

# Mister Sandman, Bring Me Good Marks!

## On the Relationship Between Sleep Quality and Academic Achievement

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**Abstract.** There is growing evidence that health factors affect tertiary education success in a causal way. This study assesses the effect of sleep quality on academic achievement at university. To this end, we surveyed

804 students about their sleep quality by means of the Pittsburgh Sleep Quality Index (PSQI) before the start of their first exam period in December 2013 at Ghent University. PSQI scores were merged with course marks in this exam period. Instrumenting PSQI scores by sleep quality during secondary education, we find that increasing total sleep quality with one standard deviation leads to 4.85 percentage point higher course marks. Based on this finding, we suggest that higher education providers might be incentivised to invest part of their resources for social facilities in professional support for students with sleep and other health problems.

**Keywords.** Belgium; economics of education; economics of health; economics of sleep; academic achievement; sleep quality.

# **1 Introduction**

For decades, economists have been studying the determinants of academic attainment. Seminal studies such as Black et al. (2005), Vardardottir (2013) and Leos-Urbel et al. (2013) have identified prior accumulated human capital, the costs and returns of higher education, social background characteristics and gender as key determinants in explaining outcomes in higher education.

Recently, academics have also focused on the role of health factors on academic achievements. For instance Ding et al. (2009), García-Gómez et al. (2013), Fletcher (2014), Sabia (2007), Balsa et al. (2011) and Pieterse (Forthcoming) identify a negative relationship between poor general health, health shocks, ADHD, body weight, alcohol usage and maltreatment respectively on the one hand and academic performance on the other hand. In addition, Bharadwaj et al. (2013) reveal a positive relationship between early health interventions and academic achievement.

A neglected factor in the economic literature on academic achievements is sleep quality. From a theoretical point of view, a positive relationship between sleep quality and academic performance can be expected. Based on research within medicine and biology, we know that night's rest is essential to helping maintain mood, attention, motivation,

memory and cognitive performance. While asleep, the brain integrates new knowledge and forms new associations (see, e.g., AlDabal & BaHammam, 2011; Alvaro, 2014; Beebe, 2011; Buckhalt et al., 2007; Gais & Born, 2004; Louca & Short, 2014; Meijer et al., 2000; Siegel, 2001; Vandekerckhove & Cluydts, 2010; Walker & Stickgold, 2004). From an empirical point of view, former contributions indeed found a positive relation between sleep quality and/or sleep duration and academic performance. Many studies report a positive association between opportune sleep habits and beneficial primary and secondary schooling outcomes such as reading, math and spatial ability test scores and school grades (see, e.g., Bruni et al., 2006; Dewald et al., 2010; Kelly et al., 2013; Meijer, 2008; Perkinson-Gloor, 2013; Short et al., 2013; Stea et al., 2014). In addition, randomised controlled trials have shown that interventions such as teaching behavioural sleep strategies and imposing minimum restrictions on sleep duration result in a positive effect on academic achievement in primary and secondary school (Beebe et al., 2010; Fallone et al., 2005; Quach et al., 2011; Quach et al., 2013; Randazzo et al., 1998; Sadeh et al., 2003). Last, some studies also report positive correlations between sleep quality or sleep duration on the one hand and grade point averages and exam passing probabilities on the other hand in tertiary education in China, Ethiopia, Germany and Portugal (Ahrberg et al., 2012;

Genzel et al., 2013; Gomes, 2011; Lemma et al., 2014; Wong et al., 2012). In contrast, Eliasson et al. (2010) and Trockel et al. (2000) find no significant results in this respect in the United States (Washington DC and Utah). For a more in-depth review of former studies on the relationship between sleep habits and academic performance, we refer to Curcio et al. (2006), Shochat et al. (2014) and Taras & Potts-Datema (2005).

In this study, we empirically test the relationship between sleep quality, sleep duration and academic achievement. To this end, we survey first-year university students on their sleep habits, by means of the Pittsburgh Sleep Quality Index (Buysse et al., 1989), before the start of their first exam period at university. In addition, these students are surveyed on general social background and health characteristics. The resulting dataset is merged with their academic achievement in terms of course marks in their first exam period. Our research question is then answered by exploring 2SLS estimations on the gathered data. To be able to correctly identify the influence of sleep quality on academic achievement, the respondents' sleep quality is instrumented by their sleep quality during secondary education.

