

DISCUSSION PAPER SERIES

IZA DP No. 14797

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Transitions to Post-secondary Education**

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ISSN: 2365-9793

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ABSTRACT

The COVID-19 Pandemic, Well-Being, and Transitions to Post-secondary Education*

This study examines the immediate and intermediate effects of the COVID-19 pandemic on the well-being of two high school graduation cohorts (2020 and 2021). We also investigate how changes in well-being at the transition to post-secondary education affect educational plans and outcomes. Our unique panel data contain prospective survey information on three dimensions of well-being: mental health problems, self-rated health, and life satisfaction for 3,697 students. Data is collected several months before (fall 2019), shortly before and soon after (spring 2020), and several months after (fall/winter 2020/21) the outbreak of the COVID-19 pandemic. Applying difference-in-differences designs, random effect growth curve models, and linear regression models, we find that school closures had a positive immediate effect on students' wellbeing. Over the course of the pandemic, however, well-being strongly declined, mainly concentrated among the 2021 graduation cohort. Finally, we show that a strong decline in mental health is associated with changes in educational and career plans and transition outcomes. As adverse life experiences in adolescence are likely to accumulate over the life course, this study is the first to exhibit potential long-lasting negative effects of the COVID-19 pandemic on education and careers of young individuals.

JEL Classification: I21, I18, J24

Keywords: COVID-19, high school graduates, mental and physical well-being, life satisfaction, school-to-work transition

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* We would like to thank Adrian Lerche and Carina Toussaint for their valuable feedback. Furthermore, we thank participants of the Annual Conference of the Verein für Socialpolitik (VfS), Annual Meeting of the Standing Field committee Economics of Education of the VfS, CIDER-LERN conference, IZA & Jacobs Center workshop, OECD-IAB-Inequality seminar, and Transition in Youth conference for their valuable comments. Moreover, we would like to thank Anton Barabasch for excellent research assistance.

1 Motivation

The COVID-19 pandemic and the related policies to stop the spread of the coronavirus, particularly school closures, present a severe shock to mental and physical well-being for millions of young individuals worldwide. These distancing measures may affect the mental and physical health and life satisfaction of young individuals, as these measures massively change the schooling and leisure activities of students, such as physical activity, social contacts, substance use, and sleep time (Hisler and Twenge 2021; Emery et al. 2021; Liu et al., 2020; Shanahan et al., 2020). Pandemic policies may especially impact the well-being of young individuals who are in their final school years because the measures not only affected schooling and leisure but also strongly reduced students' perceived career security and job and educational opportunities. As students' well-being presents a crucial resource in the process of educational decision-making and socioeconomic attainment (Haas 2006), shocks to well-being may disrupt the transition from upper secondary to post-secondary education. Such transition disruptions at this stage may have negative consequences on future educational and labour market success, lifetime earnings and later life health (e.g., Leopold 2018; Tamborini et al. 2015; Oreopoulos 2007).

However, thus far, no empirical evidence exists on how school closures and the COVID-19 pandemic affect the well-being of students in their final high school years and how effects on their mental and physical conditions relate to their educational and career plans and transition outcomes. We fill this research gap by using large-scale panel data on well-being, educational and career plans and transition outcomes of 3,697 German high school students from the 2020 and 2021 graduation cohorts. These data have two key features. First, they entail three detailed indicators of well-being, i.e., mental health (10-item Hopkins Symptom Checklist; Derogatis et al. 1974), self-rated physical health (5-point scale; e.g., Mossey and Shapiro, 1982), and life satisfaction (11-point scale; e.g., Diener 1984). Second, these data contain both pre-pandemic information and information during the pandemic, as they stem from three survey waves in fall 2019, spring 2020, and fall/winter 2020/21.

Drawing on these data enables us to investigate (i) the immediate effects of nationwide school closures on students' well-being in spring 2020; (ii) the intermediate effects of the COVID-19 pandemic in general on students' well-being in fall/winter 2020/21; (iii) the heterogeneous effects of the COVID-19 pandemic on high school graduates who transition to post-secondary education and students still enrolled in high school; and (iv) the impact of decline in mental health on career and educational plans and educational decisions of graduates.

In the first step of our analysis, we separately investigate immediate and intermediate effects of the COVID-19 pandemic on well-being. This separation is important, particularly with respect to school policies, since students may perceive school closures as holidays or health protection in the short run (Helliwell and Wang 2014), while in the long run, stressors due to adverse health, learning achievement, distancing measures or uncertainty about the future may prevail. To evaluate the immediate effects of nationwide school closures, the data allow the application of a difference-and-differences design exploiting the fact that some students within the second survey wave responded just before and other students shortly after the school closures. To elaborate on the intermediate effects of the pandemic (i.e., the developments prior to and during the crisis), we employ linear growth curve modelling.

In a second step, we investigate whether the COVID-19 pandemic has different effects on students who spend most of their two final high school years in times of the pandemic (2021 graduation cohort) and students who graduated from high school shortly after the outbreak of the pandemic (2020 graduation cohort). Differences in well-being between graduation cohorts might occur because students still enrolled in high school may face greater uncertainty about future decisions than graduates who already realized their transition to post-secondary education. However, school graduates face a completely unknown situation at their new educational institutions, since universities and vocational schools have similarly introduced distance learning (Crawford et al. 2020), which hardly enabled any interactions with new fellow students and apprentices. Additionally, the pandemic has reduced the available vocational training positions as alternative educational paths after high school as well as the number of student jobs, which may affect the financial situations of university students (Yükselen et al. 2020). Thus, it remains open whether the pandemic and related distancing measures affect students still at school or school graduates differently.

In a final analysis step, this study investigates to what extent a severe decrease in mental health leads to changes in educational and career plans and transition outcomes. Investigating such associations is important because earlier work showed that pre-transition health positively influences university enrolment decisions (Zheng 2017). Furthermore, poor mental health increases the probability of educational dropout (Cornaglia, Crivellaro, and McNally 2015). One potential mechanism explaining the importance of mental health for educational decision making might be that, for instance, depressive symptoms alter perceptions of the future (e.g., Leykin et al. 2011; Roepke and Seligman 2016). Thus, students with decreasing mental health may lose confidence in their educational and career plans or opt for transitions they would not have made with a better mental health status. As other major societal crisis, e.g. the Great Depression,

have cumulative negative effects for individuals over the life course (Hale 2017), investigating changes in mental health and educational plans and outcomes may provide important insights on the potential long-term consequences of the COVID-19 pandemic.

In analysing the immediate and intermediate effects of the COVID-19 pandemic on students' well-being in two graduation cohorts and how changes in well-being relate to educational and career plans and transition outcomes, we extend the existing and rapidly emerging research on the effects of the COVID-19 pandemic on the well-being and mental health of young adults and teenagers (e.g., Elmer et al. 2020; Emery et al. 2021; Giuntella et al. 2021; Shanahan et al. 2020). Furthermore, we contribute to the literature on how lockdowns affect educational and career plans, which until now has concentrated on university students or employed individuals (Aucejo et al. 2020; Fiaschi and Tealdi 2021). We combine these strands of literature and demonstrate that students who are close to the transition to post-secondary education are most vulnerable to shocks to their well-being due to the COVID-19 pandemic and that such shocks are related to educational and career plans and transition outcomes.

2 Methods

2.1 Participants and Data collection

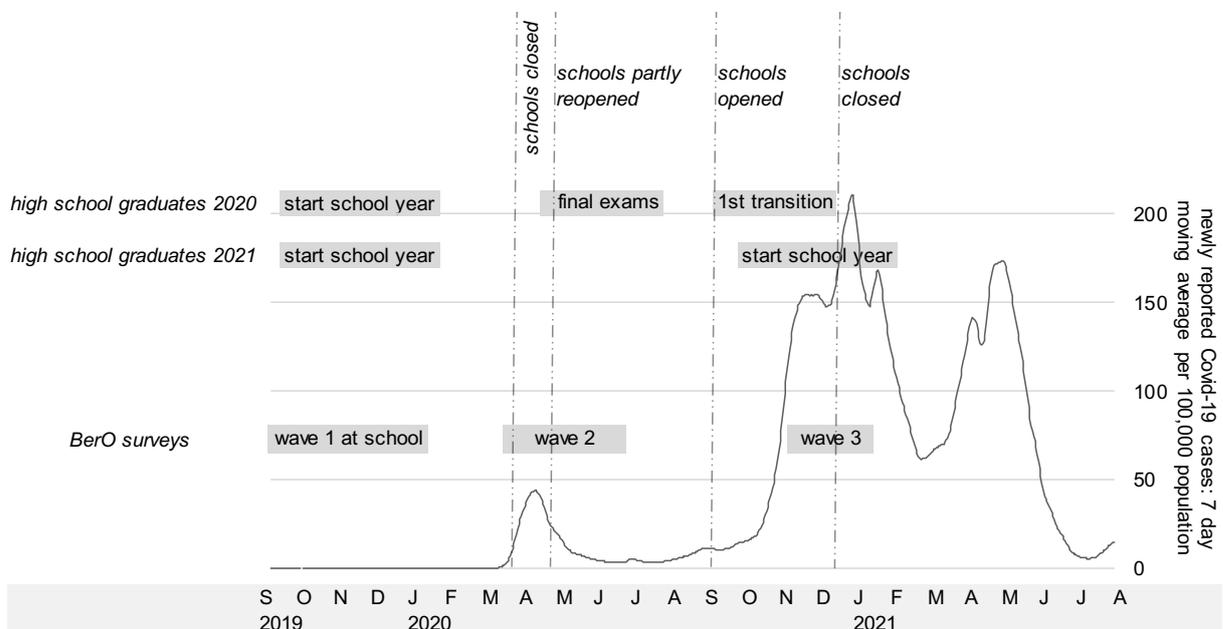
The participants of this study attended the highest track of secondary school in Germany, “*Gymnasium*”, in the final two years. The educational system comprises three tracks of secondary school: the lower and intermediate tracks prepare students for vocational training, whereas the highest (academic) track results—after successful completion—in the high school diploma “*Abitur*”, which qualifies students both to enrol at university or at a post-secondary vocational education or training. This academic track usually ends with final examinations at grade 12/13. These exams largely take place in March, and students receive their graduation diploma in the summer before they enrol at university or start vocational training in the fall.

