
Supplementary information

Global patterns of daily CO₂ emissions reductions in the first year of COVID-19

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Global patterns of daily CO₂ emission reductions in the first year of COVID-19

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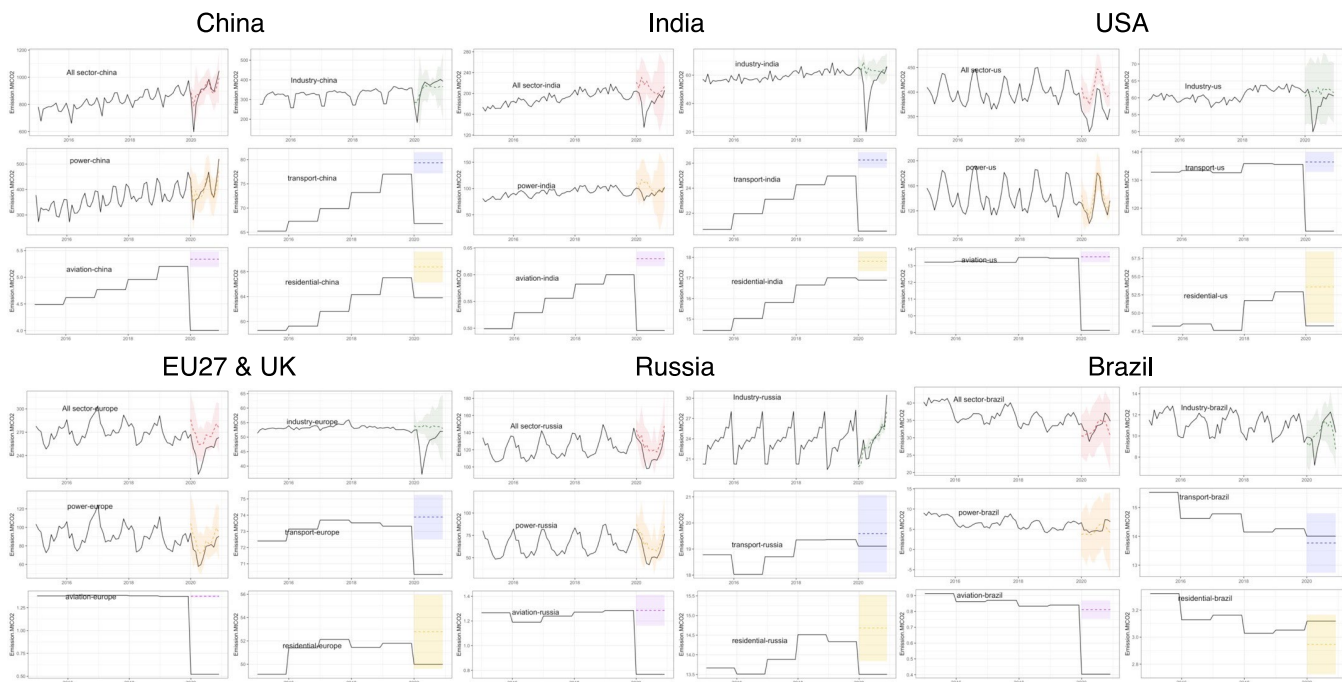
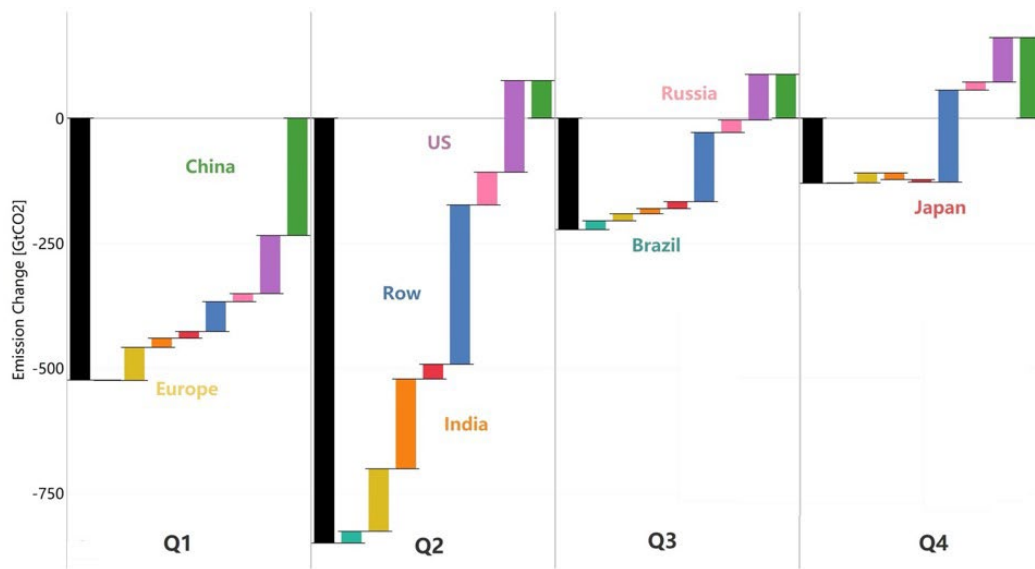
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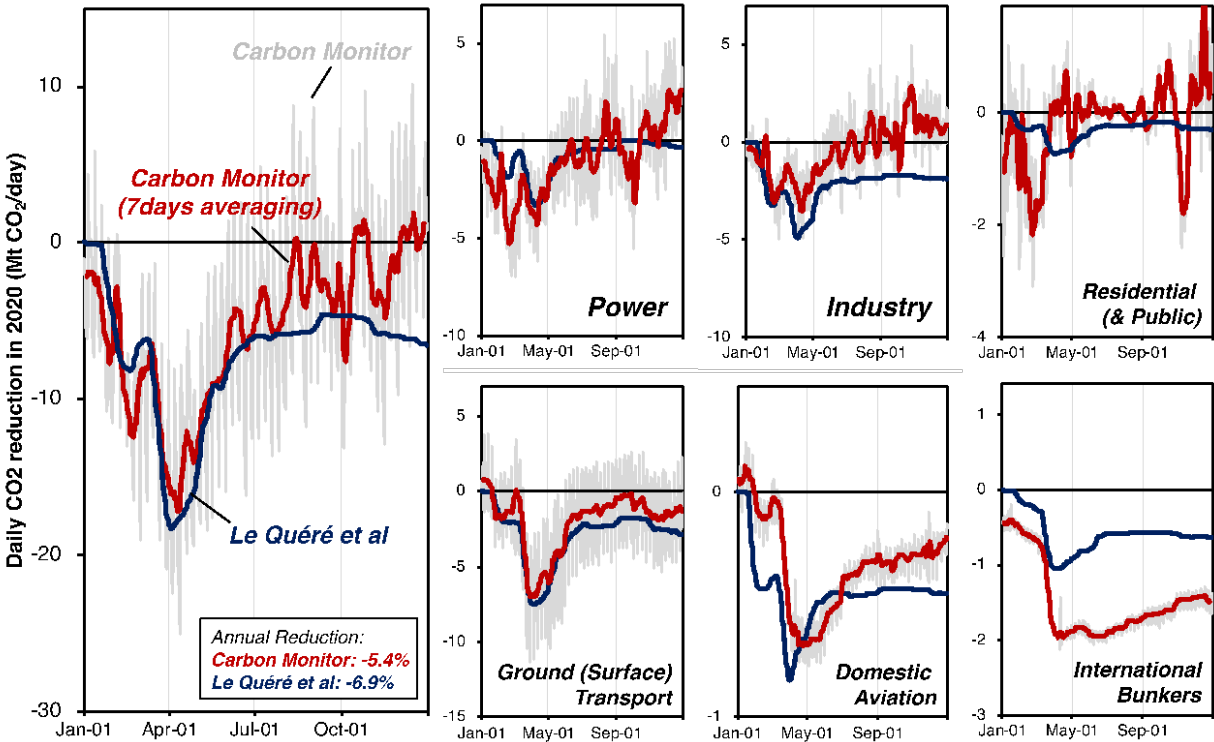
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44 **SI Figure 3 | Comparison of estimates from this study (Carbon Monitor) and Le Quéré et al. (2020) study.** The grey lines denote the
45 daily CO2 reduction in 2020 compared to the same day in 2019 from this study and the read lines denote the 7-day running mean values.
46 The dark blue lines denote the estimation of daily CO2 reduction (7-day running mean) from Le Quéré et al. (2020).

