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- Decomposing Inequality in CO₂ Emissions: the Role of Primary
 Energy Carriers and Economic Sectors
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Abstract

5 Emission inequality across countries and the contribution of the energy mix and the sectoral 6 composition of a country's energy use are of central importance to the climate debate. We 7 analyze the evolution of inequality in global CO₂ per capita emissions using both historical 8 data on energy-related CO₂ emissions and future emission scenarios generated with the 9 integrated assessment model REMIND. Within our sample of 90 countries the results 10 indicate that the Gini index declined from about 0.6 in 1971 to slightly above 0.4 in 2008. A 11 decomposition of the Gini index of total emissions into primary energy carriers and into 12 economic sectors provides additional insights. From the perspective of primary energy 13 carriers, it is indicated that this reduction is mainly attributed to declining shares of emissions 14 from coal/peat and oil in total emissions, and decreasing emission inequality within all fossil 15 primary energy sources. From the perspective of economic sectors, the decline in overall 16 inequality is almost entirely due to a pronounced decline of the contribution of emissions 17 from manufacturing & construction. Our analysis also suggests that an equally spread 18 emission reduction from any one source (i.e. primary energy carrier or economic sector) 19 would not have a major impact on overall emission inequality. The analysis of future 20 scenario data indicates that climate policy reduces absolute emissions inequality, while 21 inducing drastic progressive emission reductions in all regions.

Keywords: CO₂ emission inequality, Gini decomposition by source, economic sectors,
climate policy

24 JEL: O13, D63, Q54, Q58

26 **1** Introduction

27 The guestion of equity plays a major role in current climate negotiations, involving ethical 28 considerations and a broader analysis than the utilitarian approach to analyze the costs of 29 climate change mitigation (Gardiner, 2004). From this perspective, addressing inequality in 30 per-capita emissions across countries is central for a fair allocation of emission rights, and hence the responsibility to reduce emissions, in any future climate agreement. Currently, 31 32 global emissions are very unequally distributed between countries, with about 1 billion 33 people living in industrialized countries being responsible for roughly half of the global CO₂ 34 emissions (WEO, 2010). However, recent growth spurts have caused a sharp increase in 35 developing countries' emissions, closing part of the gap to richer countries, such that for the 36 period 2001-2008 the growth of global emission can almost exclusively be attributed to 37 developing countries (Steckel et al., 2011).

38 Previous literature has highlighted that in the process of economic development countries 39 undergo characteristic transformations of their energy systems. For instance, while countries 40 at early stages of industrialization predominantly rely on solid fuels, a large part of these fuels is replaced by grid-based, high-quality forms of energy - such as natural gas and 41 42 electricity - with proceeding industrialization (Grubler, 2008; Marcotullio and Schulz, 2007). 43 Further, rising per capita income has been found to result in a smaller share of final energy 44 use for the residential sector but larger ones for transportation and the service sector and a reversed U-shape pattern for industry (Schäfer, 2005). As convergence in per-capita 45 46 incomes is closely related to convergence in energy use patterns (Jakob et al., 2012), one 47 should expect the spectacular growth performances witnessed in a number of developing 48 countries to have major impacts on the distribution of CO₂ emissions per-capita across 49 countries.

50 Given the pivotal role of emissions inequality for climate negotiations, the issue has been 51 investigated in the literature through a variety of methods. (Hedenus and Azar, 2005) use 52 the Atkinson index (Atkinson, 1970) to find a declining trend in global emission inequality 53 between 1960 and 1999 (from 0.64 to 0.5). This is based on their finding that in 1999 the 54 20% most emitting countries emitted about 22 times the amount of carbon than the 20% least emitting countries, while in 1960 this gap had been more than twice as large. Various 55 56 authors use the Theil Index (Theil, 1972) to measure and explain global emission inequality 57 (e.g. Padilla and Duro, 2011). The advantage of the Theil index is that it can be decomposed 58 in inequality within and between country groupings (Shorrocks, 1982). Duro and Padilla, 59 (2006) use the Kaya identity (1989) to decompose emission inequality across countries into 60 the contributions of carbon intensity of energy, energy intensity and affluence. They find that during the period from 1970 to 1999 income differences were the main drivers of emission 61 62 inequality between countries, while differences in carbon intensity and energy intensity displayed lower contributions¹. Padilla and Serrano (2006) also show, using the Gini and the 63 64 Theil index for country-level emissions, that emission inequality is closely correlated with 65 income inequality and that both follow the same path over time. Finally, Duro, (2012) 66 compares a set of inequality measures such as the Gini coefficient, Theil index, the Atkinson 67 measure as well as the coefficient of variation in terms of their sensitivity to changes in the 68 distribution of CO₂ emissions over time. He concludes that different inequality indicators can 69 yield differing results due to differences in their distributive sensitivity.

A further widely employed measure of inequality is the Gini coefficient. In the case of CO_2 emissions, the cumulative share of global CO_2 emissions is plotted against the cumulative share of the countries from the lowest to the highest per capita emissions. Heil and Wodon,

¹ When Padilla and Duro (2011) apply the Theil index decomposition on emission data from the EU-27, they find that among this more homogeneous group of countries the carbon intensity of energy explains a large share of the emission inequality.

(2000, 1997) follow the methodology of Yitzhaki and Lerman (1991) to decompose the Gini coefficient in a within group and a between group component. The between group component compares rich and poor countries in order to analyze the contribution to the global emission inequality. They find that between group inequality declined slightly, but rich countries would have to reduce emissions by at least 50% to change the ranking of the highest polluting countries.

