

# Global Carbon Inequality, 1990-2019: The Impact of Wealth Concentration on the Distribution of World Emissions

Lucas Chancel (✉ [lucas.chancel@sciencespo.fr](mailto:lucas.chancel@sciencespo.fr))

Paris School of Economics

---

## Article

## Keywords:

**Posted Date:** March 17th, 2022

**DOI:** <https://doi.org/10.21203/rs.3.rs-1404683/v1>

**License:**  This work is licensed under a Creative Commons Attribution 4.0 International License.

[Read Full License](#)

---

**Version of Record:** A version of this preprint was published at Nature Sustainability on September 29th, 2022. See the published version at <https://doi.org/10.1038/s41893-022-00955-z>.

# Global Carbon Inequality, 1990-2019

## The Impact of Wealth Concentration on the Distribution of World Emissions

LUCAS CHANCEL

World Inequality Lab, Paris School of Economics

### Abstract

This paper estimates the global inequality of individual greenhouse gas (GHG) emissions between 1990 and 2019, using a newly assembled dataset of income and wealth inequality and Environmental Input-Output tables. I find that the bottom half of the world population emits 12% of global emissions, while the top 10% emits 48% of the total. The global top 1% share in world emissions rose from 14% in 1990 to 17% in 2019. While two thirds of the inequality in individual emissions was due to inequalities *between* countries in 1990, the situation has entirely reversed: in 2019, 63% of the global inequality in individual emissions was due to gaps between low and high emitters *within* countries. Emissions from investments, rather than from consumption, represent the bulk of emissions of the rich: around 70% of global top 1% emissions come from their investments. This has major implications for contemporary debates on fair climate policies. I stress the need for more systematic data on individual carbon emissions to allow informed policy debates.

### INTRODUCTION

1

2 Climate change and economic inequalities are among the most pressing chal-  
3 lenges of our times. They are also interrelated: failure to contain climate change is  
4 likely to exacerbate inequalities within and between countries [29, 41, 15, 27] and

5 economic inequalities within countries tend to slow the implementation of climate  
6 policies [20, 82]. In order to properly understand interactions between economic  
7 inequality and climate change, systematic data is needed on the distribution of green-  
8 house gases (GHG) emissions between individuals and across the globe, and its  
9 evolution. Such information is currently missing.

10 As a matter of fact, researchers, policymakers and civil society struggle to  
11 establish basic facts about individuals' carbon footprints. National carbon footprints  
12 (i.e. emissions net of the GHG content of goods and services traded with the rest of  
13 the world) are not systematically published by most statistical institutions around  
14 the world, and when this this information is released, it is with several years of delay.  
15 In addition, official publications about GHG emissions are typically blind to the  
16 distribution of these emissions: i.e. which groups of the population contribute most  
17 to GHG emissions growth or to mitigation efforts remains unknown.

18 The present paper addresses these issues by harnessing recent conceptual and  
19 empirical progress in income, wealth and GHG emissions measurement. I mobilize  
20 state-of-the-art global income and wealth inequality data from the World Inequal-  
21 ity Database [3] and GHG emissions data associated to individuals' consumption  
22 and investments, from Multi-Regional Input-Output (MRIO) tables. Compared to  
23 previous work [17, 12, 21], the novelty of this approach is to distribute the total-  
24 ity of GHG emissions to the totality of the world population, from lowest income  
25 groups to the richest, in a systematic and transparent manner and over a thirty-year  
26 period. The method developed in this paper also makes it possible to distinguish  
27 between emissions from individuals' consumption and emissions from investments  
28 and wealth ownership (see Methods).

## 29 I. RESULTS

### 30 *Global equally-split carbon budget and average emissions by regions*

31 According IPCC AR6 report [55], there are approximately 300 Giga tonnes (Gt)  
32 of Carbon Dioxide equivalent (CO<sub>2</sub>e) left to be emitted to limit global warming  
33 below 1.5°C and nearly 900 GtCO<sub>2</sub>e to limit it to 2°C.<sup>1</sup> At current global emissions

<sup>1</sup>Both budgets and given with an 83% confidence to remain under the temperature limit.

34 rates (that is, circa 50 GtCO<sub>2</sub>e in 2021), the 1.5°C budget will be depleted in six  
35 years and 2°C budget in 18 years.

36 In 2021, per capita global emissions neared 6.5 tCO<sub>2</sub>e<sup>2</sup>, significantly above  
37 the Global Equally-split Carbon (GEC) budget.<sup>3</sup> The GEC budget compatible with  
38 keeping global warming under +2°C is of 3.4 tonnes per person per annum (assuming  
39 that the entire budget is spent by 2050). The GEC budget compatible with keeping  
40 global warming under +1.5°C is of 1.1 tonne of CO<sub>2</sub> per annum per person, i.e.  
41 about six times less than the current global average.<sup>4</sup>

42 Average emissions in most world regions are above the 1.5°C and 2°C GEC  
43 budgets, as shown in Figure I (these estimates are net of imports and exports of  
44 CO<sub>2</sub>e embedded in goods and services).<sup>5</sup> Per capita emissions in Sub-Saharan Africa  
45 are around 50% above the 1.5°C GEC budget and are around half of the 2°C GEC  
46 budget. At the other end of the spectrum, emissions in North America are 21 tCO<sub>2</sub>e  
47 per capita (three times the world average and six times higher than the 2°C GEC  
48 budget). In between these two extremes stand South and South-East Asia, at 2.5  
49 tonnes per capita (80% of the 2°C GEC budget) and Latin America at 4.8 tonnes  
50 (70% of world average, 1.4 times the 2°C GEC budget), followed by the Middle East  
51 and North Africa, East Asia, Europe, and Russia and Central Asia, whose averages  
52 fall in the 7.5-10 tonnes range (between one and 1.5 times the world average, and  
53 two to three times more than the 2°C GEC budget).

54 [Figure 1 about here.]

<sup>2</sup>Unless specified, all figures are reported in CO<sub>2</sub>e and include all GHG emissions from human activity, except emissions from land use, land use change and forestry (LULUCF).

<sup>3</sup>To obtain these numbers, I divide the remaining global carbon budget by the cumulative global population that will be emitting it over the coming decades. According to the United Nations [32], there will be 265 billion individual-years between 2020 and 2050.

<sup>4</sup>Assuming that the GEC budget is shared by 2100 rather than 2050 mechanically means smaller per capita values (i.e. 0.4 tonne and 1.1 tonnes to stay below 1.5°C and 2°C, respectively).

<sup>5</sup>See also **Supplementary Information**, Table 4.5, which compares territorial emissions with carbon emissions net of imports and exports of carbon embedded in goods and services, presented here.

55 *Carbon emissions inequalities within world regions*

56 On top of large carbon inequalities between regions, significant inequalities in  
57 carbon footprints are observed within regions of the world. Figure II presents the  
58 carbon footprints of the poorest 50%, the middle 40% and the richest 10% of the  
59 population across world regions. In East Asia, the poorest 50% emit on average  
60 around three tonnes per annum, while the middle 40% emit nearly eight tonnes, and  
61 the top 10% almost 40 tonnes. This contrasts sharply with North America, where the  
62 bottom 50% emit fewer than 10 tonnes, the middle 40% around 22 tonnes, and the  
63 top 10% over 70 tonnes of carbon dioxide equivalent. This in turn can be contrasted  
64 with the emissions in Europe, where the bottom 50% emit nearly five tonnes, the  
65 middle 40% around 10.5 tonnes, and the top 10% around 30 tonnes. Emissions  
66 levels in South and South East Asia are significantly lower, from one tonne for the  
67 bottom 50% to fewer than 11 tonnes on average for the top 10%.

68 It is striking that the poorest half of the population in the US has emission levels  
69 comparable with the European middle 40%, despite being almost twice as poor  
70 as this group. Indeed, this difference is largely due to the carbon-intensive energy  
71 mix in the US (where emissions from electricity are about twice as much as in the  
72 European Union) as well as to more energy-intensive infrastructures and energy  
73 devices.<sup>6</sup>

74 [Figure 2 about here.]

75 European emissions of various income groups are indeed very high by global  
76 standards: the European middle class emits significantly more than its counterparts  
77 in East Asia (around 10.5 tonnes compared with eight tonnes, respectively) and all  
78 other regions except North America. Yet it is also remarkable that the richest East  
79 Asians and the richest 10% in the Middle East emit more than the richest Europeans  
80 (39 tonnes, 34 tonnes, and 29 tonnes, respectively). This difference results from the  
81 higher income and wealth inequality levels in East Asia and the MENA region than  
82 in Europe.

83 Turning to other regions, I find that Russia Central Asia have an emissions  
84 distribution similar to that of Europe, but with higher top 10% emissions (due to

<sup>6</sup>See for e.g. [18] for a discussion of the drivers behind these differences.

85 higher income and wealth inequalities in Russia Central Asia). Sub-Saharan Africa  
86 lags behind, with the bottom 50% emissions around 0.5 tonnes per capita and per  
87 year and top 10% emissions around 7 tonnes. Overall, it stands out that only the  
88 poorest 50% of the population in Sub-Saharan Africa and South and South-East  
89 Asia come in under the 1.5°C GEC budget. Measuring levels against the 2°C GEC  
90 budget, I observe that the bottom half of the population in each region is below or  
91 close to the threshold (apart from North America). Emissions of the poorest half of  
92 Europeans, for instance, are 50% above the 2°C GEC budget. The gap is significant  
93 but within relatively close reach.

#### 94 *Global carbon inequality between individuals*

95 Figure III presents the inequality of carbon emissions inequality between in-  
96 dividuals at the world level. The global bottom 50% emit on average 1.6 tCO<sub>2</sub>e  
97 per annum and contribute to 12% of the total. The middle 40% emit 6.6 tonnes on  
98 average, making up 40.4% of the total. The top 10% emit 31 tonnes (47.6% of the  
99 total). The top 1% emits 110 tonnes (16.8% of the total). Global carbon emissions  
100 inequality thus appears to be very large: close to half of all emissions are due to  
101 one tenth of the global population, and just one hundredth of the world population  
102 (77 million individuals) emits about 50% more than the entire bottom half of the  
103 population (3.8 billion individuals).

104 [Figure 3 about here.]

105 Table I presents more details on the global distribution of carbon emissions. The  
106 bottom 20% of the world population (1.5 billion individuals) emit fewer than 1.8  
107 tonnes per capita per annum. The entry threshold to get in the middle 40% is 3.1  
108 tonnes, and it takes 13 tonnes per capita per annum to get in the top 10%. It takes  
109 130 tonnes to break into the global top 0.1% of emitters (7.7 million individuals).

110 [Table 1 about here.]

#### 111 *The evolution of individual carbon emissions inequalities*

112 How has global emissions inequality evolved over the past decades? A simple  
113 way to represent the evolution of carbon emissions inequality is to plot the average

114 emissions growth rate of each percentile of the global distribution. Global polluters  
115 are ranked from the least emitter to the richest on the horizontal axis of Figure IV,  
116 and their per capita emissions growth rate is presented on the vertical axis. Since  
117 1990, average global emissions per capita grew by about 7% (and overall emissions  
118 grew by 58%, see Table II). The per capita emissions of the bottom 50% grew faster  
119 than the average (32%), while those of the middle 40% as a whole grew more slowly  
120 than the average (4%), and some percentiles of the global distribution actually saw  
121 a reduction in their emissions of between five and 25%. Per capita emissions of  
122 the top 1% emissions grew by 26% and top 0.01% emissions by more than 110%  
123 revealing very unequal dynamics.

124 Per capita emissions matter, but understanding the contribution of each group to  
125 the overall share of total emissions growth is critical. Groups starting with very low  
126 per capita emissions levels can increase their emissions substantially over a given  
127 period, yet still contribute very little to the overall growth in global emissions. This is  
128 in effect what has happened since 1990 (see Table II, last column). The bottom half  
129 of the global population contributed only 16% of the growth in emissions observed  
130 since then, while the top 1% (77 million individuals) was responsible for 21% of  
131 emissions growth. These values are reported in the two boxes of Figure IV.

