



YOUTH AND COVID-19

IMPACTS ON JOBS, EDUCATION, RIGHTS AND MENTAL WELL-BEING

TECHNICAL ANNEX



Global Survey on Youth and Covid-19

Technical Annex

The Global Survey on Youth and COVID-19 (the Survey), was conducted by partners of the Global Initiative on Decent Job, including the International Labour Organization, the United Nations Major Group for Children and Youth, AIESEC, the European Youth Forum, the European Union Emergency Trust Fund for Africa and the Office of the United Nations High Commissioner for Human Rights. The online survey, available in 23 languages, was conducted between 21 April and 21 May 2020 and taken by 12,605 individuals aged 18–34 years from 112 countries.¹

This technical annex describes the survey promotion in Section I, the dataset and key outcome variables in Section II, explains the survey weighting methodology and presents alternative approaches in Section III, examines the representativeness of the survey sample with regard to survey respondents' educational background and income in Section IV, and presents the results of robustness checks in Section V.

I. Survey Promotion

The Survey was translated and available in 23 languages including Arabic, Bengali, Chinese, Filipino, French, German, Hausa, Hindi, Indonesian, Italian, Japanese, Korean, Polish, Portuguese, Romanian, Russian, Sinhala, Spanish, Swahili, Thai, Turkish and Vietnamese. Translations were done by volunteers and each language version was pilot tested. The survey platform SurveyMonkey was used to conduct the Survey. Dissemination and promotion of the Survey was first done through social media and then through direct emailing of young people who were identified to be amongst the networks of the five survey partners.

II. The Dataset

The micro-survey data resulting from individual responses are merged with additional macro-level information on (i) ILO regions (Africa, Americas, Arab States, Asia and the Pacific, Europe and Central Asia)², (ii) World Bank income groups (low, lower-middle, upper-middle, high income countries)³ as well as (iii) UN Operational rates of exchange⁴ to convert self-declared gross monthly income (provided in the local currency) to US Dollar.

As shown in Table A-1 the survey included modules on employment, education and training, mental well-being, rights, social activism, policy responses and personal information. In total, the survey contained 44 questions. Survey respondents completed the survey on

¹ Only respondents at least 18 years old were included in the final dataset as for youth below the age of 18 parental consent could not be assured through the online survey. Moreover, for 416 respondents weights could not be computed based on the country, gender and age combination of the respondent (Section II for details on how weight construction). Moreover, only observations from countries with at least 10 respondents were retained (affecting 211 observations).

² Regional breakdown of ILO Member States is available at: <https://www.ilo.org/global/regions/lang--en/index.htm>

³ The most recent 2020 dataset was used and is available at: <https://datahelpdesk.worldbank.org/knowledgebase/articles/906519-world-bank-country-and-lending-groups>

⁴ Rates valid on 1 May 2020 were used and are available at: <https://treasury.un.org/operationalrates/OperationalRates.php>

average in 10 minutes and the completion rate was 65 per cent. Survey respondents are asked to indicate their “current situation” with a choice between (i) “working”, (ii) “studying”, (iii) “working and studying”, (iv) “not working or studying”. Depending on their current situation, survey respondents completed either the employment (1a) and/or education and training (1b) or the ‘not working and job loss’ (1c) module. All respondents were then directed to complete modules 2 to 6.

Description of selected outcome variables:

- **Stopped working:** This is a dummy variable that takes the value of one for all respondents who (i) declared to be “not working or studying” AND further affirmed to have lost their previous job since the start of the coronavirus outbreak; (ii) declared to be working⁵ AND further declared to be working at least one hour on a typical day *before the outbreak* AND reported to be working zero hours on a typical day *during the outbreak*. The dummy takes the value of zero for all respondents who declared to be working at least one hour on a typical day during the outbreak.
- **Hours worked:** respondents who indicated to be working were asked to numerically input both their daily hours worked before the outbreak started and during the outbreak.
- **Occupation:** Survey participants who indicated to be working or who previously worked were asked about their occupation (open-ended question). Of the 4,818 answers that were received,⁶ 3,646 could be manually matched to the International Standard Classification of Occupations (ISCO-08)⁷ at the two-digit level.⁸
- **Mental wellbeing:** the survey featured a module with the Short Warwick Edinburgh Mental Wellbeing Scale (SWEMWBS).⁹ The SWEMWBS consists of seven items that each contain one positively worded statement, with five response categories from ‘none of the time’ to ‘all of the time’. Answers are scored and summed up, resulting in an index with scores from 7 to 35 with higher scores indicating better mental wellbeing. The raw index is converted into a metric score following guidance from the Warwick Medical School.¹⁰