79 This paper adds to the existing literature by investigating how the energy mix and the 80 sectoral composition of a country's energy use determine inequality in global CO_2 emissions. Employing the decomposition of the Gini index pioneered by Lerman & Yitzhaki (1985), 81 82 economy-wide emissions are disaggregated into contributions by primary energy carriers 83 and economic sectors to estimate the contribution of each source of emission (i.e. each 84 primary energy carrier, or economic sector, respectively) to total inequality. While this 85 empirical methodology is well established in the analysis of income inequality between households (Leibbrandt et al., 2000), this is - to our knowledge - the first paper to employ it 86 87 to analyze inequality in carbon emissions across countries.

We analyze both past trends using historical data on energy-related CO₂ emissions for 90 countries (which currently account for about 90% of global CO₂ emissions) over the period 1971-2008 and, in order to provide an outlook on how climate policy could affect future emission inequality, we also apply the decomposition method on emission scenarios generated with the integrated assessment model REMIND (Kriegler et al., submitted); (Bauer et al., 2012).

The paper proceeds as follows: Section 2 presents the employed empirical method, and Section 3 the used data. Section 4 applies the methodology using both historical data, for a global sample as well as for a sub-sample of OECD and non-OECD countries, and results of the REMIND model. Section 5 presents the outcomes of a sensitivity analysis, while section 6 concludes.

99 2 Methodology

This study adopts the Gini coefficient as a measure of inequality for two main reasons. Firstly, it is arguably the most popular and widely employed inequality measure; for instance, it is adopted by the UNDP in its annual Human Development Reports (UNDP, 2010). Secondly, it allows for a straightforward decomposition of total CO₂ emission inequality into contributions of individual sources and an estimation of the marginal effect of a change in any of these sources on overall emission inequality.

The Gini index builds on the concept of Lorenz Curves, which plot the cumulative share of income earned against the cumulative share of the units from the lowest to the highest income (Gini, 1912; Lorenz, 1905). It can range between 0 and 1, where 0 represents total equality and 1 represents total inequality. The Gini Index for CO_2 emissions has the geometric interpretation as one minus twice the area below the Lorenz Curve for emission distribution across countries and the diagonal line, which represents perfect emission equality.

113 We apply the decomposition of the Gini index after Lerman and Yitzhaki (1985, 1984) to analyze the effects of each source of carbon emissions (i.e. primary energy carrier and 114 115 economic sector) on inequality in per-capita CO₂ emissions across countries. Lerman and 116 Yitzhaki (1985, 1984) demonstrate that the Gini coefficient of total emissions can be 117 expressed as a function of (i) the inequality within a given source, (ii) the share of this source 118 in total emissions, and (iii) its rank correlation with total emission inequality. With this Gini 119 decomposition we are able to determine the contribution of each emission source to the Gini 120 index of total per-capita emissions (*G*) between countries:

121
$$G = \sum_{k=1}^{K} G_k S_k R_k$$
 (1)

Here *k* is the index denoting the source of emissions (primary energy carrier, or economic sector, respectively), G_k is the Gini of component *k*, S_k is the share of component *k* in total emissions and R_k is the rank correlation between emission component k and total emissions². Consequently, the contribution of each single source k to overall inequality in per-capita emissions is given by $G_k S_k R_k$.

127 In addition, this methodology allows analyzing the effect of marginal changes in any single 128 source of emissions, which is useful to assess the impact on inequality in carbon emission 129 across countries that are brought about by policies and/or technological advances that equal 130 percentage reduction of emissions from any one source (see section 4.3). This marginal 131 effect of a change *e* in any source *k* for overall inequality *G* can then be written as:

132
$$\frac{\partial G}{\partial e_k} = S_k (R_k G_k - G)$$
(2)

The decomposition method presented above can be applied for any given time period. As we are interested in the evolution of inequality over time, Section 4.1 presents the decomposition of the overall inequality into the contribution of each source of emissions k(i.e. primary energy carrier, or economic sector, respectively) for 5-year intervals for the period 1971 to 2008.

In order to gain a deeper understanding of the factors driving the evolution of overall inequality, we employ the Laspeyres decomposition method (e.g. Sun and Ang, (2000) see section 4.2). The Laspeyres decomposition allows us to break down changes in overall emission inequality to changes in its single components *G*, *S*, and *R*. With Δ denoting the difference between the year 2008 and 1971, we can decompose the change in the Gini index of total per-capita emissions between two points in time into the joint contribution of three underlying effects *E* for every source of emissions:

145
$$\Delta G = \sum_{k=1}^{K} (R_k + \Delta R_k) (G_k + \Delta G_k) (S_k + \Delta S_k) - \sum_{k=1}^{K} R_k G_k S_k$$
(3)

² The rank correlation R_k ranges between +1 and -1. It will approach +1 (-1) if an emission source is an increasing (decreasing) function of total emissions.

146 This can be expressed as:

147
$$\Delta G = \sum_{k=1}^{K} \left(E_k^R + E_k^G + E_k^S \right)$$
(3')

The individual effects can be derived from carrying out the multiplication in (3) and dividing the residuals (i.e. changes of second and third order) evenly across factors. This is demonstrated exemplarily for E_k^S , i.e. the change in inequality that can be attributed to a change of the share of source *k* in total per-capita emissions, below:

152
$$E_k^S = \Delta S_k G_k R_k + 1/2 \Delta S_k (\Delta G_k R_k + G_k \Delta R_k) + 1/3 \Delta S_k \Delta G_k \Delta R_k$$
(4)

153 **3 Data**

We employ historical data from the International Energy Agency (IEA) on per capita CO_2 emissions from fuel combustion³ over the period of 1971 to 2008 (International Energy Agency, 2011)⁴ and emission scenario data over the period of 2005 to 2100 generated by the version 1.4 of the REMIND model (Kriegler et al., submitted); Bauer et al. 2012; Leimbach et al. (2010) under the framework of the RoSE project⁵ (Kriegler et al., submitted).

