Inequalities in the Times of a Pandemic

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Abstract
This paper summarizes some of the major inequalities that have been exacerbated by the COVID-19 pandemic and discusses avenues for policy intervention over the medium and long run.

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1 Introduction

Covid-19 has exacerbated existing inequalities. This paper reviews the evidence to date on how long-standing fractures have been put into sharp relief by the pandemic and discusses policies to address them.

The inequalities described take many forms and express themselves along various dimensions that interact with each other. Across the income distribution, pre-tax income inequalities, consumption and savings, job losses, and opportunities for remote work have evolved very differently. Across genders, and across parents and non-parents, the toll of school closures, lack of child care, and additional housework has been uneven. Across regions, sectors, and occupations, the pandemic has brought vastly different burdens and opportunities.

The policies discussed in this paper are medium and longer-term policy actions to address the long-standing inequalities that predate and have been widened by Covid-19, rather than short-run crisis mitigation responses. These policy actions need to occur at multiple levels, given the challenges and complexities of the issues, and we offer a framework to think about these various possible stages of interventions.

The paper is organized as follows. The first four sections review the evidence on the evolution of inequalities along several dimensions: across income levels (Section 2); across sectors and regions (Section 3); across genders (Section 4); and across children from different backgrounds (Section 5). Section 6 considers policy responses at different stages of the economic process. The Appendix provides additional materials on each of these topics.

2 Inequalities across the income distribution

This section reviews recent work that has investigated the evolution of inequalities across the income distribution during the pandemic, considering also remote work opportunities, employment losses, changes in consumption and savings, and the digital divide.

2.1 Evolution of income inequalities

The evolution of income inequalities since the start of the pandemic is similar for European and Anglo-Saxon countries. Accounting for government support during the pandemic through a combination of existing redistributive mechanisms and special measures taken, income inequalities have been reduced in post-tax terms. However, without governmental support, the pandemic would have increased income inequalities, hitting harder the bottom of the income distribution. Figure 1 shows that, for G7 countries, household incomes have generally increased during the pandemic, while the countries’ GDP sharply declined, suggesting important government financial support to households. The lower inequality relative to a counterfactual scenario without government intervention is thus the result of policy responses to the pandemic, which have been able to support low-income households. The pathways for the regressive impact of the pandemic will be covered in the next subsections and include regional disparities, sectoral and occupational composition, as well as unequal opportunities for remote work.

The effects of policy intervention on measures of inequality such as the Gini index are studied for a number of European countries by Palomino et al. (2020), Almeida et al. (2020), and Clark et al. (2020). O’Donoghue et al. (2020), Brunori et al. (2020), and Li et al. (2020) focus instead on the cases of Ireland, Italy, and Australia, respectively. Table 1 summarizes their findings (see also Appendix A.1). The dynamics of income inequality during the pandemic also suggest that policy responses have more than offset the regressive effect of the virus, but with lagged effects in Europe. Clark et al. (2020) show that in Germany, Spain, France, Italy, and Sweden, inequality – as measured by the shares of income going to different income groups – first widened from January to May 2020, and then decreased back to pre-crisis levels by September 2020.

Overall, statistics pointing to declining income inequality during the pandemic may be misleading. This decrease entirely relies on short-term policy responses to the pandemic outbreak, suggesting that the interventions have been effective to fight inequalities in the short run. However, the direct impact of the pandemic has been regressive and this can persist for at least two reasons. First, at the bottom of the distribution, workers that have been permanently laid off may experience the known long-term scarring effects...
of unemployment and human capital depreciation, as well as difficulties finding jobs again. Second, the establishment of remote work as a common practice might widen the gap in the long run, as the subsequent wage premium is unevenly distributed across the income distribution. Some initial insights on the possible long-term consequences of the pandemic on the income distribution are uncovered by Angelopoulos et al. (2021) who simulate the evolution of wealth accumulation over the next two decades in the U.K. Even under a short recession scenario, where full recovery is reached by 2024, the effect on inequalities will be long-lasting. Indeed, the poorest households, who were more exposed to unemployment shocks, have experienced a decrease in their level of wealth, while high-income households, who were more affected by restrictions on their consumption and less exposed to unemployment, have increased their savings. Consequently, an increase in wealth inequality is likely to persist (Figure A.4).

### Table 1: Evolution of Gini coefficients during the pandemic

<table>
<thead>
<tr>
<th>Citation Countries</th>
<th>Method</th>
<th>Without policy response</th>
<th>With policy response (Overall effect)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Almeida et al. (2020) EU (27)</td>
<td>Simulating effect of policies</td>
<td>+3.6%</td>
<td>-0.7%</td>
</tr>
<tr>
<td>Brunori et al. (2020) Italy</td>
<td>Simulating effect of policies</td>
<td>+0.67% (0.3396)</td>
<td>-0.67% (0.3396)</td>
</tr>
<tr>
<td>Clark et al. (2020) DE, ES, FR, IT, SE</td>
<td>Evolution over time</td>
<td>+2.17% (0.322)</td>
<td>-2.48% (0.322)</td>
</tr>
<tr>
<td>Li et al. (2020) Australia</td>
<td>Comparison market and post-tax and transfers income</td>
<td>+3.33% (0.539)</td>
<td>-7.57% (0.330)</td>
</tr>
<tr>
<td>O’Donoghue et al. (2020) Ireland</td>
<td>Comparison market and post-tax and transfers income</td>
<td>+20.64% (0.499)</td>
<td>-6.62% (0.317)</td>
</tr>
<tr>
<td>Palomino et al. (2020) EU (29)</td>
<td>Simulating effect of policies</td>
<td>+3.5% to +7.3%</td>
<td>NA</td>
</tr>
</tbody>
</table>

**Notes:** Figures in parentheses refer to initial level of Gini coefficient, when available. Almeida et al. (2020) and Brunori et al. (2020) simulate income distribution changes after spring 2020, under scenarios with and without policy response. They estimate the change in Gini coefficient of disposable income adjusted for household size and composition. Clark et al. (2020) estimate the change in Gini coefficient of adjusted disposable income between January and May 2020 and between January and September 2020 using longitudinal survey data from a representative sample of 7,302 respondents. Li et al. (2020) and O’Donoghue et al. (2020) simulate income distribution changes after spring 2020, accounting for the policy response, and estimate the changes in Gini coefficient of market income and of adjusted (for household size and composition) disposable income, during spring 2020. Palomino et al. (2020) simulate the wage distribution induced by a two month lockdown, with full and partial recovery and estimate the change Gini coefficient under the different scenarios.
Figure 1: Evolution of household income per capita and real GDP from 2007-Q1 to 2020-Q3 in OECD countries

Notes: Figure from OECD (2020), Household dashboard database. This figure shows the evolution of real household income per capita and real GDP per capita until the third quarter of 2020. The first quarter of 2007 is normalized to 100.
Figure 2: Loss of income from employment between 2019-Q2 and 2020-Q2 in the EU, before and after government compensation, by income

Notes: Figure from Eurostat, Impact of COVID-19 on employment income - advanced estimates (December 2020). In the top figure, each bar represents the percentage employment income loss between second quarter 2019 and second quarter 2020, by income group. Low-income group refers to the third lowest decile of income distribution, high-income to the third highest and medium to the four intermediate deciles. In the second figure, each bar (dot) represents the percentage employment loss of employees between second quarter 2019 and second quarter 2020, before (after) wage compensation. Wage compensation refers to the monetary compensation to employees absent from work due to Covid-19 restrictions that have been provided by European States in response to the pandemic. The changes in income are obtained using the microsimulation model EUROMOD, calibrated with Eurostat data.
2.2 Remote work opportunities as a vector of inequalities

Remote work has experienced a substantial surge during the pandemic. Research suggests that, while working from home induces a wage premium, the ability to work from home is unevenly distributed across the income distribution. Lower-income employees have lower potential and fewer opportunities for remote work. In addition, while higher-income employees may on average earn a wage premium from working from home, the productivity gains of lower-income employees seem much more limited. Thus, the rise and spreading of remote work might be a new vector of inequalities, especially if it becomes a norm.

In Europe, Sostero et al. (2020) estimate that 37% of employees or dependent contractors in the EU in 2018 were in occupations where remote work is possible. This is consistent with Dingel and Neiman (2020) who estimate that 37% of jobs in the U.S. in 2018 could entirely be done at home. The analysis of Sostero et al. (2020) also suggest that the potential for remote work increases as one moves higher up the wage distribution. Across European countries, 74% of employees in the top wage quintile can work remotely, compared to only 3% in the bottom quintile. In the U.S., the analysis of Bick et al. (2020) confirms that high-income workers have been more able to work from home. In May 2020, the share of employees with 2019 annual income greater than $100,000 working remotely was 45.5% compared to 18.4% for employees with 2019 annual income below $50,000.

This unequal potential for remote work is likely to reinforce existing inequalities. In Germany, Irlacher and Koch (2021) have estimated that, pre-Covid-19 in 2018, workers who are able to work remotely on average receive a 12% wage premium. Bonacini et al. (2021) show that in Italy an increase in the ability to work from home is correlated with an increase in labor income, thus deepening inequalities between workers that can work remotely and those who cannot. In addition, the wage premium for working from home is higher for workers who already earn more, so that inequalities would increase even among those able to work remotely.

Workers’ productivity has been affected heterogeneously by remote work. This seems to be at least partially driven by sectoral composition at different income levels, as different sectors have different potentials for remote work. It may also be the result of worse home working conditions for lower-income households. Based on a representative survey on self-reported productivity, conducted in June 2020 in the U.K, Etheridge et al. (2020) show that, while on average workers report being as productive as one year before the pandemic, low-income workers report being less productive. In particular, workers in the education, administrative, entertainment, or accommodation sectors report significant decreases in their productivity, while those in the financial, insurance, or information technology sectors report increases in their productivity (see Figure A.6).

Remote work may also become a driver of regional inequalities, as its feasibility varies starkly across regions. Irlacher and Koch (2021) find that poorer regions in Germany exhibit a lower share of jobs that can be done remotely. In the U.K., De Fraja et al. (2020) have also documented a highly heterogeneous potential for remote work across regions, with the share of residents that can work from home varying between 30 and 60%. These regional imbalances can also partially be seen as the result of the differential potential for remote work across sectors, and the clustering of specific sectors in some regions (see Bonacini et al. (2021) and Figure A.5 for Italy and Crowley and Doran (2020) for Ireland). Appendix A.2 summarizes further findings on remote work.

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1Cologne, Darmstadt or Upper Bavaria display the highest potential for remote work while Chemnitz, Saxony-Anhalt and Thuringia show the smallest potential.
Table 2: Potential for remote work is unequal by income

<table>
<thead>
<tr>
<th>Countries</th>
<th>Citation</th>
<th>Potential for remote work is higher at high incomes</th>
</tr>
</thead>
<tbody>
<tr>
<td>UK</td>
<td>Adams-Prassl et al. (2020c)</td>
<td>In the spring 2020, 60% of work tasks could be done at home for high-earners, against 20% for low-earners.</td>
</tr>
<tr>
<td>US</td>
<td>Bick et al. (2020)</td>
<td>In May 2020 the share of high-income remote workers were 45.5%, against 18.4% for low-income.</td>
</tr>
<tr>
<td>IT</td>
<td>Bonacini et al. (2021)</td>
<td>In 2018, average labor income for workers with low remote work potential was €24,731, against €27,320 for those with high potential</td>
</tr>
<tr>
<td>EU (27)</td>
<td>Sostero et al. (2020)</td>
<td>In 2018, 74% of high earners could work remotely, while 3% of low-earners could.</td>
</tr>
</tbody>
</table>

Notes: In Adams-Prassl et al. (2020c) high-earners refers to households with gross labor income above £70,000 and low-earners to those with less than £10,000 in 2019. The estimate is based on a representative survey of 24,924 respondents. In Bick et al. (2020) high-income households earn more than $100,000 in 2019 and low-income earn less than $50,000. The estimate is based on a representative survey of 5,000 respondents. Bonacini et al. (2021) use the index proposed by Barbieri et al. (2020) to measure remote work potential with the Italian Labor Force Survey. Workers with low (high) remote work potential are those below (above) the median value they compute. In Sostero et al. (2020), high and low-earners refer to top and bottom quintile of wage distribution. The estimate is based on the European Labor Force Survey.

2.3 Employment Loss

During the Pandemic. When it comes to the employment loss due to the pandemic, lower income workers face two contrary forces. On the one hand, they are more likely to work in “essential” occupations, which suffered less from layoffs or furloughs. On the other hand, they are also more likely to work in occupations with lower potential for remote work or which were more exposed to adverse labor demand shocks.

Overall, the unemployment shocks were heterogeneous across the income distribution, and low-income workers have been more vulnerable to them. In the U.K., Piyapromdee and Spittal (2020) show that the likelihood of being laid off or furloughed falls with higher earnings and wealth. Using data from the U.K. Household Longitudinal Study, they estimate that in April 2020, 52.7% of individuals in the bottom quintile of the earnings distribution were either furloughed or fired, while 15.1% of people in the top quintile were in this situation. This finding is consistent with those of Adams-Prassl et al. (2020b), who showed, using a representative survey, that in the U.K. from February to March 25th 2020, 12% of workers earning less than $20,000 lost their job because of Covid-19, against 5% of workers earning more than $40,000. In Australia, Li et al. (2020) estimate the propensity of dropping out of employment, conditional on being employed in the previous month, by market income quintile. Their results suggest that labor market disruptions peaked in April 2020 when the propensity is estimated at 14.17% for the bottom quintile against 2.58% for top quintile. In June this propensity was reduced to 1.78% for the top quintile, while it decreased only to 6.97% for the bottom quintile. In the U.S., the fall in employment rates has also been more pronounced for low-income workers. 37% of those in bottom wage quartile lost their job between January and April 15th 2020 compared to 14% of those in top quartile (see Chetty et al. (2020) and Appendix A.3). For workers in the top wage quartile, the evolution of employment seems to have been “V-shaped” (see Figure A.7), i.e., their employment has quickly recovered. This is not the case at the bottom of the income distribution.

