

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

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Global Carbon Inequality, 1990-2019

The Impact of Wealth Concentration on the Distribution of World Emissions

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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 (CO₂e) left to be emitted to limit global warming
33 below 1.5°C and nearly 900 GtCO₂e to limit it to 2°C.¹ At current global emissions

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

34 rates (that is, circa 50 GtCO₂e 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 tCO₂e², significantly above
37 the Global Equally-split Carbon (GEC) budget.³ 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 CO₂ per annum per person, i.e.
41 about six times less than the current global average.⁴

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 CO₂e embedded in goods and services).⁵ 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 tCO₂e
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.]

²Unless specified, all figures are reported in CO₂e and include all GHG emissions from human activity, except emissions from land use, land use change and forestry (LULUCF).

³To 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.

⁴Assuming 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).

⁵See 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.⁶

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

⁶See 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 tCO₂e
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 CO₂e 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.⁷
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.⁸

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

⁷In 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).

⁸See 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.⁹

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 CO₂e 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

⁹The 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).¹⁰

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.¹¹ 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.¹² 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]).

¹⁰In 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.

¹¹Replaced 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]).

¹²See **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 CO₂e 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.¹³ 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.¹⁴

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

¹⁴See [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

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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 2014.¹⁵

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

¹⁵also 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.¹⁶ 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

¹⁶Recent 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 CO₂ emissions, see Methods)¹⁷
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,

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

722 87, 73, 85, 2, 38, 68, 13, 71].¹⁸ 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 agents¹⁹.
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].²⁰

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

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

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

²⁰Changes 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].²¹

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:

²¹For 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 , α 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), γ 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 δ , 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 α 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 (α 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 α and γ values.

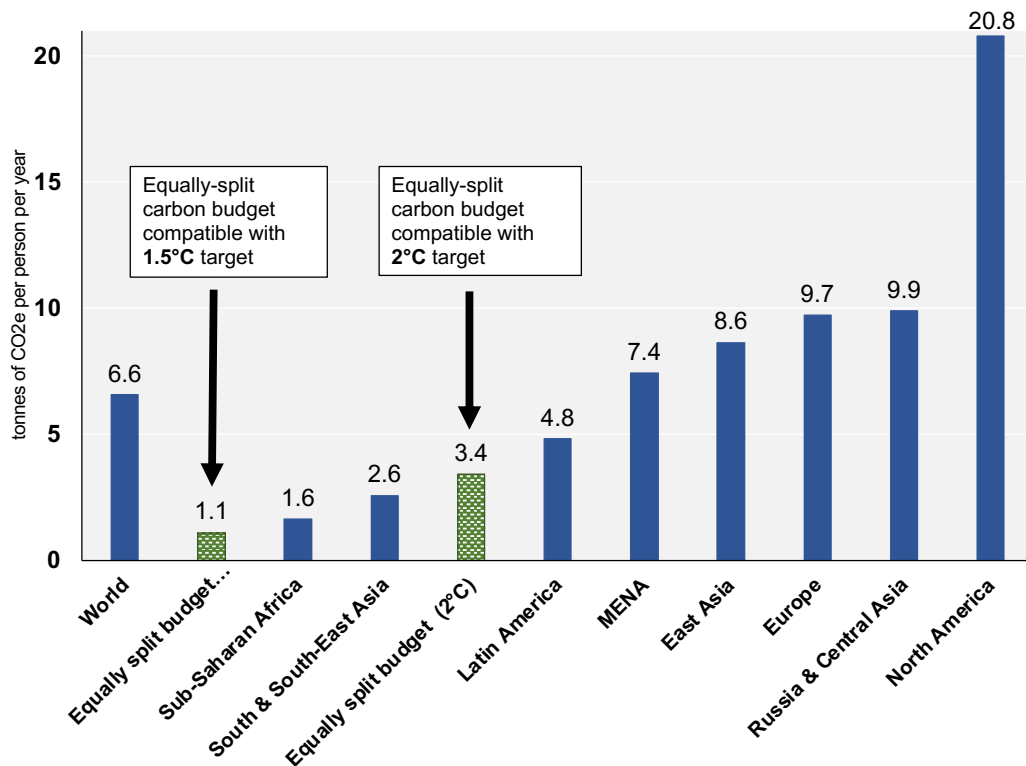
814 The flexible framework developed in this paper makes it possible to use a
815 variety of country-level studies to study the dynamics of global carbon inequality
816 over relatively long time-spans. The bottom line is that using available country-
817 level data does not modify the general conclusions 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).

