

DISCUSSION PAPER SERIES

IZA DP No. 12723

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A Literature Review**

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ISSN: 2365-9793

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ABSTRACT

Smartphone Use and Academic Performance: A Literature Review

We present the first systematic review of the scientific literature on smartphone use and academic success. We synthesise the theoretical mechanisms, empirical approaches, and empirical findings described in the multidisciplinary literature to date. Our analysis of the literature reveals a predominance of empirical results supporting a negative association between students' frequency of smartphone use and their academic success. However, the strength of this association is heterogeneous by (a) the method of data gathering, (b) the measures of academic performance used in the analysis, and (c) the measures of smartphone use adopted. The main limitation identified in the literature is that the reported associations cannot be given a causal interpretation. Based on the reviewed findings and limitations, directions for further research are discussed.

JEL Classification: I21, I23, L86

Keywords: smartphone use, smartphone addiction, academic performance, literature review, causality

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Introduction

In 2018, approximately 77 percent of America's inhabitants owned a smartphone (Pew Research Center, 2018), defined here as a mobile phone that performs many of the functions of a computer (Alosaimi, Alyahya, Alshahwan, Al Mahyijari & Shaik, 2016). In addition, a survey conducted in 2015 showed that 46 percent of Americans reported that they could not live without their smartphone (Smith, 2015). Similar numbers can be observed in other parts of the (Western) world (OECD, 2017). Therefore, it should come as no surprise that in recent years discussions about the (potential) consequences of (heavy) smartphone use have earned an important place in societal debates (see, e.g. Eliahu, 2014; OECD, 2017). Simultaneously, the possible effects of smartphone use received increasing interest from scientists in different disciplines. As such, scholars have investigated associations between smartphone use and smartphone addiction and, for example, (a) driving performance (Choudhary & Velaga, 2019); (b) sleep quality and quantity (Demirci, Akgönül & Akpınar, 2015); (c) anxiety, loneliness, and depression (Boumosleh & Jaalouk, 2017); (d) satisfaction with life (Samaha & Hawi, 2016); and, (e) social relationships (Chen & Peng, 2008).

In addition to the (potential) impact on people's private life, smartphone use is also expected to interfere with individuals' educational and professional life. In particular, it has been related to tertiary students' academic performance. Smartphone ownership is highest among people aged 18 to 29 (Pew Research Center, 2018), an age group in which students are highly represented. Moreover, there are many theoretical reasons based on which a direct effect of (heavy) smartphone use on academic performance is expected. To the best of our knowledge, 23 studies confront the theoretical expectations with the empirical reality. The present review is the first to compile the existing literature on the impact of general smartphone use (and

addiction) on performance in tertiary education.¹

We believe that a synthesis of this literature is valuable to both academics and policy makers. Firstly, as we focus on divergences in the empirical findings—ergo, aspects in which there is no consensus in the literature—and (methodological) limitations of existing studies, we explicitly provide scholars with directions for fruitful future research. Secondly, while in several countries interventions have been developed to discourage heavy smartphone use in class because it is believed to obstruct knowledge acquisition (e.g. in France, a smartphone ban was introduced into schools in 2017; Samuel, 2017), it is unclear whether these popular perceptions correspond with a consensus in the related scientific literature.

This literature review is structured as follows. In the next section, we discuss the different theoretical mechanisms reported in the literature that could lead to improved or deteriorated educational outcomes due to (heavy) smartphone use. In Section 3, we discuss a systematic overview of the empirical findings, with a focus on how the empirical research results converge concerning the overall negative association between smartphone use and academic performance but diverge according to (a) the method of data gathering, (b) the measures of academic performance used in these studies, and (c) the measures of smartphone use adopted in the research. A final section concludes with directions for future research based on the discussed findings and the limitations of the current literature.