The IEA dataset contains data on CO₂-emissions disaggregated by primary energy carriers ('coal/peat', 'oil', 'gas' and 'other') as well as economic sectors ('manufacturing & construction', 'transport', 'residential', 'other sectors', and 'agriculture'). We exclude emissions specified as coming from 'other' sources, as this source mainly includes emissions from the combustion of biomass and waste, and there is currently a lively debate

 $^{^3}$ Our analysis focuses on energy-related CO₂-emissions; land-use emissions as well as non-CO₂ greenhouse gas emissions (such as CH₄ and N₂O) are not part of the analysis.

⁴ For summary statistics refer to Table A1 and Table A2.

⁵ http://www.rose-project.org/

on how these emissions should be accounted⁶. From the sectoral perspective, the data is 164 165 organized in a way that it attributes emissions produced in transformation sectors (such as 166 electricity generation, or refining) to final use sectors (e.g. transport or residential) according 167 to the latter sector's consumption of energy from each transformation sector. With the 168 aggregate 'other sectors' being the sum of emissions from the residential and the service 169 sector, we are able to derive 'service sector emissions' by subtracting residential emissions 170 from 'other sectors'. Within the IEA dataset we exclude emissions from the agricultural 171 sector due to a lack of available data. As agricultural emissions account for only a small fraction of energy-related emissions⁷ for the large majority of countries, we do not expect this 172 173 exclusion to seriously bias our results.

In our sample, we only include countries for which there is full information (i.e. no values marked as 'missing') for CO₂ emissions on the level of primary energy carriers as well as economic sectors. We further exclude all countries for which CO₂ emissions from any primary energy carrier or economic sector have zero entries over the entire observation period, as we expect that this might indicate erroneous accounting. We do, however, include countries that display zero observations for some years⁸. For years prior to 1990, we use emission data for the Former Soviet Union and Former Yugoslavia, respectively, while from

⁶ For instance, it has been argued that biomass should be regarded as a zero-emission source of energy, as the associated emissions have been sequestered from the atmosphere during plant growth. However, if one regards the entire life-cycle, this picture changes, as also emissions related to land-use have to be taken into account (Farrell et al., 2006; Searchinger et al., 2008).

⁷ Yet, agriculture is an important source of CH_4 , N_2O as well as CO_2 -emissions from land use, all of which are not part of our analysis.

⁸ For instance, it seems plausible that some developing countries report zero emissions from services in early periods, with a formalized service sector only beginning to emerge after a certain threshold of economic development.

181 1990 onwards, we separately include each constituent state in our sample⁹. This leaves us 182 with a sample of 90 countries (see Table A3), which accounted for approximately 90% of 183 global CO_2 emissions in the year 2008.

184 The REMIND data contain information on emissions per primary energy carrier for a 185 business as usual (BAU) and a climate policy scenario (POL). REMIND is a multi-regional global integrated assessment hybrid model which couples a Ramsey-type optimal growth 186 187 model with a technology-rich detailed energy system model and a simple climate model. It represents 11 world regions (see Figure A1) and considers the time horizon of 2005-2100¹⁰. 188 In order to match the regional aggregation of the model to our data, we generate scenarios 189 190 of future emissions on the country level by means of extrapolation. That is, we apply the 191 changes in emissions from each individual primary energy carrier that are estimated by the 192 model for any of the REMIND regions to each individual country in our dataset that is 193 included in this respective region. As the model's definition of economic sectors does not 194 correspond to the one used in the IEA data, we analyze the inequality of projected future 195 emissions only from the perspective of primary energy carriers. The scenarios analyzed 196 include a BAU scenario that assumes no climate measures and a policy scenario that 197 corresponds to a 450ppm CO₂-eq. concentration stabilization target by 2100. The policy scenario allows for overshoot and full 'when-where-what flexibility' of emissions reductions 198 199 after 2010 and accounts for the radiative forcing of all radiative substances including Non-200 Kyoto gases and aerosols.

⁹ Note that different specifications are explored in the scope of the robustness checks in Section 5.

¹⁰ The spacing of time-steps is flexible: in the default case, there are five-year time steps until 2050 and 10-year time steps until 2100. The period from 2100–2150 is also calculated to avoid distortions due to end effects. Typically, we only use the time span from 2005–2100 for model applications.

201 **4 Results**

4.1 Drivers of Changes in Emission Inequality over Time

This section gives an overview of changes in per-capita emissions and the associated Gini index between 1970 and 2008 for our sample of 90 countries. Further, in order to gain a better understanding of the underlying drivers of the developments, it decomposes overall inequality in per-capita emissions in contributions attributable to emissions from specific primary energy carriers or economic sectors.

In Figure 1, that gives an overview of how per capita CO₂ emissions have developed, we differentiate between emissions from coal/peat, gas or oil combustion. Overall emissions from those three sources are rising, mostly due to an increase in emissions from coal. Additionally, when dividing the sample into OECD and Non-OECD countries we observe a diverging trend between the two groups. Per capita emissions of OECD countries follow a declining trend after 2004, while non-OECD countries show a rising trend after 2002.

214

215 Insert Figure 1 here: Global, OECD and non-OECD energy-related CO2 per-capita216 emissions over time.

217

Over the entire observation period, the Gini index of per-capita CO₂ emissions declined from almost 0.6 in 1971 to slightly above 0.4 in 2008, indicating that over time, the distribution has become more equal (Figure 2). Interestingly, and perhaps contrary to what one could expect, the largest part of this reduction is found to take place in years prior to 1990, i.e. before many emerging economies had started the spectacular growth performance witnessed in recent years. 224

Insert Figure 2 here: Contribution of individual primary energy carriers and economic sectors
 to total inequality in per-capita CO₂ emissions measured by the Gini index

227

228 Analyzing the development of the Gini index disaggregated along primary energy carriers 229 (Figure 2 left panel) we observe that the decline in total inequality in per-capita CO₂ emissions can be attributed to (a) a significant reduction in the contribution of emissions from 230 231 oil, with the most pronounced drop taking place in the period ranging from the mid-70s to the mid-80s and (b) a reduction of similar magnitude in the contribution of emissions from 232 233 coal/peat, concentrated on the period 1985-2000. Contrary to oil and coal/peat, emissions 234 from natural gas exert an upward influence on total inequality in most years. Consequently, 235 the share of total inequality explained by the contribution from natural gas has increased 236 significantly over the observation period, namely from 8% in 1971 to 23% in 2008.