47

48 **SI Table 1 | Sources of activity data from different sectors**

Sector	Region	Data type	Source
Power Sector	China	thermal production / coal consumption	China National Power Grid (thermal production) National Bureau of Statistics (thermal production) WIND (coal consumption in Zhedian Company)
	India	thermal production	Power System Operation Corporation Limited (https://posoco.in/reports/daily-reports/)
	USA	thermal production	Energy Information Administration's (EIA) Hourly Electric Grid Monitor (https://www.eia.gov/beta/electricity/gridmonitor/)
	Europe	thermal production	ENTSO-E Transparent platform (https://transparency.entsoe.eu/dashboard/show)
	Russia	total Production	United Power System of Russia (http://www.so-ups.ru/index.php)
	Japan	thermal production / total demand	OCCTO (Organization for Cross-regional Coordination of Transmission Operators, https://www.occto.or.jp/) Hokkaido Electric Power Network (http://www.rikuden.co.jp/nw/denki-yoho/index.html), Tohoku Electric Power Network (https://setsuden.nw.tohoku-epco.co.jp/download.html),

			Tokyo Electric Power Company (https://www.tepco.co.jp/forecast/html/download-j.html), Chubu Electric Power Grid (https://powergrid.chuden.co.jp/denkiyoho/), Hokuriku Electric Power Company (http://denkiyoho.hepco.co.jp/area_forecast.html), Kansai Electric Power (https://www.kansai-td.co.jp/denkiyoho/download/index.html), Chugoku Electric Power Transmission & Distribution (https://www.energia.co.jp/nw/jukyuu/index.html), Shikoku Electric Power Company (https://www.yonden.co.jp/nw/denkiyoho/download.html), Kyushu Electric Power Transmission & Distribution (https://www.kyuden.co.jp/td_power_usages/history202010.html), and Okinawa Electric Power Company(http://www.okiden.co.jp/denki2/dl/)
	Brazil	Thermal production	Operator of the National Electricity System (http://www.ons.org.br/Paginas/)
Industry Sector	China	Industrial product outputs	National Bureau of Statistics (http://www.stats.gov.cn/english/) Trading Economics (https://tradingeconomics.com)
	India	Industrial Production Index	Ministry of Statistics and Programme Implementation (http://www.mospi.nic.in) Trading Economics (https://tradingeconomics.com)
	USA	Industrial Production Index	Federal Reserve Board (https://www.federalreserve.gov) Trading Economics (https://tradingeconomics.com)
	EU	Industrial Production Index	Eurostat (https://ec.europa.eu/eurostat/home) Trading Economics (https://tradingeconomics.com)
	Russia	Industrial Production Index	Federal State Statistics Service (https://eng.gks.ru) Trading Economics (https://tradingeconomics.com)
	Japan	Industrial Production Index	Ministry of Economy, Trade and Industry Trading Economics (https://tradingeconomics.com)
	Brazil	Industrial Production Index	Brazilian Institute of Geography and Statistics (https://www.ibge.gov.br/en/institutional/the-ibge.htm) Trading Economics (https://tradingeconomics.com)
Ground Transportation	416 Cities	TomTom Congestion Level	TomTom Traffic Index (https://www.tomtom.com/en_gb/traffic-index/)
Aviation	World	Flight distance	FlightRadar24 (https://www.flightradar24.com/)
Residential	World	population-weighted heating degree days	ERA5 reanalysis of 2-meters air temperature (Copernicus Climate Change Service (C3S), 2019)
International Shipping	World	Baltic Dry Index	https://tradingeconomics.com/commodity/baltic

SI Table 2 | Data sources of the U.S. Carbon Monitor Project

Sector	Data type	Source
State-level monthly energy consumption data		
Power sector	fuel-specific consumption data	EIA (https://www.eia.gov/electricity/data/state/)
Industry sector	natural gas consumption	EIA (https://www.eia.gov/naturalgas/data.php)
		Natural Gas

Ground transportation	prime supplier sales volumes of motor gasoline and diesel	EIA	Petroleum & Other Liquids	(https://www.eia.gov/petroleum/data.php)
Aviation	prime supplier sales volumes of kerosene-type jet fuel	EIA	Petroleum & Other Liquids	(https://www.eia.gov/petroleum/data.php)
Residential and commercial	natural gas consumption	EIA	Natural Gas	(https://www.eia.gov/naturalgas/data.php)
State-level daily indicators				
Power sector	thermal generation produced by coal, petroleum and gas	EIA	Hourly Electric Grid Monitor	(https://www.eia.gov/beta/electricity/gridmonitor/)
Industry sector	pipeline deliveries to industrial end users	Genscape	Natural Gas	(https://www.genscape.com)
Ground transportation	distance traveled	Trips by Distance data from Bureau of Transportation Statistics		
		(https://data.bts.gov/Research-and-Statistics/Trips-by-Distance/w96p-f2qv)		
Aviation	kilometers flown of flights	FlightRadar24 database		
		(https://www.flightradar24.com)		
Residential and commercial	population-weighted heating degree days	ERA5 reanalysis of 2-meters air temperature	(https://cds.climate.copernicus.eu/)	

SI Table 3 | Daily CO₂ reduction in 2020 in ROW countries under 4 different levels of OxCGRT Workplace Closing Index.

