

Towards real-time verification of CO₂ emissions

The Paris Agreement has increased the incentive to verify reported anthropogenic carbon dioxide emissions with independent Earth system observations. Reliable verification requires a step change in our understanding of carbon cycle variability.

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Emissions of CO₂ from fossil fuels and industry did not change from 2014 to 2016, yet there was a record increase in CO₂ concentration in the atmosphere¹. This apparent inconsistency is explained by the response of the natural carbon cycle to the 2015–2016 El Niño event², but it raises important questions about our ability to detect a sustained change in emissions from the atmospheric record. High-accuracy calibrated atmospheric measurements, diverse satellite data, and integrative modelling approaches could, and ultimately must, provide independent evidence of the effectiveness of collective action to address climate change. This verification will only be possible if we can fully filter out the background variability in atmospheric CO₂ concentrations driven by natural processes, a challenge that still escapes us.

Recent changes in the carbon cycle

The atmospheric CO₂ increases of nearly 3 ppm in both 2015 and 2016 were record highs, raising the concentration to 402.8 ± 0.1 ppm in 2016 (ref. ¹). During the same period, CO₂ emissions from fossil fuel and industry remained approximately constant³. The much smaller but more variable CO₂ emissions from land-use change were higher than average in 2015, due to increased fires at some deforestation frontiers^{4,5}. Total CO₂ emission³ (fossil fuels, industry, and land-use change) grew 1.1% in 2015 to a record high of 41.5 ± 4.4 billion tonnes, and declined 2.1% in 2016 (Fig. 1). Despite the increase in total CO₂ emissions in 2015, the record high increase in the atmospheric CO₂ concentration in 2015 and 2016 occurred primarily due to a reduction in the uptake of carbon by terrestrial ecosystems in response to hotter and drier conditions associated with the 2015–2016 El Niño event², similar to past El Niño events⁶.

We project global fossil fuel and industry emissions to increase by about 2.0% (0.8–3.0%) in 2017, based on increased emissions in China of 3.5% (0.7–5.4%), decreased emissions in the US of –0.4% (–2.7–1.9%), increased emissions in India of 2.0%

(0.2–3.8%) and in the rest of the world of 1.9% (0.3%–3.4%) (ref. ³). The increased fossil fuel and industry emissions technically bring an end to the three years of approximately constant emissions that persisted from 2014 to 2016. Land-use change emissions in 2017 should be similar to their 2016 level⁵, based on fire observations using satellite data. When combining CO₂ emissions from fossil fuels, industry, and land-use change, we project 2017 global emissions to be 41.5 ± 4.4 billion tonnes of CO₂, similar to 2015 levels. Even though the projected 2017 emissions match those of the record year in 2015, they are not expected to increase atmospheric CO₂ concentration as much as in 2015 because of reinvigorated carbon uptake in natural reservoirs after the 2015–2016 El Niño event (Fig. 1).

Variability limits verification

CO₂ entering the atmosphere from combustion of fossil fuels, industrial processes, and land-use change is either absorbed by the carbon ‘sinks’, namely oceans (~25%) and land (~30%), or retained in the atmosphere (~45%). While measurements of atmospheric concentrations have low uncertainty, the attribution of concentration changes from year-to-year to specific sources and sinks is plagued by large uncertainties³. These uncertainties, combined with the inherent inter-annual to decadal variability in the land and ocean sinks, limit our ability to independently verify reported changes in fossil fuel and industrial emissions.

One indicator of our ability to verify global CO₂ emissions is the number of years required to detect a change in the trend of atmospheric concentration after a sustained change in global emissions takes place (Fig. 2). To quantify this detection delay, we use a well-established simple carbon-cycle model⁷ to project future atmospheric concentrations for different emission trajectories without natural inter-annual variability (Fig. 2). We estimate atmospheric concentrations for three different emission trajectories: sustained growth of 1% per year, approximately consistent with