Our contribution to the aforementioned academic literature is threefold. First, we contribute to the recent string of articles – a series characterised, as mentioned before, by mixed findings – on the effect of

sleep behaviour on academic performance among university students. Second, we are innovative in studying the effect of both sleep quality and sleep duration on academic performance at university within one empirical framework. Third, as reviewed by Curcio et al. (2006), most non-experimental studies on sleep quality and academic performance present correlation and simple (linear or mediation) regression analyses. Thereby, as we will argue in Section 3, their results cannot be given a causal interpretation. In the present study we explicitly take into account the endogeneity of sleep quality with respect to academic performance by means of the mentioned data gathering and statistical analysis method.

## **2 Data**

### **2.1 Data Gathering**

Our dataset was constructed by merging survey data on first-year university students' sleep quality and further individual characteristics with their first university exam marks.

In December 2013, we conducted a survey on the students present at the start of the last lecture of the first-semester courses of Economics and Introduction to Accountancy at Ghent University in Belgium. These courses

are part of the first year Bachelor programs of (Business) Economics – the first-year study program is the same for the Bachelor of Science in Economics and the Bachelor of Science in Business Economics at Ghent University – and Commercial Sciences respectively, but are also taken by some students of other programs. The total number of students enrolled in the course of Economics (Introduction to Accountancy) in 2013 was 685 (638). 394 (410) of them attended the last lecture of the course. The fact that the attendance rate was only about 58% (64%) should not pose a problem of external validity as it is unlikely that the relationship between sleep quality and academic achievement would be different between attenders and non-attenders. These students were asked to fill out a paper-and-pencil questionnaire.

This questionnaire comprised five sections. A first section was dedicated to general questions about the subject's social-economic background characteristics (age, gender, parental education, household composition, nationality, language at parental home, living place, prior educational attainment, relationship status and general health). These characteristics are used in our analysis to explain both sleep quality (see, e.g. Hale et al., 2013, for recent evidence on the relationship between social background and sleep quality) and academic achievement.

In a second section, we surveyed quality of sleep by means of the

validated Pittsburgh Sleep Quality Index (Buysse et al., 1989). This question module measures sleep quality (in broad sense) during the previous month. The PSQI contains 19 self-rated questions yielding seven PSQI submeasures: sleep duration, sleep disturbance, sleep latency, day dysfunction due to sleepiness, sleep efficiency, overall perceived sleep quality and need of medicines to sleep. Each component is scored from 0 to 3. For instance, the PSQI submeasure of sleep duration, the most important submeasure in the context of the present study, is based on question 4 of the Pittsburgh Sleep Quality Index: “During the past month, how many hours of actual sleep did you get at night? (This may be different than the number of hours you spent in bed.)” The answer on this question is then scored as follows: PSQI submeasure of sleep duration score 0 for a sleep duration of at least seven hours; score 1 for a sleep duration of at least six (but lower than seven) hours, score 2 for a sleep duration of at least five (but lower than six) hours and score 3 for a sleep duration lower than five hours.

Summing up all submeasures yields a total PSQI score between 0 and 21, with higher scores indicating lower sleep quality. A total PSQI score greater than 5 is classified as poor quality sleep. The official Dutch translation of the PSQI was requested from Dr. Buysse and a user agreement was signed. The reader will notice that although the ‘Quality’ in



PSQI refers to a qualitative measure of sleep, some components have a rather quantitative point of view. We will come back to this issue in Section 4 when we focus on the particularly explanatory power of the two PSQI submeasures of sleep duration and overall perceived sleep quality.

In a third section of our survey we additionally let the students rate the sleep quality of their parents (“How would you rate your mother’s/father’s sleep quality?”) and their sleep quality during secondary education (“How would you rate your overall sleep quality during secondary education?”) on a five-point Likert scale (going from “very good” to “very bad”). In addition, we asked whether or not they suffered from congenital medical problems affecting their sleep quality. These variables are important for our econometric analysis being predictors for sleep quality that cannot be determined by university outcomes.

In a fourth section, the students had to fulfil the validated Depression Anxiety Stress Scales (DASS-21) questionnaire. The Dutch translation of the DASS-21 questionnaire was downloaded from [www.psy.unsw.edu.au/dass/](http://www.psy.unsw.edu.au/dass/). DASS-21 is a set of three self-report scales designed to measure the negative emotional states of depression, anxiety and stress. Each of the three DASS-21 scales contains 7 items that are to be rated on a four-point Likert scale leading to a score between 0 and 21. The higher the score on these scales, the higher the emotional problems.