Longer term effects. Prolonged periods of unemployment are known to be harmful for human capital, reducing the potential to reintegrate into the labor market (Phelps (1972) and Blanchard and Summers (1986)). Unemployment during the pandemic can become even more challenging to escape as, in addition
to the decline in economic activity, it comes with more domestic work, such as home child care duties which can make job searching even more difficult.

These effects on unemployment might not only turn to be long-lasting, they could also widen the gap between low- and high-wage workers. As highlighted by Tyson and Lund (2021), low-income workers will face higher pressure to reallocate since they are over-represented in occupations with lower potential for remote work and in sectors more adversely hit by the pandemic. Yet, the pandemic has also accelerated automation and digitization, and these trends favor more skill-intensive sectors. In a nutshell, the pandemic has increased the supply of low, and the demand for high-skilled labor, creating a mismatch. The analysis by Stiglitz (2020) reaches the same conclusion. Modeling the macroeconomic consequences of the uncertainty surrounding the pandemic and the resulting precautionary behavior using U.S. data, the author shows that labor-intensive sectors are less attractive, thus reinforcing automation and inequalities. Moreover, as shown by Pieh et al. (2020), low-income households have suffered greater mental health deterioration, which can compound their difficulties of searching for a job.

**Figure 3: Risk of job loss in Europe between first and second quarter of 2020, by income and country**

*Notes: Figure from Eurostat, Covid-19 labor effects across the income distribution (December 2020). This figure shows the probability of losing job between first and second quarter of 2020, by country and by income. Eurostat estimates the probability of losing job on a 0-1 scale, using a logit model with controls for age, gender, skill level required by the occupation, sector of activity, and type of work contract (permanent, temporary or self-employment), with the European Labor Force Survey data. “Low-income” group refers to the third lowest decile of income distribution, “high-income” to the third highest and “medium-income” to the remaining intermediate deciles.*
Table 3: Unemployment across the earnings distribution

<table>
<thead>
<tr>
<th>Citation</th>
<th>Country studied</th>
<th>Stronger employment shock for low earners.</th>
</tr>
</thead>
<tbody>
<tr>
<td>Adams-Prassl et al. (2020b)</td>
<td>UK</td>
<td>From February to March 2020 employment decreased by 12% for low-earners against 5% for high-earners, because of Covid-19.</td>
</tr>
<tr>
<td>Chetty et al. (2020)</td>
<td>US</td>
<td>From January to mid-April 2020, employment decreased by 37% for low-earners and by 14% for high-earners.</td>
</tr>
<tr>
<td>Li et al. (2020)</td>
<td>Australia</td>
<td>In April 2020, the propensity of dropping out of employment was 14.17% for low-earners and 2.58% for high-earners.</td>
</tr>
</tbody>
</table>

Notes: In Adams-Prassl et al. (2020b) high-earners refers to households with gross labor income above £40,000 and low earners to those with less than £20,000 in 2019. The data are from a representative survey in which respondents reported the cause of their unemployment. In Chetty et al. (2020) high and low-earners refer to top and bottom quartile of wage distribution. The authors construct a detailed new database to estimate the changes in employment. In Li et al. (2020) high and low-earners refer to top and bottom quintile of market income (wage and business revenue) distribution. This paper uses the Monthly Longitudinal Labour Force Survey to estimate the change in employment.

2.4 Consumption and savings

Thanks to high-frequency, transactional data provided by banking institutions, researchers have been able to draw a picture of the main movements in spending and savings patterns at the individual level during the pandemic. At the aggregate level, the Covid-19 outbreak seems to have induced an increase in savings and a drop in consumption. There is consensus in the literature that high-income households have mainly contributed to the additional savings. There is less agreement on the extent to which low-income households have reduced their consumption.

The aggregate drop in consumption was stark since the first mobility restriction measures. In Denmark, after seven weeks of lockdown (11th March - 3rd May 2020), aggregate spending was 27% below what it was expected to be absent Covid-19 (Andersen et al. (2020)). In the U.K., Hacıoğlu-Hoke et al. (2021) find that median expenditures were almost 40% lower in April 2020 compared to April 2019. During the first lockdown period in France, Bounie et al. (2020) reported a fall in spending based on credit and debit card transactions of about 50% during the French lockdown between mid March and mid May 2020.

In several countries, this trend has been highly heterogeneous across the income distribution. Low-income households undertook smaller percent cuts to their spending, and more rapidly returned to pre-Covid-19 spending levels. Higher-income earners have reduced their spending by a higher percent, and this effect, if not permanent, seems to be longer-lasting. Table A.1 provides evidence of these patterns in several countries. In the U.S., Chetty et al. (2020) find that spending decreased by 30% between February and March 2020 for the top income quartile and by 20% for the bottom quartile. In July 2020, spending of low-income households had returned to its 2019 level, while spending of high-income households was 13% below their baseline level. A disproportionate part of the aggregate fall in spending is thus mechanically driven by higher-income households’ spending cuts. In the U.K., Hacıoğlu-Hoke et al. (2021) suggest that households earning more than £40,000 before the pandemic contributed to 45.3% of the decline in consumption in the second quarter of 2020, relative to the same period in 2019, against 9.8% for households with less than £20,000.

The mirror image of this phenomenon is the unequal distribution of the excess savings, i.e., savings that seem to be due to Covid-19. In France, Bounie et al. (2020) estimate that the pandemic has generated an excess of financial wealth by €45 billion as compared to the counterfactual (without Covid-19) levels. 55% of this amount went to households in the top decile by level of total expenditure in 2019, and 70% went to households in the top quintile.

2See Appendix A.4 for a summary of research on spending using high-frequency data.
Non-essential spending represents a higher share of the spending of high-income households with respect to low-income ones (e.g., it amounts to 67% of the spending of the top income quartile and 59% for the bottom one in the U.S.). Unsurprisingly, it was mainly non-essential spending that dropped during the pandemic, contributing 84% of the aggregate drop in the U.S. in April 2020, relative to April 2019 (Cox et al. (2020)). Low-income households are much more constrained in their ability to cut spending and a rebound in their spending is more likely to reflect the necessity to satisfy subsistence needs rather than an improvement in their condition.

Importantly, government support might have played a key role in helping low-income groups to recover their previous consumption levels and, hence, in supporting aggregate demand. As reviewed by Falcettoni and Nygaard (2021), stimulus checks in the U.S. had a larger impact on low-income household spending. By contrast, in Spain in June 2020, low-income households' spending had not recovered faster than that for other groups. García-Montalvo and Reynal-Querol (2020) argue that insufficient government assistance to the poorest households is the main reason for this lack of recovery.

2.5 Digital divides

Digital literacy and infrastructure are unevenly distributed across the income distribution, leading to highly unequal opportunities and experiences during the pandemic.

Because of the increased use of bandwidth at the global level, the least well-equipped households have suffered stronger reductions in internet speeds. In addition, households with insufficient equipment for each member have suffered higher constraints in their potential to perform online activities at home. These two effects have strongly segmented the ability to cope with the pandemic across the income distribution. For instance, Chiou and Tucker (2020) have studied the ability to self-isolate in response to state requirements in the U.S. in March 2020 and found that households living in high-income area, benefiting from higher internet speeds, were more able to respect social distancing. The digital divide has affected low-income households through additional channels, with reduced possibilities for online learning (Ong et al., 2020), as well as fewer opportunities to maintain a social life online, with adverse consequences for mental health (Beaunoyer et al., 2020). The digital divide has also played out at the regional level, with rural regions in the U.S. being more negatively affected along all these margins (Lai and Widmar, 2021).

In fact, the pandemic has further widened the digital divide. Financially constrained households were the ones most in need of investing in proper and up-to-date equipment. Yet, as they experienced adverse income shocks, many lost the means of paying for these investments. This in turn has further reduced the possibility to leverage digital and online opportunities to make up for lost in person ones. Overall, there has been a negative feedback effect between the pandemic and the digital divide, to the detriment of more vulnerable households (Beaunoyer et al., 2020).

3 Inequalities across sectors and regions

3.1 Sectoral Inequalities

The pandemic has affected distinct sectors differently. While the previous section covered labor supply shocks linked to employment and remote work, this section considers uneven demand-side shocks.

Inequalities between sectors and across firms can be traced back to the ability to i) stay open during lockdowns, ii) substitute online for in-person activity, and iii) provide critical services during the pandemic.

Regarding the first aspect, firms fared differently based on whether they were essential or non-essential businesses and the extent to which their activity requires physical proximity. In the U.S., small businesses in the leisure and hospitality sectors experienced sharper declines in revenue (57% in March 2020) than those in the retail and transportation (of goods and people) sector (26%) (see Chetty et al. (2020) and Figure B.2). During the lockdown in Denmark, consumer spending increased by 10% in sectors that could stay open and decreased by 70% in those that could not, relative to where it would have been absent Covid-19 (Andersen et al. (2020)).
Some sectors and firms have been able to maintain a sizable share of their activity by substituting in-person sales with online ones. In France, Bounie et al. (2020) show that sectors such as personal care, clothing and fashion, or jewellery have substantially increased their share of online sales during the lockdown (mid-March to mid-May 2020). Across 13 countries (Albania, Chad, Cyprus, Georgia, Greece, Guinea, Italy, Moldova, Niger, Russia, Togo, Zambia, and Zimbabwe), financially constrained firms were 22% more likely to shut down their operations and also more likely to lay off workers between May and July 2020 (Chundakkadan et al. (2020)).

Sectors that have directly been at the forefront of the fight against the pandemic, e.g., healthcare or pharmaceuticals, have seen improved financial performance. Bellucci et al. (2020) show that venture capital investment has been strongly reallocated during the pandemic, with health-related sectors experiencing increases in investments of 39% in the first half of 2020.

Sectors that have been critical to coping with the lifestyle impacts of the pandemic, i.e. IT, have also benefited from higher inflows of capital and better financial performance. Sherif (2020) shows that this sector significantly outperformed the market in the U.K. in the first half of 2020, while sectors centered around consumer discretionary services, such as transportation (of people), tourism or leisure, have significantly under-performed. Appendix B.1 provides more detailed findings on sectoral inequalities (see also Figure B.1).

3.2 Regional Inequalities in health impacts and the consequences of remote work

Poorer regions have experienced higher Covid-19-related mortality during the pandemic, in line with the fact that social distancing was less feasible and harder to implement there. In the U.S., Weill et al. (2020) have shown that mobility has decreased significantly more in wealthier counties in response to state-level measures from January to April 2020. The lower potential for implementing social distancing can be explained in part by a sectoral and occupational composition effect. As shown by Brandily et al. (2020) for France, there is a higher concentration of occupations that involve physical proximity and are “essential occupations” in poorer areas. This effect is magnified by the prevalence of poor housing conditions, air pollution, health infrastructure, and higher shares of elderly people. Overall, mortality from Covid-19 has been higher in poorer European regions (Kapitsinis, 2020), as well as in poorer U.S. regions (Siddique et al., 2020).

As already alluded to in Section 2.2, the potential for remote work is highly heterogeneous within countries. In all OECD countries, cities and urban areas benefit from higher opportunities for such type of work (see Ozguzel et al. (2020) and Figures 4 and B.3). But the pandemic may have led to a more long-lasting shift in remote work, which can in turn reshape cities and suburban areas. As productive employees, able to work remotely move to less densely populated areas and suburbs with more affordable housing, they may be rebalancing the differences in opportunities and prices between areas, possibly also giving a chance to employees who cannot work remotely to be able to move to the metropolitan areas and city centers (Delventhal et al., 2021). These patterns may have countervailing and complex long-term consequences for productivity and inequality across regions.
Figure 4: Share of jobs that can potentially be performed remotely, in cities, towns and rural areas in Europe, 2018

Notes: Figure from Ozguzel et al. (2020).
This figure shows the number of jobs in each country or region that can be carried out remotely as the percentage of total jobs. Countries are ranked in descending order by the share of jobs in total employment that can be done remotely at the national level. Regions correspond to NUTS-1 or NUTS-2 regions depending on data availability. The authors classify occupations based on their potential for remote work using the analysis in Dingel and Neiman (2020) and apply it to data from the European Labour Force Survey.

4 Inequalities across genders

This section summarizes an abundant literature on the unequal impacts of the pandemic on men and women. Women experienced higher unemployment and stronger reductions in work hours. School closures and additional child care responsibilities – while challenging for all parents – have increased unpaid home work for women especially. We will cover remote work hours and unemployment, childcare and housework, and mental health.

4.1 Disparities in remote work by gender

Conditional on not becoming unemployed, women have been more likely to work remotely during the pandemic. This is mostly due to two channels: First, their occupations are more amenable to remote work, and, second, they have shouldered a lot of the increased need for childcare due to school and kindergarten closures. Appendix C summarizes findings on disparities in remote work by gender.
4.2 Work hours and unemployment

Compared to men, women have had a higher likelihood of being laid off, furloughed, or reducing their formal work hours. The term “Shecession” (she-cession) has been employed to emphasize the unequal impacts of the pandemic-induced recession on men and women (Alon et al., 2020).

The major causes of this disparity appear to be, first, that women tend to be more represented in some of the occupations that were hardest hit by lockdowns (e.g., hospitality, in-person services). Second, a larger share of women had part-time or alternative work contracts pre-Covid-19, while firms have tended to shelter workers in permanent work contracts more. Thus, women were more represented among the first ones to absorb the negative shock of the recession (Adams-Prassl et al. (2020b)). Third, women quit their jobs or significantly reduced their working hours due to the increased need for childcare caused by school closures. Working mothers were thus hardest hit. Alon et al. (2020) argue that the main reason for this is increased child care needs due to massive closures of schools and daycare centers. Using the U.S. Current Population Survey to examine changes in mothers’ and fathers’ work hours from February through April 2020, Collins et al. (2020) find that mothers with young children have reduced their working hours four to five times as much than fathers. Andrew et al. (2020) examine how parents in England living in two-parent opposite-gender families are spending their time under lockdown. They find that mothers’ paid work has taken a larger hit than the work of fathers, on both the extensive and intensive margins. Hipp and Bünning (2020) report similar results for Germany, where mothers were more likely to work fewer hours than fathers, and this discrepancy grew over time (from March to August 2020).