¹ We concentrate on the interference of smartphone use and performance in tertiary education and, thereby, ignore the corresponding literature based on data for secondary education for two different reasons. Firstly, this strategy implies that the reviewed studies are conducted in comparable settings. Typically, differences between the organisation and regulation of secondary education institutions across and within countries exist (see e.g. Beland & Murphy, 2016 and Gao, Yan, Zhao, Pan & Mo (2014)). Therefore, many institutional modalities should have been considered in our discussion when also focussing on secondary education. Secondly, the body of research on the association between smartphone use and educational outcomes in tertiary education is more substantial than the literature on the same association in secondary education so that a literature review focussing on tertiary education is more appropriate than one focussing on secondary education.

Theoretical Mechanisms

Multiple arguments in the existing literature argue for an association—positive or negative—between smartphone use and academic performance. In this section, we review the main theoretical mechanisms.

On the one hand, (particular functions of) smartphones could—when used properly—lead to better educational performance. Smartphones' mobility allows students to access the same (internet-based) services as a computer almost anywhere, almost every time (Lepp, Barkley & Karpinski, 2014). Easy accessibility to these functionalities offers students the chance to search continuously for study-related information. Furthermore, social networking sites and communication applications may contribute to the quick sharing of relevant information. Faster communication between students and between students and faculty staff may contribute to more efficient studying and collaboration (Chen & Ji, 2015; Lepp, Barkley & Karpinski, 2015).

On the other hand, research has suggested that university students think of their smartphones as a source of entertainment, rather than as a working instrument (Lepp, Barkley, Sanders, Rebold & Gates, 2013). These findings support the idea of a time trade-off à la Becker (1965) between smartphone use and study-related activities. That is, the time spent on smartphone use is time lost for study activities. As such, university students' decision to use their smartphones could have a deleterious effect on their academic performance.

Furthermore, smartphone use may interfere with study-related activities. The proximity of the mobile device can be a tempting distraction, leading to multitasking or task-switching. A growing body of literature (see, e.g. Junco, 2012; Junco & Cotten, 2012; Levine, Waite & Bowman, 2012) has shown this behaviour's negative implications with respect to educational performance. We discuss four potential causes for this multitasking or task-switching behaviour. Firstly, visual and auditory notifications on the smartphone may draw students'

attention during class and/or during study time (Junco & Cotten, 2012). Secondly, the desire not to miss out on what is happening online and to continuously interact with the rest of the world (nowadays labelled as 'FOMO', i.e. fear of missing out) may lead to a lack of focus necessary to achieve good study performance (Chen & Yan, 2016; Firat, 2013). Thirdly, but related, smartphone use during study-related activities may be the result of addiction behaviour and cyberslacking, which can be defined as the interference of personal (online) life during working or study activities (Garrett & Danziger, 2008; Vitak, Crouse & LaRose, 2011). Finally, due to the lack of academic motivation, students can experience a sense of boredom for which smartphone applications provide a fast and tempting escape (Hawi & Samaha, 2016).

Besides the direct effect smartphone use might have on both time spent and productivity for academic-related activities, smartphone use can have an additional indirect impact on educational performance by influencing students' health. In recent years, an increasing amount of literature provides evidence for a negative relation between technology use, including smartphone use, and health indicators. As such, negative associations were uncovered between smartphone use and (a) sleep quality (Christensen et al., 2016; Demirci et al., 2015; Rosen, Carrier, Miller, Rokkum & Ruiz, 2016; Tavernier & Willoughby, 2014), (b) mental health (Lepp et al., 2014), and (c) physical fitness (Jackson, von Eye, Fitzgerald, Witt & Zhao, 2011). These health indicators have in turn been associated with educational performance (see, e.g. Baert, Verhaest, Vermeir & Omey, 2015; Galambos, Vargas Lascano, Howard & Maggs, 2013).

Taken together, the aforementioned arguments suggest good reasons to expect an association between university students' smartphone use and their academic performance. Smartphones can be supportive in the academic setting by providing easy and fast communication and to search for relevant information. However, the potential negative influence of smartphones through distraction, multitasking, and/or health implications are

dominantly advanced in the literature to date.

Empirical Findings

In this section, we summarise the literature's empirical findings on the association between general smartphone use and tertiary educational outcomes.² This review is the result of a systematic search. In the first step, all article abstracts indexed in Web of Science including a combination of 'smartphone use' with 'educational outcome' or 'academic performance' were screened for relevance, which provided an initial list of studies for our review. In the second step, we explored (a) the articles included in the studies' references and (b) the articles citing these studies in Web of Science. This second step was re-iterated whenever an additional relevant article was found.