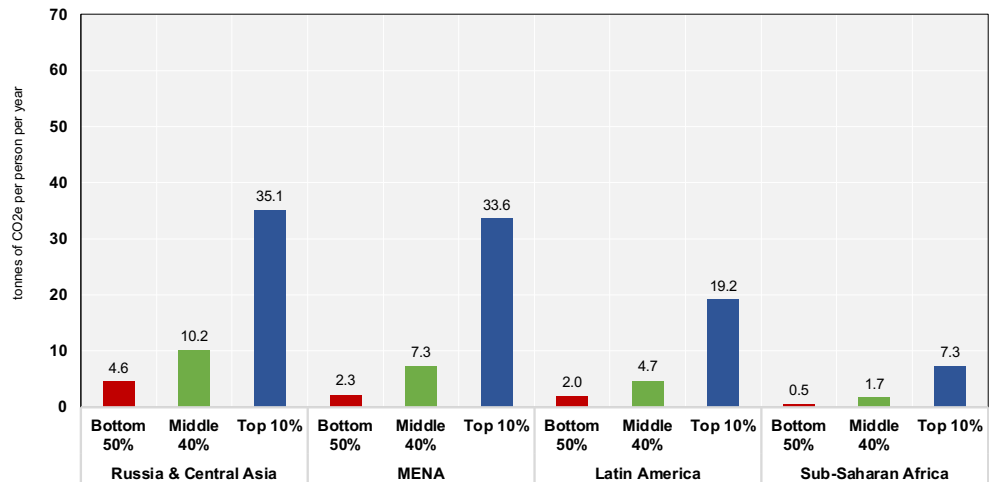
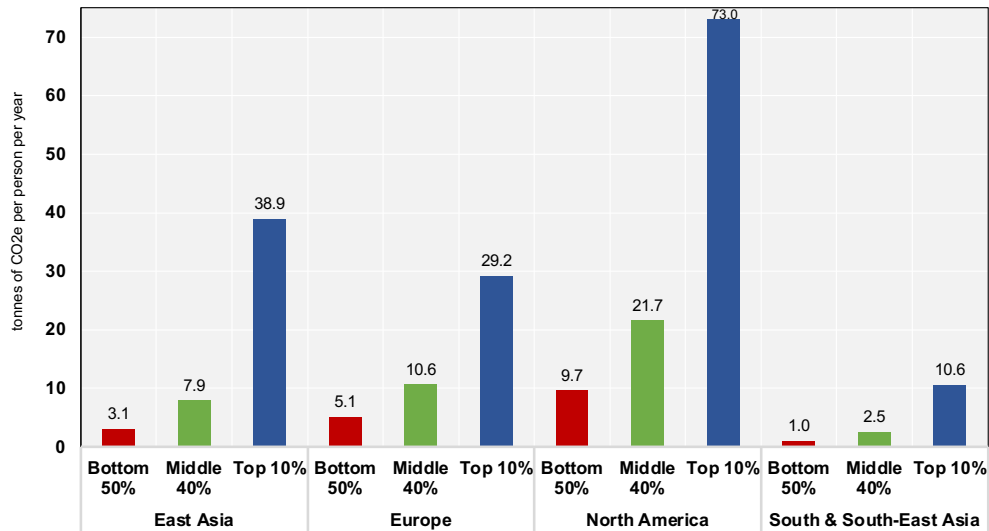
832 ACKNOWLEDGEMENTS

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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.

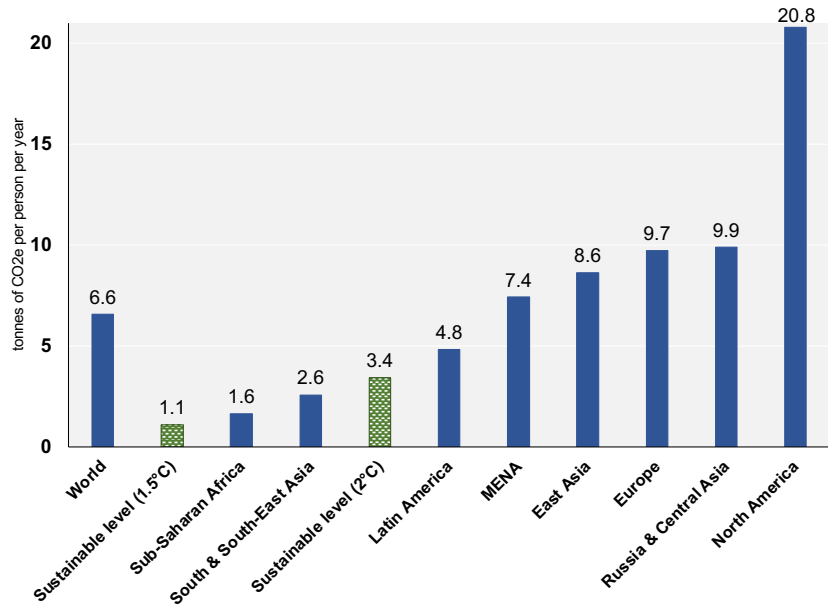
Figure I
Average GHG emissions by world region, 2019