132 [Figure 4 about here.]

133 [Table 2 about here.]

134 One of the most striking results shown in Figure IV is the reduction in the  
135 emissions of about 5-15% for percentiles p75 to p95. This segment of the world  
136 population largely corresponds to the lower and middle income groups of the rich  
137 countries. In these countries, the working and middle classes have reduced their  
138 emissions over the past 30 years resulting of a combination of improvements in  
139 overall energy efficiency and a compression of their wages as compared to richer  
140 groups of the population [4]. These reductions are insufficient to meet the goals  
141 of the Paris Climate Agreement to limit global warming to 1.5°C or 2°C, but they  
142 contrast nevertheless with the emissions of the top 1% in these countries (and at the  
143 global level), which have significantly increased. I discuss the implications of these  
144 dynamics in section II.

145 Figure V presents the evolution of the top 1% and the bottom 50% shares in  
146 total emissions between 1980 and 2019. Between 1990 and 2019, the global bottom  
147 50% increased its share of total emissions, from around 9.5% to 12%. At the same  
148 time, the top 1% share rose from 14% to close to 17%. Put differently, the gap in  
149 emissions between the top of the distribution and the bottom remained substantial  
150 over the entire period, despite relatively strong growth in emissions from the bottom  
151 50% of the world population.

152 [Figure 5 about here.]

153 Global carbon inequality dynamics are governed by two forces: the evolution of  
154 average emission levels *between* countries and the evolution of emission inequalities  
155 *within* countries. Which one of these two forces has been driving the dynamics  
156 of global carbon inequality over the past decades? Figure VI compares the share  
157 of global emissions that is due to within-country differences with the between-  
158 country differences, using a Theil-index decomposition. In 1990, most global  
159 carbon inequality (63%) was due to differences between countries: then, the average  
160 citizen of a rich country polluted unequivocally more than the rest of the world's  
161 citizens, and social inequalities within countries were on average lower across the  
162 globe than today. The situation has almost entirely reversed in 30 years. Within-  
163 country emissions inequalities now account for nearly two thirds of global emissions  
164 inequality. To be clear: this does not mean that there do not remain significant (often  
165 huge) inequalities in emissions between countries and world regions, on the contrary  
166 (see Figure I). In fact, it means that on top of the great inter-national inequality in  
167 carbon emissions, there also exist even greater inequalities in emissions between  
168 individuals. This has major implications for global debate on climate policies.

169 [Figure 6 about here.]

170 Figure VII (panel A), presents the global distribution of individual carbon emis-  
171 sions. Each color wedge is proportional to the population of a region, and the total  
172 colored area represents the global population. The Figures makes it clear that the  
173 bulk of the world population emits between 1.5 and 8 tonnes, with a mode at c. 2  
174 tonnes. Around 1 billion individuals emit less than 1 tonne of CO<sub>2</sub>e per year, 3

175 billion individuals are found to emit between 3.1 tonnes and 13 tonnes and 7 million  
176 individuals (approximately the top 0.1% of the population) emit more than 130  
177 tonnes per year. Figure VII (panel B) presents the share of population of each region  
178 in each percentile of the global carbon distribution. Sub-Saharan Africa, India and  
179 the rest of Asia excluding China represent the bulk of emitters from the bottom 30%  
180 of the global distribution. Countries like China (which represents the vast majority of  
181 East Asia), Latin America, and MENA are well represented at nearly all levels of the  
182 global distribution, from relatively low emitter groups to very high emitter groups.  
183 This reflects the dual nature of these societies, where extreme polluters live close to  
184 very low polluters. Europe and North America are essentially represented in the top  
185 half of the global distribution, to the right hand side of the graph. A important result  
186 from this graph is the small relative representation of top European emitters among  
187 very top global emitters, especially as compared to North American emitters. Also  
188 significant is the large representation of Chinese among very top global emitters.

189 [Figure 7 about here.]

190 *The weight of investments in wealthy individuals' carbon footprints*

191 The results presented above show that global carbon inequalities are currently  
192 very large and that the share of emissions of very top groups has been rising since  
193 1990. What is driving the rise of emissions at the top of the distribution? The rise in  
194 emissions at the top is due to the increase in income and wealth inequalities within  
195 countries and to the rising share of emissions from wealthy individuals' investments.<sup>7</sup>  
196 I stress that this rise can be observed in all scenarios, even when taking implausibly  
197 low assumptions on the country-level elasticity of emissions between income and  
198 emissions.<sup>8</sup>

199 Individual carbon footprints can be split into emissions from private consumption,  
200 investments or government spending. Consumption related emissions come from the

<sup>7</sup>In the benchmark scenario, the carbon content of a euro of investment is the same across income groups. Alternative assumptions are tested in the Supplementary Information. Even with an elasticity lower than one, the share of investments in wealthy groups' total footprints remains very large (see Methods).

<sup>8</sup>See Methods section (Robustness checks).

201 carbon released by the direct use of energy (i.e. fuel in a car) or its indirect use (i.e.  
202 energy embedded in goods and services consumed by individuals. Investment-related  
203 emissions are emissions associated to choices made by capital owners about new  
204 investments in the production process (i.e. the construction of new machines, new  
205 factories, etc. which will serve the production of goods and services tomorrow).  
206 These emissions are to be attributed to investors, rather than to consumers, because  
207 investors are the sole decision-makers about these investment choices. In line with  
208 earlier studies on carbon footprints, consumers are attributed emissions associated to  
209 the production of the goods they consume.<sup>9</sup>

210 Focusing on the breakdown between consumption and investment emissions,  
211 I find that a large part of emissions from the global top 1% comes from their  
212 investments, rather than from their consumption. In a world where the poorest half  
213 of the population within countries typically owns less than 5% of total wealth [22],  
214 and typically makes less than 5% of investments, it can be expected that investment-  
215 related emissions are highly concentrated, both within countries and at the global  
216 level. The question is how much exactly - and how has the weight of investments in  
217 wealthy individuals' carbon budget evolved over the past decades?

218 Figure VIII presents the share of investments in total emissions of various groups  
219 of emitters at the global level (the global bottom 50%, top 1%, top 0.1% and top  
220 0.01%). Emissions of top groups (top 1% and above) essentially come from their  
221 investments. Investments represent 70% of global top 1% emissions in 2019 vs.  
222 over 90% for the global top 0.01% in my benchmark scenario. In effect, this means  
223 that the global top 1% (77 million individuals) has an investment-related carbon  
224 footprint of 77 tonnes of CO<sub>2</sub>e per capita and per year (vs. 33 tonnes due to their  
225 private consumption), on average in my benchmark estimates. Conversely, emissions  
226 from investments of the global bottom 50% represent only 0.16 tonnes per capita  
227 in 2019 (or 10% of their total emissions of 1.6 tonnes on average). It also appears  
228 that the weight of investments in top groups' per capita footprint has been rising

<sup>9</sup>The notion of responsibility is indeed complex: many consumers are constrained in their consumption choices. Disentangling willing and constrained choices goes beyond the scope of this paper. However, the approach proposed here provides a nuanced approach to individual responsibility, in which consumers are not held responsible for investment decisions made by others, but are responsible for the carbon embedded in their own consumption choices.

229 significantly since the 1990s. This is due to the rise in wealth inequality (meaning  
230 that investments are more concentrated today than they were in 1990), as well as to  
231 the rise in overall emissions associated to investments over the periodn (because of a  
232 change in the nature of investments).<sup>10</sup>

233 [Figure 8 about here.]

## 234 II. DISCUSSION

235 The results presented in this paper reveal the very large level of concentration  
236 of individual carbon emissions that characterizes contemporary global economy:  
237 while a tenth of the global population is responsible for nearly half of all emissions,  
238 a half of the population emits no more than 12% of it.<sup>11</sup> Global carbon inequalities  
239 have been rising at the top of the distribution (Figure V) since 1990. How to explain  
240 this fast increase at the top of the distribution of world emitters? Focusing on rich  
241 countries, we observe that average per capita emissions declined in rich countries  
242 since 1990 (even when factoring in embedded emissions), but incomes and wealth  
243 have become more concentrated at the top of the distribution. In this context, the  
244 carbon footprint of wealthy individuals followed a different trend as compared to  
245 that of the rest of the population. What is observed is a "rebound effect" associated  
246 to high income and wealth level: rising income and wealth levels have been more  
247 important than gains in the GHG intensity of per capita income. I stress that this  
248 effect is very robust to different assumptions on the link between emissions and  
249 income at the household level.<sup>12</sup> In emerging countries such as China or India,  
250 average emissions levels have been growing for nearly all groups of the population,  
251 but the wealthy accounted for a disproportionate share of this growth (they account  
252 for a disproportionate share in consumption growth as well as in investments growth,  
253 see [4, 22]).

<sup>10</sup>In 1990, global emissions from the investment sector of the economy represented 25.6% of the total, vs. 32.3% in 2019, see SI Table 1.

<sup>11</sup>Replaced in perspective, carbon inequalities are lower than income or wealth inequalities (the global top 10% of earners captures 52% of total income and the global top 10% of wealth owners owns 76% of total wealth, see [22]).

<sup>12</sup>See SI Section 7.

254 The rise in emissions from top global emitters since 1990 is even more striking  
255 when compared to emission trajectories of other groups of the population. Indeed,  
256 the poorest 50% in Europe and the US saw their emissions reduced by approxi-  
257 mately 15%-20% since 1990. These reductions are due to the combined effect of  
258 compressed wages and consumption and reduced national per capita footprint in  
259 most rich countries. The consequence is that a large part of the population in rich  
260 countries already appears to be near 2030 national climate targets, when these targets  
261 are expressed in per capita terms. Nationally Determined Contributions (NDCs)  
262 established in the context of the Paris Agreement imply a per capita target of around  
263 10 tonnes of CO<sub>2</sub>e in the US in 2030, vs. around 5 tonnes for European countries  
264 (see Methods). In the US and most European countries, I find that the bottom 50%  
265 of the population meets these 2030 targets (Figure IX and SI Section 8). This is  
266 not the case for the middle 40% and top 10% of the income distribution in these  
267 countries. Wealthier groups are indeed largely above the 2030 climate target. In the  
268 US, the top 10% would need to reduce its average per capita emissions by 87% to  
269 reach the 2030 target, the value is 81% in France.

270 In emerging and developing countries, 2030 climate targets imply an increase in  
271 average per capita emissions, rather than a reduction. But there as well, inequality  
272 matters a lot: in China and India, emissions of the bottom 90% of the population  
273 is below the target, while the wealthiest 10% is already well above it. In China,  
274 the richest 10% of the population would actually need to reduce its emissions by  
275 more than 70% to reach the 2030 target, the figure is 60% in India (Figure IX and SI  
276 Section 8).

277 [Figure 9 about here.]

278 To be clear, no country currently envisages the enforcement of strict per capita  
279 targets to meet its 2030 objectives. These gaps between individual emissions levels  
280 and the implied national target nonetheless raise important questions about the  
281 design of climate policies in the years to come: how to ensure that regulations, tax  
282 instruments and other climate policies effectively address emissions of high emitters?  
283 Put it differently, how to reduce emissions in increasingly unequal societies?

284 There is no straightforward answer to such questions, but it appears that climate

285 policies over the past decades have often targeted low-income and low-emitter  
286 groups disproportionately, while leaving high emitters relatively unaffected. Carbon  
287 taxation, for instance, has been found to place a disproportionate burden on low-  
288 income and low-emitter groups [28, 23, 33]. On the contrary, the carbon price signal  
289 for high and wealthy emitters is often too low to force changes in consumption  
290 patterns. In some cases, price signals are close to nonexistent (for e.g. private jet fuel  
291 is significantly undertaxed as compared to road transport fuels in Europe or the US).

292 It also appears that emissions associated to investment decisions have attracted  
293 only little policy attention so far.<sup>13</sup> While investments represent a significant and  
294 rising share of top emitters' carbon footprints, countries do not impose taxes or  
295 regulations based on the pollution content of individual asset portfolios. This is  
296 paradoxical given that investors have a variety of options at hand to invest their  
297 wealth. This contrasts with low and middle income consumers who might not always  
298 have alternatives, in the short run, to the use of fossil fuels. Taxes on the carbon  
299 content of investments, or on the ownership of polluting assets, could in principle  
300 become attractive tools when more standard carbon taxes fail to address inequality  
301 concerns. Such options could help avoid the risk of political backlash against carbon  
302 taxation, as has been seen in several countries in the recent years [20, 78].