<Table 1 about here>

A schematic overview of the studies can be found in Table 1. The reviewed articles are ordered alphabetically based on the authors' names in column (1). Column (2) contains information on the authors' data: we provide details on how and where the data were gathered and the sample size is reported. Columns (3) and (4) list the variables used in the reviewed studies to capture academic performance and smartphone use, respectively. Finally, column (6) reports the main findings concerning the association between the variables in columns (3) and (4), based on the empirical approach listed in column (5).

² Studies that examine the association between the use of specific smartphone applications and study performance, such as Jacobsen and Forste (2011) and Bun Lee (2014), were excluded. This choice was made to realise a homogeneous literature to review. Moreover, studies investigating the association between the use of particular applications and academic performance ignore the fact that the use of an electronic device for this particular application is strongly correlated with its use for other activities (Chen & Yan, 2016), potentially resulting in an omitted variable bias.

Main findings

A quick look at Table 1 reveals the predominance of studies reporting a significantly negative association between smartphone use and academic performance in tertiary education. More specifically, 18 of the 23 included studies (i.e. 78.3 percent), relying on diverging empirical approaches elaborated upon below, conclude such a negative association. The remaining five articles find no statistically significant association between overall smartphone use and performance at university. However, no single study to date reports a positive overall association.³ This first look at the literature provides an indication⁴ that the negative mechanisms, as discussed in the previous section, contribute more to the association between smartphone use and academic outcomes than the positive mechanisms.

In column (6) of Table 1, we report significant Pearson correlation coefficients with respect to the studies' measures of these outcomes—they could be gathered for 18 of the 23 studies⁵. These correlation coefficients do not take into account any confounding variables and therefore are not the analysis end point in most reviewed articles. However, they allow us to compare a raw indicator of the magnitude of the association between smartphone use and academic performance across studies. The reported coefficients range from a correlation coefficient of -0.380 between total phone use in class and self-reported grade point average (GPA) in Kim et al. (2019) to a correlation coefficient of 0.047 between the problematic mobile phone use scale of Sert, Yilmaz, Kumsar, and Aygin (2019) and self-reported GPA. When considering only correlation coefficients of studies that concluded a statistically significantly

³ Chen and Ji (2015) report a positive association in the particular situation when personal electronic devices are (mainly) used for educational purposes.

⁴ The reported associations cannot be given a causal interpretation, though. We return to this point below.

⁵ The Pearson correlation coefficient was reported in 15 of the articles. For the other articles, we attempted to obtain this coefficient from the authors via email, or, when our email was not answered, via telephone (using both personal and institution numbers).

negative association, not surprisingly the interval is less wide. Then it moves from a very small correlation of -0.079 (Lin & Chiang, 2017) to a (rather) small correlation of -0.380 (Kim et al., 2019) In sum, the reviewed literature seems to suggest a negative association between overall smartphone use and academic performance that is small in magnitude (at most). However, this general picture may conceal interesting (further) convergences and divergences in the literature that can be observed only after investigating the studies more carefully, which is addressed in the next two subsections.

Convergences in the empirical literature

The statistical and economic magnitude of the association between smartphone use and academic performance seems to be rather homogeneous by (a) the public versus private nature of tertiary education institutions, (b) the method of data analysis as well as by (c) the region and sample size of the realised data.

Firstly, one could expect that the students in samples from private universities are positively selective with respect to non-observed characteristics like internal motivation and ability (Di Pietro & Cutillo, 2006; James, 1994; Ladd, 2002). For these students, several of the theoretical mechanisms supporting a negative association between smartphone use and academic performance might be less relevant. In 15 cases, the research population was matriculated at a publicly funded university at the time of the research. Twelve of them report a significantly negative association. In seven studies, the research was conducted at a private university or a business school. Five of these studies report a significantly negative association. Winskel, Kim, Kardash, and Belic (2019) studied students from both private and publicly funded institutions but only find a negative significant association for the subsample of students matriculated at a publicly funded institution. Thus, the empirical evidence to date suggest a negative association between smartphone use and academic performance, regardless of the public or private nature of the education institution.