After the outbreak of the COVID-19 pandemic at the beginning of 2020 and as one of the first nationwide pandemic prevention measures, all schools were closed after March 13th, 2020 in Germany. On April 23rd, 2020, the German federal states partly started reopening schools, albeit with very large regional and institutional variations: Since educational policy is the responsibility of the 16 federal states, there was no uniform school opening policy in Germany. Furthermore, local developments of the pandemic affected the closing of whole schools, class

levels, or single classes. After the summer break, schools started on a regular basis and then went gradually back to limited schooling in November and December 2020, first by allowing only alternating groups of students from each classroom and then from January 2021 switching back to complete distance schooling.

The data used in this study were collected for the BerO study, which evaluates the effectiveness of intensive job counselling for students in the highest secondary school track. The baseline survey (wave 1) was conducted as a paper-and-pencil interview (PAPI) in 214 schools in 8 of 16 German federal states. Students completed the questionnaire in school between September and November 2019 and were instructed by a professional data collection team. In addition to these data, our analyses draw on data from two follow-ups, which took place outside the school context as a computer-assisted web or telephone interview (CAWI/CATI). Students were interviewed from February to June 2020 (wave 2) during the first wave of infection with some students answering before and others after school closures. Survey wave 3 took place from November to January 2021 during the second COVID-19 wave. Figure 1 gives an overview of the timeline of the data collection.

Figure 1 Timeline of data collection and COVID-19 infections in Germany



2.2 Measures

To answer the first and second research questions on the immediate and intermediate effects of the COVID-19 pandemic on the mental and physical well-being of young individuals, this study investigates changes over time in three outcomes: (i) mental health problems, (ii) life satisfaction, and (iii) self-rated health. The first three rows of Table 1 give an overview of the descriptive statistics of these outcome variables.

First, as a widespread measure for mental health, in waves 2 and 3, this study employs data from a subscale of the well-established Hopkins Symptom Checklist (HSCL-58; Derogatis et al. 1974) to approximate individuals' risk for anxiety disorders and depression. The employed 10-item version (HSCL-10) has been shown to be a very good proxy for the longer HSCL-25 (Haavet et al 2011; Schmalbach et al 2021). This study uses a scale of four categories for each question ("No," "Yes, a little," "Yes, quite slightly," and "Yes, extremely," rated 1 to 4, respectively) and employs a binary measure, which indicates 1 if an individual's average score on the 10-item scale exceeds the widely used cut-off point of 1.85. We used this binary predictor because trials have indicated clinically relevant symptoms of anxiety and depression above this cut-off point (Strand et al. 2003).

Second, life satisfaction refers to "the degree to which an individual judges the overall quality of his/her own life as a whole favorably." (Veenhoven, 2012: 66). For waves 1, 2 and 3, we employ the established 11-point scale (e.g., Lucas 2007) and rely on answers to the following question: "How satisfied are you currently with your life in general?" Respondents could answer on a scale ranging from 0 ("totally dissatisfied") to 10 ("totally satisfied"). Prior research on life satisfaction and mortality (e.g., Diener and Chan, 2011) indicates that such cognitive evaluations of individuals' lives predict mortality.

Third, for waves 1, 2 and 3, this study examines self-rated health. Empirically, we rely on the question "How would you describe your current state of health?" Respondents could answer on a scale ranging from 1 ("poor") to 5 ("very well"). This question is a widely used item in many health studies in the social sciences and has been shown to be a strong predictor of mortality because it proxies general physical well-being (e.g., Mossey and Shapiro, 1982).

To elaborate on reasons for potential heterogeneity between graduation cohorts (i.e., between school students and school graduates), we additionally investigate whether students from the two cohorts perceive the COVID-19 situation differently. To this end, we explore whether differences in the current situation or whether worries about the future explain potential cohort

variation. To approximate young individuals' current situation, we use questions addressing enjoyment with learning and the extent to which young individuals are burdened by distancing measures. We approximate worries about the future in using questions asking about students' worries about career plans (the descriptive statistics of these outcomes are shown in table A1).

To answer the last research question, we identified severe drops in mental health between spring 2020 and fall/winter 2020/21 independent of the baseline value of mental health and the chosen cut-off point. A strong decrease is coded as 1 if values in the individuals' HSCL-10 scores changed by at least 0.4 scale points. This applies to a quarter of the sample. To investigate the influence of these severe mental health drops on educational and career plans, we rely on five different measures. For graduation cohort 2020 analysed educational and career plans comprise probabilistic beliefs about finishing the current post-secondary education. For graduation cohort 2021 analysed plans comprise probabilistic beliefs about successfully finishing potential university education, the certainty about the future educational pathway, expected final grade point average (GPA), and probabilistic beliefs of studying a science, technology, engineering and mathematics (STEM) major. While the probability of studying STEM directly measures one important educational aspiration, probabilistic beliefs and GPA expectations address—according to rational action theory (e.g., Breen and Goldthorpe 1997; Cameron and Heckman 1998)—important determinants of educational decisions.

To investigate the influence of decreasing mental health on educational transitions, we rely on three different outcomes, which were measured in the 2020 graduation cohort. Analysed measures comprise satisfaction with the overall educational decision, satisfaction with the learning institution, and satisfaction with the location of this institution. Analysing satisfaction measures appears important because research shows that satisfaction outcomes are associated with educational dropout (e.g., Sarra, Fontanella, and Di Zio 2019).

2.3 Analytical Strategy

Estimating the immediate effects of the first school closures

When investigating the effects of school closures, a pure outcome comparison between the students who answered before and after the school closures in March 2020 may be biased because the two groups of students may have different characteristics that could be related to our well-being outcomes under study. To solve this problem, we use the panel dimension of our data and

apply a difference-in-differences estimation using data from waves 1 and 2 shown in equation 1:

$$Y_{i,t} = \beta_1 SC_i + \beta_2 W_{i,t} + \beta_3 (SC_i W_{i,t}) + \varepsilon_{i,t} \quad (\text{Equation 1})$$

where $Y_{i,t}$ is the outcome of interest of individual i at wave t (life satisfaction and self-rated health, which are both available for waves 1 and 2). SC_i (School Closure) is a binary variable that takes a value of 1 for students who answered in March 2020 after the school closure and 0 for students who answered in March 2020 before the school closure, and β_1 captures the difference between those individuals. $W_{i,t}$ (Wave) contains a wave dummy for Wave 2 interviews, where β_2 captures the corresponding coefficient. $\varepsilon_{i,t}$ is a standard error term. Finally, $SC_i W_{i,t}$ is the interaction term of SC_i and $W_{i,t}$ that takes a value of 1 for students in wave 2 who answered the questionnaire after the school closure. The coefficient β_3 then measures the divergence in the outcome between those who answered after the school closure, i.e., the treatment group, and those who answered pre-event/policy, i.e., the control group, which indicates the average treatment effect on the treated (ATT). This is the effect of the school closures.

Only for the Hopkins scale we do not have information at wave 1. Therefore, for this outcome, we compare students who answered before and after the school closures, including a rich set of individual characteristics, as controls (shown in Table 1). For the analysis of all three outcomes, we restrict the time window to individuals who responded in the second survey wave prior to school closures and during closures. In doing so, the difference-in-differences analysis excludes students who participate in the interviews after school reopenings. Before the closures, all students answered within a time window of two weeks. Therefore, we argue that it is rather unlikely that pandemic factors, such as the infection rate, explain differences in well-being before and after school closures.

Estimating the development of well-being during the COVID-19 pandemic

To investigate the development of the examined well-being outcomes during the COVID-19 pandemic, we use the following specifications of linear random effects growth curve models:

$$Y_{i,t} = \alpha + \beta_n \sum_{n=2}^3 W_{n,it} + \lambda' X_i + \mu' X_{it} + \theta_i + \varepsilon_{i,t} \quad (\text{Equation 2})$$

$$Y_{i,t} = \alpha + \beta_n \sum_{n=2}^3 W_{n,it} + \gamma C_i + \delta_n \left(C_i \times \sum_{n=2}^3 W_{n,it} \right) + \lambda' X_i + \mu' X_{it} + \theta_i + \varepsilon_{i,t} \quad (\text{Equation 3})$$

In both equations, $Y_{i,t}$ represents either life satisfaction (0 to 10), self-rated health (1 to 5) or the risk for anxiety or depression (0 vs. 1). θ_i represents a person-specific error term, which is modelled as a random variable. ε_{it} constitutes an idiosyncratic error term. $W_{n,it}$ indicates dummy variables for each survey wave. X_{it} indicates a vector with time-invariant confounding variables, whereas X_i captures time-constant confounders. In equation 3, we introduce C_i indicating whether a respondent stems from the 2021 or 2020 graduation cohort. To allow for variation across graduation cohorts, we interact C_i with each wave dummy. While the multiplicative effect of γ captures heterogeneity between cohorts at wave 1, δ_n captures heterogeneity in well-being between cohorts over the course of the COVID-19 pandemic.

Finally, we apply two sets of ordinary least square regressions. In the first set, we specify a model to elaborate on differences between graduation cohorts 2020 and 2021 at wave 3 (fall/winter 2020/21). In the second set, we identify individuals with strong decreases in mental health between survey waves 2 and 3 to generate a binary variable (reference group: slight or no decrease in mental health) and regress educational and career plans and transition outcomes at wave 3 on this binary indicator. In correlating these measures, we elaborate on the potential long-term impact of the COVID-19 pandemic. Although this procedure constitutes a correlative workaround due to potential reversed causality, we can rule out large parts of endogenous selection bias by using our rich data. To this end, our model specifications condition on a vast set of individual characteristics as control variables, described in section 2.4, and they also include the baseline level (i.e., survey responses given in fall 2019 that are independent of any COVID-19-related factors) of each dependent variable and the baseline level of mental health (measured at wave 2).