303 The informational, technical and economic conditions under which policies  
304 targeting carbon investments of individuals is a matter of further research. In  
305 that respect, some developments are worth following: the recent European Union  
306 financial taxonomy on sustainable investments [81] could enable a better monitoring  
307 of the carbon content of assets - although the inclusion of gas investments has raised  
308 concerns. Critical progress will also need to be made by governments to properly  
309 monitor individual emissions, in a timely and systematic manner. Ability to produce  
310 this information will also depend on governments' ability to enforce more financial  
311 transparency to trace end-user beneficiaries of financial transactions.<sup>14</sup>

<sup>13</sup>See also [63] who discuss the need to focus on top emitters.

<sup>14</sup>See [58] for a discussion of the many issues associated to the development of financial asset registries.

312

## CONCLUSION

313 This paper mobilizes state-of-the art data on global income and wealth inequality  
314 and systematically combines it with carbon footprints estimates to track the distri-  
315 bution of individual carbon emitters between 1990 and 2019. I find that the global  
316 inequality of carbon emissions is both high and persistent, despite strong economic  
317 growth in the emerging world over the past three decades. The top 10% of global  
318 emitters are responsible for around 48% of global emissions while the entire bottom  
319 50% emits 12% of emissions in 2019. While significant inequalities in average  
320 emissions persist between countries, I find that the bulk of global inequalities in  
321 individual emissions is now due to within-country inequalities. A large and growing  
322 share of top-emitters carbon footprints come from their investments, rather than  
323 from their consumption. In rich countries, emissions of lower income groups de-  
324 clined while emissions of top groups increased significantly. In emerging countries,  
325 emissions of top income groups are now comparable to top groups in rich countries.  
326 These results highlight the need for more policy instruments specifically addressing  
327 emissions of the wealthy. While the results presented in the paper appear to be robust  
328 to a wide range of alternative estimation strategies, I stress at the outset that a lot of  
329 work still needs to be made to properly track carbon emissions inequality between  
330 and within countries. Absent such information, designing fair climate policies will  
331 remain an overly challenging task. All estimates are published online on the World  
332 Inequality Database, as well as the set of computer codes to contribute to more  
333 transparency about these important matters.

334

## REFERENCES

- 335 [1] Angel Aguiar, Badri Narayanan, and Robert McDougall. “An overview of  
336 the GTAP 9 data base”. In: *Journal of Global Economic Analysis* 1.1 (2016),  
337 pp. 181–208.
- 338 [2] Sanna Ala-Mantila, Jukka Heinonen, and Seppo Junnila. “Relationship be-  
339 tween urbanization, direct and indirect greenhouse gas emissions, and ex-

- penditures: A multivariate analysis”. In: *Ecological Economics* 104 (2014), pp. 129–139.
- [3] Facundo Alvaredo, Anthony B. Atkinson, Thomas Blanchet, Lucas Chancel, Luis Bauluz, Matthew Fisher Post, Ignacio Flores, Bertrand Garbinti, Amory Gethin, Jonathan Goupille-Lebret, Clara Martínez-Toledano, Marc Morgan, Theresa Neef, Thomas Piketty, Anne-Sophie Robilliard, Emmanuel Saez, Daniel Waldenstrom, Li Yang, and Gabriel Zucman. *Distributional National Accounts (DINA) Guidelines: Concepts and Methods used in WID.world*. World Inequality Lab, 2020.
- [4] Facundo Alvaredo, Lucas Chancel, Thomas Piketty, Emmanuel Saez, and Gabriel Zucman. “The elephant curve of global inequality and growth”. In: *AEA Papers and Proceedings*. Vol. 108. 2018, pp. 103–08.
- [5] Facundo Alvaredo, Lucas Chancel, Thomas Piketty, Emmanuel Saez, and Gabriel Zucman. “The Elephant Curve of Global Inequality and Growth”. In: *AEA Papers and Proceedings* 108 (2018), pp. 103–08.
- [6] Facundo Alvaredo, Lucas Chancel, Thomas Piketty, Emmanuel Saez, and Gabriel Zucman. *World inequality report 2018*. Harvard University Press, 2018.
- [7] A. B. Atkinson and Thomas Piketty, eds. *Top incomes: a global perspective*. Oxford: Oxford University Press, 2010.
- [8] Giovanni Baiocchi, Jan Minx, and Klaus Hubacek. “The impact of social factors and consumer behavior on carbon dioxide emissions in the United Kingdom: A regression based on input- output and geodemographic consumer segmentation data”. In: *Journal of Industrial Ecology* 14.1 (2010), pp. 50–72.
- [9] Felix Bajard, Lucas Chancel, Rowaida Moshrif, and Thomas Piketty. “Global wealth inequality on WID.world: estimates and imputations”. In: *World Inequality Lab Technical Notes* 16 (2021).
- [10] Thomas Blanchet, Luis Bauluz, Clara Martinez-Toledano, and Alice Sodano. “Estimation of Global Wealth Aggregates in WID.world: Methodology”. In: *WID.world Technical Note 2021/13* (2021).

- 370 [11] Thomas Blanchet, Ignacio Flores, and Marc Morgan. “The weight of the rich:  
371 Improving surveys using tax data”. In: *The Journal of Economic Inequality*  
372 (2022), pp. 1–32.
- 373 [12] Benedikt Bruckner, Klaus Hubacek, Yuli Shan, Honglin Zhong, and Kuishuang  
374 Feng. “Impacts of poverty alleviation on national and global carbon emis-  
375 sions”. In: *Nature Sustainability* (2022), pp. 1–10.
- 376 [13] Milena Buchs and Sylke V. Schnepf. “Who emits most? Associations between  
377 socio-economic factors and UK households home energy, transport, indirect  
378 and total CO2 emissions”. In: *Ecological Economics* 90 (2013), pp. 114–123.
- 379 [14] Milena Büchs and Sylke V Schnepf. “Who emits most? Associations between  
380 socio-economic factors and UK households’ home energy, transport, indirect  
381 and total CO2 emissions”. In: *Ecological Economics* 90 (2013), pp. 114–123.
- 382 [15] Marshall Burke, Solomon M Hsiang, and Edward Miguel. “Global non-linear  
383 effect of temperature on economic production”. In: *Nature* 527.7577 (2015),  
384 pp. 235–239.
- 385 [16] François Burq and Lucas Chancel. “Aggregate Carbon Footprints on WID.world”.  
386 In: *World Inequality Lab Technical Notes* 3 (2021).
- 387 [17] S. Chakravarty, A. Chikkatur, H. de Coninck, S. Pacala, R. Socolow, and  
388 M. Tavoni. “Sharing global CO2 emission reductions among one billion high  
389 emitters”. In: *Proc Natl Acad Sci* 106.29 (2009), pp. 11884–11888.
- 390 [18] Lucas Chancel. “Are younger generations higher carbon emitters than their  
391 elders?: Inequalities, generations and CO2 emissions in France and in the  
392 USA”. In: *Ecological Economics* 100 (2014), pp. 195–207.
- 393 [19] Lucas Chancel. “Towards Distributional National and Environmental Ac-  
394 counts”. In: *Statistical Journal of the IOAS* 36.3 (2020), pp. 597–605.
- 395 [20] Lucas Chancel. *Unsustainable Inequalities: Social Justice and the Environ-*  
396 *ment*. Harvard University Press (Belknap), 2020.
- 397 [21] Lucas Chancel and Thomas Piketty. *Carbon and inequality from Kyoto to*  
398 *Paris (1998-2013) and prospects for an equitable adaptation fund*. October.  
399 Paris School of Economics, 2015.

- 400 [22] Lucas Chancel, Thomas Piketty, Emmanuel Saez, and Gabriel Zucman. *World*  
401 *inequality report 2022*. Harvard University Press, 2022.
- 402 [23] Mireille Chiroleu-Assouline and Mouez Fodha. “From regressive pollution  
403 taxes to progressive environmental tax reforms”. In: *European Economic*  
404 *Review* 69 (2014), pp. 126–142.
- 405 [24] Maarten Christis, Koen Breemersch, An Vercalsteren, and Evelien Dils. “A  
406 detailed household carbon footprint analysis using expenditure accounts–  
407 Case of Flanders (Belgium)”. In: *Journal of Cleaner Production* 228 (2019),  
408 pp. 1167–1175.
- 409 [25] *Climate Action Tracker*. 2021.
- 410 [26] Steven J Davis and Ken Caldeira. “Consumption-based accounting of CO2  
411 emissions”. In: *Proceedings of the National Academy of Sciences* 107.12  
412 (2010), pp. 5687–5692.
- 413 [27] Melissa Dell, Benjamin F Jones, and Benjamin A Olken. “Temperature shocks  
414 and economic growth: Evidence from the last half century”. In: *American*  
415 *Economic Journal: Macroeconomics* 4.3 (2012), pp. 66–95.
- 416 [28] Francis Dennig, Mark B Budolfson, Marc Fleurbaey, Asher Siebert, and  
417 Robert H Socolow. “Inequality, climate impacts on the future poor, and  
418 carbon prices”. In: *Proceedings of the National Academy of Sciences* 112.52  
419 (2015), pp. 15827–15832.
- 420 [29] Noah S. Diffenbaugh and Marshall Burke. “Global warming has increased  
421 global economic inequality”. In: *Proceedings of the National Academy of*  
422 *Sciences* 116.20 (2019), pp. 9808–9813.
- 423 [30] Angela Druckman and Tim Jackson. “Household energy consumption in the  
424 UK: A highly geographically and socio-economically disaggregated model”.  
425 In: *Energy Policy* 36.8 (2008), pp. 3177–3192.
- 426 [31] Rosa Duarte, Alfredo Mainar, and Julio Sánchez-Chóliz. “Social groups and  
427 CO2 emissions in Spanish households”. In: *Energy policy* 44 (2012), pp. 441–  
428 450.

- 429 [32] United Nations. Dept. of Economic and Social Affairs. Population Division.  
430 *World Population Prospects: The 2019 Revision*. United Nations Publications,  
431 2019.
- 432 [33] Simon Feindt, Ulrike Kornek, José M Labeaga, Thomas Sterner, and Hauke  
433 Ward. “Understanding regressivity: Challenges and opportunities of European  
434 carbon pricing”. In: *Energy Economics* 103 (2021), p. 105550.
- 435 [34] Anders Fremstad, Anthony Underwood, and Sammy Zahran. “The environ-  
436 mental impact of sharing: household and urban economies in CO2 emissions”.  
437 In: *Ecological economics* 145 (2018), pp. 137–147.
- 438 [35] Pierre Friedlingstein, Michael O’sullivan, Matthew W Jones, Robbie M An-  
439 drew, Judith Hauck, Are Olsen, Glen P Peters, Wouter Peters, Julia Pongratz,  
440 Stephen Sitch, et al. “Global carbon budget 2020”. In: *Earth System Science*  
441 *Data* 12.4 (2020), pp. 3269–3340.
- 442 [36] Bernhard Gill and Simon Moeller. “GHG emissions and the rural-urban  
443 divide. A carbon footprint analysis based on the German official income and  
444 expenditure survey”. In: *Ecological Economics* 145 (2018), pp. 160–169.
- 445 [37] Bastien Girod and Peter De Haan. “More or better? A model for changes in  
446 household greenhouse gas emissions due to higher income”. In: *Journal of*  
447 *industrial ecology* 14 (2010).
- 448 [38] Jane Golley and Xin Meng. “Income inequality and carbon dioxide emissions:  
449 The case of Chinese urban households”. In: *Energy Economics* 34.6 (2012),  
450 pp. 1864–1872.
- 451 [39] Ian Gough, Saamah Abdallah, Victoria Johnson, Josh Ryan-Collins, and Cindy  
452 Smith. “The distribution of total greenhouse gas emissions by households in  
453 the UK, and some implications for social policy”. In: *LSE STICERD Research*  
454 *Paper No. CASE152* (2011).
- 455 [40] J. Gütschow, L. Jeffery, R. Gieseke, R. Gebel, D. Stevens, M. Krapp, and  
456 M. Rocha. “The PRIMAP-hist national historical emissions time series”. In:  
457 *Earth Syst. Sci. Data* 8 (2016), pp. 571–603.