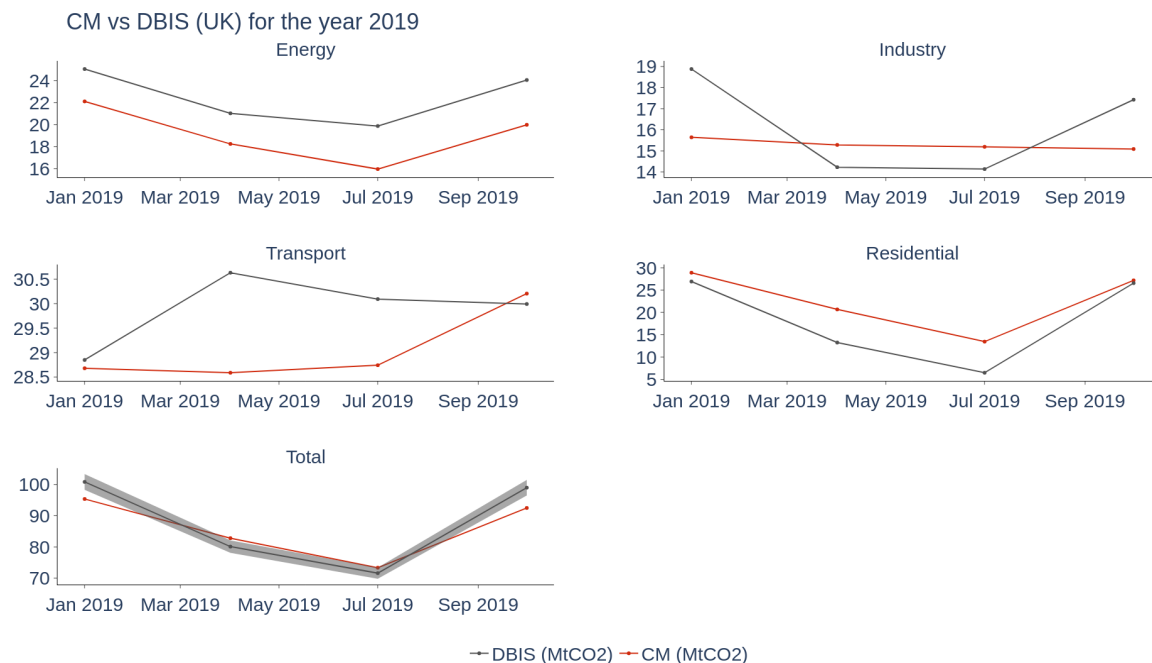
Sector	Level 0	Level 1	Level 2	Level 3
Power	0%	-3%	-4%	-16%
Industry	0%	-7%	-7%	-19%

SI Discussion | Comparison of Carbon Monitor emissions with other estimates based on energy data

To date, not many emissions reports have been made public by the environmental / inventory agencies of the world. We collected available data to assess the performance of the baseline of the Carbon Monitor (year 2019) ex post.

Department for Business, Energy & Industrial Strategy, UK Government

63 The Department for Business, Energy & Industrial Strategy (DBIS) of the UK provides an emission report by quarter up to
 64 2019¹. Although the breakdown of emissions by sector is not the same as the Carbon Monitor, they are both based on UNFCCC
 65 standards and as such, the DBIS data is a good benchmark for the Carbon Monitor.
 66