Table A-1: Overview of the survey structure

Module	Taken by
Introduction (country of residence, gender, age, main activity)	All survey respondents
1a Employment	Current situation: working OR working & studying
1b Education and training	Current situation: studying OR working & studying

⁵ Or “working and studying”.

⁶ Out of 6,259 respondents who were asked either about their current or most recent occupation.

⁷ <https://www.ilo.org/public/english/bureau/stat/isco/isco08/>

⁸ The analysis presented in the main report is limited to a breakdown at the one-digit level.

⁹ “Short Warwick Edinburgh Mental Wellbeing Scale (SWEMWBS) © NHS Health Scotland, University of Warwick and University of Edinburgh, 2008, all rights reserved.”

¹⁰ The conversion table is available at:

https://warwick.ac.uk/fac/sci/med/research/platform/wemwbs/using/howto/swemwbs_raw_score_to_metric_score_conversion_table.pdf

1c	Not working and job loss	Current situation: neither working nor studying
2	Mental wellbeing	All survey respondents
3	Social activism	All survey respondents
4	Rights	All survey respondents
5	Policy responses	All survey respondents
6	Personal information	All survey respondents

III. Weighting methods

While surveys seek to recruit representative samples of the underlying population, it is common that in the process of sampling, biases arise. According to Fotini, Evangelia and Michail (2013)¹¹, three potential sampling biases might be present in any survey: (i) bias from the under-/over-representation of some subgroups of the population, (ii) non-response bias when these are not random, and (iii) population size bias when comparing two or more countries. Biases can be corrected through the following weighting systems:¹²

- Non-response weighting: in voluntary surveys, such as the Global Survey on Youth and COVID-19, one of the major threats for the accuracy of the estimates is non-responsiveness by surveyed individuals. This leads to bias if the non-respondents are a non-random sample of the underlying population. In this case, non-response weighting is used to compensate for the fact that persons with certain (known) characteristics are not as likely to respond to the survey.¹³ Choosing which variables to use in the weighting process is crucial to correct successfully for the biased sample.¹⁴
- Population size weighting is used when examining a combination of survey data from two or more countries. It assigns the surveyed sample from each country a weight proportional to the population of the respective country. Population weights are the same for all survey participants within a country, but differ across countries.¹⁵

The weighting methodology adopted in the analysis of the sample of the Survey combines elements of non-response weighting and population size weighting. To correct for non-response bias, we weight the population by sex (women, men) and age groups (18-29, 30-34). To correct for population biases we allow weights to be a function of the youth population of each country.

Weights are constructed in three steps:

- 1) **Identification of the relevant population**: The relevant population is individuals between the age of 18 and 34 living in the 112 countries represented in the survey.

¹¹ Fotini, T., Evangelia, V., & Michail, V. (2013): Weighting of responses in the Consumer Survey : Alternative approaches – Effects on variance and tracking performance of the Consumer Confidence Indicator Table of Contents. Foundation for Economic & Industrial Research.

¹² In addition, *design weighting* can be used to correct for under-/over- representation of certain subgroups in the sample compared to the population. This is typically adapted when some sub-groups are deliberately over-/under-sampled, a case that does not apply here.

¹³ Fotini, Evangelia and Michail (2013), see footnote 9.

¹⁴ Pew Research Center (2018): For Weighting Online Opt-In Samples, What Matters Most?