4.3 Childcare and housework

The pandemic has also deepened the gender gap in unpaid work at home. School and nursery closures increased the demand for childcare, and grandparents, family members, or friends could not easily help out because of social distancing measures. Research shows that, even though fathers increased their time spent on childcare compared to pre-Covid-19 times, mothers were the ones who absorbed most of the additional childcare work and overall housework. The results are summarized in Table 4 and in more detail in Appendix C.

The pandemic seems to also have had disproportionately negative effects on women’s mental health. Appendix C.3 summarizes the research on this issue.
<table>
<thead>
<tr>
<th>Citation</th>
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<td><strong>Panel A: Increase in female transition to working from home during Covid-19 pandemic</strong></td>
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<tr>
<td>Reichelt et al. (2020)</td>
<td>US, DE, SG</td>
<td>Women are 7 percentage points more likely to work remotely than men.</td>
</tr>
<tr>
<td>Mongey and Weinberg (2020)</td>
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<td>Women are 10 percentage points more likely to be in high work-from-home occupations than men.</td>
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<td>ES</td>
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<td>IT</td>
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<td>Hatayama et al. (2020)</td>
<td>53 countries</td>
<td>Women are more likely to work remotely than men.</td>
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<td><strong>Panel B: Decline in working hours of women during Covid-19 pandemic</strong></td>
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<tr>
<td>Reichelt et al. (2020)</td>
<td>US, DE, SG</td>
<td>Women are 5 percentage points more likely to have reduced weekly hours by 10 than men.</td>
</tr>
<tr>
<td>Collins et al. (2020)</td>
<td>US</td>
<td>Mothers of young children reduced their working hours 4 - 5 times more than fathers.</td>
</tr>
<tr>
<td>Andrew et al. (2020)</td>
<td>U.K.</td>
<td>Mothers have reduced working hours by 22% compared to 16% drop for fathers.</td>
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<td>Oreffice and Quintana-Domeque (2020)</td>
<td>U.K.</td>
<td>Women have reduced working hours by 50%.</td>
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<tr>
<td>Hipp and Bünning (2020)</td>
<td>U.K.</td>
<td>Women have 4 percentage points higher likelihood to reduce working hours than men.</td>
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<td><strong>Panel C: Increase in female unemployment during Covid-19 pandemic</strong></td>
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<tr>
<td>Reichelt et al. (2020)</td>
<td>US, DE, SG</td>
<td>3 percentage points higher likelihood of having transitioned to unemployment since January.</td>
</tr>
<tr>
<td>Adams-Prassl et al. (2020b)</td>
<td>US, U.K., DE</td>
<td>7 (5) percentage points higher likelihood of losing jobs compared to men in the US (U.K.).</td>
</tr>
<tr>
<td>Sevilla and Smith (2020)</td>
<td>U.K.</td>
<td>7 percentage points higher likelihood of losing jobs compared to men.</td>
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<tr>
<td>Andrew et al. (2020)</td>
<td>U.K.</td>
<td>Mothers have 9 percentage points higher likelihood of losing jobs compared to fathers.</td>
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<tr>
<td>Oreffice and Quintana-Domeque (2020)</td>
<td>U.K.</td>
<td>Women have 4.2 percentage points higher likelihood to have lost jobs than men.</td>
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<tr>
<td>Farre et al. (2020)</td>
<td>ES</td>
<td>Women have 2.5 percentage points higher likelihood to have lost jobs than men.</td>
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<td><strong>Panel D: Increase in female childcare responsibilities during Covid-19 pandemic</strong></td>
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<tr>
<td>Adams-Prassl et al. (2020b)</td>
<td>US, U.K., DE</td>
<td>Women spend on average 1.5 hours more on childcare than men.</td>
</tr>
<tr>
<td>Alon et al. (2020)</td>
<td>US</td>
<td>Women provide 60% of childcare hours.</td>
</tr>
<tr>
<td>Sevilla and Smith (2020)</td>
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<td>Women have been doing 30.3 additional hours childcare per week compared to 19.4 done by men.</td>
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<td>Andrew et al. (2020)</td>
<td>U.K.</td>
<td>Mothers engage in childcare over 25% more hour-long slots than fathers.</td>
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<td>U.K.</td>
<td>Women take on board 9.5 extra hours of increased childcare compared to 6.9 hours done by men.</td>
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<tr>
<td>Farre et al. (2020)</td>
<td>ES</td>
<td>Mothers spend 34.6 hours on childcare per week compared to 24.9 hours spend by fathers.</td>
</tr>
<tr>
<td>Fodor et al. (2020)</td>
<td>HU</td>
<td>Women spend on average additional 5h per week more on childcare compared to men.</td>
</tr>
</tbody>
</table>

Notes: The table summarizes results from a variety of studies on the effects of Covid-19 on paid and unpaid work for women.
5 Inequalities in education

Nursery, kindergarten, school, and university closures were some of the early and widespread responses to the pandemic. In April 2020, UNESCO reported that 192 countries had closed all their schools and universities, affecting more than 90% of the world’s students. The impacts on children in the short run have been striking, and research has tried to estimate the longer-term consequences.

5.1 Time spent learning and learning delays

In the short-term, children, especially those from disadvantaged socio-economic backgrounds, have spent significantly less time on school work during school closures, which has had consequences for their learning. For younger children, the lack of face-to-face time can translate into substantially less learning. Bao et al. (2020) predict a 31% lower reading gain from January to September 2020, relative to a business-as-usual scenario for kindergarten children whose schools are closed. They also find that parental input can help, but not make up for this: children who have books read to them daily had lower, but still quite substantial learning loss.

Huber and Helm (2020) conducted a survey and tracked students’ efforts on school-related work in Switzerland, Austria and Germany for two weeks, in the very early stages of the lockdowns (March 24th to April 5th 2020). During that period, very few online classes had yet been put into place, so that students had to study on their own. The authors find that a substantial proportion of students reported learning at home during school lockdown for at most two hours per day (see Figure D.2). Only a third of students reported a relatively high level of learning commitment (more than five hours per day).

Grewenig et al. (2020) collected detailed time-use information on students before and during school closures in Germany, in June 2020, when online classes had mostly been put in place. They find that students on average reduced their daily learning time of 7.4 hours by about a half, and that the reduction was significantly larger for students with lower test scores (4.1 hours) than for those with higher test scores (3.7 hours) (see Figure D.1). Unfortunately, parental or teachers’ inputs could not substitute for school closures, and students with lower test scores appear to have substituted learning time with leisure activities such as TV or video games.

Andrew et al. (2020) confirm these results and find that there is considerable heterogeneity by family income in children’s learning experiences. Not only does the time children spend on learning differ between lower and higher income families, but so do the resources available at home and provided by schools. Bacher-Hicks et al. (2021) use high frequency internet search data to study in real time how U.S. households sought out online learning resources once schools closed. Urban areas with higher income and better internet access exhibited substantially larger increases in searches for online learning resources. The analysis of Jæger and Blaabæk (2020) for Denmark also suggests that higher income groups have enjoyed greater learning opportunities during the pandemic. Theoretically, Agostinelli et al. (2020) predict that school closures could have a large and persistent effect on educational outcomes that is highly unequal. In the U.S. Chetty et al. (2020) have focused on math lessons completed online. A divide has rapidly grown between income groups, with a much more significant drop for bottom and middle income quartile in mid March 2020 (see Figure 5). One year later this gap seems to have widen. In mid March 2021, students from high-income schools increased progress in online math coursework by 26.7% compared to January 2020, while students from low-income schools have decreased by 3.5%. In the Netherlands, Engzell et al. (2020) find that primary school closures essentially halted learning and that students were set back by almost the full time length that was spent at home (see Figure D.4). Furthermore, the learning delays were much more pronounced among students with less educated parents. These findings are echoed by those in Maldonado and De Witte (2020) for Flemish primary schools in Belgium.

5.2 Unequal technical and resource constraints

After schools closed due to the pandemic, many tried to create an online learning environment. This was not an easy transition neither for schools nor for students, especially because not all schools and students had access to adequate technical equipment for online schooling. This is another factor that amplified edu-
cational inequalities. Huber and Helm (2020) document the starkly different technical capacities for digital instruction across schools in different countries and conclude that Switzerland and Austria have significantly more resources and technical capacities for digital teaching than Germany (see Figure D.3). Another side of the problem is documented in England, where Andrew et al. (2020) find that one in ten students in primary school and one in seven students in secondary school relied only on a cell phone or had no digital device to access school materials online.

5.3 Longer-term impacts.

The longer-term impacts stem from the immediate loss of schooling time, but also from the dynamic effects of human capital accumulation, whereby earlier acquisitions improve the productivity of future ones. This implies that younger children may be more heavily impacted than older ones. Psacharopoulos et al. (2020) estimate that a four-month school interruption results in an earnings loss of $11,117 over the course of a lifetime of an individual. Using a theoretical framework and simulations matched to U.S. data, Fuchs-Schündeln et al. (2020) find that the share without a high school degree will increase by 3.8% and the share with a college degree will decrease by 2.7%. Earning losses amount to about -1% over the lifetime.

**Figure 5: Change in Math Lessons Completed in the U.S., Relative to January 2020, by Income Group**

| Notes: Figure from Chetty et al. (2020). |
| This figure shows the change in online math assignments completed by students in the U.S. relative to January 2020, by three groups of schools, ranked into quartiles based on their share of students eligible for free or reduced price lunches. Thus, the top income quartile contains the schools with students from higher-income families on average; while the bottom income quartile contains schools with students from lower-income families on average. |
6 Policy Proposals

The policy ideas proposed here are not short-term interventions to dampen the effects of the pandemic. Rather, they suggest directions for the medium and longer-term with the goal of addressing the fiscal challenges ahead and to start healing some of the inequality fractures that preexisted and have been exacerbated by Covid-19.

The challenges of inequality deepened by Covid-19 are big, and they need to be tackled at various levels. Rather than thinking about only “standard” redistribution, or only labor market or education policies, these have to be considered jointly. Redistribution is crucial, but it needs to be combined with appropriate “pre-redistribution,” i.e., with interventions to expand education and quality employment (so-called “good jobs”). This is not only a contribution to reduce inequality, but also to eventually improve productivity. Without more equal opportunities for people in different regions and with different backgrounds, new technologies and resources remain bottled up in a few companies and among a few “elite” groups of employees –mainly in urban metropolitan areas– and do not trickle down to others. Many are left behind, as documented in the first part of this paper.

A post-Covid-19 world can be inspired by the idea of a “good jobs welfare state model” that is built on three components: First, updated traditional welfare state policies that focus on education, social insurance, and progressive taxation; second, a new focus on directly fostering good jobs and labor market experiences for all through labor market policies that partner with business and industrial or innovation strategies that target quality employment more explicitly; third, a better communication between government and citizens.

This section builds heavily on Rodrik and Stantcheva (2020) and the references cited therein.

6.1 A General Framework: Intervening at several stages

A useful way to think about policy interventions is with the matrix in Figure 6. First, one can consider the income group that is mostly targeted by the policy: those at the very bottom of the income distribution, the middle class, or rather the very top? Second, one can think about the stage at which the intervention takes place: pre-redistribution policies directly influence how markets work and can usefully be split into “pre-production” policies and “production policies.” Pre-production policies shape the endowments that individuals bring to the labor market, as well as their opportunities. Production stage policies influence the functioning of the labor market, including firms’ decisions. Post-production policies are ex-post redistribution policies, i.e., government transfers or progressive taxation.

Many traditional welfare states in Europe rely heavily on the first and third columns: fostering education and training to prepare people for the labor market on the one hand, and progressive taxes and transfers, as well as social insurance against unemployment, illness, or disability on the other. Production stage policies – except perhaps the minimum wage, collective bargaining regulations, and labor protection – are not systematically geared towards reducing inequality and creating better jobs. They are instead targeted towards market competition, physical investment, and R&D, along a traditional divide between “social policies” to tackle inequality and economic policies to improve productivity, innovation, and growth.

However, such traditional welfare state systems are built upon the assumption that (almost) everyone who wants a good job can find one. Covid-19 has shown very clearly the extent of existing inequalities in the quality of jobs accessible to different groups and how unequal the opportunities are. Good jobs, which were the pillar of the welfare state in past decades have been disappearing. It is not possible to define what a “good job” is in the absolute, as it depends on local circumstances and people’s preferences (e.g., for flexibility). Nevertheless, some clear criteria are safe and reasonable work conditions, sufficient pay that enables a good living standard, access to benefits such as health care, child care, and pensions in the future, as well as adequate social insurance, and some share of career opportunities and progress.

In a world in which middle-class good jobs are disappearing due to longer-run trends such as technological change and globalization, and shocks such as Covid-19 exacerbate the cleavages, there is a need to act on all three columns in a coherent manner. Inequalities are in part perpetuated by the production stage, when firms make innovation, employment, and investment decisions without necessarily internalizing the far-reaching spillovers on current and future employees or the communities they operate in.
6.2 Pre-production policies

This section considers policies that can improve opportunities.

6.2.1 Inheritance Taxation

An important tool to dampen the persistence of wealth across generations and to level the playing field between people from families with different levels of wealth is a tax on the transfers between generations. Such taxes are currently a small source of revenue for most governments, but are projected to increase as private wealth concentration rises. They can be either beneficiary-based (e.g., the inheritance tax in France) or donor-based (e.g., the estate tax in the U.K. and the U.S.). Inheritances are in general very concentrated because wealth itself is.