237 Regarding the evolution of the Gini index of per-capita CO₂ emissions over time along 238 economic sectors (i.e. residential, transport, services, and manufacturing & construction) 239 (Figure 2 right panel), the most striking observation is that the declining inequality in CO_2 240 emissions observed between 1971 and 2008 is almost entirely explained by the pronounced 241 drop of the contribution of emissions from the manufacturing & construction sector, which 242 occurred mainly prior to 2000. While this type of emissions accounted for more than half of 243 total inequality in 1971, this figure drops to one third in 2008. Likewise, the contribution of 244 emissions from the residential sector declines by roughly one third from 1971 to 2008. In 245 contrast, the share of service sector emissions rises from 9 to 16%, and the one for 246 emissions from transportation from 14 to 28%. The contributions of emissions from the 247 service, as well as the transport, sector increase each by almost half over the observation 248 period. However, the effect is quantitatively rather small due to their relatively low initial 249 shares in total inequality.

These findings are very likely explained by the fact that (a) developing countries are increasingly relying on coal/peat and oil to meet their growing energy needs, while for industrialized countries, natural gas plays an increasing role, and (b) structural economic change has resulted in growth of manufacturing and construction in developing countries, but a relative decline of this sector (i.e. a shrinking share of total economic activity) in industrialized countries.

4.2 Laspeyres Decomposition

257 In this section, in order to give a quantitative assessment of the influence of changes in any single factor on total inequality we employ the Laspeyres decomposition. We investigate the 258 259 contribution of each of the three factors affecting the value of the Gini index to total 260 inequality (i.e. (i) changes in the Gini index within this source, (ii) changes in the share of this 261 source in total per-capita emissions, and (iii) changes in its rank correlation with total 262 emissions) for every individual source of emissions (i.e. primary energy carrier or economic sector). Table A4 presents the respective values for each of the three factors for the first and 263 264 the last year included in our sample (i.e. 1971 and 2008). The results of the application of the 265 Laspeyres decomposition are shown in Figure 3 as the percentage of the absolute value of 266 the total observed change in inequality that can be attributed to each individual factor (i.e. 267 computed over all sources k, the changes attributable to individual factors sum up to 100%).

268 For emissions from oil as well as coal/peat, the declining contribution to the Gini index of 269 per-capita CO₂ emissions is explained by the reduced shares of emissions from these 270 energy carriers in total CO₂ emissions, lower Gini indices within each of these two sources, and to a lesser extent lower rank correlations with total emissions. For natural gas, on the 271 272 other hand, the increased contribution to the overall Gini index is almost exclusively due to a 273 higher share of emissions from this source in total emissions, while it's declining Gini index works in the opposite direction and the change in the rank correlation is negligible. 274 275 Consequently, while all three primary energy carriers are characterized by declining 'within'

emissions inequality, it is the shift away from coal and oil and towards natural gas that determines the changes in the relative contribution of the three carriers described in the previous section.

279

280 Insert Figure 3 here: Laspeyres decomposition

281

282 For emissions from manufacturing & construction, both the sector's reduced share in total 283 emissions and the lower Gini index in 2008 compared to 1971 have comparable impacts on 284 its reduced contribution on overall emission inequality. For emissions from the service as 285 well as the transport sectors, increased shares in total emissions increase inequality in total 286 per-capita CO₂ emissions. On the other hand, decreased Gini indices within both sources 287 work in the opposite direction, without compensating, however, for the previous effect. 288 Finally, for emissions from the residential sector, the reduced inequality of emissions from this sector, and a lower rank correlation with total emissions in 2008 compared to 1971, 289 290 diminish its contribution of emissions. Its slightly increased share in total emissions causes 291 only a minor increase in the contribution of emissions from this source. Similarly to the 292 decomposition by primary energy carriers, also here we observe that emissions inequality 293 within each of the individual sectors is decreasing over time. However, significant differences 294 are observed between the four sectors, with the Gini coefficient declining more for the 295 manufacturing and residential sector and less for the service and the transport sectors. As a 296 consequence, both diverging changes in the Gini index within each of the four sectors, and 297 changes in the share of each sector in total per-capita emissions, are contributing to 298 differences in overall emissions inequality between countries.

299

4.3 Marginal Effects of Changing Emission Patterns

302 This section discusses the results of the application of the methodology to determine the 303 marginal effect of an equally spread percentage reduction from any single source of 304 emissions (i.e. primary energy carrier or economic sector) on the Gini index of total per-305 capita emissions. This kind of analysis could bear importance for the formulation of climate 306 policies, as it allows assessing the impacts of e.g. technological innovations that reduce 307 emissions from one given energy carrier or economic sector (such as more efficient power 308 plants or automobiles), or a global agreement calling for equal percentage reduction of 309 emissions from any particular sector (such as the manufacturing & construction sector).

310 The results of this exercise (undertaken for 2008 data) are shown in Table 1. The most 311 striking feature is that, regardless of their sign, these marginal effects are relatively small. For instance, an across the board reduction of emissions from coal/peat by 1% would 312 313 decrease the Gini index of total per-capita CO₂ emissions across countries by no more than 314 0.07%. For the remaining energy carriers and economic sectors, the respective effects are 315 even less pronounced. Hence, we conclude that an equally spread percentage reduction 316 from any one source of CO₂ emissions would not significantly alter the prevailing pattern of 317 global inequality in per-capita emissions.