Depression, anxiety and stress are regularly reported to be associated with both sleep quality (see, e.g., Breslau et al., 1996; Eller et al., 2006) and academic achievements (see, e.g., Andrews & Wilding, 2004; Eisenberg et al., 2009). Hence the importance of including them as control variables in our analysis.

In a last section, the students were asked whether they agreed with the fact that their survey answers would be merged with their first-semester exam marks by a third party. This clause was prepared in collaboration with the Chairman of the Board of Examiners of the Faculty of Economics and Business of Ghent University. In total, 382 (352) of the students in the course of Economics (Introduction to Accountancy) gave us the permission to use their survey answers together with their exam marks for our research aims. From this population we retained, for reasons of methodology (see Section 3.1) and homogeneity, all full-time first year students in the Bachelor programs of (Business) Economics and Commercial Sciences (329 and 307 students respectively).

In February 2013, the survey data were merged with the marks of the students for their first semester courses, based on the student number the participants of the survey mentioned in the last section of the questionnaire. For reasons of privacy, this was done by a third party. Due to the fact that some of the students did not bring their student card with

them and did not know their 8-digit student number by heart, the third party could only merge survey data and exam marks for 328 (293) full-time first year students in the Bachelor programs of (Business) Economics (Commercial Sciences). As a result our merged dataset contains information on 47.88% (45.92%) of the students enrolled in the course of Economics (Introduction to Accountancy). All students in (Business) Economics took the courses of Accounting, Economics, Human Sciences, Law, Mathematics and Production Technology. For the students in Commercial Sciences, marks were registered for the courses of Accountancy, Commercial and Financial Transactions, English, French, Information Technology, Law, Mathematics and Microeconomics. As English, French and Information Technology could also be taken in the second semester, not all students in our data got marks for these courses.

The data gathering process was reviewed and approved by the Ethical Committee of the Faculty of Economics and Business Administration of Ghent University.

## **2.2 Data Description**

In Table 1 we report descriptive statistics for the variables used in the econometric analysis. We separately report statistics on the total sample, on the sample of good sleepers ( $PSQI \leq 5$ ) and on the sample of bad

sleepers (PSQI > 5). In total 69.57% of our subjects are good sleepers and 30.43% are bad sleepers. The latter percentage is very comparable with the 29% individuals with a PSQI score higher than 5 in the sample of German medical students surveyed by Ahrberg et al. (2012).

#### **TABLE 1 ABOUT HERE**

Panel A provides the reader with statistics on the background characteristics of our subjects. The subsample of bad sleepers comprises, in line with Venn et al. (2013) and Jackson et al. (2014), more females and individuals with a migration background. In addition, children of parents who passed away or divorced and single individuals are overrepresented among the group of bad sleepers. There are also more individuals in this subsample with bad general and mental health characteristics. Therefore, as mentioned before, controlling for these characteristics when identifying the impact of sleep quality on study results is important. Panel B shows statistics for our potential instruments for sleep quality. These statistics provide already an indication for the relative strength of these potential instruments. The sleep quality during secondary education index seems to be the strongest predictor of poor quality of sleep. This index is 25.35% higher among bad sleepers than among good sleepers. Panel C shows that

the average PSQI score among the population is 4.80, which is quite close to the threshold for poor sleep quality. Furthermore, not surprisingly, the average scores for the PSQI submeasures of overall perceived sleep quality and sleep duration are substantially higher among the bad sleepers than among the good sleepers. For the course characteristics, presented in Panel D, we observe few differences between both subsamples by sleep quality.

Panel E of Table 1 presents the outcome variables at the individual exam level. We construct four outcomes concerning academic achievement based on the individual exam marks at the end of the first semester. The first outcome variable (“exam mark: completed exams”) is equal to the exam mark (out of 20 points) for all exams made, leaving out observations for which the students were not present at the exam and ipso facto did not pass the exam. The second outcome variable (“exam mark: potential exams”) is equal to the first one except that the exam mark when students did not show up for the exam was recoded to 0, thus increasing the number of observations slightly. We introduce this alternative outcome variable as bad sleepers might be more likely to skip exams, due to reasons of motivation and mood, than good sleepers. However, as in total only 0.74% of the potential exams were not taken, the difference between the first and the second outcome variable is very limited. The third and fourth

variable are dummies indicating whether the student passed the exam or not, i.e. whether the mark for the particular individual in the course is at least 10 out of 20 points. The third variable (“exam passed (mark  $\geq$  10): completed exams”) and fourth variable (“exam passed (mark  $\geq$  10): potential exams”) again differ in whether exams for which the student did not show up were left out or were given a 0 mark.

