The phase relation between atmospheric carbon dioxide and global temperature

Ole Humlum,1, 2 , Kjell Stordahl3 and Jan-Erik Solheim4

1Department of Geosciences, University of Oslo, P.O. Box 1047 Blindern, N-0316 Oslo, Norway
2Department of Geology, University Centre in Svalbard (UNIS), P.O. Box 156, N-9171 Longyearbyen, Svalbard, Norway
3Telenor Norway, Finance, N-1331 Fornebu, Norway
4Department of Physics and Technology, University of Tromsø, N-9037 Tromsø, Norway

Version 20120810

Abstract

Using data series on atmospheric carbon dioxide and global temperatures we investigate the phase relation (leads/lags) between these for the period January 1980 to December 2011. Ice cores show atmospheric CO2 variations to lag behind atmospheric temperature changes on a century to millennium scale, but modern temperature is expected to lag changes in atmospheric CO2, as the atmospheric temperature increase since about 1975 generally is assumed to be caused by the modern increase in CO2. In our analysis we use seven well-known datasets; 1) globally averaged well-mixed marine boundary layer CO2 data, 2) HadCRUT3 surface air temperature data, 3) GISS surface air temperature data, 4) NCDC surface air temperature data, 5) HadSST2 sea surface data, 6) UAH lower troposphere temperature data series, 7) CDIAC monthly data on release of CO2 from burning of fossil fuels, and 8) GWP data on volcanic eruptions. Annual cycles are present in all datasets except 7) and 8), and to remove the influence of these we analyze 12-month averaged data. We find a high degree of co-variation between all data series except 7) and 8), but with changes in CO2 always lagging changes in temperature. The maximum positive correlation between CO2 and temperature is found for CO2 lagging 11-12 months in relation to global sea surface temperature, 9.5-10 months to global surface air temperature, and about 9 months to global lower troposphere temperature. The correlation between ocean temperatures and atmospheric CO2 is high, but do not explain all observed changes.

Key words

Carbon dioxide, global temperature, sea surface temperature, volcanic eruptions.
1 Introduction

Ice core records indicate that the greenhouse gas CO$_2$ co-varied with the global temperature over several glacial-interglacial cycles, suggesting a link between natural atmospheric greenhouse gas variations and temperature on long time scales (IPCC AR4 2007; Lüthi et al. 2008).

Over the last 420 kyr variations in atmospheric CO$_2$ broadly followed temperature according to ice cores, with a typical delay of several centuries to more than a millennium (Lorius et al. 1990; Mudelsee 2001; Caillon et al. 2003). Atmospheric CO$_2$ is therefore not initiating the large glacial-interglacial climate changes, and presumably these are controlled by orbital Milankovitch cycles. It has however been suggested that the subsequent CO$_2$-rise may amplify or even in certain periods precede the global temperature increase initiated by Milankovitch cycles, but the interpretation of the proxy data is ambiguous with regard to this (Shackleton 2000; Alley and Clarke 1999; Toggweiler and Lea 2010; Shakun et al. 2012).

The observed time lag between atmospheric temperature and CO$_2$ from ice cores is thought to be caused by the slow vertical mixing that occurs in the oceans, in association with the decrease in the solubility of CO$_2$ in ocean water, as its temperature slowly increase at the end of glacial periods (Martin et al. 2005), leading to subsequent net out-gassing of CO$_2$ from the oceans (Toggweiler 1999).

Direct measurements of temperatures and atmospheric CO$_2$ with good time resolution are essential to understand empirically the effects of CO$_2$ on modern global temperature changes. The first in situ continuous measurements of atmospheric CO$_2$ made by a high-precision non-dispersive infrared gas analyzer were implemented by C.D. Keeling from the Scripps Institution of Oceanography (SIO). These measurements were initiated in 1958 at Mauna Loa, Hawaii, located at 19°N in the Pacific Ocean (Keeling et al. 1995). These data documented that not only was the amount of CO$_2$ increasing in the atmosphere since 1958, but also that the rise was modulated by annual cycles caused by seasonal changes in ocean surface temperature and photosynthesis in the terrestrial biosphere.

The Mauna Loa measurements were followed by other continuous in situ measurements at a limited number of other observation sites in both hemispheres (Conway et al. 1994; Nakazawa et al. 1997; Langenfelds et al. 2002). In the 1980s and 1990s, however, it was recognized that a greater coverage of CO$_2$ measurements over continental areas was required to provide the basis for estimating sources and sinks of atmospheric CO$_2$ over land as well as ocean regions, and to provide basis for calculating a good estimate of the global amount of atmospheric CO$_2$.