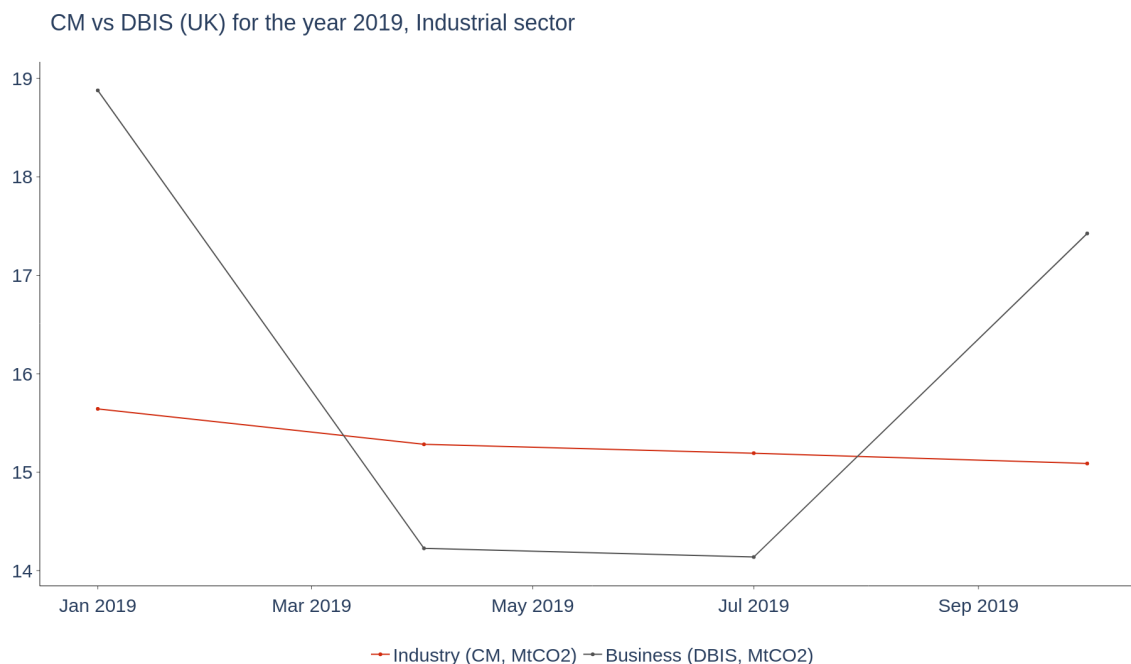
67
 68 SI Figure 5 | Carbon Monitor sectoral emissions compared to DBIS data for 2019. Sectors have been summed when appropriate.
 69 Shaded area corresponds to 95% confidence interval

70
 71 The carbon dioxide emission total of the Carbon Monitor for each of the quarters of 2019 is in line with the DBIS total, with
 72 a maximum 7% deviation. In all sectors except the Transport and Industry sectors, seasonality is comparable. The discrepancy
 73 for the Industry sector (called “Business” in DBIS) with higher emissions in quarter 1 and quarter 4 in DBIS can be explained
 74 by the fact that DBIS takes heating of businesses and industry buildings into account in this category, while the Carbon Monitor
 75 does not (see 2). The data in SI Figure 6 show that the CM allocates 15% less emissions in the first and last quarter and 7%
 76 more to the second and third quarters compared to DBIS, although the annual DBIS and CM emissions for their respective
 77 definitions of the ‘industrial’ sector differ only by 5.4%.

78
 79 The largest CM discrepancy to DBIS for the seasonality seems to be on the Transport side: DBIS takes domestic ship transport
 80 into account while CM does not. More importantly, CM is based on an emission spatial weighted average (from EDGARv4.3.1)

¹ <https://www.gov.uk/government/collections/provisional-uk-greenhouse-gas-emissions-national-statistics>

81 of temporal daily emissions from 25 cities with TomTom data, and thus ignores the contribution of other cities. This misfit of
 82 seasonality will be investigated in further work involving more precise geolocation data covering all urban areas and other
 83 roads transport.
 84