2.4 Sample Characteristics and Control variables

From the BerO baseline sample (N=7,192), we restrict our analysis sample to students who participated in all three waves, in fall 2019, spring 2020 and fall/winter 2020/21, with nonmissing information on our outcomes. Categorical control variables contain a missing information category, while missing information in metric control variables has been deleted. Furthermore, to observe the transition to post-secondary training of students from the graduation cohort 2020, we restrict the respective analyses to high school graduates 2020 who transitioned directly to post-secondary education, i.e., did not spend a “gap year” to bridge the time between high school graduation and enrolling at university or starting vocational training (Figures A2 to A4 in the Online Appendix display results from robustness checks based on a sample that does not

exclude individuals in a gap year state. This workaround reveals that the results do not depend on sample selection.). Overall, our sample consists of 3,697 students who participated in the baseline survey and in both subsequent surveys with valid information, with 2,451 students from grade 11 (graduation cohort 2021) and 1,246 students from grade 12 (graduation cohort 2020).

The set of individual characteristics that we use as controls includes sociodemographics, i.e., cohort (graduation cohort 2021 dummy), gender (male dummy), migration background (1st/2nd generation migrant), parental education (at least one parent with university education), and educational achievement (GPA better than 2.5 on average on a scale from 1 – best grade to 6 – failed). Moreover, we use a rich set of preferences, i.e., risk aversion and myopia, and personality traits, for which we use constructs based on multiple items to measure self-efficacy, grit, and the Big Five personality dimensions openness, conscientiousness, extraversion, agreeableness, and neuroticism. Furthermore, we rely on school fixed effects to account for institutional and geographical variation in the data.

Table 1 presents the sample characteristics in the three waves, with the spring 2020 wave split into individuals who answered before and after the school closures. The first rows show the means of our overall well-being measures. They show strong variation among the waves and before and after the school closures. We will investigate these changes in detail in the next sections. The next rows depict the sociodemographic characteristics, educational achievement and educational choice, as well as preferences and personality traits of the sample. As we use a balanced sample, there are no differences in the characteristics between wave 1 (fall 2019) and wave 3 (fall/winter 2020/21). However, the figures reveal that the characteristics between those students who answered before and after the school closures differ; for example, more males and slightly worse performing students answered the questionnaire after the school closures compared to students who participated in the survey before the school closures. This finding supports our strategy to apply a difference-in-differences approach to rule out biases by this selection.

Appendix Table A1 shows the descriptives of the measures, which are only available for wave 3 (fall/winter 2020/21). The variables contain information on attitudes and worries, i.e., how students deal with the COVID-19 pandemic and distancing measures, as well as on educational and career plans and further sociodemographic characteristics. Appendix Table A2 shows the aggregated values over the three waves for the variables shown in Table 1.

Table 1 Sample characteristics by wave

Means	Fall	Spring		Fall/
	2019	2020	2020	Winter 2020/21
	Oct. to Nov.	Pre-SC	Post-SC	Nov. to Jan.
Outcomes				
Life satisfaction (0–10)	7.435	7.080	7.170	6.671
Self-rated health (1–5)	3.872	3.634	3.891	3.684
Mental health problems (0–1)	-	0.425	0.334	0.485
Socio-demographics				
Graduation cohort 2021 (0 vs. 1)	0.663	0.653	0.678	0.663
Male (0 vs. 1)	0.355	0.330	0.395	0.355
1st/2nd generation migrants (0 vs. 1)	0.209	0.206	0.214	0.209
Missing information on migration status (0 vs. 1)	0.055	0.058	0.050	0.055
At least one parent with univ. education (0 vs. 1)	0.540	0.526	0.562	0.540
Missing information on parental education	0.105	0.105	0.105	0.105
GPA better than 2.5 (0 vs. 1)	0.484	0.495	0.465	0.484
Missing information on GPA (0 vs. 1)	0.010	0.008	0.013	0.010
Attending university (0 vs. 1)	-	-	-	0.284
Vocational training (0 vs. 1)	-	-	-	0.054
Preferences and Personality				
Risk aversion (0–10)	5.629	5.567	5.730	5.629
Dummy for being myopic (0 vs. 1)	0.115	0.114	0.117	0.115
Self-efficacy (1–4)	2.924	2.920	2.931	2.924
Grit (1–5)	3.472	3.488	3.446	3.472
Big Five				
Openness (1–7)	4.778	4.764	4.802	4.778
Conscientiousness (1–7)	5.222	5.258	5.165	5.222
Extraversion (1–7)	4.764	4.707	4.857	4.764
Agreeableness (1–7)	5.434	5.421	5.454	5.434
Neuroticism (1–7)	4.245	4.259	4.221	4.245
Method				
CATI interview (0 vs. 1)	-	-	0.053	0.052
PAPI/CAWI interview (0 vs. 1)	1.000	1.000	0.947	0.948
<hr/>				
N persons	3,697	2,292	1,405	3,697

Note: GPA = Grade point average, a lower GPA indicates better performance; CATI = Computer-assisted telephone interview; PAPI = Paper-and-pencil interview; CAWI = Computer-assisted web interview.

Data: BerO study wave 1, 2 and 3.

3 Results

3.1 Immediate effects of first school closures on well-being

Table 2 presents the immediate effects of school closures on the three well-being outcomes. Using the difference-in-differences approach explained in equation 1 demonstrates that while school closures did not affect life satisfaction, self-rated health significantly but weakly increased after school closures by 0.21 standard deviations (see online Appendix Figure A1 for a graphical illustration of the effects). Investigating the immediate effect of school closure on mental health based on a within-wave 2-comparison reveals that the risk of mental health problems—while controlling for the variables introduced in section 2.4—is 5 percentage points lower for the students who participated in the survey after the closures. Overall, these results indicate that school closures had a positive effect on overall health in the first weeks after their implementation, as indicated by improvements in two of the three measures.

Table 2 Immediate effects of school closures well-being: Results from difference-in-differences and OLS regressions

<i>Panel A</i>	Mean wave 1	Mean wave 2 before school closures	Mean wave 2 after school closures	DID in % of SE	p-value DID
<i>Life satisfaction (0–10)</i>	7.432	7.080	7.187	-0.005	0.872
<i>Self-rated health (1–5)</i>	3.875	3.634	3.882	0.213	0.000
<i>Panel B</i>				Mean Diff. pre and post school closures	p-value
<i>Mental Health Problems (0–1)</i>		0.434	0.354	-0.053	0.000
N persons	3,503	2,292	1,211		

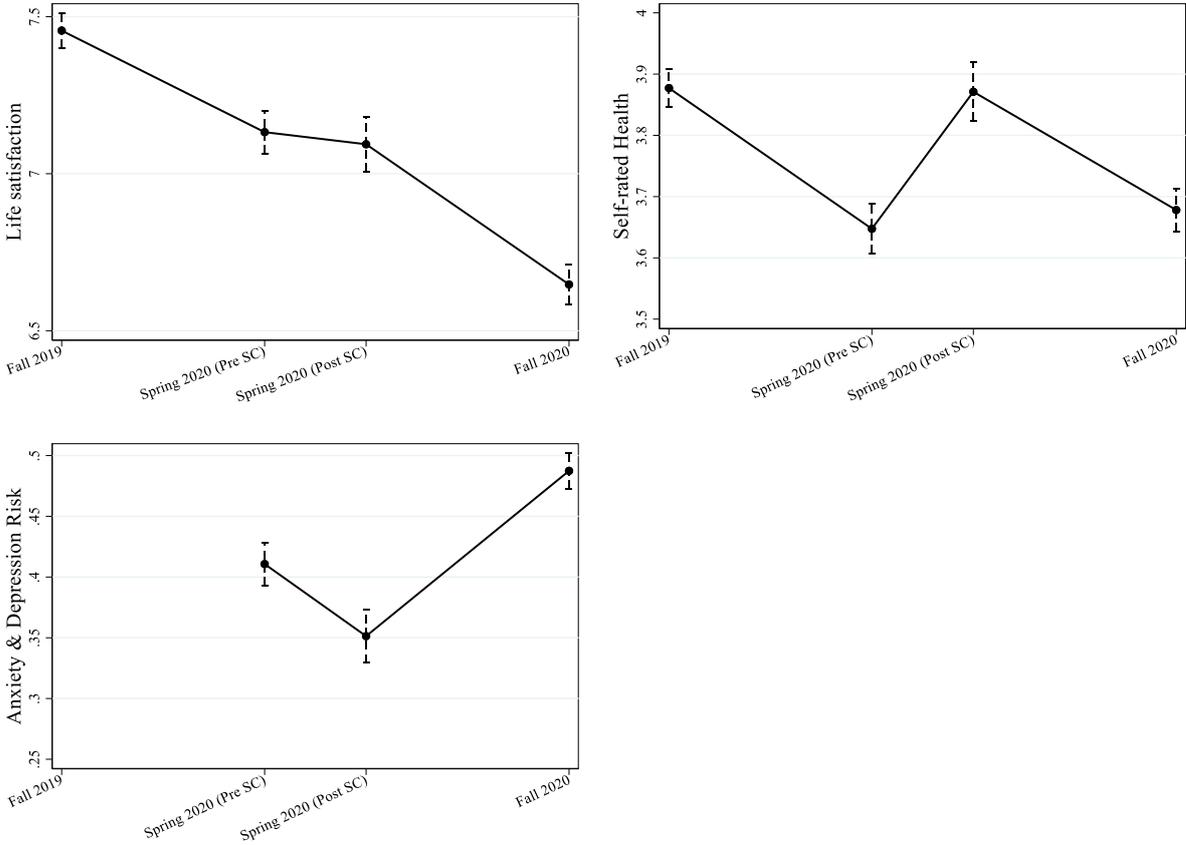
Note: Panel A presents estimates in percent of standard deviation based on difference-in-difference regressions with federal state fixed effects. Panel B presents mean differences based on an OLS regression. Controls: school fixed effects, gender, migration status, parental education, school performance at wave 1, self-efficacy, Grit, Big Five personality traits, graduation cohort, risk aversion, time preferences, self-rated health, life satisfaction, and interview mode

Data: BerO study wave 1 and 2.