318

Insert Table 1 here: Effects of a 1% decrease of per-capita CO2 emissions from any
particular source on the GINI of total per-capita emissions (in %), using 2008 data.

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325 **4.4 Emission Inequality for Different Country Groupings**

While the focus of this paper is clearly on global inequality, we repeat the above procedure for OECD as well as non-OECD countries, respectively. This allows us to assess whether the trends identified above can also be detected for these individual country groupings.

Compared to the full sample, we find a significantly lower Gini index of overall emissions for OECD countries (Figure A2). In 1971 it lies slightly below 0.4, declining to about 0.25 in 2008. Hence, OECD countries start more homogenously and seem to converge faster to more equal per capita CO_2 emission than non-OECD countries, which start with an initial Gini of 0.6 and a Gini of 0.45 in 2008 (Figure A3).

The overall pattern in terms of the contribution of each energy carrier is similar between the full sample and the two subsamples (Figure A2 and Figure A3 left panel), but OECD countries display a more pronounced reduction in the contribution of coal/peat to overall emission inequality (which declines by almost one half). Furthermore, both country groups display a slight decrease in the contribution of emissions from oil, and a slight increase in the one from natural gas.

340 From the perspective of economic sectors (Figure A2 and Figure A3 right panel), the 341 declining contribution of emissions from the manufacturing & construction sector turns out to 342 be the main driving force behind the overall decline in CO_2 emission inequality for both 343 OECD and non-OECD countries. It is more pronounced for the former than for the latter. For 344 non-OECD countries the largest part of the reduction takes place between 1990 and 1998, 345 which is possibly related to economic recession in the countries of Eastern Europe and the former Soviet Union. For OECD countries, this declining contribution is partly 346 347 counterbalanced by an increasing contribution of transport emissions, while this sector displays only little variation for non-OECD countries. The same holds for emissions from the 348 349 service as well as the residential sector for both country groups.

This analysis suggests that the observed reductions in overall emission inequality reflect global trends. That is, changing patterns of coal/peat consumption and emissions in the manufacturing & construction sector can to a large extent explain changes in inequality in per-capita emissions across countries. This conclusion not only holds on the global level, as shown in the previous section, but also for OECD and non-OECD countries separately.

355 4.5 Application to Emission Scenarios

In order to provide an outlook on the future development in global emission inequality with
and without climate policy, we build on scenario data generated with the integrated
assessment model REMIND.

359 Figure 4 contrasts the results of a BAU scenario (left panel) and a policy scenario (right 360 panel) with a stabilization target of 450ppm CO₂-eq. by 2100. The BAU scenario indicates 361 that global emission inequality, after an initial slight decline until 2025, is characterized by an 362 increasing trend from then onwards and till the end of the century (slightly above 0.48 in the 363 year 2100). Similarly, in the policy scenario inequality declines until 2020 and then rises 364 again up to 0.46 in 2070 to slightly decline again to 0.44 in 2100. Not surprisingly, a key 365 difference between the two scenarios is the contribution of emissions from the three different primary energy carriers in overall inequality¹¹. Under the BAU scenario coal is projected to 366 be the major contributor to emission inequality from 2050 onwards. Meanwhile, with the 367 368 450ppm CO₂-eq. target, emissions from coal contribute a diminishing share to overall inequality after 2035, while emissions from oil play a more important role. 369

¹¹ Note that negative emissions arising from the use of biomass in combination with CCS, which feature in the POL scenario, are not included in this analysis.

Insert Figure 4: Contribution of each primary energy carrier (coal, natural gas, and oil) to
total inequality in per-capita CO2 emissions, measured by the Gini index.

373

374 The nature of these results is explained with the use of the Laspeyres decomposition as displayed in Figure 5. For both scenarios we find that the main driver of these changes is 375 376 changes in the shares of emissions from the different carriers. In the case of the BAU 377 scenario (left panel), an increasing share of emissions from coal is the main cause for the 378 increase in inequality between 2005 and 2100. Inversely, emissions from oil work in the opposite direction, thus leading to a decline in overall emission inequality, which cannot 379 380 however offset the effect of emissions from coal. The rank correlation with total emissions and the Gini for global coal emissions play a smaller role, and so do gas emissions. For the 381 382 policy scenario (right panel), the picture is reversed. Here the decreasing share of emissions 383 from coal and its rank correlation with total emissions contribute to a decline in overall 384 emission inequality. This effect is offset mainly by the combined effect of the rise in the 385 share of emissions from oil and its increasing rank correlation with overall emissions, and 386 secondarily gas.

387

Insert Figure 5: Laspeyres decomposition to determine the influence of changes in individualcontributing factors

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The above analysis indicates that climate change policies can be expected to result in a more equal distribution compared to the BAU scenario. As we saw under the BAU scenario global emission inequality rises by about 0.05, while under the stabilization scenario it rises by only 0.01. This result is due to both a complete phase out of the most carbon intensive fossil (coal) but also due to the overall gradual dramatic reduction in all fossil fuels deployment for all regions under the policy scenario. In specific, emissions from fossil fuels at a global level are almost halved by 2030 under the policy scenario in comparison to the BAU case (about 49260 MtCO₂/yr in BAU and 26640 MtCO₂/yr in POL), reduced to slightly less than one third by 2050 (about 62140 MtCO₂/yr in BAU and 20275 MtCO₂/yr in POL), and characterized by a seven-fold reduction by 2100 (61925 MtCO₂/yr in the BAU scenario to 9200 MtCO₂/yr in POL).