The association between smartphone use and academic performance has been investigated on all continents. However, most studies were conducted in North America and Asia. No remarkable differences were noticed between the studies relying on data from these two continents. Among the North American studies, Bun Lee (2015) and Wentworth and Middleton (2014) found no significant association while a negative association was concluded in the five other articles. Similarly, two of the Asian-based investigations (Rashid & Asghar, 2016; Sert et al., 2019) did not report a significant association while a negative association was reported by the nine other studies. Furthermore, Winskel et al. (2019) found no significant association for their Korean subsample. For their Australian (Oceania) they reported a significantly negative association. Besides, Felisoni and Godoi (2018) and Jankovic, Nolic, Vukonjanski and Terek (2016) reported a significantly negative association in Brazil (South America) and Serbia (Europe), respectively. Finally, among the studies relying on African data, Olufadi (2015) found no significant association in Nigeria while Asante and Hiadzi (2018) reported a negative association in Ghana.

Except for Jankovic et al. (2016), Nayak (2018), and Winskel et al. (2019), all studies were based on surveys with students from one single educational institution. However, the sample sizes range from 43 to 2367 participants. Therefore, it could be postulated that the reported non-significant associations in Table 1 are due to a lack of statistical power in the smaller samples. Of the eleven articles with a sample size lower than the median ($N = 483$), only Olufadi (2015) and Bun Lee (2015) did not report a significant association. Likewise, two studies (Rashid & Asghar, 2016; Sert et al., 2019) with a sample size higher than the median sample size found no significant association. Finally, Wentworth and Middleton (2014), whose sample size ($N = 483$) coincides with the median sample size, reported no significant association between smartphone use and academic performance. In sum, the insignificant associations reported in the literature do not appear to be driven by (studies with) smaller

sample sizes.

Finally, the research results are rather homogeneous according to the level of control for confounding factors in the main empirical approach. Importantly, none of the empirical results summarised in Table 1 can be given a causal interpretation; that is, to date there is only evidence for heavy smartphone users performing worse at higher education institutions, with no support for heavy smartphone use causing this worse performance. None of the approaches mentioned in column (5) of Table 1 can fully control for the endogeneity of smartphone use and academic achievement because they all rely on observational, cross-sectional data. Thus, they can only control for a limited set of confounders. However, factors such as motivation, intellectual capabilities and perceived academic pressure are typically not included in their collected survey data but might influence both smartphone use and academic performance. Thereby, it is unclear whether the worse performance of heavy smartphone users reported by many of these studies is really the result of smartphone use or reflects variation in unobserved personal characteristics.

Nevertheless, the studies included in Table 1 differ in the extent to which they attempt to control for confounding variables. In this respect, the negative association between smartphone use and academic performance is not more or less outspoken when such controls are taken into account in a regression analysis. Among the nine reviewed articles that only calculate correlation coefficients, only Olufadi (2015) and Sert et al. (2019) did not find a significant association. The remaining 14 articles perform at least one form of regression analysis. Nine of them perform a linear regression analysis, four conduct a path analysis and one opts for a logistic regression as main approach. Eleven times the regression coefficient with respect to the association between overall smartphone use and academic performance is significantly negative.

Divergences in the empirical literature

By contrast, the association between smartphone use and academic performance seems to be

heterogeneous by (a) the method of data gathering, (b) the measures of academic performance used in the analysis, and (c) the measures of smartphone use adopted in the research.

Firstly, all studies in Table 1 are (mainly) based on survey data: Seven rely on an online survey and another 12 rely on a paper and pencil survey. Winskel et al. (2019) rely partly on an online survey for the Korean subsample and partly on a paper and pencil survey for their Australian subsample. In addition, Felisoni and Godoi (2018), Rosen et al. (2018), and Kim et al. (2019) combined a paper and pencil survey with objective logged data, which is a strong point for these studies as Boase and Ling (2013) reported a limited correlation between self-reported smartphone use and actual logged data—we return to this point below. Four of the five studies reporting no significant association between smartphone use and academic performance are relying on data gathered by a paper and pencil questionnaire. A possible explanation might be that those questionnaires are filled in during a course and therefore participants are surrounded by peers which might increase the tendency to social desirable answers (Krumpal, 2013).