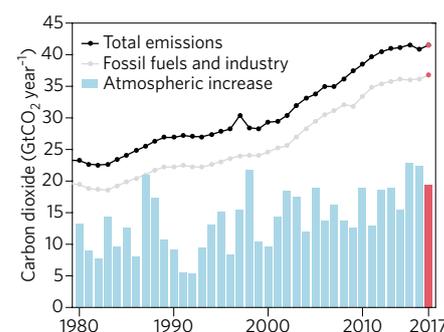


Fig. 1 | Trends in CO₂ emissions and atmospheric CO₂ concentrations. Even though CO₂ emissions from fossil fuel and industry, and total emissions including land-use change, have been relatively flat from 2014 to 2016, atmospheric concentrations saw a record increase in 2015 and 2016 (bars) due to El Niño conditions. We expect CO₂ emissions to grow in 2017 (red dots), but we expect the growth in atmospheric concentrations (red bar) to be lower in 2017 compared to 2015 and 2016, in the absence of an El Niño event.

the current pledges to the Paris Agreement⁸, constant emissions as observed from 2014 to 2016, and an arbitrary sustained reduction of 1% per year.

Our current capability to detect a change in emissions trajectory is captured by the difference between observed and reconstructed historical atmospheric concentration changes (Fig. 2). The reconstructed atmospheric growth is the difference between the reported emissions from fossil fuels, industry, and land-use change, and the estimated land and carbon sinks from models³. Over the observational period, the difference between observed and reconstructed concentrations changes, which we call the carbon budget imbalance³, has zero mean over the 1959–2016 period, but has large variability of ± 3 billion tonnes CO₂ per year (one standard deviation). With sustained changes in emission trajectories from 1% per year to 0% per year, it may take 10 years to distinguish the different emission trajectories using atmospheric observations

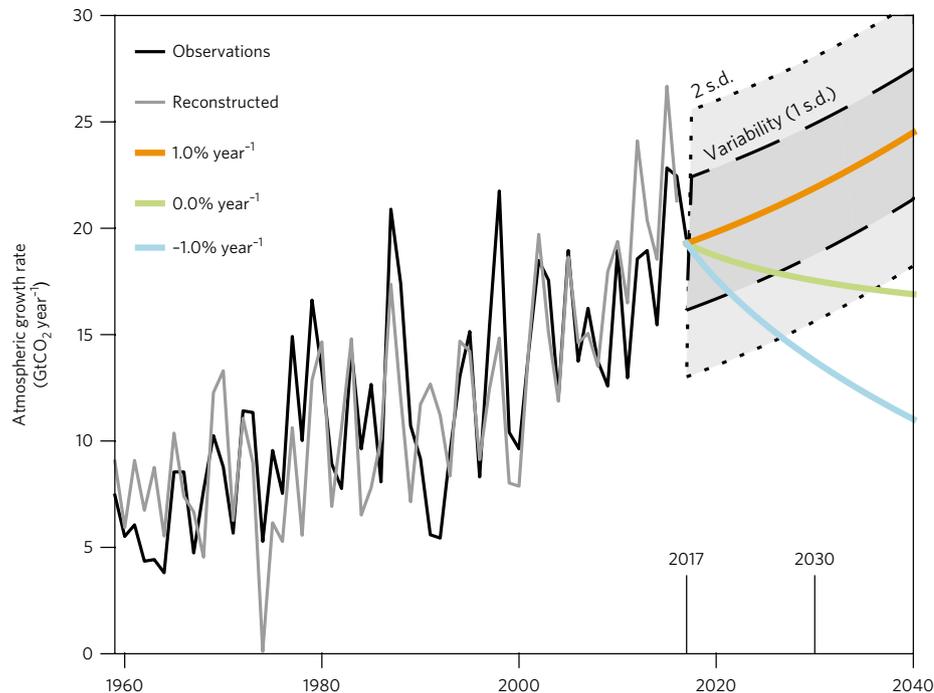


Fig. 2 | Our current ability to detect sustained changes in CO₂ emissions based on atmospheric CO₂ observations. Observations show a large inter-annual to decadal variability (black), which can be only partially reconstructed through the global carbon budget (grey; growth rate diagnosed by difference between estimated fossil fuel and industry emissions, and the simulated land and ocean sinks³). Our limited ability to fully reproduce the observed variability is quantified through the budget imbalance³ (the difference between the black and grey lines). The budget imbalance has zero mean over the 1959–2016 period, but the standard deviation (3 GtCO₂ per year) is used here to illustrate variability and our current detection delay (grey bands). If CO₂ emissions stay flat for the next decades (green; 0% annual growth), then it may take 10 years before the estimated atmospheric concentrations would exceed the budget imbalance with a probability of 68% or more (and therefore could be detected) compared to a pathway of atmospheric concentrations consistent with growth in CO₂ emissions (orange, 1% per year similar to the emission pledges submitted to the Paris Agreement). This delay increases to 20 years for a 95% probability. If emissions declined faster than expected (blue, -1% per year), then a more marked change in atmospheric growth would be expected, and a much earlier detection.