Inheritance taxes are very unpopular, but research shows that this is in large part due to misunderstandings of how they work and who actually bears them (Stantcheva (2020) and Kuziemko et al. (2015)). In particular, such taxes are made unpopular by the belief that they represent double taxation and that they hit the middle class harshly.

Most current systems suffer from being essentially myopic, in the sense that different transfers to the same person, coming from either different people or at different points in time, will be to a large extent treated independently. In addition, they typically allow for too many loopholes and, hence, suffer from abundant avoidance.

One avenue to improve the taxation of gifts, inheritances, and estates would be to move towards a
beneficiary-based regime that takes into account at once inheritances and gifts and is progressive in the cumulative amount received. Rather than taxing transfers at each “death,” such a system would tax the total transfers (gifts, inheritances, from all sources) received by the heir, and those who receive more would be taxed at higher rates (progressivity). It is possible to put the exemption threshold relatively high in that case, truly exempting the middle class, while still being able to tax very wealthy families. It is also feasible to allow for some preferential treatment of transfers based on the closeness of the family link between the donor and the heir. Such a beneficiary-based system could at once be more progressive and also much more accepted and better understood by citizens, as it addresses their concerns outlined above. A similar system currently exists in Ireland under the name of “Capital Acquisition Tax” (Nolan et al. (2020) and Irish Department of Finance (2019)).

6.2.2 Education policy

Existing disparities in the quality of education have also been put into sharp relief by the pandemic (see Section 5). Going forward, the challenge is to reduce disparities in the quality and outcomes of education for children from different socio-economic backgrounds. While EU countries in general invest substantially in free public education, there is significant variation in quality across schools in different areas and for children from disadvantaged families, including minorities (OECD (2020)). Better targeting of public investments towards disadvantaged schools and children is needed. Furthermore, children from different backgrounds may not benefit to the same extent from any given schooling inputs because of “missing family inputs.” There are on-going promising initiatives to try to level the playing field by substituting partially for inputs poorer children’s families may be lacking. These need to be expanded and fostered.

As the pandemic was unfolding, several simple and low-cost interventions were very rapidly tested. They carry some lessons for the future and could be scaled up for the longer term. Carlana and La Ferrara (2021) evaluate the effectiveness of an intervention implemented in Italian middle schools that provided free individual tutoring online to disadvantaged students during lockdown. Tutors were university students who volunteered for 3 to 6 hours per week. The study shows that the program substantially increased students’ academic performance and that it significantly improved their socio-emotional skills, aspirations, and psychological well-being. Hardt et al. (2020) evaluate the effects of remote peer mentoring at a German university that switched to online teaching due to the Covid-19 pandemic. Mentors and mentees met one-on-one online and discussed topics like self-organization and study techniques. The authors find positive impacts on motivation, studying behavior, and exam registration. Lastly, Angrist et al. (2020) provide some of the first experimental evidence on strategies to minimize learning loss when schools close. They run a randomized trial of low-technology interventions in Botswana - SMS messages and phone calls - with parents to support their school children. The results show that learning improves by 0.12 standard deviations. Thus, the value of even remote tutoring seems quite large. Such interventions could continue to be beneficial even after Covid-19, as they allow children in more remote areas or with fewer parental resources to have access to extra time, mentoring, and support.

Another pre-Covid-19 example is from France, where the program “Devoirs faits” (“Doing Homework”) since 2017 lets children spend time on homework with the supervision and the help of staff in their own school. This provides some compensating input for children whose parents cannot help with homework. The general concept behind such initiatives is critical and generalizable. It is to help children from disadvantaged backgrounds receive at least some of the inputs that other children get from their families, in terms of training and skills acquisition, as well as extracurricular activities.

Another important challenge ahead is the digital divide among children and students. This was described in Section 5 and made particularly salient by Covid-19. Unequal access to the internet and to computers hinders learning opportunities. Schools can play a role in offering quiet, safe study spaces with internet connection and computers for children who cannot access these at home. Even better would be public funding for broadband and computers for children from disadvantaged backgrounds. Access to digital technologies, in line with the Covid-19 interventions just described above, can help equalize access to learning opportunities. A successful example comes from Korea, where the “Cyber Home Learning System” is a self-study platform with the goal to reduce gaps in access to extra-curricular materials and learning between students in urban

See the report at https://assets.gov.ie/19127/bf33c368730e44dc58cc7c7930c9b8487.pdf
areas and those in more remote ones. In the U.S., the “Cognitive Tutor” program teaches mathematics with a customized software complementing a textbook. Finally, the U.K.’s Shireland Learning Gateway is a portal developed by Shireland Collegiate Academy in cooperation with Microsoft to help students and parents track student performance and behavior, facilitate contact with the school, and give access to extra-curricular materials. Regardless of the exact structure of these programs, teachers should be involved in their design, which should also be subject to regular experimentation and evaluation.

Further down the road, it is critical to facilitate the transition between school and the labor market in a way that provides opportunities for students from all backgrounds. Joblessness early in one’s career can be particularly detrimental, and yet, youth unemployment is widespread in Europe. To achieve this goal, boosting vocational and dual vocational-academic tracks seems particularly promising, as is the provision of better guidance for students on their choice of higher education.

Dual education tracks and vocational training have been shown to generate much better labor market outcomes in countries such as Germany. New initiatives are happening in France, and in the Netherlands, where the “Katapult” system brings together public-private partnerships made of businesses, R&D centers, and schools to train (mostly) Vocational Education and Training (VET) students.

To help students choose their graduate training, more information on the prospects of different majors and tracks is needed. This can also boost productivity as it may help align the demand and supply of skills. The Netherlands has launched a “graduate tracking” program in 2018, through which they analyze the careers of a large sample of graduates and make the results available to prospective students choosing their field of study.

6.2.3 Gender disparities

Inequalities across men and women in the labor force have been rendered particularly stark during Covid-19, but they are the consequences of long-standing cleavages.

Kleven et al. (2019) consider six countries that span a wide range of institutional features, policies, and social norms: Denmark, Sweden, Germany, Austria, the U.S., and the U.K. They find that the earnings of men and women evolve similarly before parenthood but diverge sharply after. Women experience a large, immediate, and persistent drop in earnings after the birth of their first child, while men are essentially unaffected. Ten years after child birth, women have not recovered and the convergence has stopped. In the Scandinavian and Germanic countries, women’s overall participation suffers less, implying that a substantial share of the earnings penalty for mothers is driven by the intensive margin of labor supply and changes in their wage. In the U.S. and the U.K., reduced participation in the labor market seems to be a key driver for the earnings penalty. Considering the causes for differences across countries, the authors show that longer paid and job-protected parental leave implies larger short-run child penalties in both earnings and employment. Nevertheless, parental leave cannot explain the longer-run effects and cross-country differences. Instead, the study points to social norms about gender roles as a major explanatory factor.

There is some evidence that the pandemic may change some of these social norms related to gender roles. Reichelt et al. (2020) analyze survey data from a representative sample of respondents in the U.S., Germany, and Singapore and ask whether Covid-19-related changes, such as becoming unemployed, reducing working hours, or working from home of respondents and their partners are associated with their views on gender roles. Figure C.2 shows that women’s egalitarian views are diminished when they become unemployed; while their partner’s attitudes are not affected. Men’s gender-role attitudes become more egalitarian when they themselves become unemployed or work from home, potentially as a result of their increased participation in household labor. Nevertheless, a deep change in attitudes takes time and needs to be supported by concrete policy action.

Expanded parental leave has been suggested as a possible buffer of the unequal consequences of the pandemic. Yet, as Kleven et al. (2019) show, there is no definite evidence that this helps mothers particularly. The literature has instead offered some early evidence on the potential effects of earmarked parental leave on women’s labor market participation and earnings. In such schemes, each parent gets allocated a given length of leave, sometimes with a requirement that it has to be taken. Dunatchik and Özcan (2019) use the introduction of a “daddy quota” – non-transferable paternity leave policies – in Quebec, Canada, as a natural experiment and study its impact on a range of mothers’ career outcomes. Mothers were 5 percentage
points more likely to participate in the labor force and to work full-time, 5 percentage points less likely to work part-time (relative to a baseline labor force participation of 75%), and 4 percentage points less likely to be unemployed (relative to a baseline unemployment rate of 9.3%, baseline full-time employment of 65%, and part-time employment of 33%). Johansson (2010) investigates the effect of parental leave - both own and spousal - on subsequent earnings in Sweden. The results show that spousal leave is important for mothers: every additional month for which the father is on parental leave has a larger positive effect on maternal earnings. Andersen (2018) tests how a father’s paternity leave affects the within-household gender wage gap among heterosexual couples in Denmark. When fathers take longer leave, mothers reduce their own leave and experience wage gains, which, on balance, increases total household earnings. Druedahl et al. (2019) estimate the causal effect of increasing the earmarked paternity leave of fathers on the relative labor income of women within couples in Denmark. They find that the reform led to shorter maternal leave and an increase in mothers’ earnings for up to eight years after birth.

Much more work is urgently needed on these issues (especially as such schemes are relatively recent) to inform the design of policies to buffer some of the stark inequalities across genders amplified by Covid-19.

6.3 Production-Stage Policies

6.3.1 Employer-focused active labor market policies

Active labor market policies (ALMPs) are defined as “all social expenditure (other than education) which is aimed at the improvement of the beneficiaries’ prospect of finding gainful employment or to otherwise increase their earnings capacity” (EU – Factsheet on ALMP). They encompass skill training and certification, employment subsidies, public sector work programs, and assistance with job search and matching with employers and are often administered through Public Employment Services (PES). In many EU countries, ALMPs cost less than one percentage point of GDP, and cover around 20-40% of people looking for employment (OECD and Eurostat via France Stratégie). Yet the evidence on their success is mixed in regard to training and wage subsidies, or public work programs. More consensus exists on the positive effects of job search assistance (Heckman et al. (1999); Kluve et al. (2002); Kluve (2010); Card et al. (2010); Caliendo and Schmidl (2016)).

Other “sectoral training programs” in the U.S. have repeatedly been shown to be quite successful. They differ from standard ALMPs in that they are specifically geared towards the needs of the local employers and cooperate tightly with them, including on the design of the training. They are managed by community organizations or private agencies; entail training in specific occupational and soft skills; follow up with services after job placement to ensure employees get paid a sufficient wage; and closely involve employers (Kruglaya (2017) and Autor et al. (2019)). Examples are Project Quest in San Antonio, Texas; Per Scholars in the Bronx, New York; Madison Strategies Group in Tulsa, Oklahoma; or Wisconsin Regional training partnerships in Milwaukee, Wisconsin.

Such successful sectoral training programs can provide inspiration for more employer-focused active labor market policies in Europe, with closer involvement of employers, especially at the local level, to ensure the appropriate training based on the skills needed. Currently, the European PES are the closest analog to such schemes, but they are in general large public bodies that cannot be as flexible and locally adaptable as may be needed. They are also not directly involved in the design of training programs, and they have weaker links with employers than successful sectoral training programs. Nevertheless, there are encouraging trends in the right directions (7).

6.3.2 Business incentives focused on good jobs

Many business incentives today take the form of subsidies for physical investment and new technologies. These are very important policies, but there is some evidence that they may not be the most direct way of helping workers.

In the U.S., there are many tax incentives provided by firms in order to attract businesses from other states. They amounted to $47 billion in 2015 (Bartik (2020a)). Slattery and Zidar (2020) summarize such incentives and find that, on average, their recipients are likely to be large firms in manufacturing, technology,
and high-skilled service industries. They find some evidence of direct employment gains from attracting a firm, but on balance conclude that there is no “strong evidence that firm-specific tax incentives increase broader economic growth at the state and local level.”

On the contrary, Criscuolo et al. (2019) study the effects of the Regional Selective Assistance (RSA) program in the U.K., a system of subsidies mainly targeted at manufacturing to prop up and expand employment in low-income, high-unemployment areas. Firms in the areas covered can submit proposals to the government with specific investment plans. The government can then finance up to a third of selected proposals. The criteria for evaluation explicitly include the creation of new employment or protection of jobs that would otherwise have disappeared – the program thus directly targets employment, unlike many others. Yet, it still subsidizes spending on physical capital and not direct job creation. The scheme has been found to have significant effects on local employment, especially on the employment of small firms with less than 50 employees. The cost per job saved or created ($3,683 at 2010 prices) is at the lower end of most programs, including the aforementioned business tax incentives. Cost per job estimates range from $18,000 for Empowerment Zones in the U.S. to more than $68,000 for investment subsidies in the South of Italy (see Criscuolo et al. (2019) and Table E.1). Indeed, as Bartik (2020a) writes, “cash incentives to encourage local job creation have high costs per job created because it takes a lot of cash to tip a business location or expansion decision.”

Guidance on how to ameliorate business incentives at the regional level comes from Bartik (2020b) and Bartik (2020a). Current systems to improve productivity growth in distressed areas are not very effective for helping workers specifically because they offer significant tax breaks that often mainly benefit large companies and are poorly designed or targeted. Instead, business incentives should focus only on areas that are distressed. Second, if job creation is the goal, such policies need to center around sectors or firms that have high potential to actually create jobs. Third, tax incentives should not be the main tool. Rather, the focus should be on specific public services needed by firms, such as customized business services, zoning or infrastructure policies, local amenities, and skills training. Related to this, business assistance should be taken as a portfolio of services that is adapted to the local conditions. The overall goal is to create a structure for assistance to local firms or potential investors, with a portfolio of services that help them increase productivity while creating good jobs. Existing networks of agencies, or new agencies could take on these roles.