- 458 [41] Stephane Hallegatte and Julie Rozenberg. “Climate change through a poverty  
459 lens”. In: *Nature Climate Change* 7.4 (2017), pp. 250–256.
- 460 [42] Gilang Hardadi, Alexander Buchholz, and Stefan Pauliuk. “Implications of  
461 the distribution of German household environmental footprints across income  
462 groups for integrating environmental and social policy design”. In: *Journal of*  
463 *Industrial Ecology* 25.1 (2021), pp. 95–113.
- 464 [43] Klaus Hubacek, Giovanni Baiocchi, Kuishuang Feng, Raúl Muñoz Castillo,  
465 Laixiang Sun, and Jinjun Xue. “Global carbon inequality”. In: *Energy, Ecol-*  
466 *ogy and Environment* 2.6 (2017), pp. 361–369.
- 467 [44] Mohammad Iqbal Irfany and Stephan Klasen. “Inequality in emissions: evi-  
468 dence from Indonesian household”. In: *Environmental Economics and Policy*  
469 *Studies* 18.4 (2016), pp. 459–483.
- 470 [45] Elisabeth T Isaksen and Patrick A Narbel. “A carbon footprint proportional  
471 to expenditure-A case for Norway?” In: *Ecological Economics* 131 (2017),  
472 pp. 152–165.
- 473 [46] Christopher M Jones and Daniel M Kammen. “Quantifying carbon footprint  
474 reduction opportunities for US households and communities”. In: *Environ-*  
475 *mental science & technology* 45.9 (2011), pp. 4088–4095.
- 476 [47] Ivan Kartha, Eric Kemp-Benedict, Emily Ghosh, Aisha Nazareth, and Tim  
477 Gore. *The Carbon Inequality Era: An assessment of the global distribution*  
478 *of consumption emissions among individuals from 1990 to 2015 and beyond*.  
479 Oxfam and Stockholm Environmental Institute Joint Research Report, 2020.
- 480 [48] Annemarie C Kerkhof, René MJ Benders, and Henri C Moll. “Determinants  
481 of variation in household CO2 emissions between and within countries”. In:  
482 *Energy policy* 37.4 (2009), pp. 1509–1517.
- 483 [49] Fabrice Lengart, Christophe Lesieur, and Jean-Louis Pasquier. “Les émissions  
484 de CO2 du circuit économique en France”. In: *L'économie Française* (2010).

- 485 [50] Manfred Lenzen, Mette Wier, Claude Cohen, Hitoshi Hayami, Shonali Pachauri,  
486 and Roberto Schaeffer. “A comparative multivariate analysis of household  
487 energy requirements in Australia, Brazil, Denmark, India and Japan”. In:  
488 *Energy* 31.2-3 (2006), pp. 181–207.
- 489 [51] Wassily Leontief. “Environmental repercussions and the economic structure:  
490 an input-output approach”. In: *The review of economics and statistics* (1970),  
491 pp. 262–271.
- 492 [52] Wassily Leontief. *Input-output economics*. Oxford University Press, 1986.
- 493 [53] Brantley Liddle. “What are the carbon emissions elasticities for income and  
494 population? Bridging STIRPAT and EKC via robust heterogeneous panel  
495 estimates”. In: *Global Environmental Change* 31 (2015), pp. 62–73.
- 496 [54] Lenzen Manfred, Daniel Moran, Keiichiro Kanemoto, and Arne Geschke.  
497 “BUILDING EORA: A GLOBAL MULTI-REGION INPUT–OUTPUT DATABASE  
498 AT HIGH COUNTRY AND SECTOR RESOLUTION”. In: *Economic Sys-  
499 tems Research* 25.1 (2013), pp. 20–49.
- 500 [55] V. Masson-Delmotte, P. Zhai, A. Pirani, S.L. Connors, C. Péan, S. Berger, N.  
501 Caud, Y. Chen, L. Goldfarb, M.I. Gomis, M. Huang, K. Leitzell, E. Lonnoy,  
502 J.B.R. Matthews, T.K. Maycock, T. Waterfield, O. Yelekçi, R. Yu, and B. Zhou  
503 (eds.) *Climate Change 2021: The Physical Science Basis. Contribution of  
504 Working Group I to the Sixth Assessment Report of the Intergovernmental  
505 Panel on Climate Change*. Cambridge University Press, 2021.
- 506 [56] Jan C Minx, Thomas Wiedmann, Richard Wood, Glen P Peters, Manfred  
507 Lenzen, Anne Owen, Kate Scott, John Barrett, Klaus Hubacek, Giovanni  
508 Baiocchi, et al. “Input–output analysis and carbon footprinting: an overview  
509 of applications”. In: *Economic systems research* 21.3 (2009), pp. 187–216.
- 510 [57] Jonas Nässén. “Determinants of greenhouse gas emissions from Swedish  
511 private consumption: Time-series and cross-sectional analyses”. In: *Energy*  
512 66 (2014), pp. 98–106.

- 513 [58] Delphine Nougayrede. “Towards a Global Financial Register? The Case for  
514 End Investor Transparency in Central Securities Depositories”. In: *Journal of*  
515 *Financial Regulation* 4.2 (2018), pp. 276–313.
- 516 [59] *OECD GDP long term forecast*. 2021.
- 517 [60] Yannick Oswald, Anne Owen, and Julia K Steinberger. “Large inequality in  
518 international and intranational energy footprints between income groups and  
519 across consumption categories”. In: *Nature Energy* 5.3 (2020), pp. 231–239.
- 520 [61] Juudit Ottelin, Jukka Heinonen, and Seppo Junnila. “Carbon and material  
521 footprints of a welfare state: Why and how governments should enhance green  
522 investments”. In: *Environmental Science & Policy* 86 (2018), pp. 1–10.
- 523 [62] Juudit Ottelin, Jukka Heinonen, and Seppo Junnila. “Carbon footprint trends  
524 of metropolitan residents in Finland: how strong mitigation policies affect dif-  
525 ferent urban zones”. In: *Journal of cleaner production* 170 (2018), pp. 1523–  
526 1535.
- 527 [63] Ilona M Otto, Kyoung Mi Kim, Nika Dubrovsky, and Wolfgang Lucht. “Shift  
528 the focus from the super-poor to the super-rich”. In: *Nature Climate Change*  
529 9.2 (2019), pp. 82–84.
- 530 [64] Shonali Pachauri. “An analysis of cross-sectional variations in total household  
531 energy requirements in India using micro survey data”. In: *Energy policy*  
532 32.15 (2004), pp. 1723–1735.
- 533 [65] Jyoti Parikh, Manoj Panda, Anand Ganesh-Kumar, and Vinay Singh. “CO2  
534 emissions structure of Indian economy”. In: *Energy* 34.8 (2009), pp. 1024–  
535 1031.
- 536 [66] Jyoti K Parikh, Manoj K Panda, and NS Murthy. “Consumption patterns by  
537 income groups and carbon–dioxide implications for India: 1990–2010”. In:  
538 *International Journal of Global Energy Issues* 9.4-6 (1997), pp. 237–255.
- 539 [67] Glen P Peters. “From production-based to consumption-based national emis-  
540 sion inventories”. In: *Ecological economics* 65.1 (2008), pp. 13–23.

- 541 [68] GP Peters, J Aasness, N Holck-Steen, and EG Hertwich. “Environmental  
542 impacts and household characteristics: an econometric analysis of Norway  
543 1999–2001”. In: *Proceedings, Sustainable Consumption Research Exchange,*  
544 *Wuppertal* (2006).
- 545 [69] Thomas Piketty and Emmanuel Saez. “Inequality in the long run”. In: *Science*  
546 344.6186 (2014), pp. 838–843.
- 547 [70] Thomas Piketty, Emmanuel Saez, and Gabriel Zucman. “Distributional na-  
548 tional accounts: methods and estimates for the United States”. In: *The Quar-*  
549 *terly Journal of Economics* 133.2 (2018), pp. 553–609.
- 550 [71] Antonin Pottier. “Expenditure elasticity and income elasticity of GHG emis-  
551 sions: A survey of literature on household carbon footprint”. In: *Ecological*  
552 *Economics* 192 (2022), p. 107251.
- 553 [72] Yannick Rehm. “Estimating the carbon content of wealth”. In: *Paris School*  
554 *of Economics Working Paper* (2021).
- 555 [73] Jordi Roca and Mònica Serrano. “Income growth and atmospheric pollution  
556 in Spain: an input–output approach”. In: *Ecological Economics* 63.1 (2007),  
557 pp. 230–242.
- 558 [74] Lutz Sager. “Income inequality and carbon consumption: Evidence from  
559 Environmental Engel curves”. In: *Energy Economics* 84 (2019), p. 104507.
- 560 [75] Moises Neil V Serioño and Stephan Klasen. “Estimation and Determinants of  
561 the P hilippines’ Household Carbon Footprint”. In: *The Developing Economies*  
562 53.1 (2015), pp. 44–62.
- 563 [76] Kaihui Song, Shen Qu, Morteza Taiebat, Sai Liang, and Ming Xu. “Scale,  
564 distribution and variations of global greenhouse gas emissions driven by US  
565 households”. In: *Environment international* 133 (2019), p. 105137.
- 566 [77] Kjartan Steen-Olsen, Richard Wood, and Edgar G Hertwich. “The carbon  
567 footprint of Norwegian household consumption 1999–2012”. In: *Journal of*  
568 *Industrial Ecology* 20.3 (2016), pp. 582–592.
- 569 [78] Thomas Sterner. *Fuel taxes and the poor: the distributional effects of gasoline*  
570 *taxation and their implications for climate policy*. Routledge, 2012.

- 571 [79] Kevin Ummel. “Who pollutes? A household-level database of America’s  
572 greenhouse gas footprint”. In: *Center for Global Development Working Paper*  
573 381 (2014).
- 574 [80] UN, WB, EC, and IMF. *System of National Accounts 2008*. 2008.
- 575 [81] European Union. *EU taxonomy for sustainable activities*. Regulation (EU)  
576 2020/85, 2020.
- 577 [82] United Nations Development Programme, Human Development Report Office.  
578 *Human Development Report 2019. Beyond income, beyond averages, beyond*  
579 *today: Inequalities in human development in the 21st century*. 2019.
- 580 [83] Christoph Weber and Adriaan Perrels. “Modelling lifestyle effects on energy  
581 demand and related emissions”. In: *Energy policy* 28.8 (2000), pp. 549–566.
- 582 [84] Christopher L Weber and H Scott Matthews. “Quantifying the global and  
583 distributional aspects of American household carbon footprint”. In: *Ecological*  
584 *economics* 66.2-3 (2008), pp. 379–391.
- 585 [85] Christopher L Weber and H Scott Matthews. “Quantifying the global and  
586 distributional aspects of American household carbon footprint”. In: *Ecological*  
587 *economics* 66.2-3 (2008), pp. 379–391.
- 588 [86] D Wiedenhofer. “Unequal household carbon footprints in China”. In: *Nature*  
589 *Climate Change* 7.1 (2017), pp. 75–80.
- 590 [87] Mette Wier, Manfred Lenzen, Jesper Munksgaard, and Sinne Smed. “Effects  
591 of Household Consumption Patterns on CO2 Requirements”. In: *Economic*  
592 *Systems Research* 13.3 (2001), pp. 259–274.