In line with our expectations, both the exam marks and the exam passing chances are somewhat lower among the bad sleepers. A simple t-test shows that this difference is significant at the 1% significance level for all academic outcomes. However, this comparison does not take selection, neither on the aforementioned observable characteristics nor on unobservable characteristics (that may correlate with both academic outcomes and sleep quality) into account. The instrumental variable regression we apply in this research takes the selection on observable characteristics into account and deals with potential problems of endogeneity. Therefore, the analyses outlined in the next section lead to a more founded answer to our research question.

## **3 Methods**

### **3.1 Sleep Quality as an Endogenous Explanatory Variable**

Sleep quality is potentially endogenous to academic achievement. Factors and events that are unobservable to the researcher may influence both academic achievement and contemporaneous sleep quality. Moreover, bad academic achievement or indications of bad academic achievement in the near future may induce sleepless nights. We aim at controlling for this problem in three ways.

First, in our analyses we control for the large set of individual background characteristics outlined in Panel A of Table 1. This set of variables includes measures for health in general and psychological health, adopting the DASS-21 scales, in particular. Thereby we aim at minimising the number of factors influencing both sleep quality and academic achievement that are omitted from the analysis and estimate the effect of sleep quality within homogeneous subgroups of individuals.

Second, and as described in Section 2.1, we measured sleep quality by means of the PSQI at the end of the first semester lectures at university, i.e. just before the start of Christmas holidays, which are used by the students to prepare the first semester exams, taking place immediately

after the Christmas holidays. This is a conscious choice as thereby sleep quality is estimated before exam stress takes place. Hence, reversed causality is not an issue in our study.

However, even if we control for a large set of individual background characteristics and use PSQI scores estimated before the exam period, sources of endogeneity might still be present. For instance, students may experience learning difficulties that were not anticipated at the start of the semester. Or, given the high failure rate during the first year, they may face strong pressure to succeed. Another potential source of endogeneity is random measurement error, because sleep quality during the month of observation may deviate from the average sleep quality during the semester. Although the medical literature indicates that sleep habits, especially sleep duration, are persistent, even at young ages and even across a 10-year period (Bruni et al., 2014; Kataria et al., 1987; Klackenberg, 1982; Roberts et al., 2008; Thorleifsdottir et al., 2002), unexpected circumstances may lead to an outlier month for at least some individuals. To deal with these endogeneity problems, we assess the impact of sleep quality on academic achievement with an instrumental variable econometric approach.

Potential instruments are, as presented in Panel B of Table 1, self-reported maternal sleep quality, paternal sleep quality, sleep quality during



secondary education and a dummy indicating (congenital) medical problems that affect sleep quality. However, most of these variables appear to be weak instruments. Table B.1 (in the Online Appendix [INSERT LINK TO ONLINE APPENDIX]) shows that the correlation rates between the PSQI measures (the submeasure for sleep duration in particular) on the one hand and all potential arguments except sleep quality during secondary education on the other hand are rather low. We also constructed various variables combining the reported maternal and paternal sleep quality but this hardly affected the low magnitude of the correlation rates. Therefore, in what follows, we will only use the sleep quality during secondary education index as an instrument for all PSQI scores. As the end of secondary education lies only a few months in the past, recall bias should not be substantial.

A key identifying assumption for the 2SLS estimator is that the chosen instrument may not be correlated with other determinants of the outcome variable, i.e. the exam marks, that are not controlled for. However, it is clear that sleep quality during secondary education may be (positively) correlated with unobserved factors that do (positively) affect university performance such as intrinsic motivation to succeed and to be healthy. Therefore, without additionally controlling for individual background characteristics, these unobserved factors could introduce an upward bias in

the 2SLS estimates of how sleep quality affects exam performance. In addition, there may also be sources of downward bias as, e.g., having an anxious or obsessive personality may result in worse sleep but better academic performance. However, we believe these sources of bias should be captured by the aforementioned additional controls outlined in Panel A of Table 1. On the one hand, we directly control for general health and the DASS-21 scales. On the other hand, other potential unobservables that affect both our instrument and academic outcomes should in the first place translate in higher or lower general end marks in secondary education, for which we control.

### **3.2 Econometric Model**

In order to answer our main research question, we regress variables capturing individual academic achievement on PSQI measures at the individual level, a set of individual-specific control variables and course dummies. The variables we include in the different regressions are the ones outlined in Panel E, Panel C, Panel A and Panel D of Table 1, respectively.