Secondly, across the reviewed studies, three measures of academic performance are used. Six studies use data on students' actual grades received from the lecturer or from the faculty or university administration. All of these studies conclude a significantly negative association between smartphone use and these outcomes. In addition, 11 studies rely on self-reported grades. Remarkably, all five articles not reporting a negative association fall within these 11 studies. Finally, The six studies using self-reported academic performance scales all found a negative association. This contrast may indicate that errors of measurement occurred in the self-reported grade variables. Indeed, these variables may be biased due to recall issues or socially desirable answering (Krumpal, 2013).⁶

⁶ In this respect, survey questions directly assessing GPA may be more sensitive than items from the broader achievement, adjustment, or engagement scales listed in Table 1.

Thirdly, we distinguish a similar difference in research results depending on how smartphone use is measured. In 12 articles, the researchers investigate the association between academic performance and total smartphone use. Except for Felisoni and Godoi (2018), Rosen et al. (2018), and Winskel et al. (2019), who used objectively tracked information, these frequencies are self-reported. Of these twelve studies, only Wentworth and Middleton (2014) did not report a significant association while Winskel et al. (2019) find a significant negative association for their Australian subsample and no significant association for the Korean subsample. In contrast, of the other 11 studies, using a scale instrument to measure smartphone attitude and addiction, only seven find a significantly negative association. In line with our explanation for the former two divergences in the literature, it might be the case that students tend more to social desirable answers (Krumpal, 2013) when inquired about attitude or addiction compared to frequency of phone use.

Conclusion

In this article, we reviewed the scientific literature to date on the relationship between smartphone use and academic performance in tertiary education. Our analysis of the literature reveals a predominance of empirical results supporting a negative association. However, this predominance is less outspoken in studies analysing data gathered by paper and pen questionnaires (compared with studies on data gathered by online surveys) and studies relying on self-reported grade point averages (compared with studies using actual grades). In general, when scholars use methods of data gathering which are more susceptible to social desirable behaviour, a non-significant association is found more often.

The main limitation identified in the literature is that the existing studies all conduct correlational analyses and/or linear or logistic regression analyses on cross-sectional data so that their results cannot be given a causal interpretation. We suggest two different forms of

analysis to counter this endogeneity problem that could be explored in further research. Firstly, longitudinal data could be collected in view of regression analyses controlling for individual fixed effects. By integrating fixed effects into the analysis, it is possible to control for time-invariant unobserved characteristics of university students that may affect both smartphone use and academic performance. Secondly, instrumental variables correlated with smartphone use but not (directly) affecting educational attainment (such as perceived quality of the WiFi network in the classrooms) may be surveyed. These can be used to capture an exogenous prediction of smartphone use. Thereafter, the causal impact on academic performance of this exogenous prediction can be estimated.

A second limitation of the literature is related to the analysed data. As pointed out in the empirical findings section, 20 out of 23 reviewed articles used self-reported measures of smartphone use in their analysis. Felisoni and Godoi (2018), Rosen et al. (2018), and Kim et al. (2019) made a first attempt to introduce actual tracked use in their analysis but they had a rather small sample size ($N = 43, 216, \text{ and } 84$ respectively). However, Boase and Ling (2013) provided evidence for only a limited correlation between actual smartphone use and that measured by self-reported instruments. So, it is recommended for future research to further investigate whether the results based on self-reported measures can be confirmed when analysing actual data.

A third shortcoming in the scientific literature so far is the lack of research investigating the empirical validity of the reviewed theoretical mechanisms for a potential impact of smartphone use on academic performance. However, uncovering the mechanisms at work is of great importance for policy making. To implement adequate policy measures on smartphone use in academic settings, we need to know what precisely causes their (potential) relationship. Tracking university students' smartphone use, as advocated above, may also help in this respect. Actual tracked data would provide more insight into the timing of students' smartphone use.