3.2 Development of well-being before and during the first and second waves of the COVID-19 pandemic

In this section, we investigate the effects of school closures and distancing regulations eight months after the pandemic started. Figure 2 shows the development of the three outcomes at wave 1, at wave 2 before and after the school closures and at wave 3 calculated by applying equation 2 (see Appendix Table A3 for point estimates and significance levels). In line with the previous section, we see an immediate increase in self-rated health and a decrease in mental health problems in wave 2 after the school closures. However, from spring to fall, we observe a strong decrease in life satisfaction and self-rated health and a particularly strong increase in mental health problems. Overall, the data suggest that after students' overall health improved in the short term, it strongly declined in the longer term.

Figure 2 Development of well-being since fall 2019. Results from random effect growth curve models



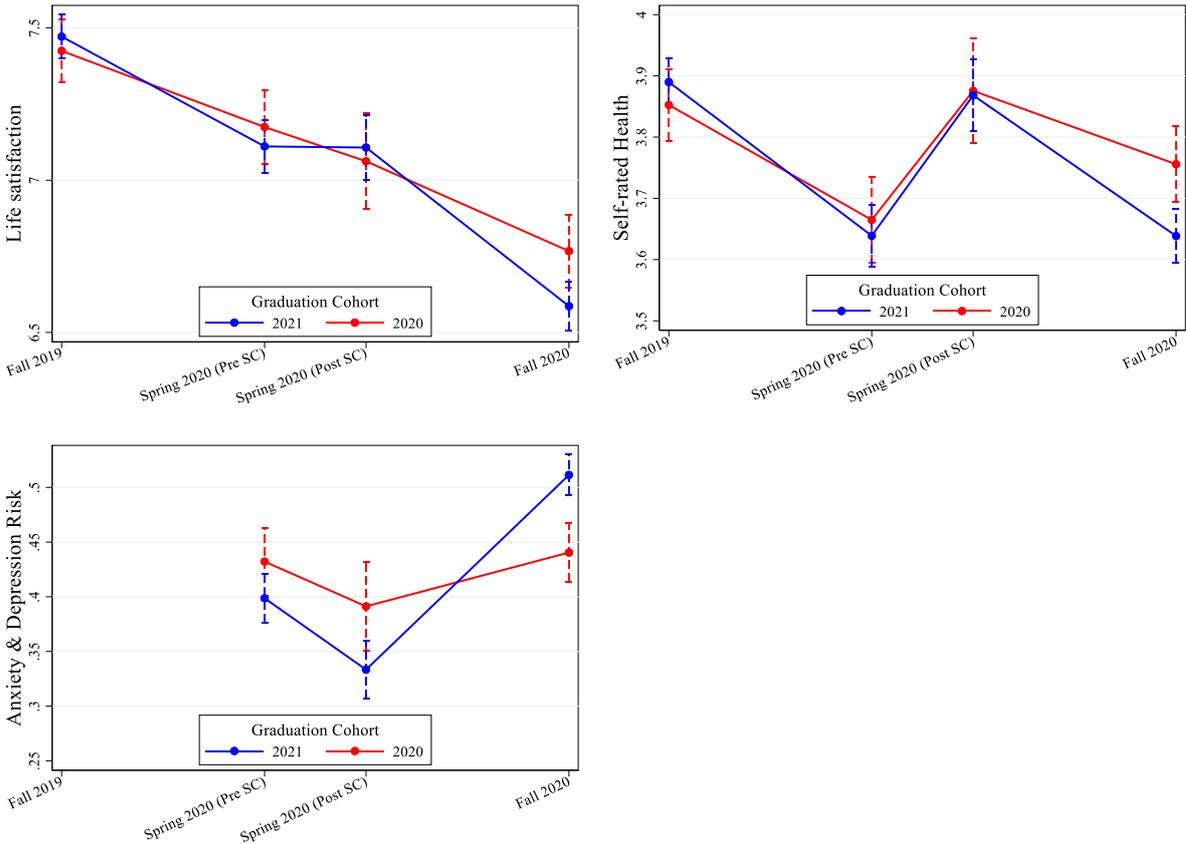
Note: Outcomes: Life satisfaction (0 to 10); self-rated health (1 to 5); Dummy for being above the clinical threshold for a high anxiety and depression risk (HSCL-10). N Life Satisfaction = 11,091; N SRH = 11,091; N SCL-10 = 7,394. Controls: see note for Table 2.

Data: BerO study wave 1 to 3.

3.3 COVID-19 effects on mental and physical well-being by graduation cohorts 2020 and 2021

We now investigate how the effects on the three well-being measures differ over time between the 2020 and 2021 graduation cohorts. Using equation 3, Figure 3 shows that none of the three measures differed significantly at wave 1 or 2 (before and after the school closures) between the two cohorts (see Appendix Table A4 for the point estimates and significance levels). However, at wave 3 (fall/winter 2020/21), the graduation cohort 2021 showed significantly worse outcomes for all three well-being measures. The difference was most pronounced for mental health problems, for which the increase for graduation cohort 2021 from wave 2 to wave 3 amounts to almost 20 pp, while the increase was 5 pp for the 2020 graduation cohort.

Figure 3 Development of mental and physical well-being by graduation cohort. Results from random effect growth curve models



Note: *Outcomes:* Life satisfaction (0 to 10); self-rated health (1 to 5); Dummy for being above the clinical threshold for a high anxiety and depression risk (HSCL-10). N Life Satisfaction = 11,091; N SRH = 11,091; N HSCL = 7,394. *Controls:* see note for Table 2 excluding graduation cohort
Data: BerO study waves 1 to 3.

After showing substantial variation between graduation cohorts in fall/winter 2020/21, we now investigate two potential mechanisms that might explain cohort differences. In particular, we test whether differences in the current enjoyment of education and burdens induced by distancing measures or whether worries about the future can explain cohort variation. We assume that the mechanisms under study are important because young individuals spend a great amount of their daily time with being at school and with school work (Anger et al. 2020). Moreover, research indicates that perceived career insecurity is detrimental to well-being (e.g., Kopasker, Montagna, and Bender, 2019).

For this purpose, Columns 1 and 2 of Table 3 present the results of two sets of questions that the students in both cohorts answered at wave 3, and Column 3 shows the differences between the two graduation cohorts. Students who had just started their final school year (graduation cohort 2021) reported higher levels of burden due to distancing measures and less enjoyment of learning in fall 2020 than those who already had left school (graduation cohort 2020) and had already attended university study (84%) or vocational training (16%). Analysing respondents' worries about their future reveals that students from graduation cohort 2021 were more worried than those in graduation cohort 2020 about their occupational futures and expected negative effects of distancing policies on their future careers. They also claim deficits with respect to receiving relevant career information.

Table 3 Attitudes of the 2020 and 2021 graduation cohorts regarding distancing measures and worries (Fall 2020)

	(1) Mean graduation cohort 2021	(2) Mean graduation cohort 2020	(3) Diff. btw. Cohorts 2021–2020
<i>Dealing with the current situation</i>			
Enjoyment of learning (1–5)	3.052 (1.052)	3.715 (0.958)	-0.662***
Burden of distancing measures (1–5)	3.036 (1.304)	2.624 (1.228)	0.412***
<i>Future worries</i>			
Impact of distancing policies on future career (0 vs.1)	0.622 (0.484)	0.447 (0.497)	0.175***
Worries about occupational future (1–5)	2.322 (1.191)	1.959 (1.150)	0.363***
Worries about too little career information (1–5)	3.058 (1.337)	2.500 (1.359)	0.558***
N persons	2,450	1,247	

Note: Standard deviations italicized in brackets. * $p < 0.10$, ** $p < 0.05$, *** $p < 0.01$ indicate significant differences (based on t-tests).

Data: BerO study wave 3.

Next, we investigate to what extent these higher concerns of the graduation cohort 2021 can explain the observed differences in well-being between the two cohorts. Table 4 shows the results of the estimations in which we regress the three well-being measures on the control variables (Column 1) and on the two sets of questions that may explain the difference between the cohorts in well-being (Columns 2 and 3). In line with the results in Figure 3, Column 1 demonstrates that graduation cohort 2021 reported significantly worse outcomes in the health measures than graduation cohort 2020. However, the results in column 2 show that including the present attitudes completely absorbs the difference between the cohorts for all outcomes and explains the largest share of the difference in life satisfaction, self-rated health, and mental health between the cohorts. The attitudes explain even more of the gap than future worries, which also reduces but does not fully absorb the effect of the graduation cohort (Column 3).

Table 4 Mechanisms explaining differences in dimensions of well-being between graduation cohorts 2020 and 2021

	(1) Baseline model	(2) + current situation	(3) + future worries
<i>Panel A: Mental Health Problems^b</i>			
Graduation cohort 2021 dummy	0.095***	0.025	0.059**
Enjoyment of current education		-0.090***	
Burden of distancing measures		0.023***	
Impact of distancing policies on future career			0.044**
Worries about occupational future			0.040***
Worries about too little career information			0.019**
R ²	0.233	0.267	0.251
<i>Panel B: Self-rated Health^a</i>			
Graduation cohort 2021 dummy	-0.058**	0.005	-0.028
Enjoyment of current education		0.187***	
Burden of distancing measures		-0.051**	
Impact of distancing policies on future career			-0.056**
Worries about occupational future			-0.065***
Worries about too little career information			-0.046*
R ²	0.147	0.178	0.160
<i>Panel C: Life Satisfaction^a</i>			
Graduation cohort 2021 dummy	-0.067***	0.016	-0.038+
Enjoyment of current education		0.273***	
Burden of distancing measures		-0.018	
Impact of distancing policies on future career			-0.045**
Worries about occupational future			-0.090***
Worries about too little career information			-0.030+
R ²	0.183	0.242	0.197
N persons	3,697	3,697	3,697

Note. ^a Standardized beta coefficients from OLS regressions. ^b Predicted probabilities from OLS regressions. Dependent variables: Life satisfaction (0–10), self-rated health (1–5), and Dummy for being above the clinical threshold for a high anxiety and depression risk (HSCL-10). Statistical significance at: * p < 0.10, ** p < 0.05, *** p < 0.01. Control variables: see note for Table 2. Additional controls: subjective household income and a dummy for an unemployed relative.