The coefficients of interest can be estimated by means of ordinary least squares (OLS). For the academic achievement outcome “exam passed”, this boils down to the choice of estimating a linear probability model instead of

a discrete choice model. As we cluster standard errors at the subject level, this linear probability model is robust to heteroskedasticity which is important given the binary nature of the outcome variable. In addition, we looked into the analogous results replacing the linear probability model with a probit model. The estimated marginal effects for the probit model were very similar to the OLS results.

However, for reasons outlined in Section 3.1, our main analyses adopts two-stage least squares (2SLS) techniques. As mentioned before, we use sleep quality during secondary education as an instrument for the PSQI-variable. In all models, standard errors are clustered at the subject level.

## **4 Results**

In this subsection, we discuss our empirical analyses. Table B.2, Table B.3, Table B.4 and Table B.5 in the Online Appendix [INSERT LINK TO ONLINE APPENDIX] present the results for basic regressions with as a dependent variable the mark on completed exams, the marks on potential exams, the indicator variable for passing completed exams and the indicator variable for passing potential exams, respectively. Each table comprises the estimation results for eight regression models labelled from (1) to (8).

Table 2 summarises the most important results of Table B.2. In model (1) and model (2) the main explanatory variable is the standardised total PSQI score. In model (1) we get by means of OLS regression a coefficient for this main explanatory variable which is not significantly different from zero. However, this estimate might be biased due to the endogeneity problem mentioned in Section 3.1. Wooldridge's (1995) robust endogeneity test (which is used given that we cluster standard errors at the subject level), presented in column (2), rejects, indeed, exogeneity of the total PSQI score with respect to the exam results. Therefore, model (2) is our preferred model. The 2SLS estimate for the effect of the total PSQI score on the exam mark is about -0.97 and significantly different from 0 at the 5% significance level. This result can be interpreted as follows. An increase of the total PSQI score with one standard deviation, i.e. with about 2.23, leads to a decrease of the exam mark with about one point out of 20 (or 4.85 percentage points).

**TABLE 2 ABOUT HERE**

A comparison of the OLS and 2SLS results presented in column (1) and column (2) of Table 2 learns that, due to the endogeneity problem, OLS estimates are biased upward. A potential omitted variable that could

explain this bias is, as mentioned in Section 3.1, the pressure a first-year student may experience as a consequence of the high failure rate. If this pressure is high, this may lead to higher marks on the one hand and lower sleep quality (and thereby a higher PSQI score) on the other hand. Another potential explanation for the downward bias of the OLS estimates is the aforementioned problem of random measurement error in average sleep quality during the semester.

We briefly discuss some secondary results reported in column (2) of Table B.2. Note that a structural interpretation of some explanatory variables is hazardous as they might be endogenous to exam outcomes. The observed (and strong) effects of ethnicity, program in secondary education and general end marks in secondary education, are generally consistent with our expectations. Somewhat surprising is the, albeit weakly significantly, negative effect of high educated fathers. This variable, however, correlates to an important extent with other drivers of exam marks such as the program in secondary education.

In model (3) and model (4) we regress the mark for each completed exam on a dummy indicating bad sleepers. Again, exogeneity of this dummy is rejected so that we focus on the results presented in column (4). We find that the average exam mark is about 2.64 points lower among the bad sleepers *ceteris paribus*, an estimate which is significantly different

from 0 at the 5% significance level.

As the total PSQI score is composed both by quantitative and qualitative indicators of sleep quality, it is interesting to test which of both dimensions affects exam results the most. Therefore, in model (5) and model (6) we substitute the total PSQI score by the PSQI submeasure of overall perceived sleep quality. In model (7) and model (8) we use the PSQI submeasure of sleep duration. The 2SLS estimation results have, for both submeasures, the expected negative sign and are significantly different from 0. However, the magnitude of the latter submeasure is somewhat higher. An increase of the overall perceived sleep quality with one standard deviation lowers the exam mark with about 0.89 points (4.45 percentage points) while an increase of the sleep duration index with one standard deviation lowers the exam score with about 1.28 points (6.38 percentage points). Interestingly, also the OLS estimate is significantly different from 0 for the latter submeasure. The reader might mention that also the test for endogeneity is less significant for this measure (compared with the test statistics in columns (2), (4) and (6)). This makes sense. While it is clear that omitted variables like indications of bad exam results or internal and external pressure may affect (overall perceived) sleep quality, this is less clear for sleep duration.