¹⁵ European Social Survey. (2014): Weighting European Social Survey Data. Available at: https://www.europeansocialsurvey.org/docs/methodology/ESS_weighting_data_1.pdf

In the final, unweighted sample, 54 per cent of those 18 to 29 old have attained tertiary education (see Table A-5). Of those young people without tertiary education, 85 per cent are studying. Within the 30-34 age cohort, 89 per cent of respondents report having attained tertiary education.¹⁶ Given the survey is overwhelmingly representing individuals who either already attained a tertiary degree or are currently pursuing one, we narrow down the sub-population to those youth who attained an advanced degree. It is important to note that not all young people in our sample have already attained tertiary level education, yet this sub-population appears as the best available proxy for the youth population captured by the sample.¹⁷

- 2) **Obtaining population weights from the ILO Department of Statistics (ILOSTAT):** As the data source, we use the ILOSTAT dataset on (youth) working-age population by sex, age and education.¹⁸ We construct population weights for four population bins, dividing the survey respondents by age group (18-29; 30-34) and sex (women, men). We use ILOSTAT datasets on (youth) working-age population by sex and age, as well as (youth) working-age population by sex, age and education.
- For the 18-29 age bin (by sex) we use data on youth working-age population for those with an advanced degree. ILOSTAT only provides data for the 15-29 cohort which is used as an approximation.¹⁹ For 28 countries the 15-29 cohort is not available. In these cases we use the two 10 year cohorts 15-24 and 30-34 and take the sum of the 15-24 cohort plus 50 per cent of the 25-34 cohort as an approximation.²⁰
 - For the 30-34 age bin (by sex) we use data on working-age population for those with an advanced degree. ILOSTAT provides data for the 25-34 cohort as well as the 25-29 age cohort. We subtract the two to obtain an estimate of the 30-34 age bracket.
 - For 18 countries, data on working-age population (by 5 and 10 year age brackets) is available but no breakdown by level of education. We impute population data by modelling the share of the population with a tertiary degree for the 15-29 age cohort and the 30-34 cohort for women and men as an average of all countries with available data in the same ILO region and same World Bank country income group.
 - *The above procedure allows us to calculate population weights for two age-cohorts (18-29, 30-34) and by sex (women, men) for 112 countries covering 98.8 per cent of all survey respondents.*²¹

- 3) **Calculation of weights:** three sets of weights are constructed

¹⁶ Among the 11 per cent of respondents in the 30-34 year old age group a significantly lower share, 29 percent, are still studying (or studying while also working).

¹⁷ Because we do not use educational background as a weighting variable (also due to a sizable non-response rate), the main reason to condition youth population to those with advanced education is to avoid imbalances where (well educated) respondents from countries with large youth population attain an outside weight.

¹⁸ Available at: <https://ilostat.ilo.org/>

¹⁹ This is unproblematic: the number of young people 15 to 17 holding an advanced degree is presumably negligible. Furthermore, we are not computing weights for sub-age groups within the 18-29 groups.

²⁰ This assumes that the population is uniformly distributed over the 10 years of the bin. Under the same assumption, the age bin 30-34 can be calculated as the remaining 50% of the 25-34 bin

²¹ 152 observations need to be dropped because no weights could be calculated.

- a) **Sub-regional weights (used throughout the analysis)** correct for gender differences between the survey sample and the population of every country. Besides, the combined weight of responses from a given sub-region corresponds exactly to the share of the relevant youth population in that sub-region relative to the global youth population. As sub-regions, we chose combinations of the five ILO geographical regions and four World Bank income groups (thus, each country is assigned to one of the 20 sub-regions). The weight of individual i who is part of bin b (four bins: 18-29, 30-34, by sex) from sub-region s is calculated as the population in the same bin and sub-region divided by the number of observation in the sample in the same bin and sub-region:

$$subregion_weight_{ibs} = \frac{population_{bs}}{N_{bs}} \quad (1)$$

- b) **Country-weights (used for robustness checks, see Section IV.)** correct for gender differences between the survey sample and the population of every country. Moreover, the combined weight of responses from a given country corresponds exactly to the share of the relevant youth population in that country relative to the global youth population. The weight of individual i who is part of bin b (four bins: 18-29, 30-34, by sex) from country c is calculated as the population in the same bin and country divided by the number of observation in the sample in the same bin and country:

$$country_weight_{ibc} = \frac{population_{bc}}{N_{bc}} \quad (2)$$

- c) **Within-country weights (used for robustness checks, see section IV.)** correct for differences in the share of women/men at the country level but not for differences in response rates across countries. The weight of individual i who is part of bin b (four bins: 18-29, 30-34, by sex) from country c is calculated as the population in the same bin and country divided by the product of the total relevant population in that country (age 18-34) and the number of observation in the sample in the same bin and country:

$$withincountry_weight_{ibc} = \frac{population_{bc}}{population_c * N_{bc}} \quad (3)$$