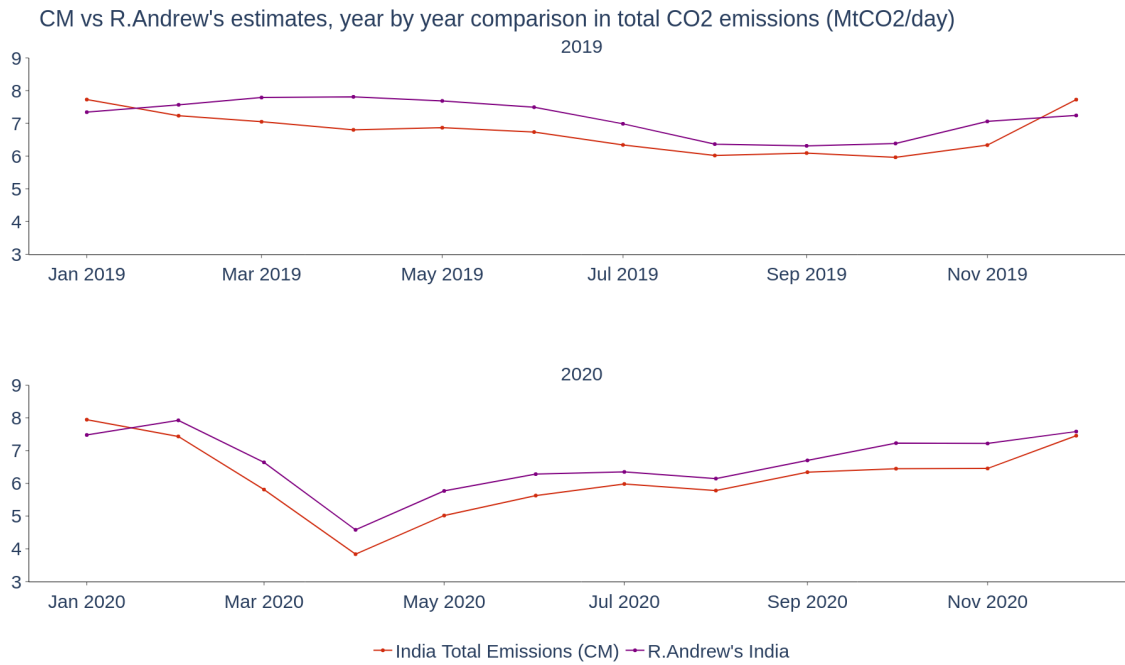
85 SI Figure 6 | Carbon Monitor “Industry” emissions compared to DBIS “Business” emission

86

87 India’s monthly fossil CO2 emissions²

88 Based on early data on fossil fuel use from the Indian government, R. Andrew [2] provided estimations of India’s emissions
 89 in 2019 and 2020, available at [2]. The 2020 estimation of CM is very much in line with RA, given an annual mean relative
 90 difference of 8.7%. The 2019 estimation of CM is however systematically lower than RA from February to June (SI Figure 7).
 91 The annual mean relative difference in 2019 is 7.9%. This negative bias of CM in the first half of 2019 can be explained by
 92 the fact that this approach uses a constant emission factor for the power production, whereas, in reality, the fuel mix is evolving
 93 on the scale of days to months. The share of coal in the Indian electricity mix (SI Figure 9) indeed rose unusually from February
 94 to June 2019, hence a greater emission factor for electricity during that period. Because CM emissions have a small bias of the
 95 same sign during both years, the difference between the two years is more consistent between CM and RA (SI Figure 8).
 96

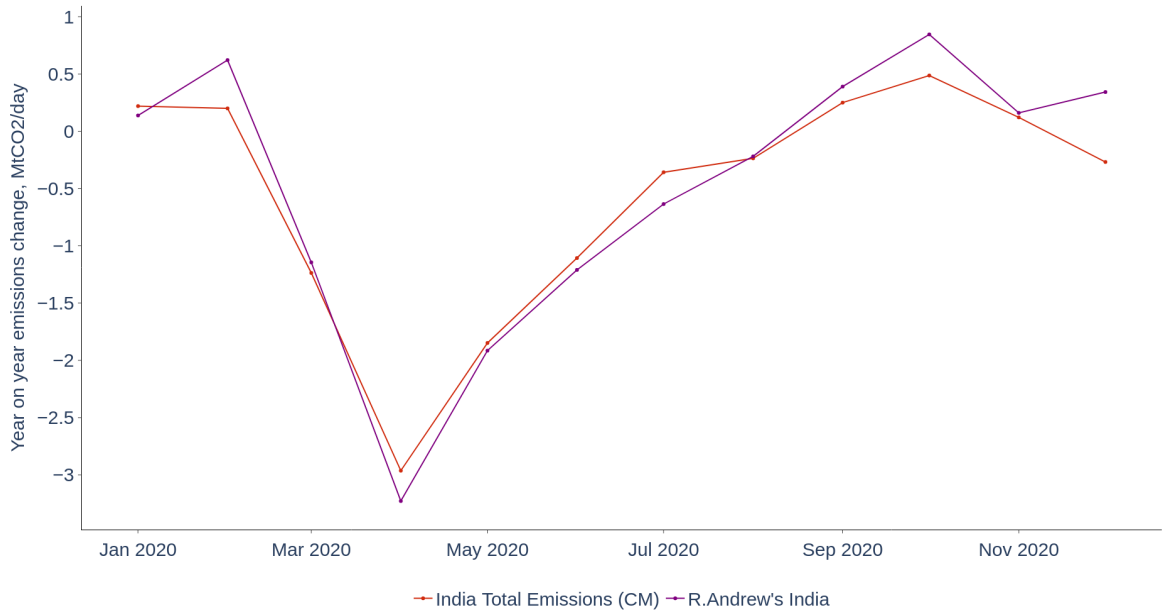
² <https://essd.copernicus.org/preprints/essd-2020-152/essd-2020-152.pdf>