2 Modern global CO$_2$ and temperature

Today, an extensive network of international air sampling sites is operated by the National Oceanic and Atmospheric Administration’s Global Monitoring Division (NOAA/GMD) in the USA. This
organization has measured carbon dioxide and other greenhouse gases for several decades at a globally distributed network of air sampling sites (Conway et al. 1994; IPCC AR4 2007). A global average is constructed by fitting a smoothed curve as a function of time to each site, and then the smoothed value for each site is plotted as a function of latitude for 48 equal time steps per year (IPCC AR4 2007). A global average is calculated from the latitude plot at each time step (Masarie 1995), based on measurements from a subset of network sites. Only sites where samples are predominantly of well-mixed marine boundary layer (MBL) air representative of a large volume of the atmosphere are considered for the global CO$_2$ data series (IPCC AR4 2007). These key sites are typically at remote marine sea level locations with prevailing onshore winds, to minimize the effects of inland vegetation and industries. Measurements from sites at higher altitude and from sites close to anthropogenic and natural sources and sinks are excluded from the global CO$_2$ estimate. The MBL data provide a low-noise representation of the global trend and allows making the estimate directly from the data without the need for applying an atmospheric transport model (IPCC AR4 2007).

Global monthly CO$_2$ data (NOAA) are available from January 1980, and is shown graphically in figure 1, along with the monthly global sea surface temperature (HadSST2) and the monthly global surface air temperature (HadCRUT3), using data published by the University of East Anglia and the Hadley Centre, UK. In addition, in the present study we also analyze global air temperature data from the Goddard Institute for Space Studies (GISS) in USA, the National Climatic Data Center (NCDC) in USA, and lower troposphere temperature data published by the University of Alabama (UAH), Huntsville, USA. At the end of the paper a list of URL’s used to obtain the data used can be found. All these monthly data series are now sufficiently long to have collected a population of climate perturbations, and they are therefore likely to reveal essentials of the coupling between atmospheric CO$_2$ and temperature in modern time.

The monthly global atmospheric CO$_2$ (Fig. 1) has increased steadily during the entire observation period since 1980. There is, however, a pronounced annual cyclic variation superimposed on this overall development, caused by seasonal changes in the magnitude of sources and sinks for CO$_2$, controlled by dynamic exchanges with oceans and vegetation (IPCC AR4 2007). The two temperature series HadSST2 and HadCRUT3 also show an overall increase over the period, but their detailed development is more complex than for CO$_2$. In addition, they only show small net changes since early 2002 (Scafetta 2011).

In general, the two temperature records are seen to vary in close concert with each other. This also applies for the three other temperature records considered in this study (GISS, NCDC and UAH), but these are not displayed in figure 1 to avoid visual congestion. All four temperature records display rhythmic annual variations because of the uneven hemispherical distribution of land and ocean, although this is not readily apparent from the diagram, as other short-term variations tend to dominate.
Resolving the degree of coupling between CO$_2$ and temperature is not visually straightforward as illustrated by figure 1, but obviously requires a more elaborate approach to the data series.

**3 DIFF12 values**

Before analyzing the monthly data, being interested in longer than annual variations, we first removed the annual cycle from the global CO$_2$ data series by calculating a 12-month running average. This implies that we here consider the annual variation as noise only, and instead are looking for the underlying longer signal, the overall CO$_2$ increase. As the signal tends to be almost the same from one monthly observation to nearby observations and the noise does not, an average of several adjacent monthly observations will tend to converge on the value of the signal alone. The most serious consequence of smoothing or filtering is the shift of peaks and troughs in the smoothed curve, relative to the original data. If several data series are to be compared, identical filtering must therefore be applied at all series, as spurious effects else may arise, perhaps even inviting a false interpretation (see, e.g. discussion in Stauning 2011).

We next calculated the difference between the average CO$_2$ concentration for the last 12 months and the average for the preceding 12 month period; referred to as DIFF12 in the following (see figure 1 for graphical explanation). In other words, DIFF 12 represents the yearly net change with a monthly time resolution. Technically DIFF12 values were plotted for the last month considered in the calculation. By this we can efficiently visualize and analyze changes in the overall atmospheric CO$_2$ increase since January 1980, without confounding influence by the annual variation.

It is important to stress that the presence of a DIFF12 peak do not indicate the presence of a concentration peak in the original CO$_2$ data, but instead a period with more rapid increase than else. It is also worth emphasizing that by studying DIFF12 values we are investigating change rates in the amount of atmospheric CO$_2$, and not the total amount of CO$_2$ itself. However, by integrating DIFF12 values over the observation period, changes in the total amount of CO$_2$ are of cause addressed. Thus, to the degree that DIFF12 values can be explained, changes in the total amount of atmospheric CO$_2$ is therefore also explained.