593

## METHODS

594 **Relationship to earlier studies.** There are two broad approaches to measure  
595 global carbon inequality, namely *bottom-up* approaches and *top-down* approaches.  
596 Both have strengths and limitations. The bottom-up approach mobilizes household-  
597 level micro-data to produce macro estimates. This is the approach followed by  
598 Bruckner et al. [12] for instance, who mobilize the large set of consumption surveys

599 available from the World Bank Global Consumption Database (WBGCD), as well  
600 as additional consumer expenditure surveys for rich countries. These surveys are  
601 linked to EMRIOs to provide estimates of emissions per consumption group. To  
602 the extent that micro-level data is available, this method is the best way to measure  
603 global carbon inequality associated to individual *consumption*. However, it requires  
604 a vast set of standardized micro data on consumption at the household level for all  
605 countries in the world, making longitudinal studies extremely difficult. This explains  
606 why Bruckner et al. [12] solely focus on year 2014.<sup>15</sup>

607 In addition, it is now well-known that household consumption surveys tend  
608 to underestimate consumption levels of richest groups, due to misreporting and  
609 sampling errors (see [11]). Using household surveys without additional datasets  
610 on inequality therefore tends to underestimate carbon emissions associated to rich  
611 individuals' consumption by construction. Another limitation of this approach is  
612 that it does not treat wealth (or investment-related emissions) particularly well. For  
613 instance, Bruckner et al. do not treat investment related emissions differently from  
614 emissions associated to household consumption. More precisely, emissions linked  
615 to investments in the car industry (construction of new factories or new machines)  
616 are distributed to individuals in proportion their consumption of cars, rather than to  
617 individuals' investments in the car industry. It is however well known that inequalities  
618 in investments are much larger than inequalities in consumption [22].

619 *Top-down* approaches to the measurement of global carbon inequality use the  
620 regularities observed in micro-level data to provide modeled estimates, typically  
621 over longer time spans. This is the approach for instance of Chakravarty et al. [17],  
622 who look at territorial emissions only and therefore miss the potentially large share  
623 of emissions embedded in international trade (see [67]). Chancel and Piketty (see  
624 [21]) follow a similar method but use the GTAP Environmental Multi Regional Input  
625 Output (EMRIO) database [1] to take into account emissions from consumption and  
626 to look at a longer period. This approach was also used by [47].

627 The basic framework developed in this study builds on the top-down approach,  
628 to be able to study global carbon inequality over relatively long time-spans. The  
629 methodology can however incorporate findings from country-level micro studies on

<sup>15</sup>also Hubacek et al. [43] who use a similar approach.

630 the link between income or consumption and emissions. In that sense, I mobilize  
631 strengths of both approaches. Departing from earlier studies, I also provide a more  
632 accurate treatment of investment-related emissions, based on novel data about income  
633 and wealth inequality within countries and on a specific treatment of emissions of  
634 the private investment sector of the economy (see below).

635 The general approach followed in this study can be summarized as follows.  
636 Based on Environmental Multi Regional Input Output Tables, I obtain country-level  
637 GHG emissions for the household sector, the investment sector and the government  
638 sector of the economy (emissions are net of imports and exports embedded in goods  
639 and services with the rest of the world). These emissions are distributed in the  
640 following way to individuals: (i) aggregate carbon footprints of the household sector  
641 in a given country are distributed following a power law of individual income, which  
642 can change from country to country or over time; (ii) aggregate emissions associated  
643 to investments and capital stock replacement are a function of the distribution of asset  
644 ownership within countries; (iii) emissions from the Government sector (emissions  
645 from the public health sector, education, infrastructure defense, etc.) are distributed  
646 as lump-sum to individuals, or as a function of individuals' income (depending on  
647 the specification chosen). I detail the main specification chosen below and present  
648 results from various parametric assumptions in the **Supplementary Information** of  
649 this paper.

650 **Income and wealth inequality data.** The methodology followed in this paper  
651 requires precise data on the distribution of income and wealth within countries. The  
652 past two decades were marked by important breakthroughs in researchers' ability  
653 to monitor global income and wealth inequality [69, 22], which I build upon. The  
654 standard source of information mobilized to track inequality within countries is  
655 via household surveys. While surveys constitute a rich source of information to  
656 track the various facets of socio-economic inequality, they do not provide statistics  
657 comparable across countries, typically fail to properly measure incomes and wealth  
658 at the top of the distribution and are typically not consistent with macroeconomic  
659 totals [7, 5].

660 The Distributional National Accounts (DINA) methodology [70, 3], developed  
661 by a large network of researchers affiliated with the World Inequality Database

662 (wid.world), in partnership with national and international statistical organizations  
663 and the United Nations, seeks to address these issues by systematically combining  
664 household surveys with additional sources of information on economic inequality.  
665 These additional sources of information include, in particular, administrative tax  
666 data and National Accounts. On the one hand, tax data offer a more reliable account  
667 of income and wealth dynamics among wealthy groups than those reported by  
668 individuals in household surveys. Tax data also enable long term comparisons,  
669 spanning over decades (and centuries, in some countries). On the other hand, the  
670 use of National Accounts concepts makes it possible to compare income or wealth  
671 levels more systematically across countries.

672 DINA made it possible to improve our collective understanding of the ultimate  
673 beneficiaries of economic growth within countries and at the global level. This  
674 body of work revealed that most societies went through a decline in inequality  
675 between the 1920-1970s and then observed a return of inequality since the 1980s [6].  
676 Such findings generated significant academic and public debates on the causes and  
677 consequences of inequality within nations. While such dynamics can have important  
678 impacts on the inequality of carbon emissions, the interactions between income,  
679 consumption, wealth and GHG emissions have attracted only a limited amount of  
680 attention to date [43, 60]. In fact, there have been no attempts to measure to dynamics  
681 of the global distribution of carbon emissions taking stock of recent progress in  
682 global inequality research in the context of Distributional National Accounts.<sup>16</sup> The  
683 purpose of this paper is to study dynamics of global carbon emissions over several  
684 decades, with a particular focus on emissions at the top of the distribution.

685 The economic inequality datasets used in this study are those we have developed  
686 in the context of the World Inequality Database (wid.world) [3]. They provide  
687 income and wealth inequality series for 174 countries over the 1990-2019 period, i.e.  
688 more than 97% of the world population and 97% of global Gross Domestic Product  
689 or global income. (See **SI** Section 2). WID.world contains reproducible inequality  
690 statistics based on the systematic combination of household surveys, tax data and  
691 national accounts, produced by an international network of researchers contributing

<sup>16</sup>Recent work by Bruckner et al. [12] focus on a single year and authors stress that their estimates do not cover emissions of top income groups particularly well (see above).

692 to the dataset. The general set of guidelines and methods underlying these data  
693 series is described in the Distributional National Accounts Guidelines [3]. Income  
694 inequality levels for all countries are presented in **SI** Table 9.10.

695 **The link between carbon emissions inequality and economic inequality.** Most  
696 countries do not publish standardized data sources on individual emissions levels.  
697 Such information can be reconstructed from household surveys and with additional  
698 data on energy. Data on individual emissions inequality have been produced for  
699 several countries and years by researchers mobilizing Input-Output tables (see below)  
700 [50, 84, 64, 30].

701 Available literature typically finds that carbon emissions associated to individual  
702 consumption depend on several factors including income and expenditure, as well  
703 as households' location, energy conversion technologies, occupation status, habits,  
704 age, national regulations and energy mixes [50, 87, 73, 85, 68, 13, 57, 71] (see  
705 also **SI** Tables 3.1 and 3.2 for a complete list of studies on the matter). While  
706 non-income factors play a significant role in determining direct individual emissions  
707 levels (i.e. emissions stemming from the direct use of energy, such as emissions  
708 associated to car driving), income is found to be the main driver of indirect emissions  
709 (emissions associated to energy mobilized to produce goods and services consumed  
710 by individuals), and of overall emissions inequalities between individuals. At a given  
711 income level, two individuals may indeed have different heating or transportation  
712 needs, implying different direct energy requirement and different direct emissions  
713 levels. However, when taking into account the carbon content of their overall  
714 consumption and of their indirect energy requirements (the energy used to produce  
715 the clothes or appliances they buy, the food they eat, the services they purchase, etc.),  
716 income differences explain most of the differences observed in carbon footprints.

717 Studies measuring the *elasticity* of individual carbon emissions (or the strength of  
718 the relationship between rising individual income and CO<sub>2</sub> emissions, see Methods)<sup>17</sup>  
719 are presented in **SI** Table 3.1 and A3.2. These studies find that the elasticity of  
720 household consumption to emissions typically falls in the 0.9-1.1 range, while the  
721 elasticity of household income to emissions typically falls in the 0.5-0.7 range [50,

<sup>17</sup>In a model of the form  $\log(CO_2) = \alpha \cdot \log(\text{income})$ , where  $\alpha$  is the elasticity

722 87, 73, 85, 2, 38, 68, 13, 71].<sup>18</sup> Using these observed regularities, and taking stock  
723 of recent progress in income inequality measurement, it is possible to estimate  
724 emissions inequalities between world individuals in a relatively straightforward and  
725 transparent manner, over long time spans.

726 **Environmental Input-Output data.** The most straightforward way to obtain  
727 internationally comparable direct and indirect emission levels of individuals is via  
728 the Input-Output (IO) framework. The IO framework is quantitative model of  
729 the economy, initially developed to represent inter-dependencies between different  
730 economic sectors (households, governments, firms) within and between a countries  
731 [52]. The framework was extended to economy-environment interactions [51] to  
732 better understand the material content of production and the impact of environmental  
733 policies and relatively recently to study international flows of carbon embodied in  
734 international trade [26, 67].

735 In the context of carbon accounting, the strength of the IO framework is to rely on  
736 a systematic representation of the world economy which avoids any double-counting:  
737 the same tonne of carbon cannot be ultimately attributed to two different agents<sup>19</sup>.  
738 The environmental IO approach is also particularly useful because it can distinguish  
739 between emissions from household consumption, investments and to government  
740 expenditures – in line with National Accounts concepts [80, 19].<sup>20</sup>

741 Let  $Z$  be the inter-industry transactions matrix (i.e. the flow of intermediary  
742 goods and services between industries, to produce final products),  $Y$  the final demand  
743 matrix (the final demand associated to the household, investment or government  
744 sector of the economy),  $Q$  the carbon emissions matrix and  $x$  as the vector of gross  
745 output by country-sector (See **SI** Section 1). Leontief's inverse (or the impact of

<sup>18</sup>See also [53] for elasticity estimations based on on macroeconomic data, rather than micro level household data.

<sup>19</sup>In other carbon accounting methodologies, such as the life-cycle analyses, the issue of double counting is omnipresent

<sup>20</sup>Changes in inventories and stocks are also reported in the dataset. Since they only represent a marginal fraction of emissions, I include them in GFCF totals so as to keep fully consistent datasets which always match with aggregate totals. I also include emissions of Non-Profit Institution Serving Households in the Household Sector as a first approximation.