402 **5 Sensitivity analysis**

In order to assess the robustness of our results with regard to how we select the countries
included in our sample, we perform three sensitivity checks.¹²

First, we repeat our estimates excluding the Former Soviet Union and Former Yugoslavia
(for years prior to 1990) and its constituent entities (1990 and after) from the overall sample.
The results are very similar to those reported above. Therefore we can conclude that the
economies in transition do not bias our results for the overall sample.

409 Second, we include all countries for which data are available, even if zeroes are reported in 410 every single year for a given energy carrier or economic sector. For this enlarged sample of 411 128 countries, the overall Gini displays a less pronounced decline, from slightly below 0.6 in 412 1971 to slightly above 0.5 in 2008. This is guite intuitive, as countries reporting zero 413 emissions can be expected to have an upward influence on overall inequality. With regard to 414 primary energy carrier, the general patterns are repeated. Yet, with the larger sample, the 415 decline in the contribution of oil is less pronounced (from 0.21 in 1971 to 0.18 in 2008) and the increase in the contribution of natural gas stronger (from 0.14 to 0.25 in 2008). For 416 417 economic sectors, we again find that the decline in overall emission inequality is almost

¹² This section only presents the main conclusions; more detailed information is available on request.

418 exclusively explained by the reduced contribution of the manufacturing & construction419 sector.

420 Third, as observations with a zero value might be considered as a sign of reporting errors, 421 we remove all countries that report zero emissions from any single source of emissions in 422 any year from our sample. This smaller sample, which includes only 46 countries, produces 423 a smaller Gini index for overall emissions (slightly below 0.5 in 1971, and slightly below 0.4 424 in 2008). This can very likely be explained by the fact that countries that report zeroes are 425 predominantly low emitters. Hence, removing these countries from the sample curtails the 426 distribution and decreases overall inequality. For the patterns explaining the observed drop 427 in the Gini index, we again find significantly declining contributions from oil and coal/peat 428 with the contribution of natural gas working in the opposite direction. With regard to the role 429 of economic sectors, the declining contribution of emissions from manufacturing & 430 construction is confirmed to be the dominant factor.

Hence, even though the choice of methods to construct our sample affects some details of the analysis, our overall conclusions appear very robust. For all samples considered, we find that the Gini coefficient declines over time, and that this decline can mainly be attributed to declining contributions from oil and coal/peat (from the perspective of primary energy carriers) or declining contributions of manufacturing & construction (from the perspective of economic sectors), respectively.

437 6 Conclusions

This paper uses the decomposition of the Gini index proposed by Lerman and Yitzhaki, (1985, 1984) to analyze the contribution of individual primary energy carriers and economic sectors on inequality in total per-capita CO₂ emissions across countries. We analyze past trends using historical data on energy-related CO₂ emissions and also provide an outlook on 442 how climate policy could affect future emission inequality using scenario data from the443 integrated assessment model REMIND.

444 For our sample of 90 countries, which represent about 90% of global emissions in 2008, we 445 find that the Gini index of per-capita CO_2 emissions has declined considerably, from about 446 0.6 in 1971 to slightly above 0.4 in 2008. From the perspective of primary energy carries, this observation can mainly be explained by a considerable reduction in the contribution of 447 448 emissions from oil and coal/peat. A Laspeyres decomposition reveals that declining shares 449 of emissions from coal/peat and oil in total emissions and the increasing share of emissions 450 from natural gas work in opposite directions. At the same time, emissions inequality is seen 451 to decrease similarly within all three individual carriers, and this is also pushing inequality 452 downwards. From the perspective of economic sectors, decreases in inequality are almost 453 entirely due to the pronounced decline of the contribution of emissions from the 454 manufacturing & construction sector. The most relevant explanatory factors are the declining 455 share of emissions from the manufacturing & construction sector in total emissions, and the 456 declining Gini index of emission from this sector. These observations highlight the 457 importance of changing energy use patterns for inequality in per-capita CO₂ emissions 458 across countries, which had not been analyzed in previous studies.

459 Our results were found to be robust for different country groupings and marginal effects 460 tests. Firstly, repeating the analysis for OECD and non-OECD countries separately revealed 461 that, while the results are quantitatively different, a similar pattern emerges. As observed 462 also for the whole sample, inequality in per-capita emissions decreased for both OECD and 463 non-OECD countries, and this was mainly due to decreasing contributions from emissions 464 from coal/peat and oil from a primary energy carrier perspective, and a decreasing contribution from manufacturing emissions from a sectoral perspective. Evaluating the 465 marginal effect of an equally spread reduction of emissions from any one source of 466 467 emissions (i.e. primary energy carries, or economic sectors) on the Gini coefficient of total 468 per-capita emissions, we find that any such reduction would only have minor impacts on469 overall emission inequality.

470 Using scenarios of future emissions generated with the integrated assessment model 471 REMIND, we find that climate policy can be expected to result in a more equal distribution of 472 global emissions across countries. The main driver behind this effect is the share of 473 emissions from coal, which under the BAU case is projected to be the major contributor to 474 emission inequality from 2050 onwards, while in the policy case its effect is completely 475 phased out by the end of the century. Changes attributed to oil use are working in the 476 reverse direction, but their effect in overall emissions inequality is not as influential as 477 changes related to coal use. Additionally, notable progressive reductions in global total fossil 478 fuel emissions indicate the considerable diminution in fossil fuel use across all countries. 479 This factor limits the possibility for significant across-country divergence in the level of fossil 480 fuel emissions, in the context of drastic emissions reductions in order to achieve the 481 stabilization target.

482 Our findings provide a more fine-grained understanding of the underlying drivers of 483 inequality in per-capita CO_2 emissions than previous studies. In particular, they underline the 484 importance of energy system and economic transformations for emissions inequality, by 485 highlighting how changes in the use of primary energy carriers and economic activity 486 between sectors propagates into across-country inequality in per-capita CO_2 emissions.