We next compared the result of the DIFF12 CO$_2$ calculation with similar DIFF12 values for the global sea surface (HadSST2) and the global surface air temperature (HadCRUT3). By this all monthly data series were exposed to identical operations and the results are therefore comparable. To ensure that no spurious effects were introduced by the data filtering we carried out a similar analysis on the unfiltered monthly data, by calculating DIFF1$_{\text{annual}}$, defined as the difference between one month and the identical month one year before, e.g. January 2000 minus January 1999, etc. Finally, all DIFF1$_{\text{annual}}$ and DIFF12 series were calculated and plotted in a time-change diagram (Fig. 2), to obtain graphical overview.
From figure 2 it is seen that the 12-month filtering process results in a displacement of DIFF12 peak and minimum values 12 months ahead in time in relation to the unfiltered DIFF1\textsubscript{annual} values, exactly as expected from definition of DIFF12. More important, however, it is also seen that the relation between peaks and lows in the different data series as to their relative timing remain unchanged. If no time displacement of peaks and lows by filtering was important for the following analysis, this might have been obtained simply by changing the plotting convention, and plotting DIFF12 at the last month in the first 12-month interval, instead of at the last month in the second interval. Thus, by using a 12-month filter as outlined above we acquire a considerably clearer picture of the underlying signal in the individual data series, compared to the unfiltered data (Fig. 2).

The difference to figure 1 in vertical scale used for displaying DIFF12 values should be borne in mind when inspecting figure 2, especially for CO\textsubscript{2}, but the visual impression is nevertheless instructive. The DIFF12 graph for CO\textsubscript{2} always remain positive, between 0.5 and 3 ppm per year, a consequence of the steady increase in atmospheric CO\textsubscript{2} over the observation period. The average DIFF12 is about 1.7 ppm/yr, which as expected adds up to the entire overall increase of 51 ppm in atmospheric CO\textsubscript{2} between 1980 and 2011. The monthly development of DIFF12 for the two temperature records are more complicated, ranging from positive to negative values, reflecting repeated periods of warming and cooling over time windows of 12 months. However, the most conspicuous observation is that all three DIFF12-graphs display variations in concert with each other, suggesting internal coupling or, alternatively, the common effect of another factor, influencing all three variables.

Figure 3 shows the result of an identical analysis, but here using global surface air temperature data from the Goddard Institute for Space Studies (GISS), USA. The general visual impression is identical to that conveyed by figure 2: all three DIFF12 graphs vary in close concert with each other.

4 DIFF12 surface temperature analysis

A detailed inspection of figure 2 and 3 provides some important clues to what appears to be an important relation between global temperature and atmospheric CO\textsubscript{2}. For example, the prominent temperature change peak associated with the well-known 1998 El Nino event is seen to precede the corresponding CO\textsubscript{2} change peak by almost one year, and similar lags of several months are seen for most of the other DIFF12 CO\textsubscript{2} peaks.

One additional intriguing detail can be observed in both figure 2 and 3. The sea surface temperature change peaks typically occurs shortly before the corresponding global surface air temperature change peak, suggesting a characteristic sequence of events: 1) change of the global sea surface temperature, 2) change of global surface air temperature, and 3) change of global
atmospheric CO\textsubscript{2} content. In other words, on this background global temperature changes appear to be initiated at the surface of the oceans.

Figure 4 shows the correlation coefficient calculated for the DIFF12 surface temperature (GISS, HadCRUT3 and HadSST2) and global CO\textsubscript{2} data, to investigate the lag of atmospheric CO\textsubscript{2} in relation to the three different surface temperature records. The maximum positive correlation between CO\textsubscript{2} and temperature is found for CO\textsubscript{2} lagging about 9.5 months after GISS, 10 months after HadCRUT3 and 11 months after HadSST2.

In other words, a change of atmospheric CO\textsubscript{2} typically follows about 11 months after the corresponding change of sea surface temperatures, and 9.5-10 months later than the global surface air temperature, again showing that changes in sea surface temperature typically occur a little (1-1.5 months) before corresponding changes in the global surface air temperature. The strongest positive correlation coefficient (0.45) between CO\textsubscript{2} and temperature is found towards the HadSST2 sea surface temperature. However, the difference to surface air temperatures is relatively small, and the maximum positive correlation coefficient is 0.40 for HadCRUT3 and 0.43 for GISS.

5 DIFF12 land and ocean temperature analysis

The NCDC data series on surface air temperatures makes it possible to analyze the phase relation between atmospheric CO\textsubscript{2} and subsets of surface air temperatures calculated for land, ocean, and global surface areas, respectively. To achieve this, all three NCDC datasets were exposed to the DIFF12 numerical procedure described above. The result is shown in figure 5.

The DIFF12 NCDC data for land areas show more marked variations than the corresponding ocean and global surface air temperature records. As expected, the ocean record is showing the smallest temperature change rates, reflecting the large difference in heat capacity between land and ocean surfaces. However, all three DIFF12 NCDC series are seen to vary in concert with each other, as well as with the DIFF12 CO\textsubscript{2} series. As was the case for the previously investigated temperature series, changes in the global atmospheric CO\textsubscript{2} data series are seen to lag behind corresponding changes in the NCDC temperatures.