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 (tCO₂ / year)



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

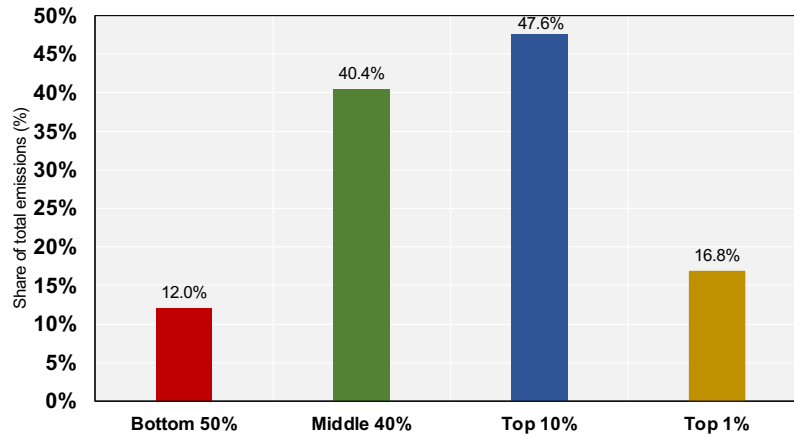
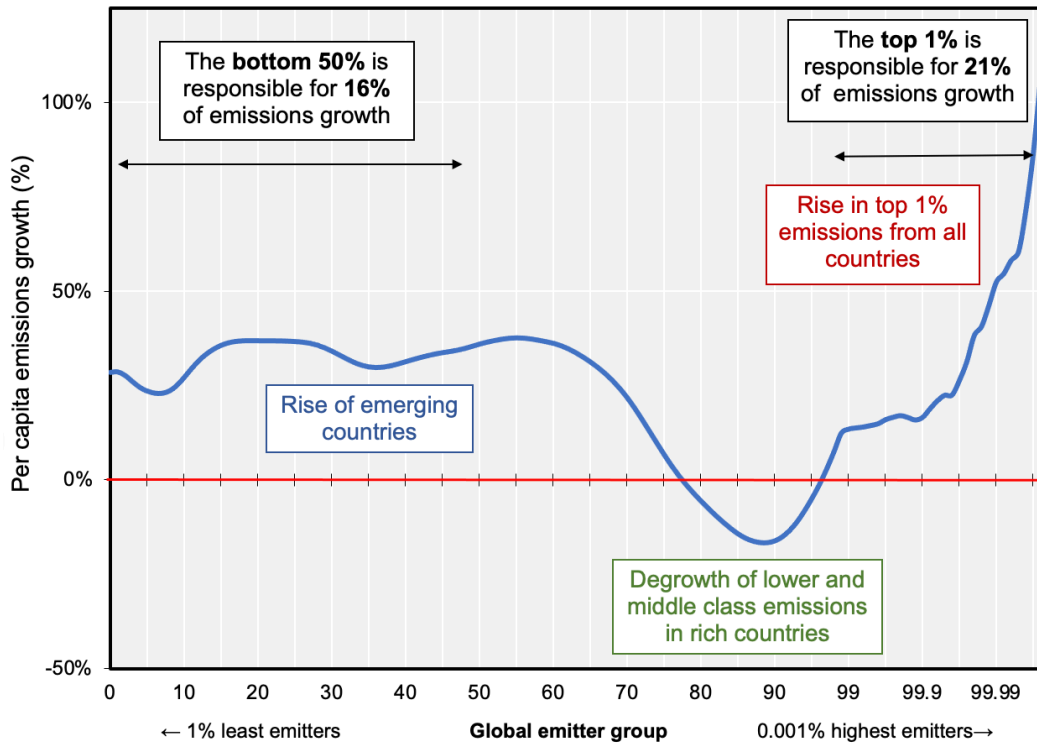