Data: BerO study wave 3.

3.4 Associations between mental health decreases and educational and career plans and transition outcomes

In this section, we analyse whether and to what extent the decrease in mental health that we observe from spring 2020 to fall/winter 2020/21 is related to students' educational and career plans and transition outcomes. We focus on mental health, for which we find the strongest decrease between spring 2020 and fall/winter 2020/21. As described in section 2.3, we regress educational and career plans and transition outcomes, measured in fall/winter 2020/21, on a dummy that takes a value of 1 if a student showed a strong decline in mental health (i.e., an increase on the HSCL-10 scale above a value of 0.4, which represents the upper quartile) from spring to fall/winter. The first model controls for student characteristics, while the second model additionally uses the panel dimension of our data and includes mental health values at wave 2 (i.e., the first measure of mental health available) and the dependent variable of the model, i.e., the educational and career plans at wave 1 (i.e., the first measure that is independent of the COVID-19 situation) as control variables. Models investigating transition outcomes include life satisfaction at wave 1 as a control. These two additional controls exclude the possibility that our estimates merely capture the effect of a student's generally low mental health and that those students with a decrease in mental health would have already stated less ambitious educational and career plans and dissatisfaction before the decrease.

Table 5 Associations between strong decreases in mental health and educational and career plans and transition outcomes in fall/winter 2020/21

Outcomes in fall/winter2020/21	Cohort 2020		Cohort 2021	
	(1) Base model	(2) + Base- line value of DV & SCL- 10	(3) Base model	(4) + Baseline value of DV & SCL-10
<i>Success probability¹</i>				
Strong HSCL-10 increase	-0.093**	-0.118***	-0.064**	-0.067***
R ²	0.269	0.308	0.172	0.271
N persons	1244	1244	2447	2447
<i>Security of educational path</i>				
Strong HSCL-10 increase	-	-	-0.038*	-0.043**
R ²	-	-	0.155	0.254
N persons	-	-	2449	2449
<i>Expected GPA</i>				
Strong HSCL-10 increase	-	-	0.050**	0.044**
R ²	-	-	0.523	0.674
N persons	-	-	2372	2372
<i>Probability of STEM Studies</i>				
Strong HSCL-10 increase	-	-	-0.035*	-0.020
R ²	-	-	0.286	0.684
N persons	-	-	2414	2414
<i>Satisfaction with decision²</i>				
Strong HSCL-10 increase	-0.133***	-0.158***	-	-
R ²	0.275	0.305	-	-
N persons	1242	1242	-	-
<i>Satisfaction with location²</i>				
Strong HSCL-10 increase	-0.081**	-0.089**	-	-
R ²	0.258	0.268	-	-
N persons	1177	1177	-	-
<i>Satisfaction with institution²</i>				
Strong HSCL-10 increase	-0.124***	-0.141***	-	-
R ²	0.256	0.282	-	-
N persons	1,181	1,181	-	-

Note. Standardized beta coefficients. Statistical significance at * $p < 0.10$, ** $p < 0.05$, *** $p < 0.01$.

Dependent variables: Success probability (11-point Likert scale), security of educational path (5-point Likert scale), probability of STEM Studies (11-point Likert scale), satisfaction measures (11-point Likert scale). GPA = Grade point average, a lower GPA indicates better performance.

Overall, 26.81% of the sample experienced a strong decrease in mental health (i.e., an increase on the HSCL-10 scale of more than 0.4 points). The share of strong decreases was stronger among the 2021 graduation cohort, in which 29.92% of respondents exhibited a strong decline in mental health, while the corresponding share from the 2020 graduation cohort was only 20.69%.

¹ Success probability for the 2020 cohort refers to the likelihood of finishing the current post-secondary education. For the 2021 cohort, success probability refers to the likelihood of successfully finishing a potential future academic study.

² The baseline value here constitutes overall happiness at survey wave 1.

Control Variables: graduation cohort (Panel A), gender, migration status, school performance, school fixed effects, parental education, subjective household income, parental unemployment in last 6 months, onsite education, self-efficacy, grit, time preference, risk aversion, Big Five personality traits, and interview mode.

Data: BerO study wave 3.

Table 5 shows the results of the estimations for the 2020 graduation cohort (Columns 1 and 2) and for the 2021 graduation cohort (Columns 3 and 4). In both cohorts, a strong decline in mental health was related to a lower success probability of the current educational path or potential future university education. The next rows present the results for educational and career plans among graduation cohort 2021 and transition outcomes among graduation cohort 2020. For the 2020 graduation cohort, the results reveal that students with a strong decline in mental health were less satisfied with their overall current educational decision, location, and institution than students with a lower or no decline. In line with these results, students with a strong decline in mental health in the 2021 graduation cohort stated that they felt less secure about their future career paths, they expected a worse GPA, and they had a lower probability of wanting to study STEM subjects. Columns 2 and 4 demonstrate that these results hold in the very tight specification, which includes lagged mental health and lagged educational and career plans and satisfaction. The sizes of the effects (between 10 and 15% of a standard deviation) are higher for the 2020 graduation cohort than for the 2021 cohort. However, for the 2021 cohort, the effects were in a relevant range, with approximately five percent of a standard deviation.

4 Discussion

We start the discussion of the results with the question of why school closures create positive effects on well-being in the short run. Our first explanation for the positive short-run effects refers to the idea that students perceived school closures as a relief, hence resembling additional holidays. This explanation is based on the finding that individuals show higher well-being on weekends and during holidays (e.g., Ryan et al. 2010), which in the case of students may be caused by the fact that studying provides less well-being than other leisure activities (Helliwell and Wang 2014). Additional studies have shown that if high school students are not at school, they feel less stressed because of reduced pressure and bullying, which in extreme cases even leads to less suicide during the holiday months (Hansen and Lang 2011, Kim and Leventhal 2008). The finding that students spend much fewer hours studying during school closures than in normal times supports the holiday explanation (Anger et al. 2020 for the present sample; Grewenig et al. 2020, Grätz and Lipps, 2021).

An alternative explanation for the increase in well-being might be that students perceive school closures as a measure to protect their health and relieve their fear of becoming infected with COVID-19. The finding that a positive effect exists for self-rated health and mental health

problems while the effect is absent for life satisfaction may support this explanation. In addition, the decline in self-rated health and life satisfaction from Wave 1 to Wave 2 for the students who answered the questionnaire before the school closures supports the health protection explanation, as the decline may have resulted from COVID-19 fear. Finally, self-rated health for students who answered the questionnaire after the school closures remained at the same level as in wave 1 and did not increase. In the case of a holiday effect, we may have expected an increase in this outcome. However, interview mode or honesty-in-reporting effects may also explain the decline in life satisfaction and self-rated health from wave 1 to wave 2, which supports the holiday explanation (Chadi 2013; Warren and Halpern-Manners 2012). Irrespective of the final explanation for the increase in well-being shortly after school closures, the findings are policy relevant, as they demonstrate that short-term school closures are not harmful to students' well-being.

Next, we discuss why well-being declines in the long run and why this decline is stronger for students who are still enrolled at high schools. In the longer run, the burdens of school closures and other distancing measures may accumulate because students suffer more from social distancing and home schooling and may be afraid about a loss of human capital. Additionally, students increasingly perceive the pandemic not only as a short-term event but also as a long-term condition. This is an important finding, as little research exists on how mental and physical well-being develops over the course of a pandemic. For example, Sachser et al. (2021) also found positive immediate effects of lockdown measures on mental health in a representative sample of the German population without assessing longer-term outcomes. Therefore, our results indicate that physical and mental well-being developed very dynamically during the pandemic and that measuring well-being at one point in time during the crisis may be misleading.

As an explanation for the observed long-run decline in well-being of the 2021 graduation cohort, we find that students who were still in school were more worried about their future careers and were more burdened by the current COVID measures, in particular school closures, social distancing measures during phases of reopened schools and perceived future career uncertainty. These differences in perceptions explain the difference in decline in overall well-being almost completely.

The existence and explanation of the difference between the cohorts are surprising, as the high school graduates from the 2020 cohort were also strongly affected by distancing measures, such as having online lectures or prohibitions of freshman events, and uncertainties about the

future. However, in the first post-high school year, individuals from graduation cohort 2020 appeared to cope better with the situation. One explanation for the finding may be that freshmen were not familiar with universities, vocational schools or training firms without distancing measures and that therefore, they did not miss anything, for example, on-site lectures. Additionally, after having made a successful transition, worries about the future may have decreased. Overall, the results indicate that students who were locked down while still in school are most vulnerable to an overall health decline, which should be considered in prevention or support services.

Finally, it is important that the decline in well-being is related to educational and career plans and satisfaction with the chosen educational path. As students from the 2020 graduation cohort with a decline in mental health reported less satisfaction with their choices, they were more likely to drop out, causing high costs for the individual and society. The same is true for the 2021 graduation cohort: our results, i.e., that these students feel less secure about the future, suggest that they will make decisions that they would have not made without the pandemic, which may also lead to high individual and societal costs.

Although our study has many strengths, it also shows limitations. First, we cannot state whether the COVID-19 distancing measures causally generated the decline in physical and mental well-being, as we do not have a control group who was not affected by the pandemic policies. However, it is very unlikely that only time, seasonal effects or any other event caused the reduction, as the decline was too strong for these explanations. Similar declines only occur in regions where a war started or in individuals who have experienced a stroke of fate, such as becoming widowed or disabled (Coupe and Obrizan 2016; Infurna et al. 2017; Oswald and Powdthavee 2008). Furthermore, other studies showed that well-being remained stable during the final years of high school education for cohorts who graduated before the pandemic (Herke et al. 2019). A second limitation might be that the size of the relation between the decline in mental health and the transition outcomes in our estimates is not very large. Nevertheless, the relations are meaningful because a rich body of research based on the notion of cumulative (dis)advantage shows that even small changes in this critical stage of the life course can have long-lasting accumulating effects over the life course (e.g., DiPrete and Eirich 2006).