Figure 6 show statistics of the CO\textsubscript{2} lag in relation to the three different NCDC surface air temperature series, when seen over the entire study period. The maximum positive correlation (0.45) between temperature and CO\textsubscript{2} is found for changes in atmospheric CO\textsubscript{2} lagging 12 months behind ocean surface air temperature. The lag is shorter (9.5 months) in relation to land surface air temperature, with a maximum positive correlation coefficient of only 0.29. In relation to the global (land and ocean) NCDC temperature atmospheric CO\textsubscript{2} has a maximum positive correlation (0.39) for a lag of 10 months after corresponding changes in temperature.

Thus, summing up for the analysis of the NCDC data, changes in atmospheric CO\textsubscript{2} is lagging 9.5-12 months behind changes in surface air temperatures calculated for the two main types of
planetary surface, land and ocean, respectively. The strongest correlation (0.45) between atmospheric CO$_2$ and NCDC temperature is found in relation to ocean surface air temperatures, suggesting a rather strong coupling from changes in ocean temperature to changes in atmospheric CO$_2$.

6 DIFF12 troposphere temperature analyses

The above observations all suggest the ocean surface temperature to be an initiating factor in a planetary sequence of events, coupling temperature and atmospheric CO$_2$. Therefore, maintaining a special focus on the oceans (covering about 71% of the planetary surface), we proceed to investigate the lag between the changes of temperatures in the lower troposphere above oceans only, compared to associated changes in global atmospheric CO$_2$. This is shown in figure 7, using A) data on global atmospheric CO$_2$ and B) data on lower troposphere temperatures above oceans (UAH). Here tropical oceans are defined as oceans within 20 degrees north and south of the Equator. To ensure that our analysis is comparable with the previous analysis, all four data series were exposed to the DIFF12 yearly change procedure described above.

In general, the DIFF12 temperature change peaks for these three important ocean regions are seen to be more or less simultaneous (Fig. 7), but with the lower troposphere temperature change above the Northern Oceans lagging slightly behind corresponding changes above the Southern and the Tropical Oceans. This suggests that ocean surface temperature changes generally are initiated in the Southern Hemisphere or near the Equator. However, notwithstanding these comparatively small internal lags in the ocean temperature changes as reflected in the lower troposphere, the corresponding peaks and lows in the DIFF12 change rate of atmospheric CO$_2$ are clearly lagging several months after the lower troposphere temperature change above all three main ocean regions.

Figure 8 show the correlation coefficient between DIFF12 for global atmospheric CO$_2$ and DIFF12 for lower troposphere temperature (UAH) over land, ocean and global, respectively, for different lags of CO$_2$ in relation to temperature. The time lag is 8 months for land areas, 10 months for oceans, and 9 months for both land and oceans. The time lag of CO$_2$ is slightly shorter (one month) than found for the surface temperature records, but changes in atmospheric CO$_2$ clearly lag the associated changes in all three lower troposphere temperatures analyzed. The correlation coefficient for the global 9-month lag is 0.48.

The overall shorter time lag of CO$_2$ in relation to lower troposphere temperatures, compared to surface temperatures, suggest a typical sequence of temperature change events starting at the planet surface and propagating to the lower troposphere with about one month delay, given the time resolution of the data series.
7 DIFF12 hemispheric temperature analyses

The above observations suggest the ocean surface temperature to be an initiating factor in a planetary sequence of events, coupling temperature and atmospheric CO$_2$. This motivates an analysis of the relation between global atmospheric CO$_2$ and surface air temperature in the two hemispheres separately (Fig. 9), as especially the surface of the southern hemisphere is dominated by oceans, and the lower troposphere temperature change above the Northern Oceans appears to be lagging slightly behind corresponding changes above the Southern and the Tropical Oceans.

Here we show hemispherical temperatures according to NCDC, but other temperature databases (HadCRUT3 and GISS) show essential an identical picture. The visual impression gained from figure 9 is similar to what is shown by the analysis above: DIFF12 variations for atmospheric CO$_2$ are tracking behind both sets of hemispherical surface air temperatures. However, because of the dominance of oceans in the southern hemispheres, the temperature change rates are smaller here than in the northern hemisphere.

Figure 10 shows the correlation coefficient between DIFF12 for global atmospheric CO$_2$ and DIFF12 for northern and southern hemisphere temperature (NCDC), respectively, for different lags of CO$_2$ in relation to temperature. Positive correlation is found only for DIFF12 CO$_2$ lagging DIFF12 temperature. The maximum positive correlation is found for atmospheric CO$_2$ DIFF12 lagging about 9.5 months behind northern hemisphere temperature, and 11 months behind southern hemisphere temperature, suggesting the southern hemisphere to be leading the dynamics shown in figure 9. The correlation coefficient is considerably higher (0.56) for the southern hemisphere than for the northern hemisphere (0.26), indicating the association between changes in hemispherical temperature and changes in global atmospheric CO$_2$ to be especially strong for the southern hemisphere. Thus, both observations suggest a mainly southern hemisphere origin of observed DIFF12 changes for atmospheric CO$_2$.