Figure III

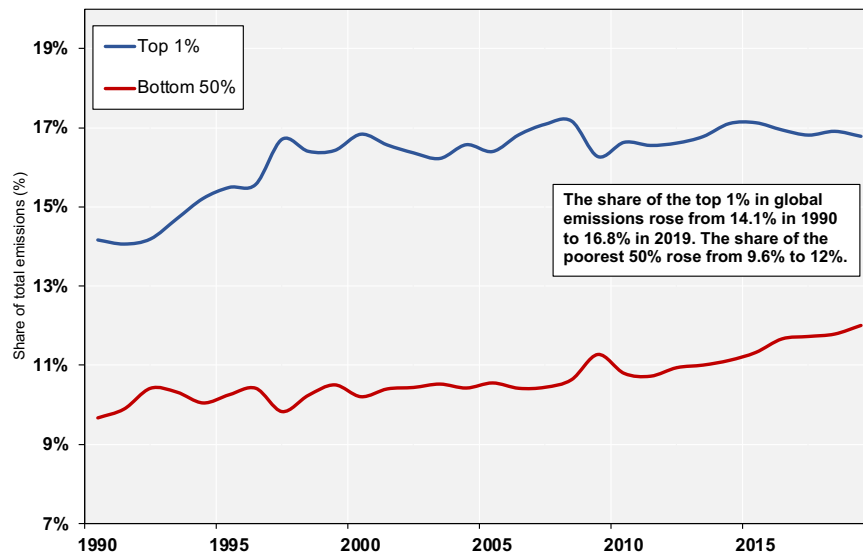
Global inequality in individual carbon emissions, 2019

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.



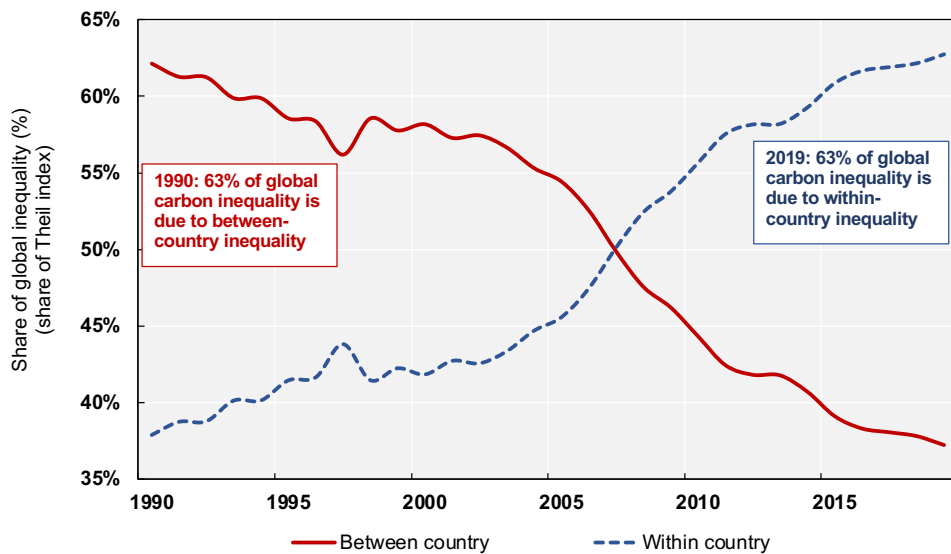
Interpretation: Emissions of the global bottom 50% rose by around 20-40% between 1990 and 2019. Emissions notably declined among groups above the bottom 80% and below the top 5% of the global distribution, these groups mainly correspond to lower and middle income groups in rich countries. Emissions of 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.