This could reveal indications for the multitasking mechanism and/or the time trade-off argument behind the association between smartphone use and academic performance.

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Table 1

SUMMARY OF THE LITERATURE

(1) Study	(2) Data	(3) Outcome variable	(4) Explanatory variable	(5) Main empirical approach	(6) Association reported between (3) and (4)
Alosaimi, Alyahya, Alshahwan, Al Mahyijari and Shaik (2016)	Online survey, public university, Saudi-Arabia (N = 2367).	Academic achievement subscale. ^a	Total phone use (hours/day).	Linear regression analysis.	Negative association. ^m
Asante and Hiadzi (2018)	Paper and pen survey, public university, Ghana (N = 150).	Self-reported GPA.	Total phone use in class (minutes/class).	Linear regression analysis.	Negative association. ⁿ
Bun Lee (2015)	Paper and pen survey, public university, United States (N = 304).	Self-reported GPA.	Smartphone addiction scale. [§]	Linear regression analysis.	No significant association. ⁿ
Chen and Ji (2015)	Paper and pen survey, private university, Taiwan (N = 506).	Actual GPA.	Total personal electronic device use (minutes/day).	Linear regression analysis.	Negative association (r = -0.09). ^{o,p}
Felisoni and Godoi (2018)	Paper and pen survey combined with logging data, business school, Brazil (N = 43).	G-MNPS score. ^b	Total phone use (minutes/day).	Linear regression analysis.	Negative association (r = -0.301).
Ibrahim et al. (2018)	Paper and pen survey, public university, Saudi-Arabia (N = 610).	Self-reported GPA.	Problematic mobile phone use questionnaire. ^h	Correlational analysis.	Negative association. ⁿ
Hawi and Samaha (2016)	Online survey, private university, Lebanon (N = 293).	Actual GPA.	Smartphone addiction scale. [§]	Logistic regression analysis.	Negative association (r = -0.2). ^o
Jankovic, Nolic, Vukonjanski and Terek (2016)	Paper and pen survey, two public universities, Serbia (N = 485).	College adjustment scale. ^c	Total phone use (hours/day).	Linear regression analysis.	Negative association (r = -0.111).
Kim et al. (2019)	Paper and pen survey combined with logging data, public university, South Korea (N = 84).	Self-reported GPA.	Total phone use in class (minutes/day).	Linear regression analysis.	Negative association (r = -0.380). ^p
Lepp, Barkley and Karpinski (2014)	Paper and pen survey, public university, United States (N = 536).	Actual GPA.	Total phone use (minutes/day).	Path analysis.	Negative association (r = -0.203).
Lepp, Barkley and Karpinski (2015)	Paper and pen survey, public university, United States (N = 536).	Actual GPA.	Total phone use (minutes/day).	Linear regression analysis.	Negative association (r = -0.234).
Li, Lepp and Barkley (2015)	Online survey, public university, United States (N = 516).	Self-reported GPA.	Total phone use in class (minutes/day).	Path analysis.	Negative association (r = -0.173).