8 DIFF12 release of CO$_2$ from anthropogene sources

We next investigated the degree of correlation between atmospheric CO$_2$ and the release of CO$_2$ from different anthropogene sources, using monthly data from the Carbon Dioxide Information Analysis Center (CDIAC). DIFF12 values were calculated as above, and the result is shown in figure 11. Since 2009 only annual data are at hand, and these were therefore extrapolated on a flat average monthly basis. The inclusion of these data in the present analysis is important, as the increase in atmospheric CO$_2$ since 1958 usually is ascribed to anthropogene release of CO$_2$ from anthropogene sources. It is generally assumed that about half of the emitted anthropogene CO$_2$ is absorbed by the oceans and on land by various physical and chemical processes, leaving about the remaining half to explain the rise in global atmospheric CO$_2$. 
Figure 11 shows visually that the coupling between changes in CO\textsubscript{2} released by anthropogene sources and changes in atmospheric CO\textsubscript{2} is weak, especially when compared to what was the case for temperatures (Fig. 2, 3, 5 and 7). Several of the change rate peaks in anthropogene CO\textsubscript{2} actually appear to lag behind peaks in atmospheric CO\textsubscript{2}. In other cases, the direction of change is opposed to each other. For example, this is illustrated by the pronounced 1998 El Nino peak in atmospheric CO\textsubscript{2} which corresponds to almost zero change in anthropogene CO\textsubscript{2}. The recent 2011 change peak for atmospheric CO\textsubscript{2} is another example, as this occurs simultaneously with the negative change rate for anthropogene CO\textsubscript{2}, caused by the recent economical crisis. A similar contrast has previously been pointed out (Jaworowski et al. 1992) for Mauna Loa CO\textsubscript{2} data during the petroleum crisis 1973-74.

As anthropogene CO\textsubscript{2} is released directly into the atmosphere at many locations, although mainly in the northern hemisphere, no long time lag between the two CO\textsubscript{2} records is to be expected if anthropogene CO\textsubscript{2} is important for atmospheric CO\textsubscript{2}. In contrast to this, figure 11 appears to suggest the opposite relation, if anything. This might partly be due to the extraordinary high use of fossil fuels especially in the Northern Hemisphere for heating purposes in relatively cold periods, which associates with DIFF12 CO\textsubscript{2} lows (Fig. 2). However, a detailed analysis of this falls beyond the present investigation.

In general the coupling between the two CO\textsubscript{2} data series appears visually weak compared to the previous analyzes of temperatures. The relation also appears partly conflicting, and it is difficult to determine precisely on a visual basis alone. However, it is evident from the visual analysis that changes in atmospheric CO\textsubscript{2} are generally not tracking changes in anthropogene emissions, which is contrary to expectation, if anthropogene CO\textsubscript{2} is the main driver for the observed rise in global atmospheric CO\textsubscript{2}.

The release of CO\textsubscript{2} from fossil fuels into the atmosphere is strongly asymmetrical, with the large majority being released in the northern hemisphere around 40°N (Fig. 12), which motivates a hemispherical analysis of the relation between changes in anthropogene CO\textsubscript{2} released and changes in atmospheric CO\textsubscript{2}. The influence of anthropogene CO\textsubscript{2} might be expected to be stronger and more direct in the northern hemisphere, compared to the southern hemisphere.

Figure 13 shows the relation between DIFF12 for anthropogene CO\textsubscript{2} and DIFF12 values for atmospheric CO\textsubscript{2} recorded at four different stations representing a transect from the High Arctic to the South Pole. Here we are using the period from July 1991 to December 2006, as data (NOAA) only are available from all four stations for this time window. Once again, there is no direct visual coupling from the DIFF12 values for anthropogene CO\textsubscript{2} to any of the four individual records, but the DIFF12 values for anthropogene CO\textsubscript{2} appears to lag somewhat after DIFF12 values for all individual records.

Comparing the dynamic behaviour of the four DIFF12 series for atmospheric CO\textsubscript{2}, a systematic sequence however stands out: Peaks and lows in DIFF12 values for Ascension Island (8°S) tend to occur shortly before the DIFF12 values for Mauna Loa (20°N), which leads before the
DIFF12 values for the South Pole (90 °S), which leads before the DIFF12 values for Alert (82 °N) in the High Arctic. From this, changes in atmospheric CO\textsubscript{2} appears to be initiated near or shortly south of the Equator, and from there spreads towards the two Poles within a year or so. En route, the signal presumable is modulated by local and regional effects, as is indicated by the much larger annual CO\textsubscript{2} variation (not shown here) in the High arctic, compared to that recorded at the South Pole. There is however no indications of the main signal originating at mid-latitudes in the northern hemisphere as would be expected from figure 12. The above visual analysis might be taken further by subsequent analyses, but here is suffices to investigate if correlation analyses between the four DIFF12 atmospheric CO\textsubscript{2} records and the DIFF12 anthropogene CO\textsubscript{2} record support the result of the visual analysis.