Throughout the analysis, sub-regional weights, equation (1), are used. This is because in the survey sample there are large differences in response rates between countries. The trade-off in the analysis is that including countries with relatively few observations, on the one hand, improves the representativeness at the global level as more countries are included, but on the other hand, introduces increased uncertainty as observations from countries with few participants will receive a relatively large weight.

Sub-regional weights are constructed to address this trade off. Aggregating responses and weights at the sub-regional level reduces considerably the variance of the resulting weights compared to simply using country-level weights (equation (2)).²²

IV. Sample representativeness

The section aims to assess the representativeness of the Global Survey on Youth and COVID-19, detailing educational and income profiles of survey respondents. The survey includes observations from 112 countries. In these countries there are altogether 1.47 billion young people (15 to 29 years old) which represent around 92 per cent of the world's youth population (based on 185 countries for which population data is available).²³

Table A-2 presents descriptive statistics on self-reported monthly income before tax in US dollars for a total of 2,111 survey respondents.²⁴ As is common for income data, the distribution is skewed to the right, with an average of USD 1,038/month and a median of USD 444/month.²⁵ Average and median income correlate with the country income group classification. Table A-2 also lists the mean and median GDP per capita (per month) for the same sample. Importantly, GDP per capita and personal gross income cannot be directly compared.²⁶ However it is insightful to observe that the ratio of GDP (column 2) to income (column 1) decreases as the income level of countries rises. This can be seen as an indication that the survey reaches *relatively* more affluent youth in low- and middle-income countries compared to high-income countries.

²² Sub-regional weights are thus meant to reduce the overall volatility of estimates. Indeed, in the sample at hand the ratio between the observation with lowest weight and the observation with the highest weight is 1:18 (standard deviation: 19.3) while when using weights at country level the ratio is 1:9760 (standard deviation 37.2)

²³ We download the data on Working-age population by sex and age (in thousands), from the ILOSTAT explorer. For Jamaica, working-age population is only available for two cohorts 15-24 and 25-34. Then, in this case, we imputed the working-age population for the cohort 15-29 by assuming a uniformly distributed population across age, also within gender. Source:

https://www.ilo.org/shinyapps/bulkexplorer47/?lang=en&segment=indicator&id=POP_XWAP_SEX_AGE_NB_A

²⁴ The analysis is based on the question: "What is your monthly income, before tax, in your country's currency?". In total, 2,111 out of 4,861 (43.4 per cent) of survey respondents aged 20 to 29 who completed the personal information section also provided data on their monthly income. Underreporting might vary across education level, so results should be regarded as indicative only.

²⁵ Huge variations in monthly income across countries and regions suggest that measurement errors might be significant – median values should be more reliable than means.

²⁶ Among other factors, GDP captures all income generated by economic activity in a country, including for example, capital income.

Table A-2: Gross income per month (Global Survey) vs. GDP per capita per month

	(1) Gross income p. month		(2) GDP p. capita p. month		N
	Mean	Median	Mean	Median	
Global	1,038	444	1,175	379	2,111
<i>Country income group</i>					
Low income	292	207	66	75	52
Lower-middle income	645	330	225	179	439
Upper-middle income	680	332	596	578	821
High income	2,119	1,407	3,386	3,528	799

Table A-3 compares the educational profile of the survey to that of included countries for the youth cohort aged 20 to 29.²⁷ The first column lists the distribution by education level using ILOSTAT data for 94 of the 112 countries represented in the survey²⁸ which is compared to the self-reported level of education of survey respondents. Survey respondents are considerably more likely to have attained tertiary level education.