Lin and Chiang (2017)	Online survey, private university, Singapore (N = 438).	Self-reported GPA.	Smartphone dependency symptom scale. ⁱ	Path analysis.	Negative association (r = -0.079). ^p
Nayak (2018)	Paper and pen survey, public university, India (N = 429).	Academic performance scale. ^d	Total phone use (minutes/day).	Correlational analysis.	Negative association (r = -0.276).
Olufadi (2015)	Paper and pen survey, two public universities, Nigeria (N = 286).	Self-reported GPA.	Mobile phone use behaviours scale. ^j	Linear regression analysis.	No significant association (r = -0.060).
Rashid and Asghar (2016)	Online survey, private university, Saudi-Arabia (N = 761).	Self-reported GPA.	Media and technology use and attitude scale. ^k	Path analysis.	No significant association (r = -0.01). ^o
Rosen et al. (2018)	Paper and pen survey combined with logging data, public university, United States (N = 216).	Social science course grade. ^e	Total phone use (minutes/day).	Correlational analysis.	Negative association (r = -0.13). ^o
Samaha and Hawi (2016)	Online survey, private university, Lebanon (N = 293).	Actual GPA.	Smartphone addiction scale. ^g	Linear regression analysis.	Negative association (r = -0.143).
Sert, Yilmaz, Kumsar, and Aygin (2019)	Paper and pen survey, public university, Turkey (N = 743).	Self-reported GPA.	Problematic mobile phone use scale. ^h	Correlational analysis.	No significant association (r = 0.047).
Uzun and Kilis (2019)	Paper and pen survey, public university, Turkey (N = 631).	Self-reported GPA.	Media and technology use and attitude scale. ^k	Correlational analysis.	Negative association (r = -0.107).
Wentworth and Middleton (2014)	Paper and pen survey, private university, United States (N = 483).	Self-reported GPA.	Total phone use (minutes/week).	Linear regression analysis.	No significant association (r = 0.01). ^o
Winkel, Kim, Kardash and Belic (2019)	Online survey, one private and one public university, South Korea and Australia (N = 389).	Self-reported GPA.	Total phone use (minutes/day).	Correlational analysis.	No significant association (South Korea, r = 0.04) and negative association (Australia, r = -0.30). ^o
Witecki and Nonnecke (2015)	Online survey, public university, Canada (N = 972).	Student course engagement scale. ^f	Information technology habits during lectures scale. ^l	Correlational analysis.	Negative association. ^q

The following abbreviation is used: GPA (grade point average).

^a The academic achievement subscale is part of the authors' problematic use of mobile phones scale. This subscale relates to the item that assesses whether the respondent believes her/his smartphone use adversely affects her/his academic achievement, rated on a 5-point Likert scale. To allow the reader to compare column (6) row by row, we refer to an implicitly reverse coded version of the measured subscale in the article.

^b The G-MNPS score is the student's weighted average score calculated by the business school, based on the student's rank for every course included in the curriculum.

^c The college adjustment scale comprises three items that assess self-related academic motivation, academic performance, and social adjustment to college, rated on a 5-point Likert scale.

^d The academic performance scale comprises three items that assess self-related academic participation, academic performance, and time management, rated on a 5-point Likert scale.

^e The final course grade is the final grade obtained for the specific course in which the survey was held, as provided by the lecturer of this course.

^f The student course engagement scale comprises 23 items that assess four factors found to relate to student course engagement (skills engagement, emotional engagement, participation and interaction engagement, and performance engagement), rated on a 5-point Likert scale.

^g The smartphone addiction scale comprises 10 items that assess smartphone use primarily to identify the level of smartphone addiction risk, rated on a 6-point Likert scale.

- ^h The problematic mobile phone use scale comprises 27 items that assess smartphone use primarily to identify the level of problematic mobile phone use, rated on a 5-point Likert scale.
- ⁱ The smartphone dependency symptom scale comprises 16 items that assess smartphone use primarily to identify the level of smartphone dependency, rated on a 5-point Likert scale.
- ^j The mobile phone use behaviour scale comprises 13 items that assess the smartphone behaviour in different circumstances, rated on a 3-point Likert scale.
- ^k The media and technology usage scale comprises 44 items clustered in 11 subscales on smartphone use for different activities, rated on a 9- or 10-point frequency scale.
- ^l The information technology habits during lectures scale comprises 9 items that assess the frequency and the purpose of technology use during lectures, rated on a 5-point Likert scale and by open answers.
- ^m This negative association represents the reported *positive* association between total phone use and the full problematic use of mobile phones scale ($r = 0.311$). The academic achievement subscale (see above) is highly correlated to the other subscales of the full problematic use of mobile phones scale (correlation between academic achievement subscale and full problematic use of mobile phones scale: $r = 0.474$).
- ⁿ The author(s) could not be reached via both email or telephone communication to obtain the Pearson correlation coefficients.
- ^o The article and/or electronic communication provided Pearson correlation coefficients with fewer than three decimals.
- ^p The reported Pearson correlation coefficient was obtained via personal communication with the author(s).
- ^q The authors indicated that based on the available data they were not able to provide a Pearson correlation coefficient.