97 SI Figure 7 | Carbon Monitor compared to estimations provided by RA

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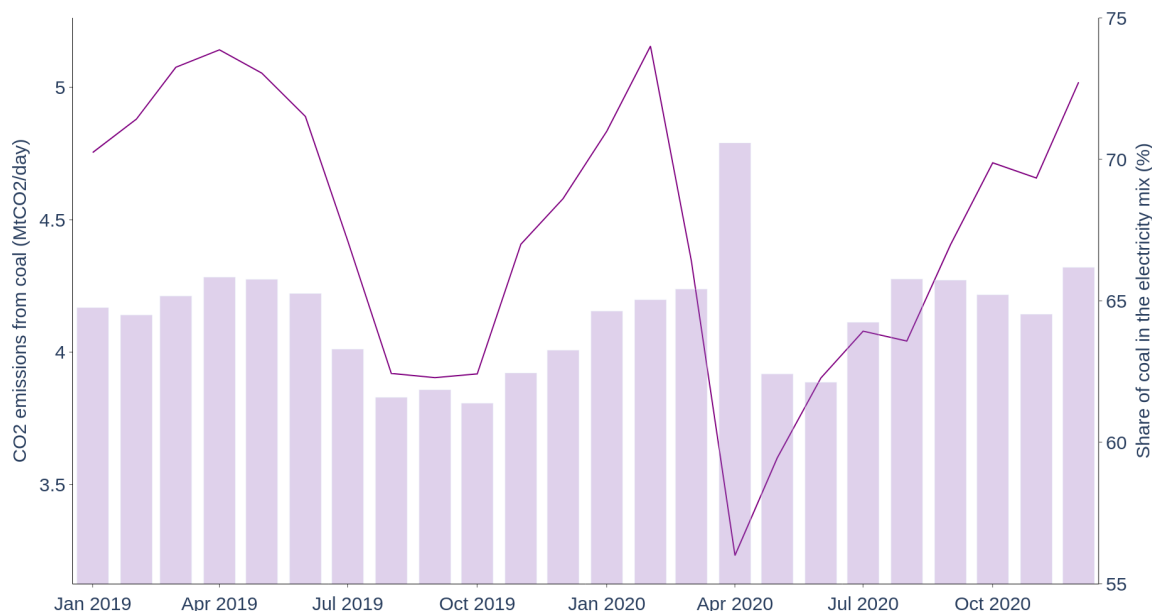
Carbon Monitor (Old and New) vs R. Andrew's estimations for India



99 SI Figure 8 | Carbon Monitor compared to estimations provided by RA, year-on-year changes 2020 minus 2019.

100
101 The estimation of CM is close to RA, given an annual mean relative difference of 8.7% in 2020 and 7.9% in 2019. The small
102 systematic errors between the two estimations are likely due to differences in methodology. While RA accounts for varying
103 fuel mix, CM uses a constant average emission factor. Indeed, India's fuel mix is changing on the scale of months and the
104 higher emissions in the first half of 2019 in Fig S3 can be explained by the higher share of coal in SI Figure 9.

R.Andrew's estimation of CO2 emissions in India



SI Figure 9 | Share of coal in the electricity mix of India. The bars denote the CO2 emissions from coal and the line denotes the share of coal in the electricity mix.

Energy Information Agency, USA

The EIA regularly publishes CO2 emissions estimates for the US in its Monthly Energy Review³ publication, and associated data. Recently the Carbon Monitor was enhanced⁴ by using more of the EIA data as a baseline in 2017 and 2018, especially the sectoral breakdown of emissions (Peng et al. in preparation). As a result, the latest version of the Carbon Monitor matches EIA data for 2019 and 2020 used in this study much more closely than the older version of Liu et al., with a mean relative bias going from 34.8% MAE in 2019-2020 changes (older version) to 12.9% (latest version). Computations of absolute levels of emissions are also greatly improved as well. The table below presents these improvements.