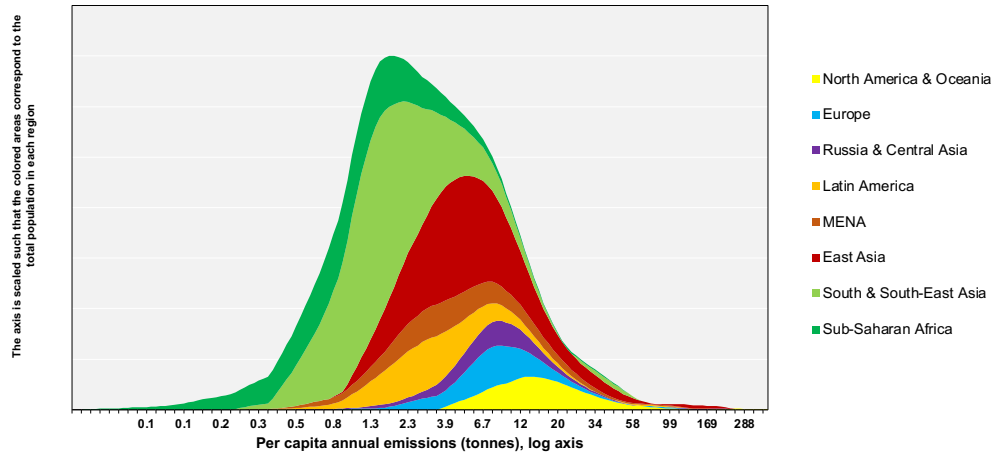
Figure V
Top 1% and bottom 50% shares in global carbon emissions, 1990-2019



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

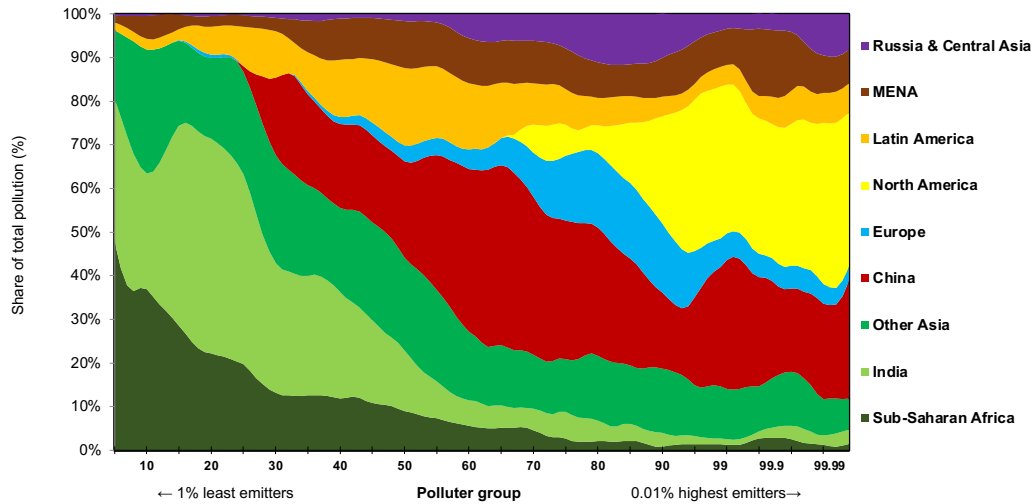
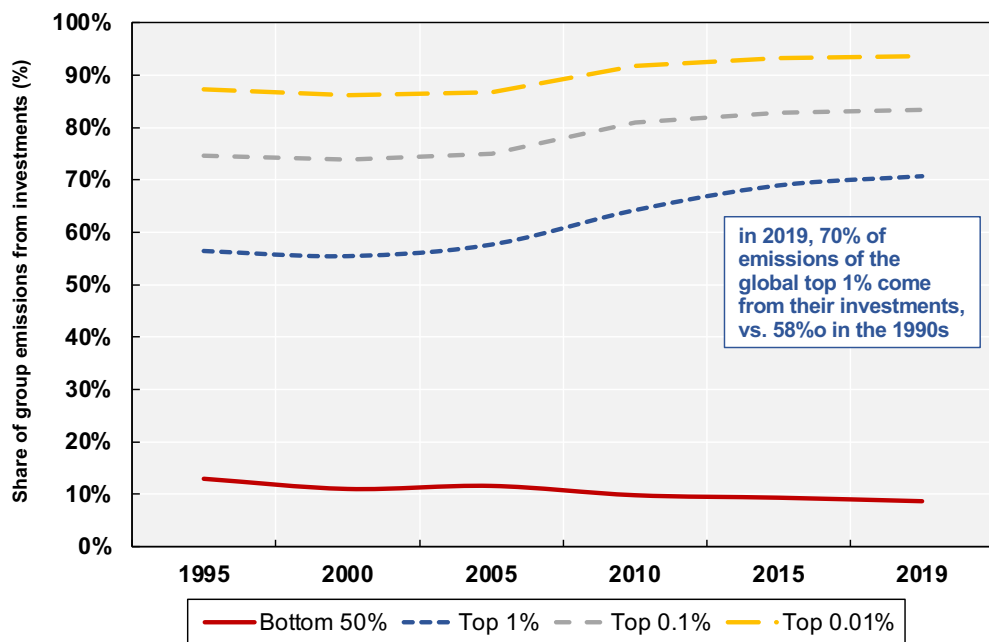