This higher predictive power for sleep duration compared with overall

perceived sleep quality can be explained by the types of sleep that a person goes through during a regular sleep period of about seven hours. The first half of the sleep period is dominated by a deep sleep, the slow-wave sleep (SWS). The second half of the sleep period is characterised by longer periods of rapid-eye-time-sleep (REM), during which more brain activity occurs. Individuals who score low in terms of sleep duration, will typically get less REM-sleep. It is known, however, that this type of sleep is important for storing knowledge in a more permanent way. In other words, the REM-sleep leads to memory consolidation. Moreover, it associates new information with existing knowledge (Diekelmann & Born, 2010; Smith & Lapp, 1991). It goes without saying that the latter mechanisms are important in the context of rehearsing, understanding, reproducing and applying new academic knowledge in preparation of an exam.

Table B.3 presents comparable results for the same academic outcome, i.e. exam scores, but now for all potential exams, recoding exams for which students did not show up to 0. Unsurprisingly, given the small number of potential exams that were not taken, this leads to results that are completely similar to those of Table B.2.

Table B.4 shows the regression results when using the dummy indicating exam success, i.e. indicating an exam mark of at least 10 points out of 20. First, we get that increasing the total PSQI score with one

standard deviation lowers the probability of exam success with about 9.22 percentage points. This is a strong effect. However, our regression results show that characteristics such as ethnicity, program in secondary education and general end marks in secondary education are still better predictors for exam success than sleep quality. Second, the effect of the dummy indicating bad sleepers is even higher. At the same time, the standard errors are quite high in regression model (4). Last, based on columns (6) and (8) we find again suggestive evidence for sleep duration being a better predictor for exam success than overall perceived sleep quality. The results presented in Table B.5, based on all potential exams, lead to the same conclusions.

Last, we conducted additional regressions with the mark on completed exams as dependent variable and alternative specifications for the main explanatory variables. The results of these analyses, which are in line with our benchmark results, are discussed in the Online Appendix [INSERT LINK TO ONLINE APPENDIX].

## **5 Conclusion**

In this study we empirically tested the impact of sleep quality on



educational achievement at university. This research complements recent contributions looking into the role of other health factors on university achievements. Furthermore, our hypothesis of a positive relationship between sleep quality and academic success was supported by former research within medicine and biology indicating that night's rest is essential to helping maintain mood, motivation, memory and cognitive performance.

In view of our research aims, we surveyed first-year university students on their sleep quality, by means of the Pittsburgh Sleep Quality Index (PSQI). In addition, these students were surveyed on general social background and health characteristics. The resulting dataset was merged with the marks they scored in their first examination period. To be able to correctly identify the influence of sleep quality on academic achievement, we used an instrumental variable econometric approach.

We found that an increase of one's PSQI score with one standard deviation, which implies a deterioration of his/her overall sleep quality, leads to a decrease of the exam mark with 0.97 out of 20 points (or with 4.85 percentage points). Moreover, this result seems to be to a large extent driven by aspects of sleep duration captured by the PSQI measure (rather than by qualitative aspects).

From a policy perspective, our results seem to indicate that students should be encouraged to not sleep (systematically) too little. While sleep quality is a factor which is not fully under control of students, their average sleep duration is to a large extent a choice. In this respect we follow Mindell et al. (2011) that “sleep should be a standard component of school curriculums, with an emphasis on the importance of the need for sleep, the impact of sleep loss, awareness of sleep problems, and the basics of sleep and sleep architecture.” In addition, higher education providers might be incentivised to invest part of their resources for social facilities in professional support for students with health – including sleep – problems.

With this paper, we aimed at taking an important step forward in the causal interpretation of the relationship between sleep quality and academic achievement. While we carefully discussed the validity of our instrumental variable approach, complete exogeneity cannot be guaranteed. Therefore, we believe that randomised controlled trials of sleep interventions with students in tertiary education would be a fruitful direction for further research. Last, we look forward to future research on the impact of sleep quality and sleep duration on other dimensions of academic achievement such as the length and success of doctoral studies.