6.3.3 Labor-friendly innovation policies

Contrary to what is sometimes believed, the direction of technological change is not ineluctably determined or exogenous to policy incentives. Tony Atkinson already called for policy makers to pay attention to the direction of policy change. Innovation could potentially be encouraged to augment labor, improve the productivity of workers, and help generate good jobs, rather than purely and simply replace labor. This is a relatively understudied area. But there are clear directions for policy makers to explore. First, existing fiscal regimes may be systematically creating an excessive incentive for automation (see Section 6.4). Second, existing tax incentives for R&D and innovation could possibly be made contingent on employment outcomes. Third, when determining which innovation areas to fund, the government could use an “employment test” as a criterion. E.g., AI systems could potentially complement low and middle skills, rather than only high skills. Fourth, the government can encourage learning organizations that empower workers, with teamwork; development of cognitive, social, and soft skills; workers’ autonomy and continuous learning instead of the Taylorist or “lean” models.

6.4 Post-production policies

Raising sufficient tax revenues has been a perennial challenge, but is particularly binding given the large sums of government funds deployed during the pandemic. In many OECD countries, taxation has gradually shifted from capital to labor, including through social security and payroll taxes. This may be correlated with deepening inequalities, contribute to labor market rigidity and polarization, and exacerbate social tensions with feelings of unfairness. More generally, letting the tax gap between capital and labor income grow too wide could cause a range of problems, starting from the creation of incentives to shift income
between these tax bases (e.g., pass-through entities incorporating, or wages in private businesses being paid out as dividends). In addition, financing the bulk of social programs (bound to grow due to demographic change and aging) by labor taxation and social security contributions on labor income, rather than through general taxation on all income, can create an incentive to shift away from labor income, reduce labor market participation, increase labor market duality (e.g., between standard and non-standard employment), and diminish labor productivity and growth. Finally, over the longer run, a disproportionate burden on labor income taxation can create incentives to direct innovation towards labor-substituting technologies and invest more in automation. This can reinforce the shift away from labor (Acemoglu et al., 2020).

To a large extent, this shift has been the consequence of the increasing mobility of capital and corporations with the rise of globalization. Yet, pressure to tax better and more fairly is likely to intensify post-Covid-19. Inequality and polarization have further increased, which will deepen dissatisfaction with globalization and the inequality in treatment of capital and labor incomes. Public tolerance for tax evasion and avoidance can be limited given the stark demand for revenue generated by the pandemic, and the secular rise in public revenue needs stemming from aging populations and demographic changes. At the same time, there are more opportunities now than in the last decades to tax capital income thanks to the exchange of information between countries, as well as improved information technologies and data analytics. Some of the fatalistic views that it is impossible to tax internationally mobile capital and corporations in a globalized world may need to be reconsidered.

6.4.1 Exchange of information and tax cooperation

On capital income. There have been major improvements in international cooperation through the Automatic Exchange of Information (AEOI) implemented and pushed by the “Global Tax Forum,” which offer renewed opportunities to tax capital more efficiently and improve compliance. In addition, it may make progressive taxes on capital more conceivable, given the feasibility to track the capital income of a taxpayer across countries.

On labor income and people. People, like capital, can be internationally mobile. This is especially true for higher-income, highly skilled professionals with non-location-specific human capital. In a review of recent evidence on international tax mobility, Kleven et al. (2020) conclude that preferential tax regimes for foreigners – whereby foreigners coming to the country are given tax breaks for a few years – are widespread. While this may be individually beneficial for a country, it is a beggar-thy-neighbor policy. At the moment, retaliation is not widespread, presumably because the countries with such schemes have relatively high taxes, but there is the possibility of a race to the bottom. Such preferential tax treatments may become more attractive to individual countries if their revenue needs increase, but they nevertheless want to remain attractive destinations for highly skilled foreigners, and if remote work, digital communications, and travel continue to become easier. There is a lot to gain from at least some amount of international cooperation on labor income taxation at the EU level.

6.4.2 Reducing fiscal leakages by reducing avoidance and evasion

Raising revenues to finance much-needed public spending, particularly after the pandemic, entails not only thinking about tax rates and bases, but also about tax compliance. To reduce fiscal leakages, sufficient resources for tax authorities are needed, so that they can leverage new data analytics and digital tools. Third-party reporting can also be expanded.

Expanding third-party reporting. Third-party reported income is barely evaded, yet, there are income flows not subject to it. Top incomes may have a larger share of their incomes that are not third-party-reported, e.g., private business and proprietorship income. Sarin and Summers (2020) propose a way to resolve some of the challenges around private businesses and partnerships. In their proposal, business owners and entrepreneurs with income above a given threshold would be required to report to the tax authority the bank accounts that host their business income. Banks would then act as third parties and be bound to report the flows – deposits and disbursements – on those accounts to the tax authority.
**Data Analytics to Identify Non-Compliance.** Tax authorities could leverage big data and analytics to improve methods of detecting fraud or mistakes among taxpayers and profit from substantial progress made in predictive algorithms, machine learning, and AI. Combined with the tax data available in many countries, these data analytics methods could help track and enforce compliance more cost-efficiently and better allocate the tax authority’s scarce time and resources.4

This may go hand in hand with making data available to and cooperating with researchers. Many academic papers on avoidance and evasion by individual taxpayers, small businesses, or large corporations identify behaviors and markers for noncompliance that can be used by the tax authority. They also estimate models of taxpayers’ or firms’ behaviors that can help predict noncompliance.

**Giving resources to tax enforcement.** Resources are needed by tax administrations in many countries in order to leverage the aforementioned big data and analytics opportunities. Tax administrations need investment in their technology infrastructure (software and hardware) and advanced analytical capacities, as well as regular training of their staff. Sarin and Summers (2020) lament the outdated information technology of the Internal Revenue Service (IRS). A recently piloted return review program (RRP) mentioned by the authors improves the matching of taxpayer filings with information returns to flag and freeze fraudulent refunds and showcases a gigantic 50-1 return, much higher than that of traditional enforcement programs. Outdated systems are not peculiar to the U.S., so the EU can also learn from this experience.

6.4.3 Corporate taxation and multinational taxation.

A source of revenue shortfall in many countries are the shortcomings in the taxation of corporations, particularly multinationals. In addition, there is a fairness argument in the eyes of many citizens: multinationals and their shareholders are considered not only to have benefited tremendously from globalization, but also to have received extensive government help during Covid-19. The goal is not to just tax “foreign” companies, but also national companies that operate abroad. A company’s “nationality” is complex, as a domestic company can be partially owned by foreigners, whether it operates domestically or not and domestic shareholders can own stock in foreign companies too.

The important Base Erosion and Profit Shifting (BEPS) initiative by the G-20 and the OECD in the Global Tax Forum has produced and pushed a set of recommendations to ensure a better taxation of multinationals. In a nutshell, it is based on two pillars. Pillar 1 focuses on the allocation of taxing rights and aims for the coherence of the profit allocation and tax nexus rules. Pillar 2 – called “GloBE” (Global Anti-Base Erosion) proposal – focuses on multilateral “backstop” rules that would give countries the right to “tax back” in cases where other jurisdictions have not exercised their primary taxing rights, or where the tax payment by the company has been “too low” according to some benchmark. It shares many features with GILTI implemented by the U.S. (which would be considered a “compliant income inclusion rule” under Pillar 2, so that both mechanisms can co-exist). It is important to not ring-fence “digital companies” in these initiatives. Digital technologies pose particular challenges, but they are not limited to digital companies per se, given that many companies now have sizable shares of digital activities.

These comprehensive initiatives are critical and valuable and should be pushed forward. The EU has to play a major role in being a leader and supporter of this tax cooperation.

6.5 Surveys as a key tool for understanding citizens and designing policies

Implementation of the policies described here will hinge on data collection, experimentation, and rigorous evaluation. Data is needed also on what people think. “Surveys” are a way of getting into citizens’ minds and to elicit perceptions, knowledge, understanding, attitudes, and views. Large-scale surveys should become a continuously used, well-designed, and interactive policy tool that the government leverages to better understand citizens, as well as employers and companies. They are not simple “opinion polls,” but rather in-depth studies and scaled-up versions of town halls and direct debates. They can complement the dialogue between constituents at different levels, using mobile phone and internet technologies to rapidly reach a large

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4For instance, for small- and medium-sized enterprises it is possible to combine datasets across years and agencies and use predictive models to estimate the expected revenues of businesses, in order to flag those that fall short. Private sector methods (e.g., from finance and private equity), could be used to value even relatively illiquid assets for tax purposes.
and diverse set of people. They can also help give a voice to citizens who may otherwise not always be heard. As an iterative policy design and test tool, surveys can be used to collect valuable input and feedback from constituents, test reform ideas, detect implementation challenges, as well as study the impacts of policies in real-time.

It will be key for such surveys to have a serious reputation, a regular and predictable schedule, as well as the sense that views will be heard, even if not all directly implemented. This, too, is an area in which the involvement of researchers can be very fruitful. Indeed, this approach has been deployed to a variety of issues by researchers, as exemplified by the studies of the Social Economics Lab at Harvard (socialeconomicslab.org).

References


Phelps, E. S. (1972). Inflation policy and unemployment theory.


APPENDIX

A Inequalities across income distribution

A.1 Evolution of income inequalities

- Almeida et al. (2020)
  The authors use the micro-simulation model EUROMOD to estimate the effect of changes in GDP and employment induced by the pandemic during the spring 2020 on household income for 27 European countries. They compute the distribution of disposable income, equivalised for household size and composition, taking into account the discretionary measures that have actually been implemented by the 27 European countries they study. They also compute the counterfactual distribution, under a scenario where no discretionary measures are taken in response to the pandemic. Then they estimate the Gini coefficient of these two distributions. They estimate that without policy measures, the Gini coefficient would have increase by 3.6%, while it has decreased by 0.7% with the policy response.

- Brunori et al. (2020)
  The authors use the micro-simulation model MicroReg tax microsimulation from IRPET to estimate the change in equivalised disposable income distribution induced by the pandemic in Italy during the spring 2020. They consider the measures that have actually been implemented, in addition to a counterfactual scenario without policy support. They estimate that without policy measures, the Gini coefficient would have increase by 0.67%, while it has decreased by 0.67% with the policy response.

- Palomino et al. (2020)
  The authors estimate the effect of legal restrictions during lockdown periods in the spring 2020 on wage distribution for 29 European countries. After estimating the ability of workers to perform their jobs during the lockdown period, they compute wage distribution under four different scenarios: The first one with a two-month lockdown and three others, where the lockdown period is followed by six month of partial functioning of closed activities, at 60%, 70% or 80% of full capacity. This leads to a range of Gini coefficients associated with each counterfactual wage distribution. In the more stringent scenario with partial functioning at 60% the Gini coefficient increases by 7.3%, against 3.5% in the more optimistic scenario.

- O’Donoghue et al. (2020)
  The authors use European Union Survey on Income and Living Conditions to calibrate their own microsimulation model for Ireland and estimate the change in income distribution induced by the pandemic in Ireland during the spring 2020. They estimate both the distribution of market income and disposable income, equivalised for housing, work-related expenses and capital losses. They compute that the Gini coefficient of market income has increased by 20.64% as compared to before the pandemic, while the Gini of disposable income has decreased by 6.62%.

- Li et al. (2020)
  The authors use a semi parametric approach to estimate the impact of the pandemic on Australian income and employment distribution during the spring 2020. They estimate both the distribution of market income and equivalised disposable income. They compute that between February and June 2020, the Gini coefficient of market income has increased by 3.33% as compared to before the pandemic, while the Gini of disposable income has decreased by 7.57%.

- Clark et al. (2020)
  The authors use longitudinal high-frequency data from a representative survey of 7,302 respondents to directly estimate the changes in income distribution over the year 2020 for France, Germany, Italy,
Spain, and Sweden. They compute the changes in the Gini coefficient of equivalised disposable income. Between January and May 2020, the Gini increased by 2.17% on average; between January and September 2020, it decreased by 2.48%.

- Angelopoulos et al. (2021)

The authors use a Bewley model to analyse the effect of the pandemic of wealth accumulation of heterogeneous households in the U.K. The model is calibrated using the Wealth and Asset Survey. Authors model the crisis as an increase in labor income risk, a drop in mean income and an upper bound on consumption to capture the restrictions associated with lockdown periods. They consider a short and a long recession scenario with a recovery in 2024 and 2027, respectively. In both scenarios, the consequences on wealth accumulation propagate over time and increases wealth inequalities persistently. The first driver of inequality is the labor income shock, which hits low-income households harder, leading them to decrease their savings. The second are the consumption restrictions, which lead high-income households to increase their savings.

**Figure A.1: Estimated changes in income and spending between spring 2019 and 2020 in the U.S., by income quartiles**

![Bar chart showing changes in income and spending by income quartile](image)

**Notes:** Figure from Cox et al. (2020).

This figure shows the change in income and spending by income quartile. The change in income compares March, April, and May 2020 to average quarterly income in the prior year. The change in income reflects the decline in labor income, the EIPs, and unemployment benefits. The change in spending compares April 15-May 30 to spending at the same time in the prior year and is computed using accounts data from the bank Chase.
Figure A.2: Evolution of income inequality measures in Europe over 2020

Notes: Figure from Clark et al. (2020).
These figure shows the evolution of Gini coefficient of equivalised disposable income distribution as well as three other measures of inequalities, the mean logarithmic deviation, the Theil index and the half square of the coefficient of variation. They are estimated for January, May, and September 2020, using a longitudinal representative survey data of 7,302 respondents.
Figure A.3: Impact of spring 2020 lockdowns on income inequality (Gini index) in EU countries, with and without policy answers

Notes: Figure from Almeida et al. (2020). This figure shows the change in the Gini coefficient of equivalised disposable income induced by the lockdowns put in place by European countries during the spring 2020. The authors estimate the changes in GDP and employment induced by lockdown measures and use a micro-simulation model to estimate the effect of these changes on income distribution, under a counterfactual scenario without policy response (blue bars) and under a scenario accounting for actual policy response (red bars). Lines illustrate 95% confidence interval.
**Figure A.4: Evolution of mean wealth and wealth inequalities post-Covid-19 in the U.K. Angelopoulos et al. (2021)**

Notes: Figure from Angelopoulos et al. (2021).
The authors calibrate a Bewley model to analyse the effect of the pandemic of wealth accumulation of heterogeneous households in the U.K. They consider a short and a long recession, with an increase in labor income risk, a drop in mean income and consumption restrictions, where recovery is reached in 2024 and 2027 respectively. The model is calibrated using the Wealth and Asset Survey. On the top-left panel, mean wealth pre-Covid-19 is normalized to 1. The two lower panels show the evolution of the share of wealth owned by the bottom 40% and the top 20% of wealth distribution.
A.2 Remote work opportunities as a vector of inequalities

- Sostero et al. (2020)
  The authors use occupational task descriptions to build a classification of occupations in terms of potential for remote work. They apply it to European countries with 2018 data on labor force and estimate that 37% of employees or dependent contractors (“dependent employment”) in the E.U. are in occupations where remote work is possible.