Figure VII
Global inequality in individual carbon emissions, 2019

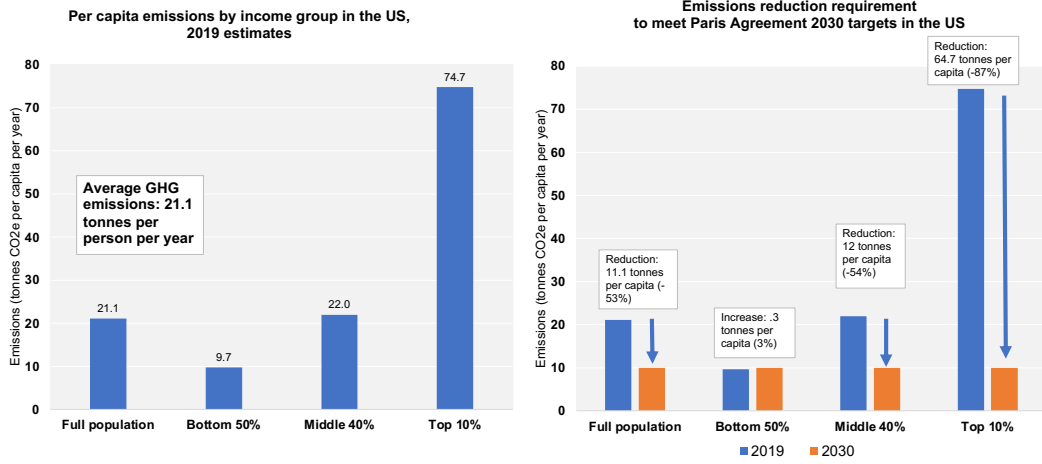
Interpretation: **Panel A** The graph shows the share of world regions in each group of global emitters, from the lowest 1% to the highest 0.1%. **Panel B** shows the global distribution (density) of individual emitters in 2019. GHG emissions measured correspond to individual footprints, i.e. they include 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

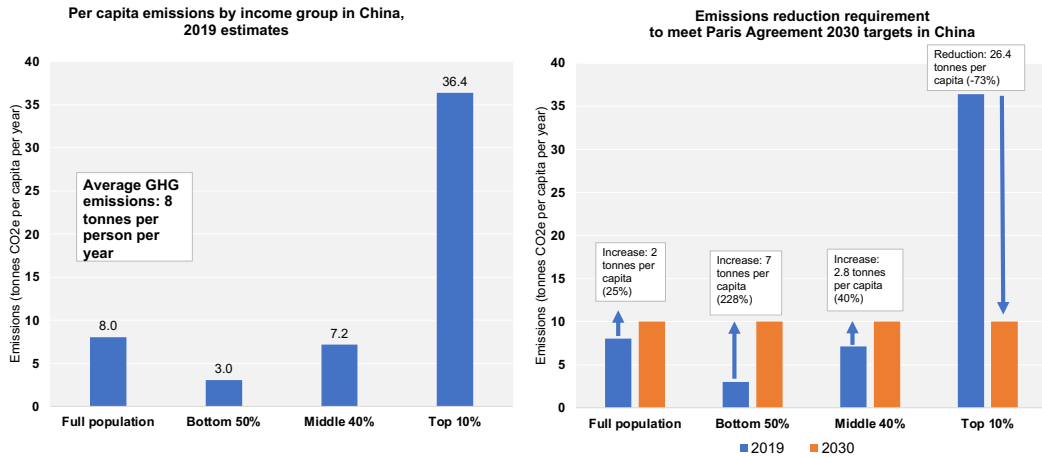


Figure IX

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)
Full population	7710	6.6	<0.1	100%
Bottom 50%	3855	1.6	<0.1	12.0%
<i>incl. Bottom 20%</i>	<i>1542</i>	<i>0.8</i>	<i><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>
Middle 40%	3084	6.6	3.1	40.4%
Top 10%	771	31	13	47.6%
<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			
Full population	6.2	6.6	32.0	50.5	7%	58%	100%
Bottom 50%	1.2	1.6	3.1	6.1	32%	96%	16%
Middle 40%	6	6.6	13.3	20.4	4%	54%	39%
Top 10%	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

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