746 final demand on the output of a given sector) is given by:

$$L = (I - A)^{-1} \quad (1)$$

747 With:

$$A = Zx^{-1} \quad (2)$$

748 The carbon intensity of production is then given by:

$$C = (Qx^{-1})L \quad (3)$$

749 Carbon emissions associated to final demand is obtained as follows:

$$N = CY \quad (4)$$

750 Our benchmark MRIO data source is the Global Carbon Project (GCP) [35]. In  
 751 certain cases, GCP does not provide data for a given country or for a given type of  
 752 emissions. In order to cover all countries and all types of emissions, I also rely on  
 753 the EORA dataset [54].<sup>21</sup>

754 **Distributing emissions to individuals.** In line with the National Accounts  
 755 Methodology, I decompose national-level distributions (of income, wealth or carbon  
 756 emitters) in 127 generalized-percentiles: 99 percentiles from  $p = 0\%$  to  $p = 99\%$ ,  
 757 9 tenths of a percentile from  $p = 99\%$  to  $p = 99.9\%$ , 9 hundredths of a percentile  
 758 from  $p = 99.9\%$  to  $p = 99.99\%$ , 10 thousandths of a percentile from  $p = 99.99\%$   
 759 to  $p = 100\%$ . In order to determine carbon emission levels associated to each of  
 760 these generalized-percentiles of income, in each country of the world, I proceed as  
 761 follows. Average per capita emissions at percentile  $p$ , in a given year and country  
 762 are defined as:

$$E_p^{tot} = E_p^{cons} + E_p^{inv} + E_p^{gov} \quad (5)$$

763 Where  $E_p^{cons}$ ,  $E_p^{inv}$ ,  $E_p^{gov}$  are individual average footprints at percentile  $p$ , asso-  
 764 ciated to consumption, private investment and public spending, respectively. More  
 765 precisely:

<sup>21</sup>For details on the construction of aggregate series used in this study, see SI Section 1 and [16].

$$E_p^{cons} = f(E^{cons}, Y_p, \alpha) \quad (6)$$

$$E_p^{inv} = f(E^{inv}, W_p, \gamma) \quad (7)$$

$$E_p^{gov} = f(E^{gov}, y_p, \delta) \quad (8)$$

766 Where  $E^{cons}$  is the average carbon footprint associated to consumption in the  
 767 country,  $Y_p$  the average income level of individuals in percentile  $p$ ,  $\alpha$  the elasticity  
 768 of household consumption carbon emissions to income (in a model of the form  
 769  $E_p^{cons} = E^{cons} \times Y_p^\alpha$ );  $E^{inv}$  is the average emissions level associated to investments  
 770 (or asset ownership, in our framework),  $\gamma$  the elasticity of wealth to investment  
 771 emissions;  $E^{gov}$  is the average emission level of the government sector (associated to  
 772 in-kind redistribution) and  $\delta$ , is the elasticity of government emissions to individual  
 773 income.

774 The results presented above are based on  $\alpha = 0.6$ ,  $\gamma = 1$ ,  $\delta = 0$ . The benchmark  
 775  $\alpha$  value is based on the large regularity observed in available studies focusing  
 776 on *income* and carbon emissions. This is also the value that corresponds to a  
 777 *consumption*-carbon elasticity near 1, which is also what Bruckner et al.[12] find  
 778 for most countries in 2014. I also produce results for *alpha* values varying country  
 779 by country and corresponding to available *alpha* from micro studies. Given that  
 780 changing *alpha* does not significantly impact results, and given that there are no  
 781 available *alpha* for countries over the time period considered, I opt for a constant  
 782 elasticity to ensure a greater consistency.

783 In the benchmark scenario, investment-related emissions are attributed in pro-  
 784 portion to individuals' wealth (that is, the share of investment related emissions of  
 785 the top 1% in a given country is equal to its share of wealth in that country). This  
 786 implies that  $\gamma=1$  in the benchmark scenario. This choice is probably conservative as  
 787 Rehm [72] finds that emissions incorporated in wealth ownership could rise more  
 788 rapidly than wealth (i.e. the carbon intensity of high net wealth is higher than at  
 789 low or moderate levels of wealth).  $\delta = 0$  amounts to distributing collective con-  
 790 sumption expenditure of governments equally to individuals, as a lump-sum. This

791 has been a relatively standard choice in earlier studies. In alternative scenarios,  
792 I distribute emissions in proportion to individuals' consumption. This alternative  
793 choice mechanically increases top emitters' contributions.

794 I produce results for the following range of parameters:  $\alpha = (0.4; 0.5; 0.6; 0.7; 0.8; 0.9; 1)$ ;  
795  $\gamma = (1; 1.2; 1.4)$ ;  $\delta = (0; 1)$ . In all countries, I assume that emissions are split  
796 equally within households.

797 **Robustness checks.** The Supplementary Information document provides results  
798 for different parametric assumptions at the global, regional and country level. Overall,  
799 my main results appear to be robust to a wide set of parameters. In the extreme  
800 lower-bound scenario (i.e. a scenario which leads to a very low level of emissions  
801 inequalities within countries, with  $\alpha=0.4$ ,  $\gamma=1$ ), I find that the global top 10%  
802 emissions' share nears 45% in 2019. In my extreme upper-bound scenario, the  
803 global top 10% emissions' share is of 56%. Using parameters closer to what is  
804 observed within countries ( $\alpha$  around 0.6-0.8) yields values in the 46%-52.5% range,  
805 that is within a 5-10% range of our benchmark estimate for the global top 10% share.

806 I also observe that the global dynamics between 1990-2019 are robust across  
807 these different scenarios, and are not particularly sensitive to changes in choices  
808 of parameters over time. In **SI** Figure 7.5, I reproduce Figure **IV** across dozens of  
809 scenarios and find that the pattern and levels are consistent with benchmark results  
810 presented above (see also **SI** Fig. 7.6 and **SI** Table 7.8). Investment share at the  
811 top of the distribution also appear to be very large, irrespective of the assumptions  
812 made parameters. The top 0.1% has between 65% and 90% of its emissions from  
813 investments, even when using very different  $\alpha$  and  $\gamma$  values.

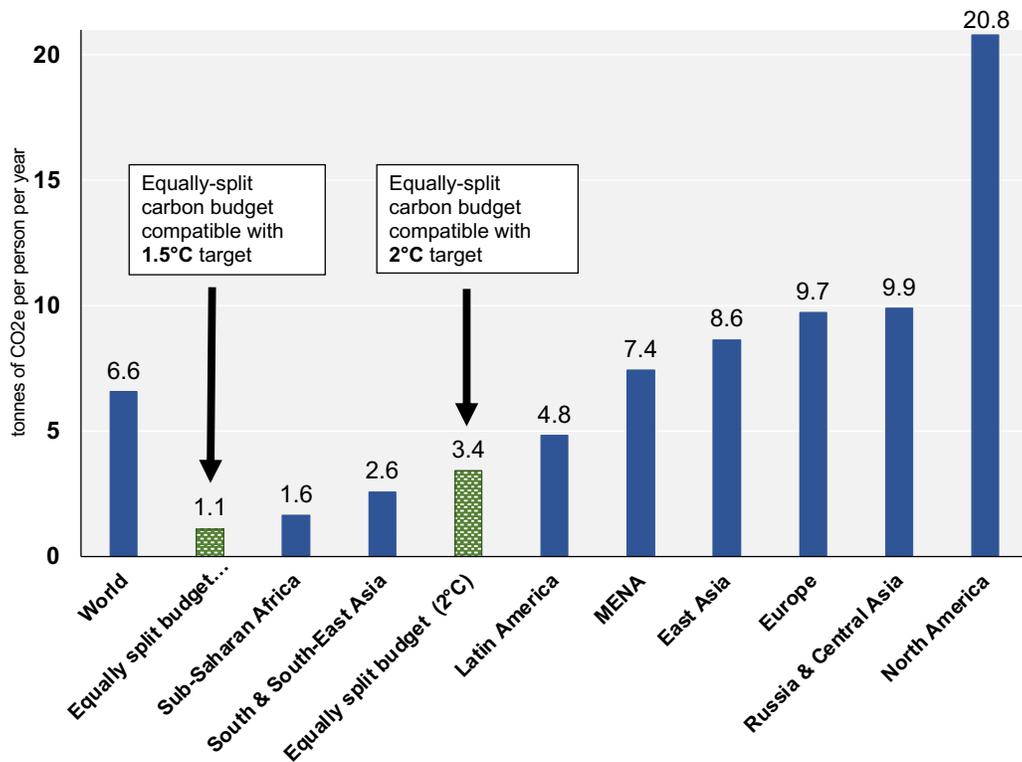
814 The flexible framework developed in this paper makes it possible to use a  
815 variety of country-level studies to study the dynamics of global carbon inequality  
816 over relatively long time-spans. The bottom line is that using available country-  
817 level data does not modify the general conclusions presented above. All the data  
818 mobilized for this study, as well as the computer codes are to be posted online on  
819 the World Inequality Database, making it possible for researcher to make alternative  
820 assumptions in the future.

821 **National 2030 targets.** Per capita national targets are based on countries'  
822 Nationally Determined Commitments as of July 2021. Values are obtained from [25].

823 Rich countries typically express their targets in terms of percentage reduction by a  
824 certain date, as compared to a benchmark date. For instance, the US objective is to  
825 cut its total emissions by 50-52% as compared to 2005 level. In order to obtain per  
826 capita targets in 2030, I divide the implied 2030 emissions total by the US population  
827 in 2030, using UN Population Prospects [32]. In emerging countries, targets are  
828 typically expressed as changes in the carbon intensity of GDP. In that case, I use  
829 GDP forecasts produced by the OECD [59] and calculate the implied per capita  
830 emission target, based on carbon intensity targets announced by the country (see  
831 also **SI** Section 8).

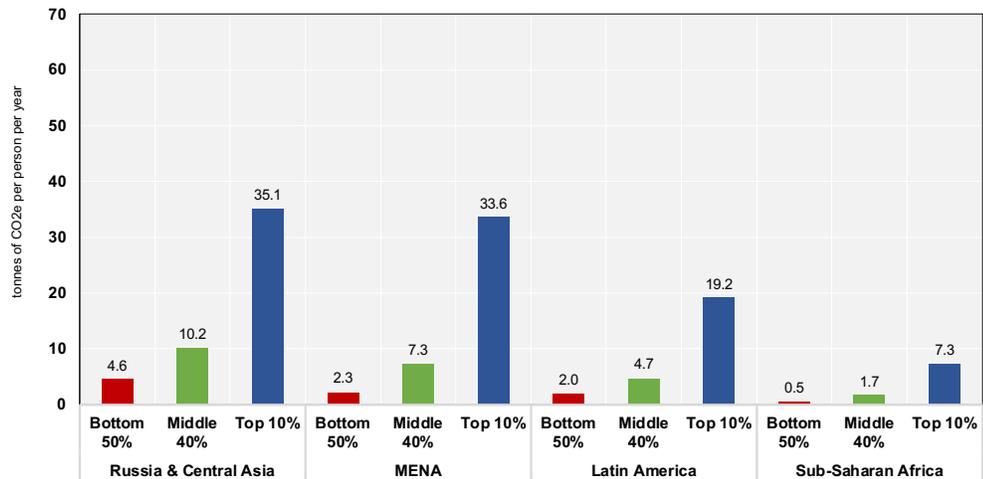
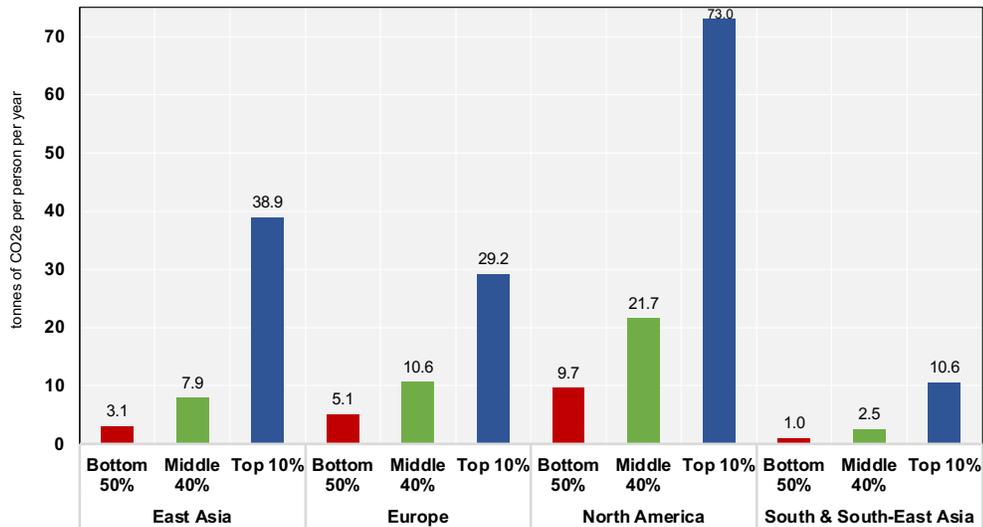
#### 832 ACKNOWLEDGEMENTS

833 The author thanks Felix Bajard, François Burq Aymeric Capitaine for research  
834 assistance as well as Thomas Piketty, Tancrède Voituriez, Thomas Blanchet, Rowaida  
835 Moshrif, the UNPD HDRO team and participants at the Paris School of Economics,  
836 the London School of Economics and Sciences Po seminars, for valuable comments.