- Dingel and Neiman (2020)
  The authors use surveys describing occupational tasks to build a classification of occupations in terms of potential for remote work. They apply it to the U.S. with 2018 data on labor force and estimate that 37% of US jobs can be worked from home.

- Bick et al. (2020)
  The authors use data from a representative survey of over 5,000 respondents to estimate the share of workers that have worked remotely in the U.S. during the spring 2020, keeping track of individual characteristics. They estimate that 35.2% of the workforce worked from home in May 2020.

- Irlacher and Koch (2021)
  The authors use data from the German Qualifications and Career Survey, conducted between 2017 and 2018, to run a Mincer regression and estimate the effect of remote work on wages. Using a rich set of controls (worker and firm characteristics, industry, and region fixed effects), they estimate that people who work from home benefit from a 12% wage premium. They also investigate regional disparities and show that the poorest regions display smaller potential for remote work.

- Bonacini et al. (2021)
  The authors use 2018 employment data from the Italian Labor Force Survey to estimate the consequences of a permanent increase in the feasibility of remote work on the labor income distribution. Using the remote work feasibility index proposed by Barbieri et al. (2020), they classify the population between employees with low and high remote work feasibility. They estimate with unconditional quantile regression the effect of a shift of a 10 percentage points share of employees from the low feasibility level to the high one. They find that it would increase the mean labor income up to €259 and that employees in the 8th decile of income distribution benefit from the higher wage premium, by about €500.

- Etheridge et al. (2020)
  The authors use data from a representative survey conducted in June 2020 with 3,411 respondents on self-reported productivity of people working from home during the first lockdown period in the U.K. They estimate that on average, workers report the same productivity as one year before the pandemic. They find that the self-reported productivity of workers at the bottom of the earnings distribution have significantly decreased.

- De Fraja et al. (2020)
  The authors assess the effect of the geographic shift of productive activities from office to home, which they call the "Zoomshock", in the U.K. They find a strong heterogeneity in the Zoomshock between regions, with a share of residents that can work from home varying between 30% and 60%.

- Crowley and Doran (2020)
  The authors use occupational level data from O*NET in Ireland to construct a measure of social distancing potential and a measure of remote work potential. They find higher potential for remote work in Dublin region and provincial city regions, where economic activity is concentrated.
Adams-Prassl et al. (2020c)

The authors use data from a representative survey of 24,924 respondents during Spring 2020 to estimate the share of tasks that can be done from home within occupations and industries in the U.K. and in the U.S.

**Figure A.5: Potential for Working From Home in Italy in 2018, by sector and income**

![Graph showing potential for working from home in Italy by sector and income.](image)

*Notes: Figure from Bonacini et al. (2021). The authors use the index proposed by Barbieri et al. (2020) to measure the feasibility of Working From Home (WFH) of Italian employees, with the 2018 Italian Labour Force Survey. They separate employees between those with high and low WFH attitude. This figure shows the share of employees with high WFH attitude by sector (left scale) and the average gross labor income (right scale) for both groups of employees, by sector.*

A-7
Figure A.6: Mean productivity change between January and June 2020, in the U.K., by Industry and by Occupation

Notes: Figure from Etheridge et al. (2020).
The authors use data from a representative survey performed in June 2020 with 3,411 respondents to estimate the changes in self-reported productivity between January and June 2020 in the U.K. A value of 0 indicates that the respondent reports no productivity change. Values are expressed in terms of standard deviation away from 0. Lines illustrate 95% confidence intervals.

A.3 Employment Loss

- Piyapromdee and Spittal (2020)
The authors use data from the U.K. Household Longitudinal Study to evaluate the labor market impacts of the pandemic in the U.K. They investigate the change in employment status between February and April 2020 and find that the likelihood of being furloughed or laid off fall along earnings and wealth distribution.

- Adams-Prassl et al. (2020b)
The authors conducted a representative survey on March 25th, 2020 on 3,974 respondents to collect employment data and evaluate the labor market impacts of the pandemic in the U.K. during the spring 2020. Causes of unemployment were self-reported by the respondents.

- Li et al. (2020)
The authors use a semi parametric approach to estimate the impact of the pandemic on Australian income distribution during the spring 2020. They use the Monthly Longitudinal Labour Force Survey to estimate the probability of dropping out of employment conditional on being employed in the previous period across income quintiles. In April 2020, they estimate the propensity of dropping out of employment in the bottom income quintile was 14.17% and 2.58% in top income quintile.

- Chetty et al. (2020)
  The authors build a real-time and granular level database to keep track of the impact of the pandemic on key indicators in the U.S. Their economic tracker is publicly available here.

- Pieh et al. (2020)
  The authors conduct a representative survey to collect data on the mental health consequences of the pandemic in Austria during the spring of 2020. Considering several measures of depressive symptoms, stress, anxiety, or sleep quality, they found that after four weeks of lockdown, mental health burden was significantly higher for low-income households.

- Stiglitz (2020)
  The authors study a two-sector model with mobility constraints and uncertainties on the depth, duration, and long-term consequences of the pandemic. As labor intensive production becomes more costly with infection risk, the uncertainty on the duration of the pandemic trigger a shift toward automation, reinforcing inequalities.

- del Rio-Chanona et al. (2020)
  The authors estimate the supply and demand shock induced by the pandemic in the early spring 2020 and analyse the effect on employment and wages in the U.S. The supply shock is thought as a labor supply shock and is estimated by constructing a measure of remote work feasibility at the occupation and sector level. The demand shock is estimated using the estimates of the U.S. Congressional Budget Office (2006) who aimed at evaluating the effect on an influenza pandemic. Authors aggregate the two shock and estimate the impact on employment and wages.
**Figure A.7: Changes in employment by wage quartiles from January to November 2020, in the U.S.**

*Notes:* Figure from Chetty et al. (2020).
This figure shows the changes in employment by wage quartile relative to January 2020. Each quartile is computed based on pre-Covid-19 wage distribution.

### A.4 Consumption and savings

- **Andersen et al. (2020)**
  
The authors use daily transaction data in Denmark to estimate the effect of lockdown on consumer spending, between January and March 2020. They estimate that overall, aggregate spending decreased by 27% as compared to a counterfactual situation without Covid-19.

- **Hacıoğlu-Hoke et al. (2021)**
  
The authors use fintech app high frequency data on transactions in the U.K. from January to June 2020 to estimate the change in spending behaviours. They estimate that in April 2020, median expenditure declined by 40% relative to its 2019 level.

- **Bounie et al. (2020)**
  
The authors use high frequency data from French bank card transactions during the spring 2020 to evaluate the effect of lockdown on consumption and savings behaviour. They estimate that overall, net financial wealth has increased by €45 billions compared to a counterfactual situation without Covid-19. They refer to this increase as “excess savings”.

- **Cox et al. (2020)**
The authors use U.S. household-level bank account data to estimate the change in spending and savings behaviour during Spring 2020. They estimate that by the end of March 2020, total spending per household declined by 35% relative to its 2019 level.

- **Falcettoni and Nygaard (2021)**
  The authors review the literature developed in 2020 on Unemployment Insurance and stimulus checks in the U.S.

- **García-Montalvo and Reynal-Querol (2020)**
  The authors use fintech app high frequency data on transactions in Spain from March to June 2020 to estimate the change in spending behaviours. They estimate that in April 2020, total expenditure declined by more than 40% relative to its 2019 level.

### Table A.1: Patterns of spending behaviour during spring 2020 across income distribution

<table>
<thead>
<tr>
<th>Citation</th>
<th>Higher cuts in spending at the top of the distribution</th>
<th>Steeper return to pre-Covid-19 level of spending at the bottom of the distribution</th>
</tr>
</thead>
<tbody>
<tr>
<td>Bounie et al. (2020) FR</td>
<td>High-income group accounts for 55% of excess savings.</td>
<td>By June, the 2020/2019 expenditure ratio reach 1 for high income and attained 1.2 for low-income group.</td>
</tr>
<tr>
<td>Cox et al. (2020) US</td>
<td>High-income group accounts for 50% of total cut in spending.</td>
<td>By May high income spending remains 20% below 2019 level while low-income almost return to pre-Covid-19 level.</td>
</tr>
<tr>
<td>Hacıoğlu-Hoke et al. (2021) UK</td>
<td>High-income group accounts for 45% of total cut in spending.</td>
<td>By June, high income spending remains 35% below 2019 level while low-income were 15% below.</td>
</tr>
</tbody>
</table>

**Notes:** In Bounie et al. (2020) households are ranked according to their level of expenditure in 2019. High and low income refer to top and bottom decile respectively. The excess savings refers to the savings accumulated due to the spring lockdown. In Cox et al. (2020) high and low-income groups refer to top and bottom quartile of 2019 income distribution. In Hacıoğlu-Hoke et al. (2021) these groups refer to households with after tax income above £40,000, or below £20,000 respectively.

### A.5 Digital skills and learning opportunities

- **Chiou and Tucker (2020)**
  The authors use mobile devices data to keep track of physical mobility in the U.S. They found that in March 2020, people in high-income regions were more likely to stay at home. They show that this correlation is driven by the access to high internet speed.

- **Ong et al. (2020)**
  The authors use data from the U.S. Census Bureau’s weekly Household Pulse Survey to evaluate the effect of the pandemic on the digital divide in virtual learning, on the fall 2020 semester. They find that low-income households are more impacted with limited access to computers or internet.

- **Beaunoyer et al. (2020)**
The authors analyse the reciprocal effects of the pandemic and the digital divide, arguing that the two phenomena are mutually reinforcing.

- Lai and Widmar (2021)
  The authors use publicly available data on internet performance in the U.S. between December 2019 to June 2020 to evaluate the digital divide along a geographic dimension. They find that rural areas are associated with lower Internet speed at the county level.

B Inequalities across sectors and Regions

B.1 Sectoral Inequalities

- Chundakkadan et al. (2020)
  The authors use firm-level data from 13 different countries (Albania, Chad, Cyprus, Georgia, Greece, Guinea, Italy, Moldova, Niger, Russia, Togo, Zambia, and Zimbabwe) to evaluate the consequences of the pandemic on financially constrained firms. They use a probit model including firms controls, countries and regions fixed effects, to estimate the effect of financial constraints on the probability of firms closures and on the probability that firms layoff workers.

- Bellucci et al. (2020)
  The authors use data on Venture Capital (VC) investments in 160 countries, between 2018 and July 2020. Using difference in difference estimation, they show that the pandemic triggered a significant reallocation of VC flows toward health-related sectors.

- Sherif (2020)
  The author use U.K. Dow Jones data between January 20 to May 20 to evaluate the effect of the pandemic on the stock market performance of different sectors. The finding suggests that the information technology sector (defined using the classification proposed by Bloomberg) performed significantly better than the market. On the contrary, consumer discretionary sector, which refers to people transportation, beverages, tourism and leisure significantly underperformed the market.
Figure B.1: Expenditure changes in France, before and after the spring lockdown, by sector

Notes: Figure from Bounie et al. (2020).
This figure shows the evolution of aggregate weekly credit card expenditures in France during the lockdown period (from mid March to mid May, in blue) and post-lockdown period (from mid May to August, in red). Each bar measures the overall losses or gains made during each period as a fraction of annual 2019 expenditures, controlled for seasonality and trend.
**Figure B.2: Percentage change in small business revenue in the US**

*Notes: Figure from Opportunity Insights | Economic Tracker.*

This figure shows the evolution of the percentage change in small business revenue in the U.S. relative to January 2020, until March 2021, for retail and transportation sector (in red), education and health services (in green) and for leisure and hospitality (in blue). Data on business revenues are seasonally adjusted.
B.2 Regional Inequalities in health impacts and consequences of remote work

- **Weill et al. (2020)**
  The authors use mobile device location data to measure the changes in human mobility in the U.S. between January and April 2020. Controlling for county fixed effects, they found that mobility declined significantly more in wealthier areas.

- **Siddique et al. (2020)**
  The authors use county and state level data to evaluate the relation between the pandemic and poverty in the U.S. between January and July 2020. Controlling for state fixed effects, they found that at the county level, higher poverty leads to higher infection and fatality rates.

- **Brandily et al. (2020)**
  Authors use municipalities level data to evaluate the heterogeneous impact of the pandemic on mortality in France until June 2020. They estimate that mortality due to Covid-19 was twice higher for municipalities in the bottom quarter of the national income distribution than in other municipalities. They show that the higher concentration of occupations involving physical proximity plays a key role, and with poor housing conditions, explains 77% of the difference between poor and rich municipalities.

- **Kapitsinis (2020)**
  The authors use territorial level data (NUTS 2) to evaluate the uneven spread of the virus between regions in nine EU countries, from January to May 2020. They identify several key factors shaping regional inequalities in terms of Covid-19 mortality, including air pollution, previous health expenditure, the share of elderly, and urbanization rates.

- **Ozguzel et al. (2020)**
  The authors use data from the European Labor Force Survey to evaluate transition to remote work at a regional level in 27 EU countries, Switzerland, Turkey, and the U.S. during lockdown periods. They find a strong heterogeneity in the potential for remote work between and within regions as well as important disparities between urban and rural regions.