The issue of inequality in per-capita emissions can be expected to occupy a top spot in the agenda of future climate negotiations. Any future climate agreement faces the challenge of achieving a distribution of emission rights that is recognized as equitable by all participants. Our results indicate that reducing emission inequality is compatible with climate stabilization only if high emitters accept more substantial cuts in their per-capita emissions compared to low emitters. A clear understanding of historical trends and patterns, combined with the

- 493 integrated assessment of policy scenarios, can form the basis to evaluate the implications of
- 494 future climate policies. This is where this paper intends to make a contribution.

495

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568

570 8 Appendix

571 Table A1: Summary Statistics IEA Data: Emissions per capita for primary energy carriers

Variable	Obs	Mean	Std. Dev.	Min	Max
coal_pc	2935	1.91	3.01	0.00	32.47
gas_pc	2935	0.94	1.26	0.00	6.13
oil_pc	2935	2.62	2.46	0.01	17.09
manufact_pc	2935	1.99	2.60	0.00	38.66
service_pc	2832	0.65	0.74	0.00	3.77
transport_pc	2935	1.12	1.32	0.00	14.73
residential_pc	2935	1.09	1.13	0.00	5.17

572 and economic sectors in metric tons of CO_2 (t CO_2).

573 Source: (International Energy Agency, 2011)

574

576 Table A2: Summary Statistics Scenario Data: Emissions per capita for primary energy

Variable	Scenario	Obs	Mean	Std. Dev.	Min	Max
oil_pc	BAU	1248	2.95	2.94	0.09	22.56
gas_pc	BAU	1248	3.22	3.59	0.00	22.52
coal_pc	BAU	1248	2.86	5.79	0.00	67.65
oil_growth	BAU	1170	0.51	1.65	-0.87	13.09
gas_growth	BAU	1170	4.75	17.29	-0.79	164.46
coal_growth	BAU	1170	4.72	16.61	-0.87	119.56
oil_pc	POL	1248	2.16	2.24	0.06	17.09
gas_pc	POL	1248	1.47	1.74	0.00	13.18
coal_pc	POL	1248	0.48	1.15	0.00	10.49
oil_growth	POL	1170	-0.07	0.55	-0.91	3.26
gas_growth	POL	1170	1.34	5.64	-0.85	39.76
coal_growth	POL	1170	-0.71	0.55	-1.00	5.33

577 carriers in metric tons of CO_2 (t CO_2) and their respective growth rate (in %).

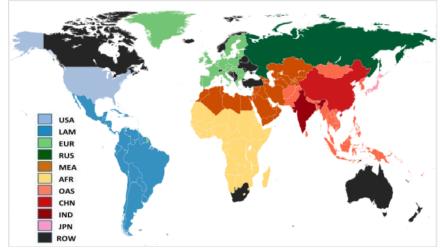
578 Source: Kriegler et al. (submitted)

579

581 Table A3: Country List

Country List IEA					
Albania	Georgia	Peru			
Algeria	Germany	Philippines			
Argentina	Greece	Poland			
Armenia	Hong Kong, China	Portugal			
Australia	Hungary	Romania			
Austria	India	Russian Federation			
Azerbaijan	Indonesia	Senegal			
Bangladesh	Islamic Republic of Iran	Serbia			
Belarus	Ireland	Singapore			
Belgium	Israel	Slovak Republic			
Bolivia	Italy	Slovenia			
Bosnia and Herzegovina	Japan	South Africa			
Brazil	Korea	Spain			
Bulgaria	Latvia	Sweden			
Canada	Lithuania	Switzerland			
Chile	Luxembourg	Syrian Arab Republic			
People's Republic of China	FYR of Macedonia	Tajikistan			
Chinese Taipei	Malaysia	United Republic of Tanzania			
Colombia	Mexico	Thailand			
Democratic Republic of Congo	Republic of Moldova	Tunisia			
Croatia	Mongolia	Turkey			
Cuba	Morocco	Ukraine			
Czech Republic	Mozambique	United Kingdom			
Denmark	Myanmar	United States			
Dominican Republic	Netherlands	Uruguay			
Egypt	New Zealand	Uzbekistan			
Estonia	Nigeria	Venezuela			
Finland	Norway	Vietnam			
France	Pakistan				

582 Source: IEA (2011)



- 583
- 584 Figure A1: REMIND Regions
- 585 Source: Luderer et al. (2013) Note: The regional acronyms are as follows: *USA* USA; LAM
- 586 Latin America; EUR Europe; RUS Russia; MEA Middle-East; AFR Africa; OAS –
- 587 other Asia; CHN China; IND India; JPN Japan; ROW rest of the World.

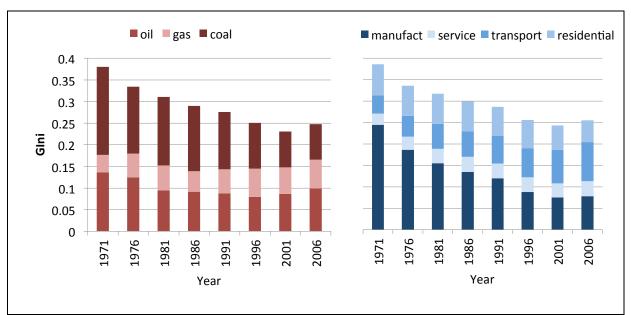
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	1971			2008		
	Share	GINI	Rank Corr.	Share	GINI	Rank Corr.
Oil	0.53	0.55	0.90	0.43	0.46	0.85
Coal	0.39	0.77	0.93	0.32	0.65	0.81
Gas	0.08	0.84	0.75	0.25	0.53	0.74
Manufact	0.52	0.65	0.95	0.34	0.45	0.94
Service	0.09	0.69	0.93	0.16	0.54	0.90
Transport	0.17	0.53	0.85	0.28	0.50	0.89
Residential	0.21	0.65	0.98	0.22	0.48	0.91