Figure 14 shows the calculated correlation coefficients between changes in anthropogene CO\textsubscript{2} and changes in atmospheric CO\textsubscript{2} for the four stations shown in figure 13. In all four cases there is a negative correlation from the time of release and 17-24 months later between changes in anthropogene CO\textsubscript{2} and changes in atmospheric CO\textsubscript{2}, showing that changes in anthropogene CO\textsubscript{2} are not causing changes in atmospheric CO\textsubscript{2}. The maximum correlation coefficient ranges from -0.30 to -0.52, and the corresponding lag from 4 to 19 months, depending on latitude. The maximum absolute value of the negative correlation coefficient is larger for the southern hemisphere than for the northern hemisphere. On the other hand, the lag of DIFF12 for anthropogene CO\textsubscript{2} in relation to DIFF12 for atmospheric CO\textsubscript{2} is larger in the northern hemisphere than in the southern. Thus, there is no indication of a northern mid-latitude origin for DIFF12 changes in atmospheric CO\textsubscript{2}. The lag is about 4 months for Ascension (8°S), 9 months for the South Pole (90°S), 14 months for Mauna Loa (20°N), and about 19 months for Alert (82°N) in the High Arctic. As each station is correlated with identical values (DIFF12 for anthropogene CO\textsubscript{2}) the observed lags also suggest a sequence of events starting near Equator in the Southern Hemisphere, and from there propagating towards the two Poles.

Figure 15 (panel a) shows the calculated correlation coefficient between changes in anthropogene CO\textsubscript{2} and changes in global atmospheric CO\textsubscript{2} for the entire period 1980-2011. It is seen that the correlation generally is low as was suggested by the visual analysis above, and noticeably smaller than found for correlations with temperatures (Fig. 4, 6 and 8). From the time of release and to about one year later there is a negative correlation between DIFF12 changes in anthropogene CO\textsubscript{2} and DIFF12 changes in atmospheric CO\textsubscript{2}. The maximum positive correlation is found by correlating DIFF12 changes in CO\textsubscript{2} from fossil fuels with atmospheric CO\textsubscript{2} DIFF12 changes 11 months earlier. However, this is opposite to what would be expected if changes in anthropogene CO\textsubscript{2} had significant impact on changes in the amount of atmospheric CO\textsubscript{2}.

However, on shorter time scales the relation is highly variable as the two panels b and c in figure 13 show. Dividing the entire observation period into the two sub-periods 1980-1998 and 1998-2011 yields a different result as to the correlation between DIFF12 changes in anthropogene CO\textsubscript{2} and DIFF12 changes in global atmospheric CO\textsubscript{2}. For the period 1980-1998 the correlation is
found to be positive for the period at and following release, while it becomes negative for the time window 1998-2011. A positive correlation would indeed be expected, if changes in the release of anthropogene CO$_2$ were controlling changes in atmospheric CO$_2$. It is however contrary to expectation that the maximum positive correlation is found one month before the time of release (Fig. 15, panel b).

This demonstrates that the relation between changes in the release of CO$_2$ from anthropogene sources and changes in atmospheric CO$_2$ is not stable, but undergoes significant changes on a decadal time scale. Had changes in the release of anthropogene CO$_2$ represented the main control on changes in atmospheric CO$_2$, the relation would be stable. A possible explanation for the unstable correlation might be due to the combined El Niño – La Nino event 1997-2001, affecting the two investigated sub periods (Fig. 15) differently. Another possible explanation was provided by Njau (2007), suggesting that anthropogene CO$_2$ has so far been added to the atmosphere in an amount which would have otherwise been added naturally, especially by the warming oceans, had the emission of anthropogene CO$_2$ been absent. However, a detailed analysis of such and other possibilities falls beyond the present investigation, and here it suffices to conclude that the absence of a stable, positive correlation suggests other processes than the emission of anthropogene CO$_2$ to control the observed changes in atmospheric CO$_2$.

Summing up, our analysis suggests that changes in atmospheric CO$_2$ appear to occur largely independently of changes in anthropogene emissions. By this we have not demonstrated that CO$_2$ released by burning fossil fuels is without influence on the amount of atmospheric CO$_2$, but merely that the effect is insignificant compared to the effect of other processes. Our previous analyzes suggest that such other more important effects are related to temperature, and with ocean surface temperature near or south of the Equator pointing itself out as being of special importance for changes in the global amount of atmospheric CO$_2$.

9 Volcanic eruptions

Volcanic eruptions are obviously relevant also to consider here, as they may contribute not only with tephra (e.g. ash), but also with substantial amounts of greenhouse gases such as H$_2$O and CO$_2$ to the atmosphere, influencing both temperature and the total amount of greenhouse gasses. In addition, a gas like SO$_2$ is injected into the atmosphere during eruptions. Oxidation and gas-to-particle conversion transforms SO$_2$ to sulphuric acid (H$_2$SO$_4$), which may effect cloud formation.

Each year several volcanic eruptions occur globally, but the majority are small, and not significant within the present global context.