- **Delventhal et al. (2021)**
  The authors develop a quantitative model of the Los Angeles metropolitan area to evaluate the effect of a permanent increase in remote work. Their model predicts that more productive workers move to the periphery, reducing congestion and commuting time but also housing prices in core locations. Workers who continue to work on-site stay in core locations.
Figure B.3: Share of jobs that can potentially be performed remotely, between and within countries in Europe, 2018

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<tr>
<th>Country or Region</th>
<th>Minimum</th>
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<td>Turkey</td>
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Notes: Figure from Ozguzel et al. (2020).
This figure shows the number of jobs in each country or region that can be carried out remotely as the percentage of total jobs. Countries are ranked in descending order by the share of jobs in total employment that can be done remotely at the national level. Regions correspond to NUTS-1 or NUTS-2 regions depending on data availability. The authors built a classification of occupations in terms of potential for remote work based on the one proposed by Dingel and Neiman (2020) and applied it to the European Labor Force Survey.

C Gender inequality

C.1 Remote work, Working hours and unemployment

- Reichelt et al. (2020)
  The authors want to answer how men’s and women’s employment status, working hours and working arrangements (main place of work) changed during the pandemic in the US, Germany and Singapore. For this purpose they have collected a representative sample of the adult population in these countries using the survey platform YouGov. The data is representative by age, gender, and education level in all three countries, while in the US, the sample is also representative by race and region. From the complete sample of 5,008 respondents, the authors have used a subsample of 2,594 respondents who reported that they worked full- or part-time in January 2020, answering retrospective questions about their employment status (49.7% US, 21.1% Germany, 29.2% Singapore). Main dependent variables for the questions regarding the gendered impact of Covid-19 on employment transitions are threefold: (i) whether the respondents transitioned to working from home between January and the time of the interview; (ii) whether the respondents experienced a significant reduction in working hours - meaning
a reduction of at least 10 hours per week compared to their working hours in January; and (iii) whether
the respondents have lost their full- or part-time employment since January 2020. Main independent
variable for the analysis is the respondent’s gender. To assess whether and to what extent women’s
employment outcomes are more affected by labor market disruptions related to Covid-19, the authors
first compare the unconditional mean differences across groups and use simple t-tests. They also
calculate linear probability models for each of the three transitions in the respondents employment
outcomes, where they are interested in the statistical association between the gender dummy and the
outcomes of interest. This allows to assess whether transition probabilities are significantly different
between men and women. The results suggest that women have a 7 percentage points higher likelihood
of having experienced a transition to working from home, a 5 percentage points higher likelihood of
having reduced their weekly hours by more than 10, and a 3 percentage points higher likelihood of
having transitioned to unemployment since January (figure C.1). Additional analyses reveal that gender
differences are not due to differences in labor market participation across countries. Gender differences
in the transition to working from home and in reducing hours mostly hold even after accounting for
male-female differences in socio-demographics, income, and pre-Covid employment relations. Gender
differences in transitions to unemployment, however, disappear, which can mostly be ascribed to women
having worked part-time and having had lower incomes in January, factors associated with a higher
risk of becoming unemployed.

• Farre et al. (2020)
The authors investigate how the social distancing measures and stay-at-home orders affected the gender
inequality in employment and time devoted to childcare and household chores in Spain. For this
purpose, they ran a survey on a representative sample of the Spanish population aged 24 - 50 in early
May 2020, during the lockdown driven by the spike in COVID-19 cases in mid-March. The survey was
carried out by a survey company Ipsos and it was implemented with quotas by regions, education, and
family composition to ensure the representatives along these dimensions. The final sample size was
5,001 individuals. Main dependent variables to assess gender differences in labor market outcomes are:
(i) whether the respondents became unemployed during the lockdown; and (ii) whether the respondents
transitioned to working from home during the lockdown. Main independent variable for the analysis is
the respondent’s gender. The authors employed linear probability models for each of the two transitions
in the respondents employment outcomes, and were interested in the statistical association between the
gender dummy and the outcomes of interest. This allowed to assess whether transition probabilities are
significantly different between men and women. They find that previously employed women (before the
lockdown) are significantly more likely to be furloughed and unemployed than men. Also, women were
6.5 percentage points more likely than men to work from home during the lockdown. To assess changes
in the distribution of unpaid work at home - childcare and housework, authors focus on two-parent
household of opposite sex with children. This results in the sample of 3,894 individuals. Authors find
that even though men increased their participation in housework and childcare, still most of the burden
fell on women, who were already doing most of the housework before the lockdown, and who on average
spend around 10 hours more per week on childcare compared to fathers. Overall, they conclude that
the Covid-19 crisis appears to have reinforced gender inequalities in both paid and unpaid work in the
short-term in Spain.

• Mongey and Weinberg (2020)
The authors categorize occupations in the US by a measure that captures the likelihood that jobs can
be conducted from home (Dingel and Neiman (2020)), as well as a measure of low personal proximity
in the workplace. The former relates to how well work can be done under social distancing policies, the
latter relates to how quickly occupations might come back offline. Understanding how individuals vary
across these occupations is important for targeting economic policies designed to assist workers. The
authors combine the Bureau of Labor Statistics’ Current Population Survey (CPS) with a version of the
Dingel and Neiman (2020) classification of occupations’ capacity to work from home and add a measure
of personal proximity in the workplace. These two measures were constructed using the data from the
Department of Labor’s Occupational Information Network (O*NET) data. The results suggest that
individuals in occupations that score highly in terms of work-from-home and occupations with high
personal proximity are more likely to be women. This suggests that the employment effects of broad social distancing policies may be less severe for women who work from home, but later integration into the economy may be more difficult due to the high share of women who work in occupations with high personal proximity.

- **Bonacini et al. (2021)**
  The authors start from the survey on the Participation, Labor and Unemployment (PLUS), which provides reliable statistics on labor market phenomena and has 45,000 individuals, and merge it with the 2013 wave of the Italian Survey of Professions (ICP) which reports information on the nature and content of the work for about 16,000 Italian workers. Adopting the work-from-home feasibility index from Barbieri et al. (2020) the authors apply it to the detailed data set and find that women experience 5 percentage points higher feasibility to work from home than men in Italy.

- **Del Boca et al. (2020)**
  The authors use a representative sample of 800 Italian working women interviewed in April and July in 2019 and April 2020. The questionnaire gathers rich information on changes in the respondents' employment status, working hours, childcare, income, and satisfaction regarding their work and family during the pandemic. By employing linear probability models, the authors confirm that Italian women have higher likelihood to have continued working from home during the pandemic compared to Italian men.

- **Hatayama et al. (2020)**
  The authors use skill surveys in 53 countries to estimate jobs’ amenability to work from home. The results suggest that across most countries women are more amenable to working from home. This is because they are less likely to have jobs intensive in physical/manual work than men.

- **Adams-Prassl et al. (2020b)**
  Using real time surveys, the authors document that women were 7 (5) percentage points more likely to lose their jobs (compared to men) in the US and U.K., while in Germany gender does not predict job loss significantly. They found that the share of tasks that can be done from home within occupation and industry is a powerful predictor of the share of workers that lost their jobs, as it alone can explain more than 50% of the variation in job loss due to COVID-19 across occupations in the US, the U.K. and Germany (figure C.4). Additionally, workers with permanent, salaried fixed hour contracts were less likely to be affected compared to workers who were on temporary contracts, non-salaried and whose hours varied (and later groups better describe women).

- **Sevilla and Smith (2020)**
  The authors collected real-time data for 4,250 respondents aged 18 - 60 in the U.K., in May 2020. They confirm the results from Adams-Prassl et al. (2020b) and show that women in the U.K. are 7 percentage points less likely than men to still be working, 4 percentage points more likely to be furloughed, and 2 percentage points more likely to be not working.

- **Oreffice and Quintana-Domeque (2020)**
  The authors use an online sample of approximately 1,500 respondents in Prolific, representative of the U.K. population along the dimensions of age, sex, and ethnicity. They show that women are 4.2 percentage points more likely than men to have lost their job due to COVID-19. They argue that the main driver of the results is the fact that women are concentrated in sectors which are disproportionately affected by the crisis.

- **Del Boca et al. (2020)**
  The authors use a representative sample of 800 Italian women interviewed in April and July 2019 and in April 2020. The questionnaire gathers information on changes in the respondents’ employment status, working hours, income, and satisfaction regarding their work and family during the pandemic. They discover that women in Italy are less likely to have kept working in their usual workplace than men (figure C.3).
• Kristal and Yaish (2020)
Based on a longitudinal survey of 2,040 adult Israeli men and women, who were employed or self-employed in the first week of March 2020, the authors study the impact of the pandemic on the Israeli labor market. Results are in line with other countries and confirm that women’s employment has been more harshly affected than men’s. The explanations they offer are that a high fraction of women were employed in low- and medium-wage jobs before the pandemic; more Israeli women than men have non-standard employment relations; fewer women hold managerial positions in the labor market; women are more likely than men to work in occupations and sectors of the economy that were hit the hardest; and mothers of young children are more likely than fathers to leave the employment and care for the children.

C.2 Childcare and housework

• Adams-Prassl et al. (2020b)
Using real-time surveys, the authors ask respondents in the US, the U.K., and Germany about the time spent on active childcare and homeschooling. After restricting the sample to individuals with children and who report working from home, the authors find that women spend a lot more time on active childcare and homeschooling than men. On average, that is around 1.5 hours per day more than men (figure C.5).

• Alon et al. (2020)
The authors combine several data sets for the US (American Time Use Survey; American Community Survey; and US Census Bureau) and find that among the couples with children who both work full time, women provide about 60% of childcare hours. When children are younger, the share is even higher. They also argue that it is likely that the situation will persist throughout the crisis, as the factors which initially led to this arrangement (e.g. relative income, relative bargaining power, influence of traditional social norms and models) will continue to hold.

• Sevilla and Smith (2020)
The authors collect real-time data on daily lives of U.K. families with young children. They document that women are doing the majority of the childcare, irrespective of their employment. On average, women have been doing 30.3 additional hours of childcare per week, compared to 19.4 hours done by men. The amount of additional childcare provided by men is very sensitive to their employment status - the allocation is more equal in households where men are working from home or where they have been furloughed / lost their jobs.

• Andrew et al. (2020)
The authors use real-time data on 4,915 parents in England who live with at least one child in one of eight different school years (which roughly corresponds to having a child between 4 and 15). They note that mothers are spending much longer hours in childcare and housework than their partners, and that they are spending a larger fraction of their paid work hours having to juggle work and childcare. They conclude that gender differences in the allocation of domestic work cannot easily be explained by differences in employment or earnings.

• Dijst et al. (2021)
The authors focus on a survey of 344 individuals in Luxembourg that are in stable relationships with a child younger than 13 years old. By comparing the time spent on 9 activities per week (work outside; work at home; housework; childcare; commuting; leisure; helping; nothing; other) they show that that women have on average spent 12 hours per week more than men on childcare.

• Hupkau and Petrongolo (2020)
The authors use the Covid-19 supplement to the Understanding Society longitudinal survey in the U.K.. They also find that women shouldered a higher share than men of the increased childcare needs (9.5 extra hours for women as opposed to 6.9 extra hours for men, figure C.6).
• **Oreffice and Quintana-Domeque (2020)**
The authors collected new data on mental well-being, perceptions, employment, health, and behaviours via an online survey of 1,500 Prolific respondents in the U.K.. Authors find that between February and June 2020 women have decreased their working hours, but increased housework and childcare much more than men. These gender differences are not driven by differences in age, ethnicity, education, family structure, income in 2019, current employment status, place of residence or living in rural/urban areas.

• **Hipp and Büning (2020)**
The authors use three waves of an online survey conducted between March and August 2020. They find that parenthood might be a driver of increased gender inequality during Covid-19 in Germany. Parents were more likely than non-parents to work fewer hours during the pandemic than before, and mothers were more likely than fathers to work reduced hours once lockdowns were lifted. Moreover, mothers shouldered more of the childcare work than fathers did, while the division of the remaining housework was unchanged.

• **Zoch et al. (2020)**
The authors use novel data from two panel surveys of the National Educational Panel study and its supplementary Covid-19 web survey in Germany. They examine the short-term consequences for care arrangements among working parents, who were affected by the school and nursery closures in Germany. The study finds that mothers continue to play a key role in the care-arrangements during the pandemic, and that about one-third of families relied exclusively on maternal care, compared to 4-6% of families which relied on exclusive paternal care.

• **Del Boca et al. (2020)**
The authors use a representative sample of 800 Italian women interviewed in April and July 2019 and in April 2020. In Italy, they register similar patterns of gender inequality in unpaid work - most of the additional workload associated to Covid-19 falls on women.

• **Fodor et al. (2020)**
The authors have use data collected in May 2020 in Hungary and land with a sample size of 1,900 adult Hungarians, representative along the dimensions of gender, age, education, type of settlement and administrative regions. They investigate the impact of Covid-19 on the gender division of childcare work in Hungary and conclude that women’s contributions grew significantly more than men’s and the gap between men and women has increased in absolute work hours by close to 5 hours for women.

### C.3 Mental health

Covid-19 crisis has not only affected women’s professional life, but also their work-life balance and well-being. Thus far, the studies have documented a stark reduction in women’s well-being compared to men. Women seem to me more worried and pessimistic about the current situation, they also report to feel lonely because of the social distancing measures and separation from family, friends, and colleagues. Women also feel much more anxious, overwhelmed with childcare and housework and with distorted work-life balance.

• **Adams-Prassl et al. (2020a)**
The authors document the impact of state-wide stay-at-home measures on mental health using real time surveys in the US. They show that the lockdown measures have lowered mental health by 0.085 standard deviations, and that this large negative effect is entirely driven by women. This also means that the existing gender gap in mental health has increased by 66%. The explanation for this is not clear.