5 Conclusion

This paper analyses how the COVID-19 pandemic and the related measures to reduce the spread of the coronavirus have affected the well-being, educational and career plans and transition outcomes of students from the 2020 and 2021 high school graduation cohorts. The results show that after an immediate increase in physical and mental health around the time of the first school closures, well-being strongly declined in the longer run, particularly for students in the 2021 graduation cohort, who were still in school at the time of the survey in fall/winter 2020/21. Additionally, our results demonstrate that this decline in well-being was related to several educational and career plans and transition outcomes. The results clearly indicate that the COVID-19 pandemic, including school closures and distancing measures, has had negative effects on current graduation cohorts, which may cause life-long harm. Next, disentangling the effects of different pandemic policies, e.g., school closures or shutdown of leisure activities, would be crucial to evaluate the true costs and benefits of such policies. This is particularly important; for example, the effectiveness of school closures as one of the main policies to prevent COVID-19 infections is disputed (Courtemanche et al. 2021; Isphording et al. 2021; van Bismarck-Osten et al. 2021). Finally, we address the question of intergenerational equality, as our findings demonstrate that the younger generation is likely to bear the long-term costs of pandemic policies, while the benefits of distancing measures in terms of lower infections are likely to be higher for older individuals.

References

- Anger, S., Bernhard, S., Dietrich, H., Lerche, A., Patzina, A., Sandner, M., & Toussaint, C. (2020). Schulschließungen wegen Corona: Regelmäßiger Kontakt zur Schule kann die schulischen Aktivitäten der Jugendlichen erhöhen, *LAB-Forum* vom 23. April.
- Aucejo, E. M., French, J., Araya, M. P. U., & Zafar, B. (2020). The impact of COVID-19 on student experiences and expectations: Evidence from a survey. *Journal of public economics*, *191*, 104271.
- Breen, R., & Goldthorpe, J. H. (1997). Explaining educational differentials: Towards a formal rational action theory. *Rationality and society*, *9*(3), 275-305.
- Cameron, S. V., & Heckman, J. J. (1998). Life cycle schooling and dynamic selection bias: Models and evidence for five cohorts of American males. *Journal of Political economy*, *106*(2), 262-333.
- Chadi, A. (2013). The role of interviewer encounters in panel responses on life satisfaction. *Economics Letters*, *121*(3), 550-554.
- Cornaglia, F., Crivellaro, E., & McNally, S. (2015). Mental health and education decisions. *Labour Economics*, *33*, 1-12.
- Coupe, T., & Obrizan, M. (2016). The impact of war on happiness: The case of Ukraine. *Journal of Economic Behavior & Organization*, *132*, 228-242.
- Courtemanche, C. J., Le, A. H., Yelowitz, A., & Zimmer, R. (2021). *School Reopenings, Mobility, and COVID-19 Spread: Evidence from Texas*. NBER Working Paper No. 28753
- Crawford, J., Butler-Henderson, K., Rudolph, J., Malkawi, B., Glowatz, M., Burton, R., Magni P. & Lam, S. (2020). COVID-19: 20 countries' higher education intra-period digital pedagogy responses. *Journal of Applied Learning & Teaching*, *3*(1), 1-2
- Derogatis, L. R., Lipman, R. S., Rickels, K., Uhlenhuth, E. H., & Covi, L. (1974). The Hopkins Symptom Checklist (HSCL): A self-report symptom inventory. *Behavioral science*, *19*(1), 1-15.
- Diener, E. (1984). Subjective Well-Being. *Psychological Bulletin*, *95*(3), 542-575.
- Diener, E., & Chan, M. Y. (2011). Happy people live longer: Subjective well-being contributes to health and longevity. *Applied Psychology: Health and Well-Being*, *3*(1), 1-43.

- DiPrete, T. A., & Eirich, G. M. (2006). Cumulative advantage as a mechanism for inequality: A review of theoretical and empirical developments. *Annual Review of Sociology*, 32, 271-297.
- Elmer, T., Mepham, K., & Stadtfeld, C. (2020). Students under lockdown: Comparisons of students' social networks and mental health before and during the COVID-19 crisis in Switzerland. *Plos one*, 15(7), e0236337
- Emery, R. L., Johnson, S. T., Simone, M., Loth, K. A., Berge, J. M., & Neumark-Sztainer, D. (2021). Understanding the impact of the COVID-19 pandemic on stress, mood, and substance use among young adults in the greater Minneapolis-St. Paul area: Findings from project EAT. *Social Science & Medicine*, 276, 113826.
- Fiaschi, D., & Tealdi, C. (2021). *Young People between Education and the Labour Market during the COVID-19 Pandemic in Italy*. IZA DP No. 14479.
- Giuntella, O., Hyde, K., Saccardo, S., & Sadoff, S. (2021). Lifestyle and mental health disruptions during COVID-19. *Proceedings of the National Academy of Sciences*, 118(9), e2016632118.
- Grewenig, E., Lergetporer, P., Werner, K., Woessmann, L., & Zierow, L. (2020, October). *COVID-19 and educational inequality: how school closures affect low-and high-achieving students*. CESifo Working Paper No.8648.
- Grätz, M., & Lipps, O. (2021). Large loss in studying time during the closure of schools in Switzerland in 2020. *Research in Social Stratification and Mobility*, 71, 100554.
- Haas, S. A. (2006). Health selection and the process of social stratification: the effect of childhood health on socioeconomic attainment. *Journal of health and social behavior*, 47(4), 339-354.
- Hale, J. M. (2017). Cognitive disparities: The impact of the great depression and cumulative inequality on later-life cognitive function. *Demography*, 54(6), 2125-2158.
- Hansen, B. and Lang, M. (2011). Back to school blues: Seasonality of youth suicide and the academic calendar. *Economics of Education Review*, 30(5),850–861.
- Haavet, O. R., Sirpal, M. K., Haugen, W., & Christensen, K. S. (2011). Diagnosis of depressed young people in primary health care—a validation of HSCL-10. *Family practice*, 28(2), 233-237.

- Helliwell, J. F., & Wang, S. (2014). Weekends and subjective well-being. *Social indicators research, 116*(2), 389-407.
- Herke, M., Rathmann, K., & Richter, M. (2019). Trajectories of students' well-being in secondary education in Germany and differences by social background. *European journal of public health, 29*(5), 960-965.
- Hisler, G. C., & Twenge, J. M. (2021). Sleep characteristics of US adults before and during the COVID-19 pandemic. *Social Science & Medicine, 276*, 113849.
- Infurna, F. J., Wiest, M., Gerstorf, D., Ram, N., Schupp, J., Wagner, G. G., & Heckhausen, J. (2017). Changes in life satisfaction when losing one's spouse: Individual differences in anticipation, reaction, adaptation and longevity in the German Socio-economic Panel Study (SOEP). *Ageing & Society, 37*(5), 899-934.
- Isphording, I. E., Lipfert, M., & Pestel, N. (2021). Does re-opening schools contribute to the spread of SARS-CoV-2? Evidence from staggered summer breaks in Germany. *Journal of Public Economics, 198*, 104426.
- Kim, Y. S., & Leventhal, B. (2008). Bullying and suicide. A review. *International journal of adolescent medicine and health, 20*(2), 133-154.
- Kopasker, D., Montagna, C., & Bender, K. A. (2019). Insecure Lock-In: The Mental Health Effects of Anticipating Insecure Employment.
- Leopold, L. (2018). Education and physical health trajectories in later life: a comparative study. *Demography, 55*(3), 901-927.
- Leykin, Y., Roberts, C. S., & DeRubeis, R. J. (2011). Decision-making and depressive symptomatology. *Cognitive therapy and research, 35*(4), 333-341.
- Liu, C. H., Zhang, E., Wong, G. T. F., & Hyun, S. (2020). Factors associated with depression, anxiety, and PTSD symptomatology during the COVID-19 pandemic: Clinical implications for US young adult mental health. *Psychiatry research, 290*, 113172.
- Lucas, R. E. (2007). Adaptation and the set-point model of subjective well-being: Does happiness change after major life events?. *Current directions in psychological science, 16*(2), 75-79.
- Mossey, J. M., & Shapiro, E. (1982). Self-rated health: a predictor of mortality among the elderly. *American journal of public health, 72*(8), 800-808.

- Oreopoulos, P. (2007). Do dropouts drop out too soon? Wealth, health and happiness from compulsory schooling. *Journal of public Economics*, 91(11-12), 2213-2229.
- Oswald, A. J., & Powdthavee, N. (2008). Does happiness adapt? A longitudinal study of disability with implications for economists and judges. *Journal of public economics*, 92(5-6), 1061-1077.
- Roepke, A. M., & Seligman, M. E. (2016). Depression and prospection. *British Journal of Clinical Psychology*, 55(1), 23-48.
- Ryan, R. M., Bernstein, J. H., & Brown, K. W. (2010). Weekends, work, and well-being: Psychological need satisfactions and day of the week effects on mood, vitality, and physical symptoms. *Journal of social and clinical psychology*, 29(1), 95-122.
- Sachser, C., Olaru, G., Pfeiffer, E., Brähler, E., Clemens, V., Rassenhofer, M., Witt A. & Fegert, J. M. (2021). The immediate impact of lockdown measures on mental health and couples' relationships during the COVID-19 pandemic-results of a representative population survey in Germany. *Social Science & Medicine*, 278, 113954.
- Sarra, A., Fontanella, L., & Di Zio, S. (2019). Identifying students at risk of academic failure within the educational data mining framework. *Social Indicators Research*, 146(1), 41-60.
- Schmalbach, B., Zenger, M., Tibubos, A. N., Kliem, S., Petrowski, K., & Brähler, E. (2021). Psychometric properties of two brief versions of the Hopkins Symptom Checklist: HSCL-5 and HSCL-10. *Assessment*, 28(2), 617-631.
- Shanahan, L., Steinhoff, A., Bechtiger, L., Murray, A. L., Nivette, A., Hepp, U., Ribeaud, D. & Eisner, M. (2020). Emotional distress in young adults during the COVID-19 pandemic: evidence of risk and resilience from a longitudinal cohort study. *Psychological medicine*, 1-10.
- Strand, B. H., Dalgard, O. S., Tambs, K., & Rognerud, M. (2003). Measuring the mental health status of the Norwegian population: a comparison of the instruments SCL-25, SCL-10, SCL-5 and MHI-5 (SF-36). *Nordic journal of psychiatry*, 57(2), 113-118.
- Tamborini, C. R., Kim, C., & Sakamoto, A. (2015). Education and lifetime earnings in the United States. *Demography*, 52(4), 1383-1407.