Year	MAE (%)	RMSE (%)
2019, older method	5.1	6.1

³ <https://www.eia.gov/totalenergy/data/monthly/#environment>

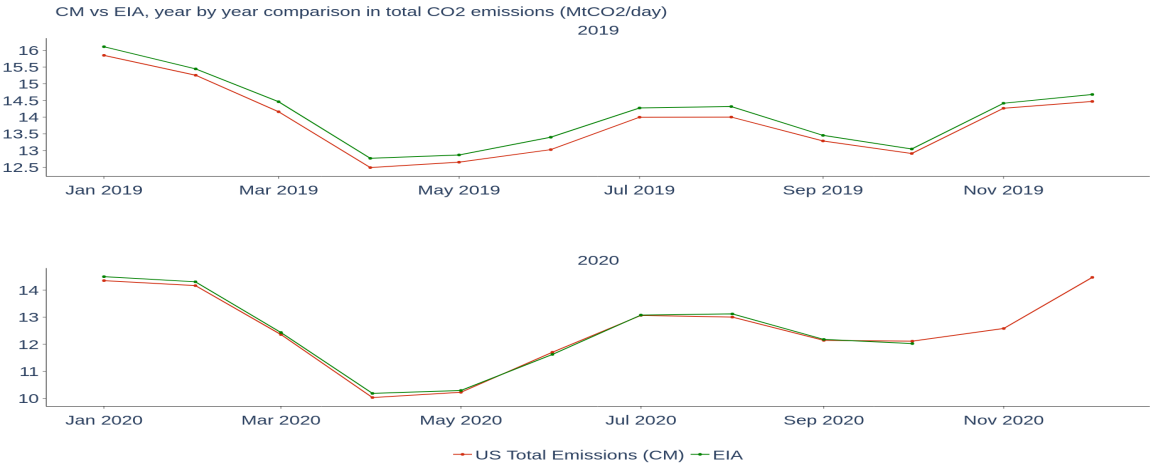
⁴ "Carbon Monitor, a near-real time daily dataset of global CO2 emissions from fossil fuel and cement production", revised 20-02-2020

2019, latest method	1.7	0.74
2020, older method	6.7	7.1
2020, latest method	1.78	0.83

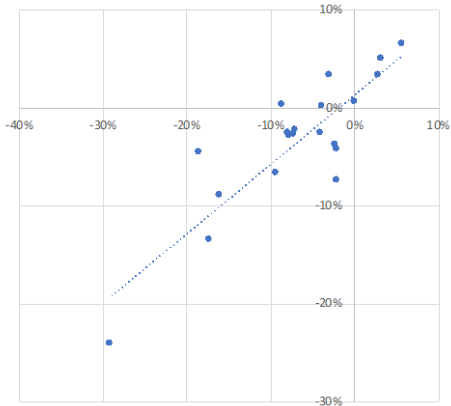
SI Table 6 | Mean absolute error (MAE) and Root Mean Square Error (RMSE) between EIA emissions and CM older (Liu et al. 2020) and latest versions (this study)

SI Figure 10 | Latest vs older method estimates for Carbon Monitor US





SI Section 2 | Comparing CO₂ decreases and the GDP changes.



SI Figure 12 | The relationship between quarterly CO₂ emission reductions (x-axis) in China, USA, India, EU27 and Russia and their quarterly GDP growth (y-axis) in 2020 compared to the same period of previous year. Detailed data are shown in SI Table 7.

SI Table 7 | Quarterly CO₂ emission reductions in China, USA, India, EU27 and Russia and their quarterly GDP growth in 2020 compared to the same period of previous year. GDP data are collected from OECD Data (<https://data.oecd.org/gdp/quarterly-gdp.htm>)

Country	Category	Q1	Q2	Q3	Q4
China	CO ₂	-9.3%	+3.0%	+3.3%	+5.8%
	GDP	-6.8%	+3.2%	+4.9%	+6.5%
USA	CO ₂	-8.6%	-16.0%	-7.2%	-7%
	GDP	+0.3%	-9.0%	-2.8%	-2.4%
India	CO ₂	-2.8%	-29.0%	-1.9%	+0.2%
	GDP	+3.2%	-24.2%	-7.5%	+0.5%

EU27	CO2	-7.9%	-17.2%	-2.2%	-1.9%
	GDP	-2.6%	-13.6%	-3.9%	-4.3%
Russia	CO2	-3.7%	-18.5%	-7.6%	-3.9%
	GDP	+0.1%	-4.6%	-3.0%	-2.7%