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Table 1 – Summary Statistics

	(1)	(2)	(3)	(4)	(5)	(6)	(7)
	All observations		PSQI ≤ 5		PSQI > 5		Difference (5) – (3)
	Mean	SD	Mean	SD	Mean	SD	
<b>A. Background characteristics</b>							
Age	18.03	0.431	18.01	0.363	18.08	0.552	0.070*** [4.696]
Female Sex	0.503	0.500	0.467	0.499	0.583	0.493	0.115*** [6.684]
Highest diploma mother							
Tertiary education: university	0.277	0.448	0.289	0.453	0.252	0.434	-0.036** [2.353]
Tertiary education: outside university	0.411	0.492	0.424	0.494	0.382	0.486	-0.042** [2.459]
No tertiary education	0.311	0.463	0.287	0.453	0.366	0.482	0.078*** [4.900]
Highest diploma father							
Tertiary education: university	0.392	0.488	0.387	0.487	0.405	0.491	0.018 [1.030]
Tertiary education: outside university	0.309	0.462	0.322	0.467	0.280	0.449	-0.042*** [2.597]
No tertiary education	0.298	0.458	0.291	0.454	0.315	0.465	0.024 [1.523]
At least one of parents passed away	0.029	0.168	0.014	0.116	0.064	0.245	0.051*** [8.807]
Parents divorced	0.199	0.399	0.168	0.374	0.269	0.444	0.101*** [7.360]
Grandmother on mother's side foreign nationality	0.079	0.269	0.067	0.249	0.106	0.308	0.040*** [4.269]
Number of siblings							
None	0.086	0.280	0.076	0.265	0.109	0.312	0.033*** [3.413]
One	0.537	0.499	0.548	0.498	0.511	0.500	-0.037** [2.155]
Two	0.298	0.458	0.288	0.453	0.322	0.468	0.035** [2.184]
More than two	0.079	0.270	0.089	0.284	0.058	0.234	-0.030*** [3.261]
Living in a student room	0.460	0.498	0.452	0.498	0.478	0.500	0.027 [1.552]
Program in secondary education							
Economics - languages/sports	0.286	0.452	0.268	0.443	0.327	0.469	0.059*** [3.783]
Economics - maths	0.267	0.442	0.286	0.452	0.223	0.417	-0.062*** [4.082]
Ancient languages	0.166	0.372	0.167	0.373	0.164	0.371	-0.002 [0.191]
Exact sciences - maths	0.179	0.383	0.182	0.386	0.172	0.377	-0.010 [0.756]

General secondary education: other	0.056	0.230	0.061	0.239	0.046	0.209	-0.015* [1.867]
Technical secondary education	0.046	0.210	0.037	0.188	0.067	0.251	0.031*** [4.226]
General end marks in secondary education							
Less than 70%	0.013	0.113	0.016	0.127	0.005	0.070	-0.011*** [2.939]
Between 70% and 80%	0.393	0.488	0.372	0.483	0.440	0.497	0.068*** [4.051]
More than 80%	0.594	0.491	0.612	0.487	0.555	0.497	-0.057*** [3.349]
In a relationship	0.413	0.492	0.428	0.495	0.379	0.485	-0.049*** [2.877]
General health							
Very good	0.359	0.480	0.408	0.492	0.247	0.431	-0.161*** [9.848]
Good	0.526	0.499	0.519	0.500	0.544	0.498	0.025 [1.468]
Moderate, bad or very bad	0.115	0.319	0.073	0.261	0.209	0.407	0.136*** [12.583]
DASS-21 depression scale	3.148	3.519	2.543	3.005	4.553	4.166	2.010*** [16.937]
DASS-21 anxiety scale	2.982	3.027	2.371	2.496	4.355	3.612	1.984*** [19.843]
DASS-21 depression scale	5.178	3.964	4.390	3.572	6.958	4.222	2.569*** [19.515]
<b>B. Sleep quality predictors</b>							
Maternal sleep quality index	2.753	0.970	2.681	0.951	2.919	0.995	0.238*** [7.049]
Paternal sleep quality index	2.423	0.959	2.355	0.950	2.586	0.960	0.231*** [6.884]
Sleep quality during secondary education index	2.079	0.749	1.929	0.669	2.418	0.807	0.489*** [19.767]
Congenital medical problems that affect sleep quality	0.014	0.119	0.014	0.117	0.015	0.122	0.001 [0.295]
<b>C. Sleep quality</b>							
PSQI: total measure	4.802	2.228	3.624	1.149	7.408	1.779	3.784*** [79.149]
Poor sleep quality (PSQI > 5)	0.306	0.461	-	-	-	-	-
PSQI: submeasure overall perceived sleep quality	0.937	0.626	0.710	0.493	1.452	0.590	0.742*** [40.887]
PSQI: submeasure sleep duration	0.129	0.363	0.059	0.245	0.287	0.508	0.227*** [18.932]
<b>D. Course characteristics</b>							
Number of ECTS-credits in program	26.704	1.338	26.760	1.416	26.579	1.132	-0.181*** [3.910]
Program of BE	0.500	0.500	0.490	0.500	0.522	0.500	0.031* [1.799]
Program of BE: Accounting	0.083	0.276	0.085	0.279	0.080	0.271	-0.005 [0.542]
Program of BE: Economics	0.083	0.276	0.085	0.279	0.080	0.271	-0.005 [0.542]
Program of BE: Human Sciences	0.083	0.276	0.085	0.279	0.080	0.271	-0.005 [0.542]