- van Bismarck-Osten, C., Borusyak, K., Schönberg, U. (2021). *The role of schools in transmission of the SARS-CoV-2 virus: Quasi-experimental evidence from Germany*. Forthcoming: *Economic Policy*
- Veenhoven R. (2012) Happiness: Also Known as “Life Satisfaction” and “Subjective Well-Being”. In: Land K., Michalos A., Sirgy M. (eds) *Handbook of Social Indicators and Quality of Life Research*. Springer, Dordrecht.
- Yükselen, I., Sandner, M., Jugenheimer, N., & Christoph, B. (2020). To work or not to work? Eine Fallstudie zur Beschäftigungsstruktur von Studierenden und möglichen Auswirkungen der Corona-Krise, *IAB-Forum* vom 16.Juli.
- Warren, J. R., & Halpern-Manners, A. (2012). Panel conditioning in longitudinal social science surveys. *Sociological Methods & Research*, 41(4), 491-534.
- Zheng, H. (2017). Why does college education matter? Unveiling the contributions of selection factors. *Social science research*, 68, 59-73.

Online Appendix

Table A1 Variables only included in fall/winter 2020/21

	Mean	Std. Dev.	Min	Max
Dealing with the current situation				
Fun with current education	3.276	1.068	1	5
Burden distancing measures	2.897	1.294	1	5
Worries about occupational future	2.200	1.190	1	5
Worries about too less career information	2.870	1.370	1	5
Distancing policies have impact on future career	0.563		0	1
Educational plans				
Security educational path	3.666	1.134	1	5
Success probability of finishing pot. study	8.258	1.803	0	10
Success probability of finishing current post-sec. edu.	7.742	2.272	0	10
Satisfaction with decision	7.407	2.055	0	10
Satisfaction with location	7.381	2.234	0	10
Satisfaction with institution	7.500	1.915	0	10
Strong increase in anxiety & depression risk	0.268		0	1
Control variables for wave 3-specific analyses				
Dummy for an unemployed relative	0.257		0	1
Subjective household income				
<i>1 much less than one needs for a decent life</i>	0.012			
2	0.062			
3	0.229			
4	0.549			
<i>5 much more than one needs for a decent life</i>	0.140			
Missing information on subj. HH income	0.009			
Onsite education at time of interview	0.641			
Baseline fun with education	3.300	0.890	1	5
Baseline security with educational path	3.545	1.135	1	5
Baseline success probability of finishing pot. study	8.517	1.573	0	10
Baseline happiness	7.435	1.940	0	10
N persons	3,697			

Table A2 Overall Sample characteristics

	Mean	Std. Dev.	Min	Max
Outcomes				
<i>Life satisfaction</i>	7.074	2.040	0	10
<i>Self-rated health</i>	3.763	1.065	1	5
<i>Anxiety & depression risk</i>	0.438	0.496	0	1
Socio-demographics				
Graduation cohort 2021	0.663		0	1
Male	0.355		0	1
1st/2nd generation migrants	0.209		0	1
Missing information on migration status	0.055			
At least one parent with university education	0.540		0	1
Missing information on parental education	0.105			
GPA better than 2.5	0.484		0	1
Missing information on GPA	0.010			
Personality and Preferences				
Self-efficacy	2.924	0.407	1	4
Grit	3.472	0.615	1	5
Dummy for being myopic	0.115		0	1
Openness	4.778	1.220	1	7
Conscientiousness	5.222	1.023	1.333	7
Extraversion	4.764	1.381	1	7
Agreeableness	5.434	0.947	1	7
Neuroticism	4.245	1.234	1	7
Risk aversion	5.629	2.183	0	10
Method				
CATI interview	0.024		0	1
Observations	11,091			
N persons	3,697			

Table A3 Development of mental and physical health since fall 2019. Results from random effect growth curve models

	Life-satisfaction	Self-rated Health	Anxiety & Depression risk
Spring 2020 (Pre SC)	-0.323*** (0.036)	-0.230*** (0.022)	
Spring 2020 (Post SC)	-0.362*** (0.045)	-0.006 (0.026)	-0.059*** (0.014)
Fall 2020	-0.808*** (0.037)	-0.199*** (0.021)	0.077*** (0.010)
Graduation cohort 2021	-0.053 (0.056)	-0.033 (0.029)	0.014 (0.014)
At least one parent with university education	0.147*** (0.056)	-0.015 (0.029)	-0.000 (0.014)
Missing information on parental education	0.109 (0.099)	-0.039 (0.049)	-0.018 (0.024)
CATI interview	0.861*** (0.099)	0.224*** (0.061)	-0.137*** (0.027)
Male	0.163*** (0.056)	0.214*** (0.028)	-0.185*** (0.015)
GPA better than 2.5	0.197*** (0.052)	0.040 (0.027)	-0.010 (0.014)
Missing information on GPA	-0.127 (0.272)	-0.092 (0.116)	-0.101* (0.060)
1st/2nd generation migrants	-0.222*** (0.069)	0.017 (0.034)	0.072*** (0.017)
Missing information on migration status	-0.361*** (0.124)	-0.016 (0.061)	0.087*** (0.030)
Dummy for being myopic	0.035 (0.081)	-0.027 (0.039)	0.006 (0.020)
Risk aversion	-0.042*** (0.014)	-0.022*** (0.007)	0.014*** (0.003)
Openness	-0.018 (0.022)	-0.049*** (0.011)	0.038*** (0.005)
Conscientiousness	0.002 (0.035)	0.041** (0.017)	0.005 (0.009)
Extraversion	0.121*** (0.021)	-0.004 (0.011)	0.000 (0.005)
Agreeableness	0.269*** (0.030)	0.079*** (0.014)	-0.032*** (0.007)
Neuroticism	-0.190*** (0.024)	-0.120*** (0.012)	0.083*** (0.006)
N observations	11091	11091	7394

Note: Robust standard errors in parentheses; Statistical significance: * $p < 0.10$, ** $p < 0.05$, *** $p < 0.01$.

School fixed effects are included. As mental health problems were not included in the first survey wave, the baseline value in the anxiety and depression model is “Spring 2020 (Pre SC)” and not “Fall 2019”.

Data Source: Bero-Study wave 1 to 3

Table A4 Development of mental and physical health by graduation. Results from random effect growth curve models

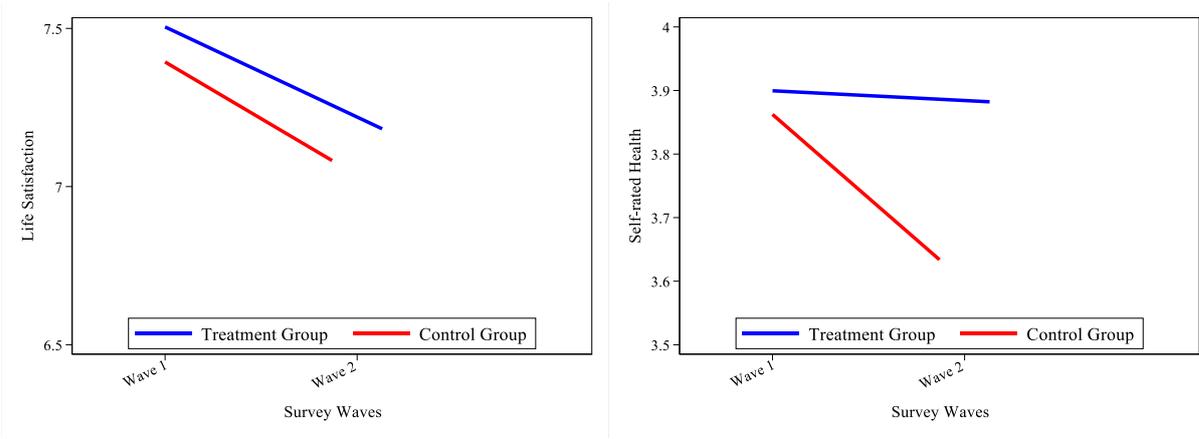
	Life-satisfaction	Self-rated Health	Anxiety & depression risk
Spring 2020 (Pre SC)	-0.250*** (0.061)	-0.188*** (0.038)	
Spring 2020 (Post SC)	-0.363*** (0.080)	0.023 (0.047)	-0.041* (0.025)
Fall 2020	-0.657*** (0.066)	-0.097*** (0.036)	0.008 (0.017)
Graduation cohort 2021	0.047 (0.066)	0.037 (0.037)	-0.033* (0.020)
Spring 2020 (Pre SC) x Graduation cohort 2021	-0.111 (0.076)	-0.063 (0.047)	
Spring 2020 (Post SC) x Graduation cohort 2021	-0.002 (0.096)	-0.045 (0.056)	-0.024 (0.030)
Fall 2020 x Graduation cohort 2021	-0.228*** (0.079)	-0.155*** (0.044)	0.104*** (0.021)
At least one parent with university education	0.146*** (0.056)	-0.015 (0.029)	-0.000 (0.014)
Missing information on parental education	0.110 (0.099)	-0.039 (0.049)	-0.018 (0.024)
CATI interview	0.865*** (0.099)	0.226*** (0.061)	-0.139*** (0.027)
Male	0.163*** (0.056)	0.214*** (0.028)	-0.185*** (0.015)
GPA better than 2.5	0.198*** (0.052)	0.040 (0.027)	-0.010 (0.014)
Missing information on GPA	-0.129 (0.272)	-0.092 (0.116)	-0.100* (0.060)
1st/2nd generation migrants	-0.221*** (0.069)	0.017 (0.034)	0.072*** (0.017)
Missing information on migration status	-0.362*** (0.124)	-0.016 (0.061)	0.087*** (0.030)
Self-efficacy	0.930*** (0.080)	0.305*** (0.039)	-0.200*** (0.019)
Grit	0.301*** (0.060)	-0.004 (0.029)	-0.027* (0.015)
Dummy for being myopic	0.035 (0.081)	-0.027 (0.039)	0.006 (0.020)
Risk aversion	-0.042*** (0.014)	-0.022*** (0.007)	0.014*** (0.003)
Openness	-0.017 (0.022)	-0.049*** (0.011)	0.038*** (0.005)
Conscientiousness	0.002 (0.035)	0.041** (0.017)	0.005 (0.009)
Extraversion	0.121*** (0.021)	-0.004 (0.011)	0.000 (0.005)
Agreeableness	0.269*** (0.030)	0.079*** (0.014)	-0.032*** (0.007)
Neuroticism	-0.190*** (0.024)	-0.120*** (0.012)	0.083*** (0.006)
Constant	2.630*** (0.408)	3.265*** (0.210)	0.700*** (0.105)
N observations	11091	11091	7394