Table A-3: Educational profile, youth population (age: 20-29) (%)

	(1) Youth population*	(2) Global Survey on Youth and COVID-19 (weighted)
None	12	0
Primary	30	2
Secondary	36	22
Tertiary	21	76

Notes: * 94 countries out of the 112 in the survey sample, data: ILOSTAT. Sub-regional weights apply (see Section II).

Next, table A-4 depicts the share of the youth-to-adult population with advanced education both for the general population (aged 20 – 29) and the survey sample. Overall, 21 per cent of the youth population (aged 20-29), and 24 per cent of those aged 30-34, have attained an advanced education level. Among the (weighted) survey population 76

²⁷ For the purpose of this section we limit the sample to youth aged 20 to 29 years as we can reliably calculate the share for this age range but not for the 18-19 age range (as only data on 15-19 is available).

²⁸ Data on youth working-age population by sex, age, and education (in thousands) are obtained for a total of 143 countries, and matched with the list of 112 countries included in the survey, resulting in 94 countries for which data are available.

per cent (age group: 20-29) to 90 per cent (30-34) have attained tertiary education.²⁹ For the general youth population, a much greater share of young people (aged 20-29) in high-income countries has attained tertiary education (43 per cent) than, for example, in lower middle-income countries (16 per cent). This pattern does not hold for the survey population. In fact, among the survey population young people from lower middle-income countries are most likely to have attained tertiary level education (86 per cent). This is a further indication that in low- and middle-income countries the survey reaches those *relatively* more educated and privileged compared to in high-income countries.

Overall, these results show that young people (age group: 20-29) in the survey sample are around three times as likely to have attained tertiary education as compared to the general youth population of the same age bracket. In high-income countries, young people with tertiary education are over-represented at a rate of two to one. This is because in these countries the overall level of education among the youth population is higher and because the survey respondents in high-income countries show a more diverse educational profile.

Table A-4: Share of youth-to-adults with advanced education:

	General youth population (%)		Global Survey on Youth and COVID-19 (weighted) (%)	
	20-29*	30-34	20-29	30-34
Global	21	24	76	90
<i>Country income groups</i>				
Low-income	5	5	68	87
Lower middle-income	16	15	86	98
Upper middle-income	25	26	77	94
High-income	38	47	80	89

Notes: the age-cohort 20-29 is constructed with 93 out of the 112 included in the survey for which population data is provided by ILOSTAT.

V. Robustness checks

This section presents the results of two different robustness tests. First, we test the sensitivity of the results to changes in the sample – leaving out observations from one country at a time. Second, we compare results when using different weights (as described in Section III). The robustness checks are examining three key outcome variables: the share

²⁹ These descriptive measures were constructed by survey weights (see Section III). This is also why the weighted sample shows a higher share of young people with tertiary education compared to the unweighted sample (see Table A-5). Generally young people from countries that are under-represented in the survey report on average a higher level of education.

of survey respondents who (i) stopped working since the onset of the pandemic; (ii) reported that their education will be delayed or might fail and (iii) who were assessed to be probably affected by anxiety or depression.

The first test implies performing the so-called “jackknife procedure”. The method is typically used in statistical regressions but can easily be applied to the framework of the survey. Intuitively, the procedure re-calculates results – here the average of a weighted variable in the sample – leaving observations from one of the 112 countries in the sample out at a time. This tests shows whether the results are primarily driven by one country or a group of countries.³⁰

The second robustness check compares global estimates for the three different weighting approaches described in Section II: (i) the sub-regional weights (used throughout the report), (ii) country-weights and (iii) within-country weights. Country-weights correct for differences in the shares between women and men at the country level and would be the preferred approach in a survey where each country is represented by roughly a similar number of observations (which is not the case in this survey). The within-country weights only correct for within-country gender imbalances but do not correct for different population sizes. Results from these weights can be regarded as baseline estimates (see also table A-6 for an overview on survey participation per country).

The results of the jackknife procedure can be shown in many different ways. Figure A-1 presents them as boxplots, which allows identifying outliers.