591 Table A4: Contributing factors to changes in the Gini index of total per-capita emissions by 592 source

593 Source: (International Energy Agency, 2011)

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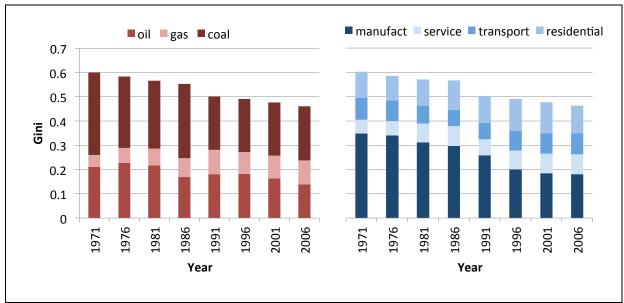


596 Figure A2: Contribution of emissions from individual primary energy carriers and economic

597 sectors to total inequality in per-capita CO₂ emissions measured by the Gini index, for OECD

598 countries only

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601 Figure A3: Contribution of emissions from individual primary energy carriers and economic

sectors to total inequality in per-capita CO_2 emissions measured by the Gini index, for non-

- 603 OECD countries only

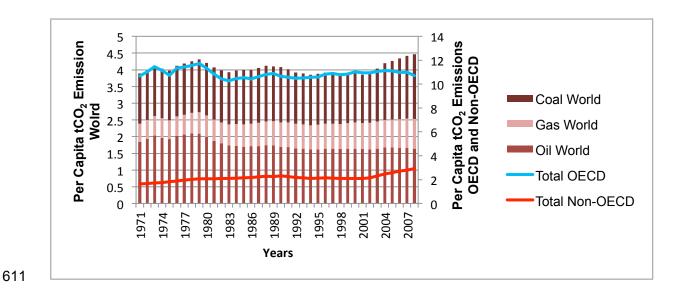
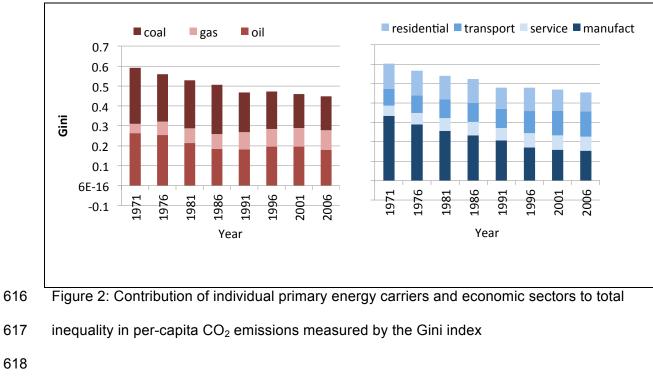


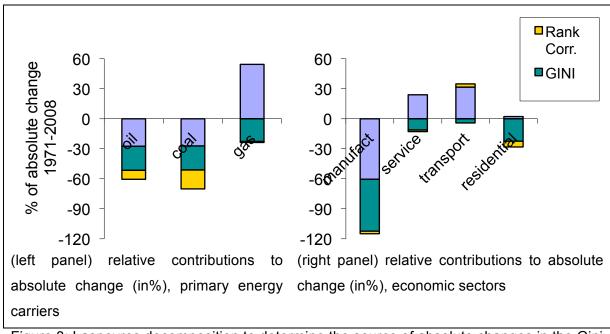
Figure 1: Global, OECD and non-OECD energy-related CO₂ per-capita emissions over time.

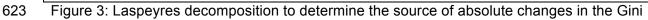
- 613 Source: IEA (2011)
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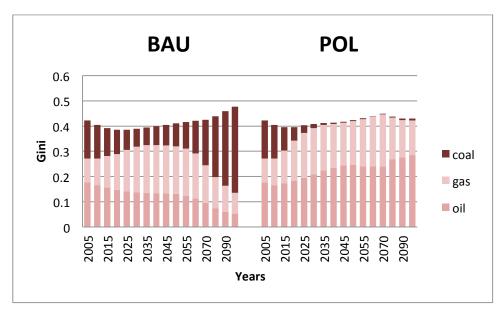
610 9 Figures and Tables



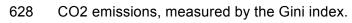


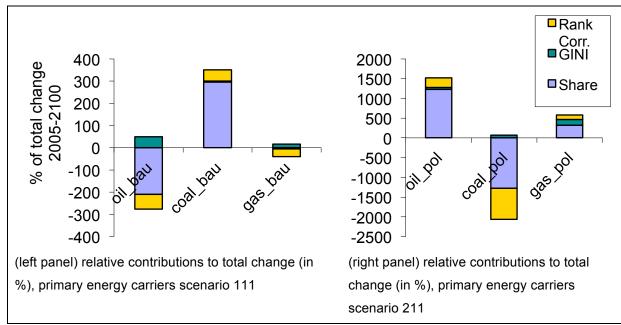


624 index of total per-capita emissions.



627 Figure 4: Contribution of individual primary energy carriers to total inequality in per-capita





633 Figure 5: Laspeyres decomposition to determine the source of changes in the GINI index of

634 total per-capita emissions.

- Table 1: Effects of a 1% decrease of per-capita CO2 emissions from any particular source
- on the Gini of total per-capita emissions (in %), using 2008 data.

Primary Energy Carriers		Economic Sectors	
Oil	0.0465	Manufact.	0.0142
Coal	-0.07	Service	-0.0144
Natural Gas	0.0235	Transport	-0.0014
		Residential	0.0017

641