*Interpretation:* Sharing the remaining carbon budget to have 83% chances to stay below 1.5°C global (see [55]) temperature increase implies an annual per capita emissions level of 1.1 tonnes per person per year between 2021 and 2050 (and zero afterwards). Emission levels present regional per capita emissions and include all emissions from domestic consumption, public and private investments as well as imports and exports of carbon embedded in goods and services traded with the rest of the world (LULUCF emissions are excluded). *Source and series:* Author, see Methods and Supplementary Information.

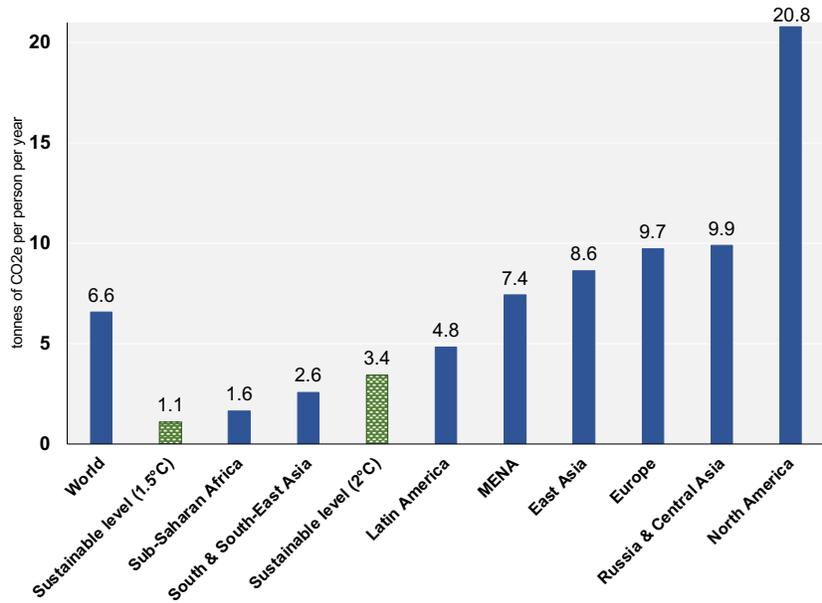
Figure I  
Average GHG emissions by world region, 2019



**Interpretation:** Personal carbon footprints include emissions from domestic consumption, public and private investments as well as imports and exports of carbon embedded in goods and services traded with the rest of the world. Modeled estimates based the systematic combination of tax data, household surveys and input-output tables. Benchmark scenario. Emissions split equally within households. *Source and series:* Author, see Supplementary Information.

Figure II  
Carbon footprints by income group across the world, 2019

(a) Per capita emissions by group (tCO<sub>2</sub> / year)



(b) Group share (%) in world total emissions

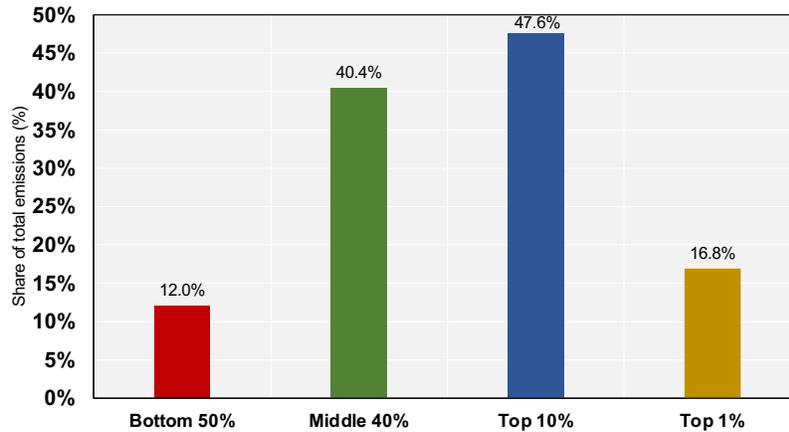
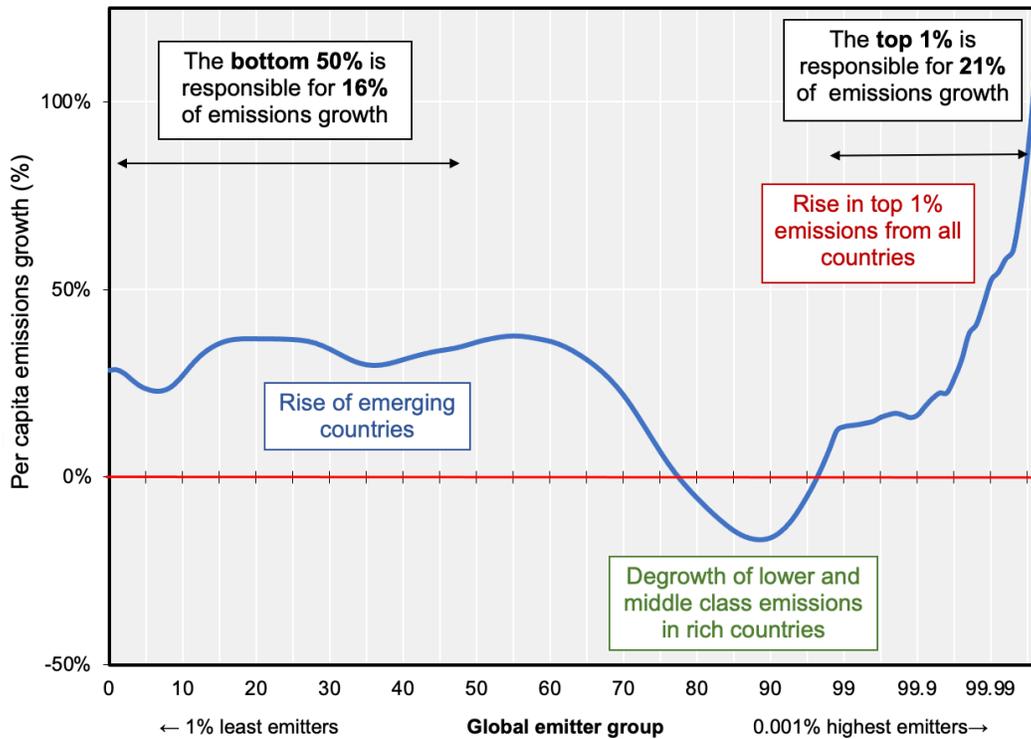


Figure III

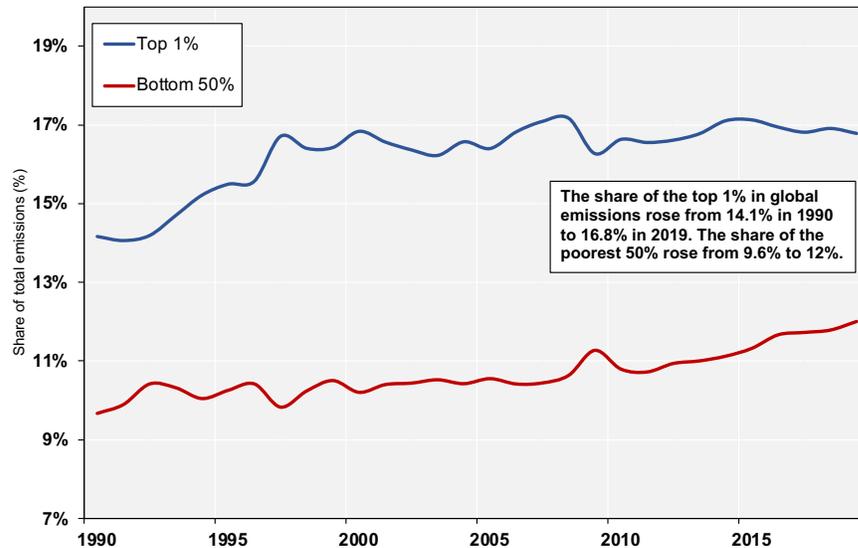
Global inequality in individual carbon emissions, 2019

**Interpretation:** Personal carbon footprints include emissions from domestic consumption, public and private investments as well as imports and exports of carbon embedded in goods and services traded with the rest of the world. Modeled estimates based the systematic combination of tax data, household surveys and input-output tables. Benchmark scenario. Emissions split equally within households. **Source and series:** Author, see Supplementary Information.



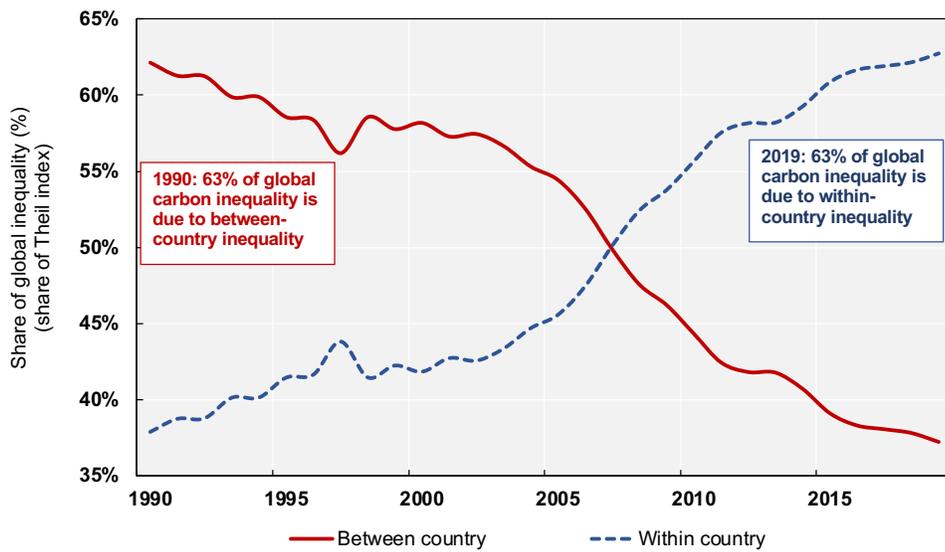
*Interpretation:* Emissions of the global bottom 50% rose by around 20-40% between 1990 and 2019. Emissions notably declined among groups above the bottom 80% and below the top 5% of the global distribution, these groups mainly correspond to lower and middle income groups in rich countries. Emissions of the global top 1% and richer groups rose substantially. Personal carbon footprints include emissions from domestic consumption, public and private investments as well as imports and exports of carbon embedded in goods and services traded with the rest of the world. Modeled estimates based the systematic combination of tax data, household surveys and input-output tables. Benchmark scenario. Emissions split equally within households. *Source and series:* Author, see Supplementary Information.

Figure IV  
Global inequality and carbon emissions, 1990-2019



*Interpretation:* This figure presents the share of global GHG emissions by the top 1% and bottom 50% of the global population between 1990 and 2019. GHG emissions measured correspond to individual footprints, i.e. they include indirect emissions produced abroad and embedded in individual consumption. Modeled estimates based on the systematic combination of tax data, household surveys and input-output tables. Benchmark scenario. Emissions split equally within households. *Source and series:* Author, see Supplementary Information.

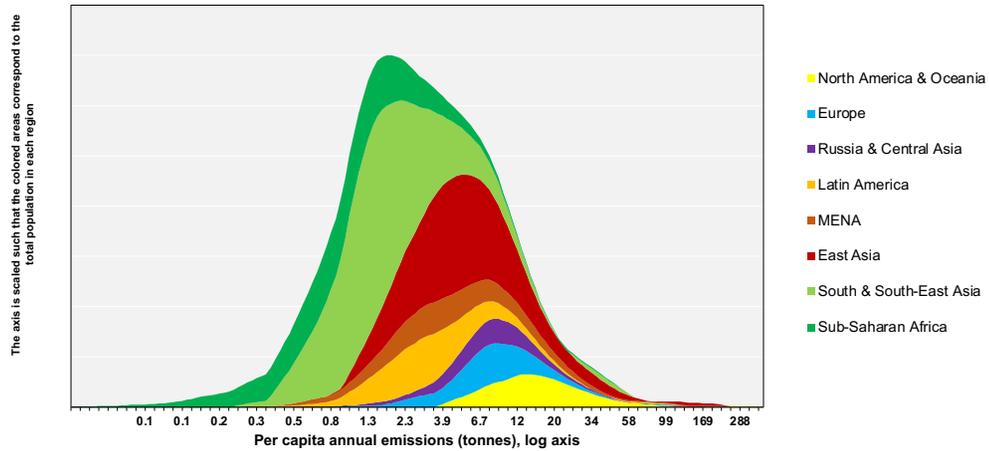
Figure V  
Top 1% and bottom 50% shares in global carbon emissions, 1990-2019



*Interpretation:* 37% of global carbon inequality between individuals is due differences in emissions levels between countries while 63% is explained by inequality within countries in 2019. Modeled estimates based the systematic combination of tax data, household surveys and input-output tables. Benchmark scenario. *Source and series:* Author, see Supplementary Information.