The reported magnitude of volcanic eruptions depends much on both the experience and vantage point of the observer. To meet the need for a meaningful magnitude measure that can be easily applied to eruption sizes, Newhall and Self (1982) therefore integrated quantitative data with the subjective descriptions of observers, resulting in the Volcanic Explosivity Index (VEI) to
provide a relative measure of the explosiveness of volcanic eruptions. The VEI scale is a simple 0-8 index of increasing explosivity, based on volume of ejected mass, eruption cloud height, and other qualitative observations. The scale is logarithmic, with each interval on the scale representing a tenfold increase in observed ejecta criteria, with the exception of between VEI 0, VEI 1 and VEI 2. The VEI-divisions indicate the amount of tephra (volcanic ash) ejected into the atmosphere, but do not inform directly about the amount of water vapour and CO₂ emitted, but usually the amount of these is expected to follow roughly the VEI class, although with individual variations (see below). The most important characteristics of the individual VEI-classes are shown in Table 1.

Here we consider only eruptions greater than VEI size class 4, as these eruptions all are classified as large and therefore may have ejected significant of tephra and other eruption products to high altitudes, across the tropopause and into the stratosphere. The timing of such major volcanic eruptions since 1981 is shown in figure 16, together with the DIFF12 changes in the amount of atmospheric CO₂. Data on volcanic eruptions were obtained from the Global Volcanism Program (GWP).

As the VEI-data are qualitatively different from all previous data types used in the present analysis, being somewhat subjective and representing discrete events, we shun from attempting any kind of statistical analysis of these data, and will only consider briefly the visual impression gained from figure 16.

It is well known that major volcanic eruptions may influence global temperatures 4-5 years (see, e.g., Lockwood and Fröhlich 2008; Thompson et al. 2009) after the culmination of the eruption, although with variations reflecting the position of the volcano in relation to the Equator. The variable temperature effect of volcanic eruptions is illustrated by comparing the timing of major eruptions shown in figure 16 with the temperature graphs shown in figures 1, 2, 3, 5 and 7. For example, these diagrams all show global cooling after the very large Mount Pinatubo eruption that began in earnest June 1991. However, figure 16 show this very large eruption not to have resulted in any increase of atmospheric CO₂. The main DIFF12 change observed is actually a period of low growth rate for atmospheric CO₂. The same observation applies for the group of 2008 eruptions. This effect on CO₂ may be explained by volcanic gasses/aerosols and airborne debris affecting cloud cover, leading to cooling of the uppermost part of the oceans and therefore increased ability to take up atmospheric CO₂, according to Henry’s law.

Summing up, although there has been several volcanic eruptions of at least VEI 4 during the study period, the direct effect on atmospheric CO₂ in general appears to be limited on the time scale investigated, and only contributes little to explain directly observed changes in atmospheric CO₂ within the study period. Perhaps eruptions larger than VEI=6 are required to affect the global atmospheric composition and climate significantly, but there has been no eruptions of this magnitude during the study period. The analysis might be taken further by also considering the type of volcanism, and from which depth in the crust it originates. In general, deep-seated volcanism is
relatively rich in CO₂, while tholeiitic eruptions originating from more shallow depths usually are relatively rich in H₂O.

As was the case for anthropogene CO₂, we can therefore not conclude that volcanic activity since January 1980 is without influence on atmospheric CO₂, but only that the effect of eruptions in the study period apparently is small compared to other factors. On the time scale investigated here, the dominant effect of volcanic eruptions appears to be increased removal of CO₂ from the atmosphere by the oceans, presumably caused by volcanic eruptions affecting the global cloud cover, and thereby resulting in lower ocean surface temperature. On longer (geological) time scales the relation may however well be different.

10 Fourier frequency analyzes

Visual inspection of figures 2, 3, 5, 7 and 9 suggests that some DIFF12 series may be changing according to one or several rhythmic variations. Therefore, to investigate if these records are in fact characterised by such periodic variations, a Fourier frequency analysis was carried out on four key data series, 1) atmospheric CO₂ (NOAA), 2) HadCRUT3, 3) HadSST2 and 4) CO₂ released by use of fossil fuels (CDIAC). The results are shown in figure 17.

The DIFF12 atmospheric CO₂ record is seen to be dominated by two periods of about 3.8 and 2.5 yr length, both with amplitude greater than 0.25 ppm. The two DIFF12 temperature spectra are rather similar to each other, dominated by a period of about 3.7 yr length with amplitude greater than 0.09°C. The spectrum for the DIFF12 CO₂ fossil fuel record is different. In particular, there is no frequency peak suggesting the existence of an important period in the range from 3.7 to 3.8 yr, which is a principal period length in the other three records. However, the CO₂ fossil fuel record does show a less pronounced peak around 4 yr. The potential implication of this peak will be commented later.