• **Oreffice and Quintana-Domeque (2020)**
The authors explore the effect of lockdown on mental health in the U.K. and find that women’s mental health in mid-June 2020 is worse than men’s along several dimensions: general anxiety disorder, depression, anxiety attacks, and loneliness.
• Banks and Xu (2020)
The authors confirm the results for the U.K. using the 12-item Generalised Health Questionnaire (GHQ-12) measure of mental health, distress and well-being and conclude that the effects of Covid-19 on mental health are substantial and are greater for women.

• Etheridge and Spantig (2020)
The authors document twice as large decline in mental well-being for women as for men in the U.K. after the onset of the Covid-19 pandemic. They seek to explain this gender gap by exploring gender differences in family and care responsibilities; financial and work situations; social engagement; health situations and health behaviors. Differences in family and care responsibilities play some role, but most of the gap is explained by social factors. Women reported having more close friends before the pandemic than men, and increased loneliness after the pandemic’s onset. Other factors are similarly distributed across genders and do not play a significant role.

• Czymara et al. (2020)
The authors confirm that women are much more worried about social contact and childcare compared to men who are most worried about paid work and economy. Another research, this time from Switzerland, demonstrates that women process the pandemic harder than men.

• Brülhart and Lalive (2020)
The authors use helpline calls in Switzerland to measure psychological and social suffering in the population at a daily frequency. They find that women call the helpline twice more frequently than men.

• Yamamura and Tsutsui (2020)
The authors document that Covid-19 has stronger impact on women than on men in Japan, as they feel more anxiety and fear due to the on-going situation.

• Gao and Sai (2020)
Lastly, the author shares a personal reflection on how Covid-19 pandemic affects the working lives and well-being of single female academics who live alone in the U.K.. The author says that "Spending almost all time alone aggravates feelings of disconnectedness and loneliness".
**Figure C.1: Gender differences in Covid-related changes in labor market outcomes in the US, Germany and Singapore, Reichelt et al. (2020)**

*Notes:* First dependent variable is ‘transition to working from home’ which presents whether the respondents transitioned to working from home between January 2020 and the time of the interview (May or June 2020). Second dependent variable ‘reduction of hours > 10’ stands for whether the respondents experienced a significant reduction in working hours, meaning a reduction of at least 10 hours per week compared to their working hours in January 2020. Lastly, dependent variable ‘transition to unemployment’ assesses whether the respondents have lost their full- or part-time employment since January 2020. Main independent variable for this analysis is the respondent’s gender. 95% confidence intervals, survey weights used $N = 2,589$, *$p < 0.05$; **$p < 0.01$; ***$p < 0.001$.

**Figure C.2: Gender egalitarian attitudes, Reichelt et al. (2020)**

*Notes:* The authors assess whether Covid-19-related changes in transitions to unemployment, reduction in working hours, or transitions to working from home of respondents and their partners are associated with men’s and women’s gender-role attitudes. The graph shows estimates for the relationship between the respondents’ and/or their partners’ transitions on all three measures for gender-role attitudes (based on the arithmetic mean, iterated principle factor analysis (IPF), and principal component analysis (PCA)). Bars around the point estimate show the 95% confidence intervals for each estimate. Sample conditioned to respondents who worked full- or part-time in January and whose cohabiting partners worked full- or part-time in January; Survey weights used; Bars signify 95% confidence intervals.
Notes: Figure presents the working arrangements of working women and their partners during the COVID-19 emergency in Italy. N = 520.

D Inequalities in Education
Figure C.4: Job loss probability to Covid-19 by % tasks that can be done from home in the US, the U.K. and Germany Adams-Prassl et al. (2020b)

Notes: The figure displays the percentage of people who lost their job due to the coronavirus outbreak by the percentage of tasks respondents report being able to do from home (summarized into quintiles represented on the x-axes). The thin black bars represent the 95% confidence intervals. The figure shows the share of individuals who were in paid work four weeks before data collection that lost their job due to Covid-19.
**Figure C.5:** Hours spent on a "typical" work day during the past week on active childcare and homeschooling in the US and the U.K. **Adams-Prassl et al. (2020b)**

![Bar chart showing hours spent on childcare and homeschooling in the US, UK, and Germany.](image)

*Notes:* The figure shows average number of hours that men and women reported spending on childcare and homeschooling. The authors restrict the sample to individuals with children who report working from home, and whose answers to the time use questions combined to not exceed 24 hours. The thin black bars represent the 95% confidence intervals.

**Figure C.6:** Gender gaps in housework hours, before and during Covid-19 in the U.K. **Hupkau and Petrongolo (2020)**

![Bar chart showing gender gaps in housework hours before Covid-19 (2016-2017) and during Covid-19 (April - May 2020).](image)


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A-25
Figure D.1: Activities of students with low and high test scores before and during the school closures in Germany, Grewenig et al. (2020)

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<th>School activities</th>
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<th>Learning for school</th>
<th>Hours per day</th>
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<td>During Corona</td>
<td>2.3</td>
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<td>-3.7</td>
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Conducive activities

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<td>1.6</td>
<td>3.5</td>
<td>+0.4</td>
</tr>
<tr>
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<td>1.7</td>
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Detrimental activities

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<th>Social media</th>
<th>Online media</th>
<th>Hours per day</th>
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<td>1.2</td>
<td>0.9</td>
<td>4.6</td>
<td>+1.7</td>
</tr>
<tr>
<td>During Corona</td>
<td>1.5</td>
<td>1.0</td>
<td>1.8</td>
<td>6.3</td>
<td></td>
</tr>
<tr>
<td>High-achievers</td>
<td>1.2</td>
<td>1.0</td>
<td>0.8</td>
<td>4.8</td>
<td>+1.5</td>
</tr>
<tr>
<td>During Corona</td>
<td>1.4</td>
<td>1.0</td>
<td>1.2</td>
<td>4.8</td>
<td></td>
</tr>
</tbody>
</table>

Notes: The figure shows average hours spent on different activities on a typical workday reported by students. During Corona is a period of school closures due to Covid-19. Before Corona stands for a period before the school closures. Students with low versus high test scores are defined as students with an average grade in mathematics and German below versus at-or-above the median for their respective school type.

Figure D.2: Students’ weekly effort for school matters in Switzerland, Austria and Germany – Huber and Helm (2020)

Notes: This figure shows the number of hours spent each week on school matters reported by students in Switzerland, Austria and Germany. 13% of the students are 6 - 12 years old, 32% are 13 - 15 years old and 55% are 16 - 20 years old. N = 2.063.
**Figure D.3: Technical capacities for web-based formats by country – Huber and Helm (2020)**

<table>
<thead>
<tr>
<th>Country</th>
<th>Authors</th>
<th>Phenomenon</th>
<th>Direction</th>
</tr>
</thead>
<tbody>
<tr>
<td>Switzerland</td>
<td>11%</td>
<td>8%</td>
<td>25%</td>
</tr>
<tr>
<td>Austria</td>
<td>10%</td>
<td>17%</td>
<td>19%</td>
</tr>
<tr>
<td>Germany</td>
<td>31%</td>
<td>25%</td>
<td>19%</td>
</tr>
</tbody>
</table>

*Notes:* The figure represents the answers by school staff to the question whether technical capacities in the school are sufficient for web-based formats. The school staff are distributed according to school level as follows: 5% from kindergartens, 24% from primary schools, 35% from lower secondary schools, 21% from upper secondary schools, 11% from vocational schools and 5% from other schools.

**Table D.1: Children and school closures due to Covid-19**

<table>
<thead>
<tr>
<th>Country</th>
<th>Authors</th>
<th>Phenomenon</th>
<th>Direction</th>
</tr>
</thead>
<tbody>
<tr>
<td>192 countries</td>
<td>Psacharopoulos et al. (2020)</td>
<td>Loss in earnings</td>
<td>Increase</td>
</tr>
<tr>
<td>US</td>
<td>Fuchs-Schündeln et al. (2020)</td>
<td>Loss in earnings</td>
<td>Increase</td>
</tr>
<tr>
<td>US</td>
<td>Fuchs-Schündeln et al. (2020)</td>
<td>High school degree</td>
<td>Decrease</td>
</tr>
<tr>
<td>DE, AT, CH</td>
<td>Huber and Helm (2020)</td>
<td>Learning</td>
<td>Decrease</td>
</tr>
<tr>
<td>DE</td>
<td>Grewenig et al. (2020)</td>
<td>Learning</td>
<td>Decrease</td>
</tr>
<tr>
<td>U.K.</td>
<td>Andrew et al. (2020)</td>
<td>Learning</td>
<td>Decrease</td>
</tr>
<tr>
<td>US</td>
<td>Agostinelli et al. (2020)</td>
<td>Learning</td>
<td>Decrease</td>
</tr>
<tr>
<td>NL</td>
<td>Engzell et al. (2020)</td>
<td>Learning</td>
<td>Decrease</td>
</tr>
<tr>
<td>BE</td>
<td>Maldonado and De Witte (2020)</td>
<td>Learning</td>
<td>Decrease</td>
</tr>
<tr>
<td>US</td>
<td>Bacher-Hicks et al. (2021)</td>
<td>Search intensity</td>
<td>Decrease</td>
</tr>
<tr>
<td>US</td>
<td>Bao et al. (2020)</td>
<td>Reading ability</td>
<td>Decrease</td>
</tr>
<tr>
<td>DE, AT, CH</td>
<td>Huber and Helm (2020)</td>
<td>Technology</td>
<td>Insufficient</td>
</tr>
<tr>
<td>U.K.</td>
<td>Andrew et al. (2020)</td>
<td>Technology</td>
<td>Insufficient</td>
</tr>
<tr>
<td>IE</td>
<td>Doyle (2020)</td>
<td>Technology</td>
<td>Insufficient</td>
</tr>
</tbody>
</table>

*Notes:* The table summarizes results from a variety of studies on the effects of Covid-19 and school closures on education-related outcomes.
**Figure D.4: Estimates of Learning Loss for the Whole Sample and by Sub-group and Test in the Netherlands — Engzell et al. (2020)**

*Notes:* The graph shows estimates of learning loss from a difference-in-differences specification that compares learning progress between the two testing dates in 2020 to that in the three previous years. Statistical controls include time elapsed between testing dates and a linear trend in year. Point estimates with 95% confidence intervals, robust standard errors accounting for clustering at the school level. One percentile point corresponds to approximately 2.5% of a standard deviation. Where not otherwise noted, effects refer to a composite score of Maths, Spelling, and Reading.
## E Policy Proposals

### Table E.1: Cost per job estimates of Regional Selective Assistance

<table>
<thead>
<tr>
<th>Program</th>
<th>Country</th>
<th>Program Description</th>
<th>Method</th>
<th>Unit</th>
<th>Cost per job (2010 USD)</th>
<th>Source(s)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Regional Selective Assistance</td>
<td>UK</td>
<td>Investment subsidies to businesses in disadvantaged areas.</td>
<td>IV</td>
<td>Area (wards)</td>
<td>3,541</td>
<td>Criscuolo et al. (2019)</td>
</tr>
<tr>
<td>Regional Selective Assistance</td>
<td>UK</td>
<td>Investment subsidies to businesses in disadvantaged areas.</td>
<td>DD</td>
<td>Area (wards)</td>
<td>24,662</td>
<td>Criscuolo et al. (2019)</td>
</tr>
<tr>
<td>Regional Selective Assistance</td>
<td>UK</td>
<td>Investment subsidies to businesses in disadvantaged areas.</td>
<td>DD</td>
<td>Area (wards)</td>
<td>26,572</td>
<td>Criscuolo et al. (2019)</td>
</tr>
<tr>
<td>Empowerment Zones</td>
<td>US</td>
<td>Grants, hiring credits and other benefits for businesses in distressed urban areas.</td>
<td>DD</td>
<td>Area (tract)</td>
<td>18,295</td>
<td>Bartik (2010), Busso et al. (2010)</td>
</tr>
<tr>
<td>Empowerment Zones</td>
<td>US</td>
<td>Grants, hiring credits and other benefits for businesses in distressed urban areas.</td>
<td>DD</td>
<td>Area (tract)</td>
<td>63,100</td>
<td>Glaeser and Gottlieb (2008), Busso and Kline (2008)</td>
</tr>
<tr>
<td>New Markets Tax Credit</td>
<td>US</td>
<td>Subsidised capital investment in low-income neighborhoods.</td>
<td>RDD</td>
<td>Area (tract)</td>
<td>50,820</td>
<td>Freedman (2012)</td>
</tr>
<tr>
<td>Small Business Administration loans</td>
<td>US</td>
<td>Guaranteed and partially-guaranteed loans up to $5.5m for small businesses.</td>
<td>IV</td>
<td>Firm</td>
<td>22,781</td>
<td>Brown and Earle (2017)</td>
</tr>
<tr>
<td>Law 488/91</td>
<td>Italy</td>
<td>Capital subsidies to businesses in least-developed regions.</td>
<td>RDD</td>
<td>Firm</td>
<td>42,638</td>
<td>Pellegrini and Muccigrosso (2017)</td>
</tr>
<tr>
<td>Law 488/91</td>
<td>Italy</td>
<td>Capital subsidies to businesses in least-developed regions.</td>
<td>RDD</td>
<td>Firm</td>
<td>68,409</td>
<td>Cerqua and Pellegrini (2014)</td>
</tr>
</tbody>
</table>

**Notes:** This table is extracted from Criscuolo et al. (2019), the three first estimates refer to their paper. Cost per job estimates have been converted from original units to U.S.$ using yearly average exchange rates for the year that costs were reported for and then deflated to 2010 using a U.S. consumer price index from the World Bank. Midpoints are taken where cost per job is reported as a range. In cases where base year is not stated, the last year of reported expenditure is taken. In the methods column: IV is instrumental variable, DD is differences-in-differences and RDD is regression discontinuity design. If more than one source is cited, the first source provides the cost per job estimate based on the effects on employment that are cited in the second source.