Each boxplot shows the same descriptive statistics and can be interpreted similarly:

- (1) The average global estimate is represented by a red diamond – in the middle of each boxplot
- (2) Each circle represents the re-calculated global estimate leaving aside one country at the time. For better readability, only countries at the maximum and minimum are labelled.
- (3) The horizontal lines represent the median (long horizontal line in the middle of each boxplot), the 25th percentile, and the 75th percentile (shorter horizontal lines above and below the median line).

The first column of Panel A in figure A-1 depicts the results of applying the jackknife procedure to the outcome variable “Stopped working” using the preferred weight estimation system (sub-regional weights). There are two clear outliers: China and Sri Lanka. It implies that leaving observations from China aside, the global average for “stopped working” is estimated to be around 18.3 per cent. The opposite is true in the case of Sri Lanka: without the observations from the country, the global average is estimated to be around 15.8 per cent. However, we observe that results when leaving out one of the other 110 countries cluster tightly around the global average of 17.4 per cent (that considers all 112 countries). Panel B and C lead to similar conclusions: leaving one country aside at the time never changes the global estimate by more than 3 percentage points (Panel B), and 1.5 percentage points (Panel C). There are four notable outliers in Panel C – all above the

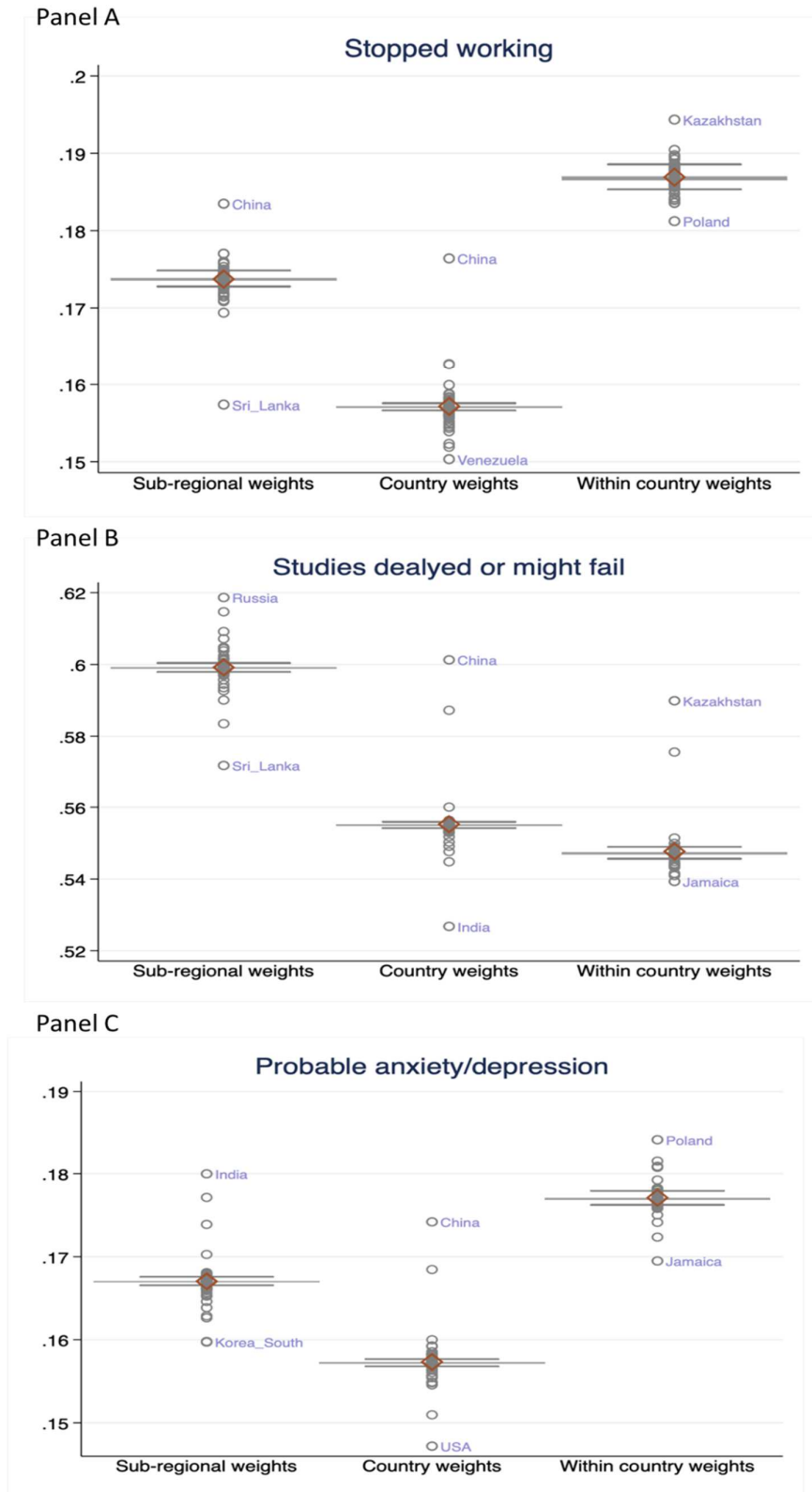
³⁰ For a more detailed explanation in this procedure see Nisbet R., Miner G., Yale, K. (2018): Chapter 11 - Model Evaluation and Enhancement, Handbook of Statistical Analysis and Data Mining Applications (Second Edition), Academic Press, Pages 215-233.

