.024	.019
Chronic disease diagnosis			
No	2,358 (73.4)	Ref.	Ref.
Yes	854 (26.6)	−0.50 [−1.41, 0.42]	−0.50 [−1.42, 0.43]

Note. No results reached a significance level of $p < .001$. SAS-SV: Smartphone Addiction Scale – Short Version; SD: standard deviation; CI: confidence interval.

^aWeighted by age, sex, and education attainment distribution of the Hong Kong general population. ^bAdjusted for sex, age, marital status, highest education level, employment status, and monthly household income. ^cAdditionally adjusted for other variables in the table.

* $p < .05$. ** $p < .01$.

differences in sampling methods while it likely reflects the geographical disparity in ICTs development and penetration in Hong Kong. Specifically, Hong Kong is significantly more technologically advanced than other countries with affordable (0.18% of gross national income per capita) mobile-cellular services (International Telecommunication Union, 2016), promoting mobile phone use. Furthermore, as mobile phone use has been rapidly increasing in the past few years, our more recent data in 2017 would perhaps be expected to show higher prevalence of PSU. Cultural difference may also contribute to the discrepancy, as suggested by a cross-cultural study, which found markedly higher rate of PSU in convenience samples of young adults in China (63%) than Germany (7.0%) (Lachmann et al., 2018). Note that the suggested SAS-SV cut-off scores used to identify subjects with PSU were established by receiver-operating characteristic analyses on Korean adolescents (Kwon, Kim, et al., 2013), which may be less applicable to adults. Further studies are needed to determine the optimal cut-offs for other age groups. Nevertheless, the high prevalence of PSU in Hong Kong underscores the need to examine its adverse health consequences at the population level.

Consistent with previous studies, we found that female had more dependent use of smartphone, which may relate to their greater sociability and feminine ideal of collaboration nurturing (Billieux, 2012; De-Sola Gutiérrez et al., 2016; Lopez-Fernandez et al., 2017). We also observed older age strongly predict lower PSU, as younger respondents may have lower self-control ability and seem more reliant on using smartphone for daily life activities (Billieux, 2012; De-Sola Gutiérrez et al., 2016). Previous research did not find clear relationship between SES and problematic use of mobile phone (Billieux, 2012). We found lower risk of PSU

in respondents with higher levels of education, possibly explained by their greater health literacy and thus lower risk of developing unhealthy behaviors (DeWalt & Pignone, 2005). The roles of other indicators of SES (employment status and household income) in PSU were less clear. Lower household income seemed to predict PSU, but the results were inconsistent, as we found SAS-SV score was not associated with the poorest category of household income. This contrasted to the strong and consistent associations between higher SES and Internet health information seeking and smartphone health apps possession (the inverse ICT law: those most in need of healthcare have less access to health-related ICTs) observed in our earlier studies (Shen et al., 2017; Wang et al., 2013). It is possible that respondents in the lowest income group may avoid heavy use of smartphones to reduce the expense or cannot afford to buy expensive prepaid data packages for mobile Internet services. This contributes to the null association between low income and PSU. Of note, smartphones have many functions (e.g., social networking, Internet browsing, online shopping, and video gaming), all of which may have distinct relations to SES. Therefore, the relationship between SES and PSU may be unique with complex diversity depending on the use of specific smartphone functions and costs. Future studies may benefit from more detailed measurements of the pattern of smartphone usage including cost barriers and the type of smartphone functions most frequently used by owners.

We found intriguing associations between higher risk of PSU and both married/cohabitated and divorced/separated respondents when compared to the never-married. Being married/cohabitated is associated with more frequent use of telephone calls, instant messaging, and e-mail for family communication compared to those who are single

(Wang et al., 2015), which may contribute to more dependent use of smartphones. On the contrary, problematic use of smartphone functions unrelated to family communication such as online video gaming and gambling may worsen family relationships and contribute to marital breakdown (Black, Shaw, McCormick, & Allen, 2012; Wang et al., 2014). Overuse of social networking sites (e.g., Facebook) may also contribute to family conflict due to reduced time spent with family or inattention during face-to-face conversations (Sharaievska & Stodolska, 2017). Disrupted communication pattern may also mediate the relationship between problematic use of media (including smartphone use, online gaming, and video watching) and marital dissatisfaction (Spencer, Lambertsen, Hubler, & Burr, 2017).

We found current smoking and weekly or more frequent alcohol drinking are associated with higher risk of PSU. Similar findings in adolescent samples of Spanish and Chinese from Taiwan were reported (Sánchez-Martínez & Otero, 2009; Yang, Yen, Ko, Cheng, & Yen, 2010). These correlations may be explained by a shared underlying psychosocial process that contributes to overuse and/or addictions. Established risk factors of PSU, including personality traits like neuroticism and lower conscientiousness, mental health problems like anxiety and depression, and other attributes like low self-esteem and low self-control ability (Billieux et al., 2015; De-Sola Gutiérrez et al., 2016; Hussain, Griffiths, & Sheffield, 2017; Park & Choi, 2017), may also contribute to substance use (Kotov, Gamez, Schmidt, & Watson, 2010). Future research that includes measures of these psychosocial factors would be helpful to understand the interrelationship among PSU, health behaviors, and comorbidities. We also found an association between PSU and physical activity in our sample. More dependent use of smartphone may reduce the time spent on leisure time exercise, and this has significant potential negative health impacts. The coexistence of PSU with current smoking, frequent drinking, and physical inactivity also likely reflects higher health risks – all of which are amenable to effectual health promotion interventions.

This study had several limitations. First, longitudinal associations between PSU and health behavioral and other risk factors cannot be determined due to the cross-sectional design of the Survey. Second, we did not collect detailed information on the pattern of smartphone use among the users, which would have been helpful for interpreting the complex relationship between PSU and SES and other factors. Third, measures on health behaviors were self-reported. Although self-reported smoking and alcohol consumption are widely used and valid, the potential of under-reporting might reduce the strength of associations observed in this study (Del Boca & Darkes, 2003; Gorber, Schofield-Hurwitz, Hardt, Levasseur, & Tremblay, 2009). Future research may also be benefited from more comprehensive measures of these behaviors such as nicotine dependence and alcohol use disorder. Fourth, although we adjusted for several potential confounders, residual or unmeasured confounders (e.g., personality traits) might explain the observed associations between PSU and health behaviors. Finally, the debatable conceptualization of PSU precludes a strong theoretical foundation on which to develop a universally accepted measure (Billieux et al., 2015; Panova &

Carbonell, 2018). This also applies to SAS-SV as a measuring tool for PSU. Nevertheless, the SAS-SV is among the most widely used and translated instruments assessing smartphone dependence, which can help unify approaches to collecting empirical data to understand and address this rapidly emerging public health issue.

CONCLUSIONS

The Chinese adaptation of SAS-SV was found valid and reliable in Chinese adults in Hong Kong. The high prevalence of PSU in the Hong Kong adult population suggests that further research on its health consequence is required. The identified sociodemographic factors and health behaviors associated with PSU are important findings, which should be further investigated to help developing potential prevention programs for PSU.

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