Figure VI  
Theil index decomposition of global carbon inequality

(a) Global carbon emissions density function



(b) Share of each region in the emissions of global emitter groups

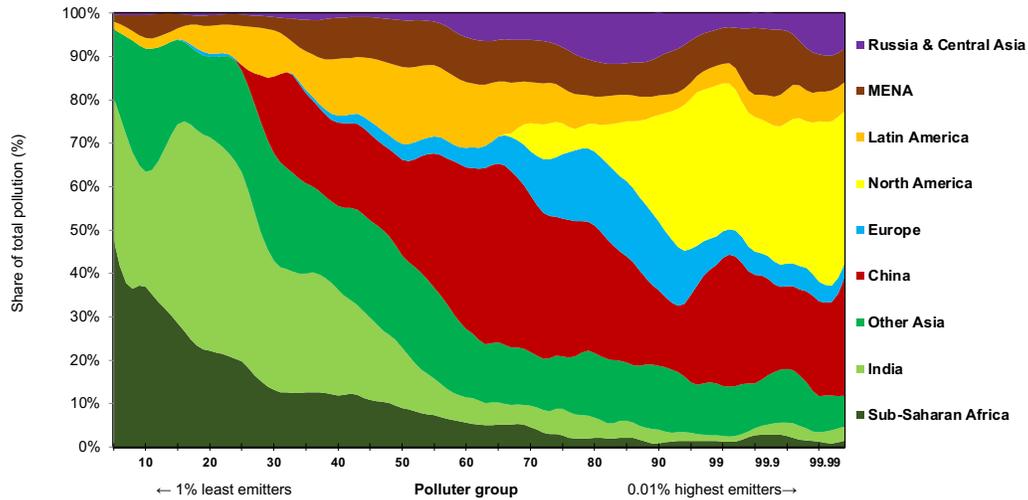
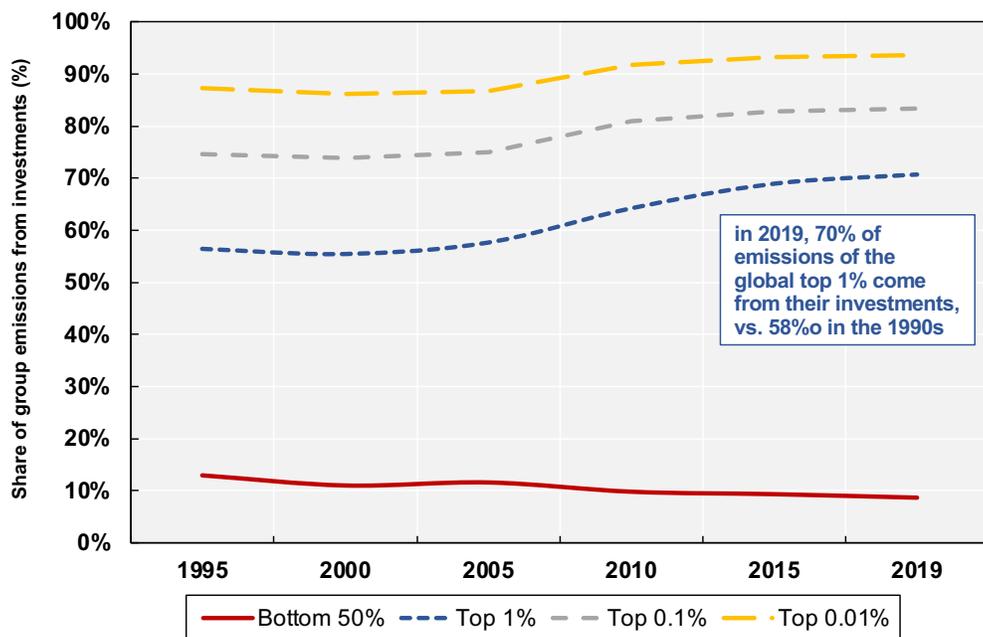


Figure VII  
Global inequality in individual carbon emissions, 2019

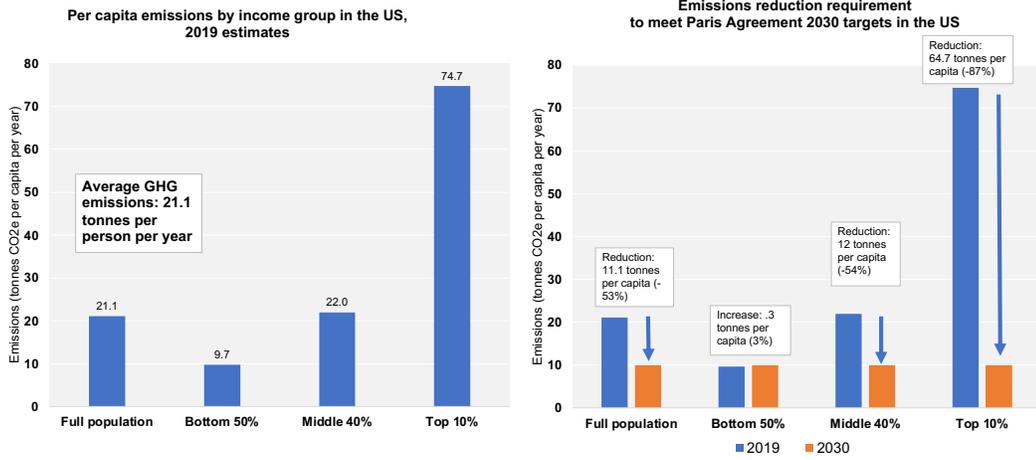
*Interpretation:* **Panel A** The graph shows the share of world regions in each group of global emitters, from the lowest 1% to the highest 0.1%. **Panel B** shows the global distribution (density) of individual emitters in 2019. GHG emissions measured correspond to individual footprints, i.e. they include indirect emissions produced abroad and embedded in individual consumption. Modeled estimates based on the systematic combination of tax data, household surveys and input-output tables. Benchmark scenario. Emissions split equally within households. *Sources and series:* Author, see Supplementary Information.



*Interpretation:* This figure presents the share of GHG emissions by different groups of emitters that can be traced to their investments, rather than to their consumption. Modeled estimates based the systematic combination of tax data, household surveys and input-output tables. Benchmark scenario. *Source and series:* Author, see also Supplementary Information.

**Figure VIII**  
The share of investments in emissions of various global emitter groups, 1995-2019

(a) Emissions inequality and climate targets in the US



(b) Emissions inequality and climate targets in China

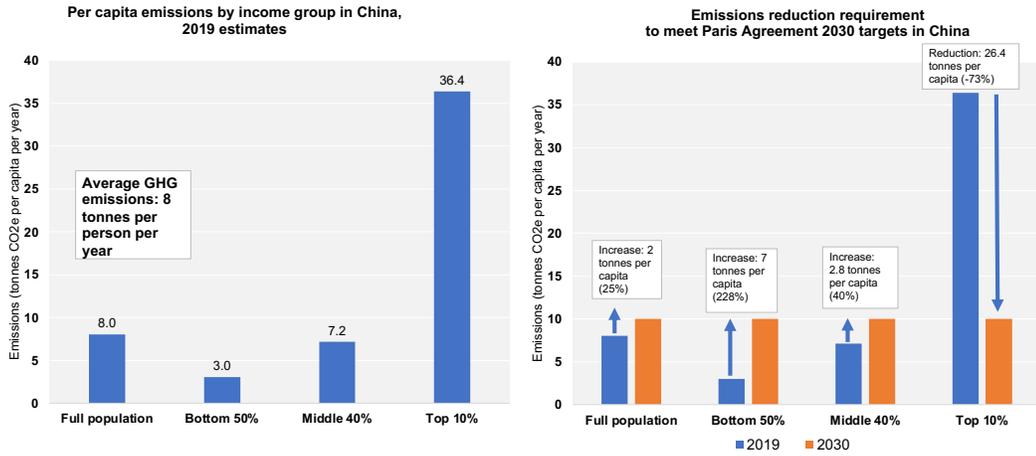


Figure IX

Emissions inequality and climate targets

*Interpretation:* The graph shows national emissions targets (NDCs) expressed in per capita terms, and compares these with current emission levels of different income groups in the US and in China. For China, targets are expressed in efficiency terms so we use GDP projections to obtain overall emissions levels. *Sources and series:* Author, see Supplementary Information.

	Number of individuals (million)	Average (tonne CO2 per capita)	Threshold (tonne CO2 per capita)	Share (% total)
<b>Full population</b>	<b>7710</b>	<b>6.6</b>	<b>&lt;0.1</b>	<b>100%</b>
<b>Bottom 50%</b>	<b>3855</b>	<b>1.6</b>	<b>&lt;0.1</b>	<b>12.0%</b>
<i>incl. Bottom 20%</i>	<i>1542</i>	<i>0.8</i>	<i>&lt;0.1</i>	<i>2.5%</i>
<i>incl. Bottom 30%</i>	<i>2313</i>	<i>2.1</i>	<i>1.8</i>	<i>9.5%</i>
<b>Middle 40%</b>	<b>3084</b>	<b>6.6</b>	<b>3.1</b>	<b>40.4%</b>
<b>Top 10%</b>	<b>771</b>	<b>31</b>	<b>13</b>	<b>47.6%</b>
<i>incl. Top 1%</i>	<i>77.1</i>	<i>110</i>	<i>46</i>	<i>16.8%</i>
<i>incl. Top 0.1%</i>	<i>7.71</i>	<i>467</i>	<i>130</i>	<i>7.1%</i>
<i>incl. Top 0.01%</i>	<i>0.771</i>	<i>2531</i>	<i>569</i>	<i>3.9%</i>

Table I  
Global inequality of individual carbon emissions, 2019

*Interpretation:* Individual carbon footprints include emissions from domestic consumption, public and private investments as well as imports and exports of carbon embedded in goods and services traded with the rest of the world. Modeled estimates based the systematic combination of tax data, household surveys and input-output tables. Benchmark scenario. Emissions split equally within households. *Source and series:* Author, see Supplementary Information.

	Per capita emissions (tonnes CO2e per capita)		Total emissions (billion tonnes CO2e)		Growth in per capita emissions (1990-2019)	Growth in total emissions (1990-2019)	Share in emissions growth (1990-2019)
	1990	2019	1990	2019			
<b>Full population</b>	6.2	6.6	32.0	50.5	7%	58%	100%
<b>Bottom 50%</b>	1.2	1.6	3.1	6.1	32%	96%	16%
<b>Middle 40%</b>	6	6.6	13.3	20.4	4%	54%	39%
<b>Top 10%</b>	30	31	15.7	24.0	4%	54%	45%
<i>Top 1%</i>	87	110	4.5	8.5	26%	87%	21%
<i>Top 0.1%</i>	323	467	1.7	3.6	45%	114%	10%
<i>Top 0.01%</i>	1397	2531	0.7	2.0	81%	168%	7%

Table II  
Emissions growth and inequality, 1990-2019

*Interpretation:* Individual carbon footprints include emissions from domestic consumption, public and private investments as well as imports and exports of carbon embedded in goods and services traded with the rest of the world. Modeled estimates based the systematic combination of tax data, household surveys and input-output tables. Benchmark scenario. Emissions split equally within households. *Source and series:* Author, see Supplementary Information.

## Supplementary Files

This is a list of supplementary files associated with this preprint. Click to download.

- [Chancel2022SI.pdf](#)