global estimate suggesting that without those four countries, the global average would be around 5 percentage points higher.

To assess the robustness of our weighting system we compare the global averages (red diamonds) in the three boxplots of each panel. Global estimates vary somewhat by weighting system while not changing results in a qualitatively meaningful manner. As expected outliers in the second column (country-level weights) mostly represent countries with large youth populations while outliers in the third column (within-country weights) represent countries with a large number of respondents in the survey (see also table A-6). Overall, the distribution of country-weights is more spread than for regional-groupings highlighting the sensitivity of this alternative weighting approach.

Overall, survey results appear robust to a variety of sensitivity checks, including different weighting approaches.

Figure A-1: Robustness tests (jackknife procedure and three different weighting methods) for selected outcome variables, (%)



Note: Robustness tests are conducted for the youth sample (age range: 18 to 29) of the survey. Scales in all three panels represent population shares from 0 to 1, corresponding to 0 to 100 per cent.

Table A-5: Sample characteristics: weighted vs. unweighted sample

		Weighted sample				Unweighted (raw) sample			
		18-29		30-34		18-29		30-34	
Category		No.	%	No.	%	No.	%	No.	%
Labour market status	Working	3,170	28.5	1,027	69.6	2,705	24.3	971	65.8
	Studying	5,028	45.2	109	7.4	5,729	51.5	115	7.8
	Study and work	1,775	15.9	230	15.6	1,645	14.8	229	15.5
	NEET	1157	10.4	109	7.4	1051	9.4	160	10.8
Gender	Female	5,958	53.5	788	53.4	7,136	64.1	874	59.3
	Male	5,172	46.5	687	46.6	3,994	35.9	601	40.7
Age group	18-24	7,354	66.1			8,071	72.5		
	25-29	3,776	33.9			3,059	27.5		
	30-34			1,475	100.0			1,475	100.0
Region	Africa	765	6.9	78	5.3	785	7.1	298	20.2
	Americas	2,045	18.4	320	21.7	2,070	18.6	314	21.3
	Arab States	141	1.3	10	0.7	178	1.6	29	2.0
	Asia and the Pacific	6,018	54.1	734	49.7	2,131	19.1	285	19.3
	Europe and Central Asia	2,161	19.4	334	22.6	5,966	53.6	549	37.2
Country income group	Low income	149	1.3	15	1.0	244	2.2	86	5.8
	Lower middle income	4,076	36.6	381	25.8	1,966	17.7	412	27.9
	Upper middle income	4,142	37.2	620	42.0	4,553	40.9	558	37.8