and carbon cycle models with a probability of 68% (Fig. 2). This detection delay is too long to inform the stocktake of the Paris Agreement, which occurs every five years.

Steps to reduce key uncertainties

A step-change in our ability to understand and quantify the inter-annual to decadal variability in emissions and sinks of CO₂ is needed before reported emissions can be challenged by Earth system observations. On top of continuous atmospheric measurements essential for verification, we propose several ways to better constrain each component of the global carbon budget.

Emissions from fossil fuels and industry.

Global fossil fuel and industry emissions are the sum of those countries with declining emissions (for example, US and Europe) and those countries with rising emissions (for example, China and India), indicating the

importance of tracking country level changes¹⁰. They are also the sum of the declines in coal use, growth in oil and natural gas use, and the growth in renewables which displaces some fossil fuel use, indicating the importance of tracking changes in the energy system^{9,10}. Economic growth and new policies will play an important role in determining short-term emission pathways¹⁰. Emission uncertainty persists at the country level¹¹, limiting our ability to accurately understand emission trends and drivers¹⁰. Considerable improvements are needed in estimating recent emission trends and their drivers, particularly in rapidly emerging economies and developing countries. High-precision measurements of ¹⁴CO₂ could quantify, objectively and transparently, the contribution of fossil and biogenic CO₂ sources¹².

Emissions from land-use change. Whereas emissions from land-use change are only

about 10% of the global anthropogenic total, land-use change emissions are highly uncertain³. The two dominant fluxes that make up the net flux from land-use change are emissions from land clearing and sinks from regrowth, such as afforestation, reforestation, land abandonment and shifting cultivation practices¹³. Major improvements in emission estimates will come from better estimates of standing biomass carbon and changes in carbon density across landscapes that include land degradation and disturbances currently poorly understood or not captured, and from better quantification of emissions associated with land management such as harvesting, afforestation, and shifting cultivation^{13,14}.

Land sink. Variability in the land sink is estimated from terrestrial ecosystem models driven by observed changes in environmental conditions. However, understanding of the land sink is limited by the lack of spatially explicit observations of changes in carbon in vegetation and soils¹³. Major improvements can come from systematic benchmarking of these models against the increasing availability of observations of key components of the biosphere (for example, biomass, productivity, and leaf area), and also taking advantage of emerging constraints from atmospheric CO₂ data to reduce uncertainties in the sensitivity of fluxes to climate variability, CO₂, and nutrients^{15,16}.

Ocean sink. Our understanding of the ocean sink is limited primarily by the insufficiency of physical, chemical and biological observations that would allow for quantitative understanding of the causes of inter-annual to decadal variability^{17–19}. To reduce the uncertainty in the ocean sink and quantify its variability sufficiently so as to make a material contribution to the five-year-or-less detection goal, two types of observations are critical: an optimized system of long-term, sustained observations to directly monitor the ocean carbon sink, and targeted field studies that elucidate critical processes driving inter-annual to decadal variability. These observations will allow both for direct estimation of the sink and support improvements in model-based estimates.

Now that we see signs of a sustained change in emission trajectory away from the high growth rates of the first decade of this millennium, independent verification of global emissions takes on a new imperative. Providing independent verification in the context of the Paris Agreement, with its global stocktake every five years,

leads to a new urgency for the scientific community to focus on reducing key uncertainties and quantifying natural variability in all components of the carbon cycle so that it can collectively meet the demands of policymakers and society. □

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Author contributions

G.P.P., C.L.Q., and J.G.C. designed the research; G.P.P. made Fig. 1; G.P.P., C.L., P.F., F.J. made Fig. 2. All authors analysed the data, figures, and contributed to the text.

Competing interests

The authors declare no competing financial interests.