Program of BE: Law	0.083	0.276	0.085	0.279	0.080	0.271	-0.005 [0.542]
Program of BE: Mathematics	0.083	0.276	0.085	0.279	0.080	0.271	-0.005 [0.542]
Program of BE: Production Technology	0.083	0.276	0.085	0.279	0.080	0.271	-0.005 [0.542]
Program of CS: Accountancy	0.074	0.263	0.073	0.261	0.077	0.267	0.004 [0.444]
Program of CS: Commercial and Financial Transactions	0.074	0.263	0.073	0.261	0.077	0.267	0.004 [0.444]
Program of CS: English	0.059	0.236	0.057	0.231	0.066	0.248	0.009 [1.085]
Program of CS: French	0.039	0.193	0.034	0.181	0.049	0.216	0.015** [2.25]
Program of CS: Information Technology	0.030	0.170	0.034	0.180	0.021	0.143	-0.013** [2.199]
Program of CS: Law	0.074	0.263	0.073	0.261	0.077	0.267	0.004 [0.444]
Program of CS: Mathematics	0.074	0.263	0.073	0.261	0.077	0.267	0.004 [0.444]
Program of CS: Microeconomics	0.074	0.263	0.073	0.261	0.077	0.267	0.004 [0.444]
<b>E. Academic Achievement</b>							
Exam mark: completed exams	10.82	3.637	10.98	3.620	10.46	3.651	-0.515*** [4.081]
Exam mark: potential exams	10.74	3.740	10.91	3.711	10.36	3.779	-0.551*** [4.266]
Exam passed (mark ≥ 10): completed exams	0.651	0.477	0.669	0.471	0.611	0.488	-0.058*** [3.515]
Exam passed (mark ≥ 10): potential exams	0.646	0.478	0.665	0.472	0.605	0.489	-0.060*** [3.637]
Number of subjects	621		432		189		-

All statistics are presented at the individual exam level. Used abbreviations: SD: standard deviation; BE: (Business) Economics; CS: Commercial Sciences. t-tests are performed to test whether the difference presented in column (7) is significantly different from zero. \*\*\*(\*\*)(\*) indicates significance at the 1%(5%)(10%) significance level. t-statistics are between brackets.

Table 2 – Main Results

Regression number	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
Estimation method	OLS	2SLS	OLS	2SLS	OLS	2SLS	OLS	2SLS
Dependent variable	Exam mark: completed exams							
<b>A. Main explanatory variables</b>								
PSQI: total measure (normalised)	-0.103 (0.118)	-0.972** (0.386)						
Poor sleep quality (PSQI > 5)			-0.004 (0.227)	-2.639** (1.138)				
PSQI: submeasure overall perceived sleep quality (normalised)					0.071 (0.163)	-0.889** (0.358)		
PSQI: submeasure sleep duration (normalised)							-0.236** (0.103)	-1.275** (0.516)
<b>B. Control variables</b>								
Background characteristics	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Course dummies	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Number of exam outcomes	3607	3601	3658	3652	3651	3645	3652	3646
Number of subjects	573	572	581	580	580	579	580	579
Wooldridge's (1995) robust endogeneity test (p-value)	-	0.016	-	0.009	-	0.004	-	0.029
First stage: effect of instrument on sleep quality measure	-	0.398*** (0.059) [44.94]	-	0.146*** (0.027) [29.66]	-	0.430*** (0.058) [54.19]	-	0.303*** (0.061) [24.74]

All PSQI scales are normalised by subtracting the sample mean and dividing the result by the sample standard deviation. Used instrumental variable for sleep quality measures in 2SLS: sleep quality during secondary education index. Standard errors are between parentheses and clustered at the subject level. F-statistics for the significance of the instrument in the first stage regressions are between brackets. \*\*\*(\*\*)(\*) indicates significance at the 1%(5%)(10%) level. The various numbers of exam outcomes and subjects can be explained by a different number of missing explanatory and instrumental variables across the regression models.