	High income	2,764	24.8	459	31.1	4,367	39.2	419	28.4
	Total	11,130	100.0	1,475	100.0	11,130	100.0	1,475	100.0
Marital status	Single	5,783	80.0	405	43.4	5,312	73.5	412	44.1
	Married/Have a partner	1068	14.8	493	52.7	1258	17.4	482	51.6
	Prefer not to say	379	5.2	37	3.9	660	9.1	41	4.4
Children	No	6,797	94.0	632	67.5	6,782	93.8	600	64.2
	Yes	433	6.0	303	32.5	448	6.2	335	35.8
Area	Urban	4,277	59.2	614	65.6	4,059	56.1	623	66.6
	Suburban	1,574	21.8	217	23.2	1,317	18.2	193	20.6
	Rural	1379	19.1	104	11.2	1854	25.6	119	12.7
Highest education level attained	None	64	0.9	8	0.8	131	1.8	8	0.9
	Primary	348	4.8	12	1.2	1056	14.6	8	0.9
	Secondary	2,057	28.5	78	8.4	2,178	30.1	83	8.9
	Tertiary	4,761	65.8	837	89.6	3,865	53.5	836	89.4
Identity	Minority	1461	20.2	172	18.4	1130	15.6	173	18.5
	Refugee and migrant	195	2.7	32	3.4	218	3.0	45	4.8
	Person with disability	168	2.3	36	3.8	156	2.2	27	2.9
	LGBTI	491	6.8	46	4.9	466	6.4	42	4.5
	Total	7,230	100.0	935	100.0	7,230	100.0	935	100.0

Table A-6: Observation per country (unweighted sample)

Country	18-29	30-34	Country	18-29	30-34	Country	18-29	30-34	Country	18-29	30-34
Afghanistan	25	5	Croatia	11	1	Kenya	88	22	Romania	81	20
Albania	21	5	Cyprus	7	4	Korea, South	158	9	Russia	59	21
Algeria	15	4	Czech Rep.	12	0	Kyrgyzstan	37	9	Rwanda	10	4
Angola	9	6	DRC	23	7	Latvia	26	3	Saudi Arabia	19	3
Argentina	59	19	Denmark	14	2	Lebanon	73	17	Senegal	32	20
Armenia	38	19	Dom. Rep.	6	4	Malawi	6	5	Sierra Leone	5	11
Australia	31	7	Ecuador	17	6	Malaysia	18	5	Singapore	30	1
Austria	48	19	Egypt	56	9	Maldives	10	3	Slovenia	17	10
Azerbaijan	65	32	El Salvador	47	14	Mali	105	31	South Africa	82	36
Bangladesh	34	31	Fiji	21	8	Mauritania	53	55	Spain	98	31
Barbados	71	13	Finland	18	2	Mexico	307	46	Sri Lanka	248	43
Belgium	88	21	France	197	36	Moldova	25	11	Suriname	24	11
Belize	13	7	Georgia	26	4	Mongolia	21	14	Sweden	20	3
Benin	3	7	Germany	324	45	Morocco	20	2	Switzerland	73	34
Bhutan	12	1	Ghana	15	15	Myanmar	19	11	Tanzania	14	8
Bolivia	47	8	Greece	25	7	Nepal	16	7	Thailand	25	23
Brazil	262	12	Guatemala	200	32	Netherlands	72	5	Trini. & Toba.	101	37
Bulgaria	13	4	Guyana	11	7	New Zealand	9	2	Tunisia	20	13
Burkina Faso	11	19	Honduras	8	4	Nicaragua	5	8	Turkey	140	25
Burundi	9	3	India	559	94	Nigeria	86	63	UK	139	22
Cambodia	33	12	Indonesia	373	64	N. Macedonia	8	2	USA	170	100
Cameroon	7	6	Iraq	60	4	Norway	8	2	Uganda	17	13
Canada	53	7	Ireland	12	4	Pakistan	16	5	Ukraine	82	39

Chile	12	3	Italy	150	34	Panama	55	4	Uruguay	8	3
China	216	22	Jamaica	424	41	Peru	54	12	Venezuela	6	8
Colombia	95	34	Japan	94	41	Philippines	152	34	Vietnam	11	3
Costa Rica	15	15	Jordan	26	25	Poland	2,168	79	Zambia	28	5
Cote d'Ivoire	30	35	Kazakhstan	1,815	222	Portugal	29	7	Zimbabwe	41	9



SCALING UP ACTION & IMPACT ON YOUTH EMPLOYMENT

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