Note: Robust standard errors in parentheses; Statistical significance: * $p < 0.10$, ** $p < 0.05$, *** $p < 0.01$.

School fixed effects are included. As mental health problems were not included in the first survey wave, the baseline value in the anxiety and depression model is “Spring 2020 (Pre SC)” and not “Fall 2019”.

Data Source: Bero-Study wave 1 to 3

Figure A1 Immediate effects of school closures on mental and physical health: Results from difference-in-difference regressions



Comparison of analytical sample (truncated by transition) and untruncated sample

Figure A2 Immediate effects of school closures on mental and physical health: Results from difference-in-difference regressions. (Left hand side shows results from the manuscript. Right hand side shows results based on an untruncated sample, i.e., with individuals in gap years.)

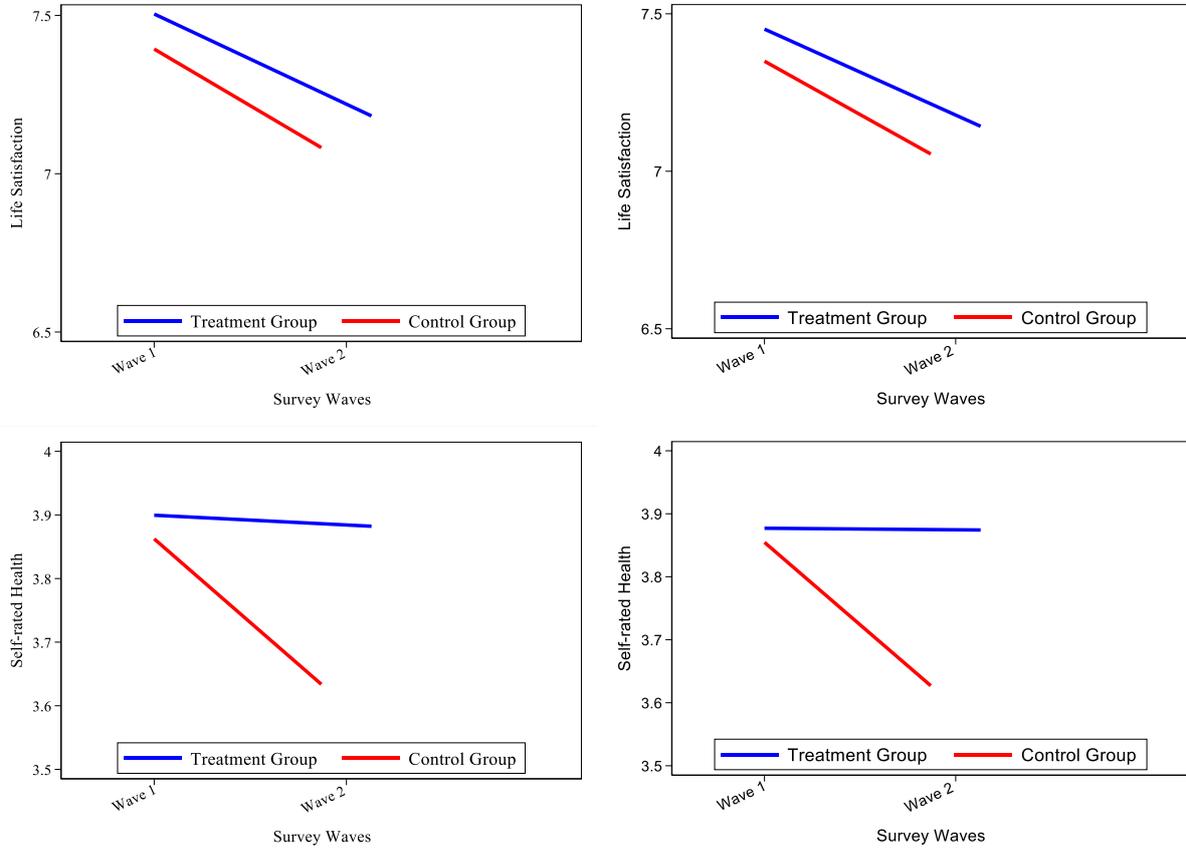


Figure A3 Development of mental and physical well-being since fall 2019. Results from random effect growth curve models. (Left hand side shows results from the manuscript. Right hand side shows results based on an untruncated sample, i.e., with individuals in gap years.)

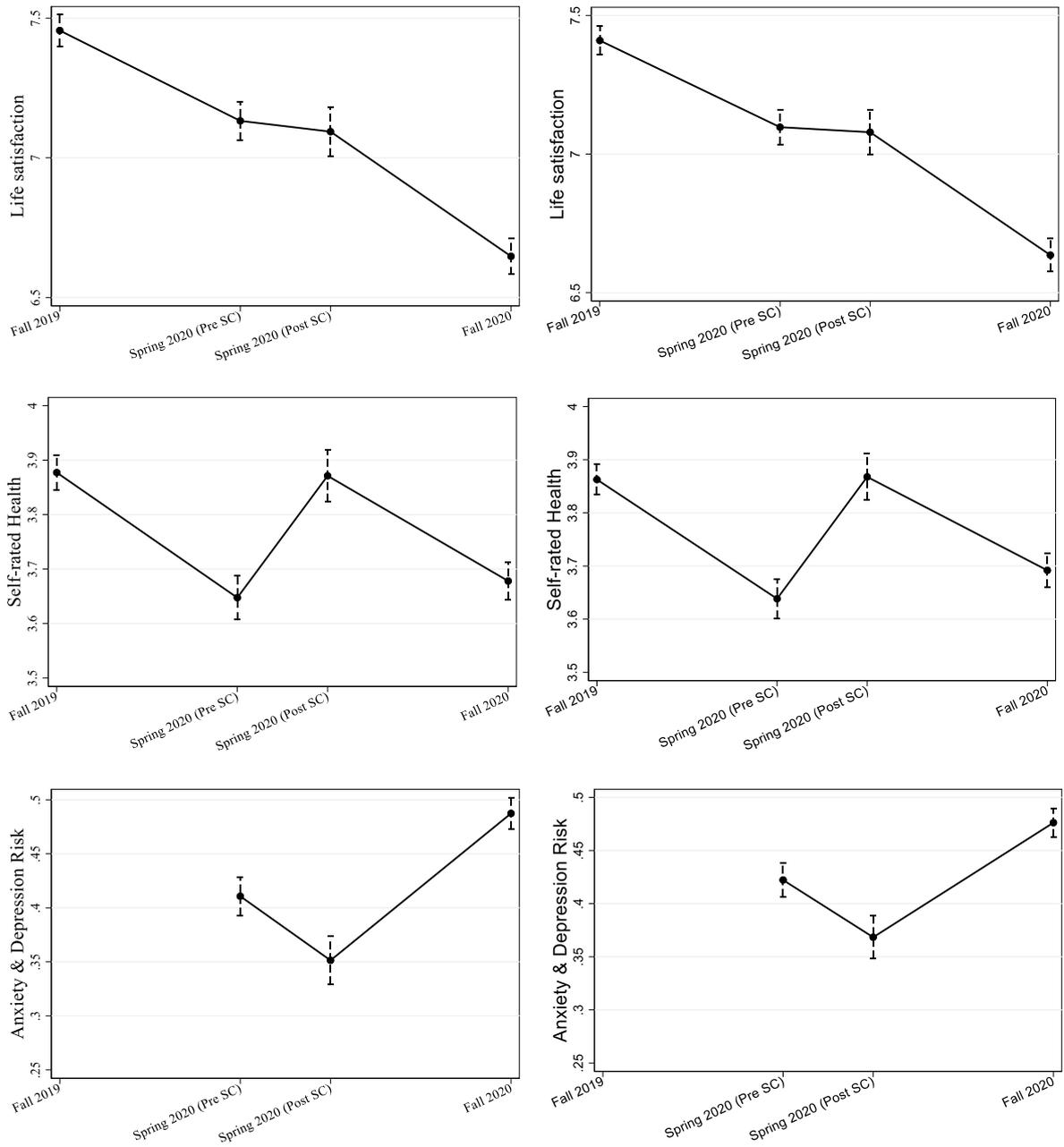


Figure A4 Development of mental and physical well-being by graduation cohort. Results from random effect growth curve models. (Left hand side shows results from the manuscript. Right hand side shows results based on an untruncated sample, i.e., with individuals in gap years.)