The significance levels shown in figure 17 are peak-based critical limit significance levels, which are of particular merit in ascertaining the significance of the largest spectral component (SeaSolve 2003; Humlum et al. 2012). In this type of test, one seeks to disprove the null hypothesis postulating either a white noise signal with no autocorrelation (AR(1) = 0.0), or a red noise signal with autocorrelation (AR(1) > 0.0). Red noise is present when the background power decreases with increasing frequency, and the autoregressive value is a measure of the similarity between observations in a time series as a function of the time separation between them. All four DIFF12 series are characterised by high autocorrelation, with AR(1) coefficients ranging from 0.97 to 0.99.

With the background set, the peak spectral power was then compared against the various critical limits. A 99% critical limit is that level where in only 1 of 100 separate random noise signals the highest peak would achieve this height strictly due to random chance. Likewise, for a peak reaching the 50% critical limit there is a 50-50 probability that this could have arisen strictly from chance. The critical limit test used here is different from traditional confidence or significance
levels that apply to a single data set only. For example, a standard 95% confidence limit would specify a level where 5% of the points in a single spectrum would be expected to lie above this height strictly due to random chance.

Exposed to the critical limit test all four DIFF12 series display several frequency peaks exceeding the 95% critical limit. For that reason the alternative null hypothesis, that the observed oscillations represent red noise only, can be rejected using a 95% peak-based critical limit test, and the significant frequency peaks are positively worthy of interpretation. Several peaks are seen to exceed even the 99% critical limit.

Comparing the different Fourier frequency spectra shown in figure 17, the degree of similarity between the two temperature records (HadCRUT3; HadSST2) and the atmospheric CO₂ record appear higher than the similarity between anthropogene CO₂ and atmospheric CO₂. This lends support to the result of the previous analyzes, that changes in atmospheric CO₂ appear to be coupled especially to changes in temperature, and not to the same degree to changes in CO₂ released by anthropogene sources.

The lower panel (d) in figure 17 indirectly shows the main rhythms of the world economy. However, the amount of anthropogene CO₂ must to some degree also reflect the production of heat and electricity for heating purposes, especially during winter in the Northern Hemisphere, as this is where most people live. Thus, it is not entirely surprising that the prominent temperature DIFF12 peak of about 3.7 yr (HadCRUT3 and HadSST2) may find itself represented in the broad frequency peak around 0.25 yr⁻¹ (4 yr) in the DIFF12 data for anthropogene CO₂. The visual inspection of figure 9 leads to a similar interpretation.

11 Discussion

By the above analysis we have demonstrated that there exists a clear phase relationship between the change of different global temperature indices, and corresponding changes in the global amount of atmospheric CO₂. We have used data on CO₂ and temperature with a monthly time resolution for the time window January 1980 – December 2011, but instead of analysing the monthly data, we have removed the influence of annual cycles by analysing DIFF12 values. The average DIFF12 CO₂ value for the entire observation period is about 1.7 ppm/yr, which adds up to the entire observed increase of 51 ppm of atmospheric CO₂ since January 1980. Therefore, to the degree that the observed DIFF12 atmospheric CO₂ values (Fig. 3, 5, 7 and 8) can be explained by temperature changes, the overall increase of atmospheric CO₂ since January 1980 (Fig. 1) is also explained.

The coupling from changes in atmospheric CO₂ to changes in temperature appears to be weak. The rate of temperature increase often decreases when the amount of atmospheric CO₂ increases (Fig. 2, 3, 5 and 7). There are even examples where the DIFF12 temperature change becomes
negative, corresponding to decreasing temperatures, even though the contemporary DIFF12 CO₂ change is positive.

In general, we find that changes in atmospheric CO₂ are lagging behind changes in any of the different temperature records considered. The typical lag is 9.5-12 months for surface temperatures and about 9 months for lower troposphere temperatures, suggesting a temperature sequence of events from the surface to the lower troposphere.

As cause always must precede effect, this observation demonstrates that modern changes in temperatures are generally not induced by changes in atmospheric CO₂. Indeed, the sequence of events is seen to be the opposite: temperature changes are taking place before the corresponding CO₂ changes occur.

As the theoretical initial temperature effect of changes in atmospheric CO₂ must materialize first in the troposphere, and then subsequently at the planet surface (land and ocean), our diagrams 2-8 reveal that the common notion of globally dominant temperature controls exercised by atmospheric CO₂ is in need of reassessment. Empirical observations indicate that changes in temperature generally are driving changes in atmospheric CO₂, and not the other way around.

Numerical global climate models generally assume atmospheric CO₂ in combination with alleged feed-back effects on atmospheric humidity and cloud cover to have a clear net warming effect, and that changes in atmospheric CO₂ therefore represent a main driver for global temperature changes. For that reason changes in temperature should therefore be lagging behind corresponding changes in CO₂. Figures 4, 6 and 8 however show correlation between changes in temperature and CO₂ to be negative for negative offsets (temperature lagging CO₂), indicating that changes towards higher concentrations of atmospheric CO₂ then empirically would associate with less rapid temperature increase or even a temperature decrease. However, as this would invalidate the basic assumption of CO₂ having a clear net warming effect, the perception of temperature lagging behind CO₂ must therefore be rejected. A visual inspection of the data displayed in figures 2, 3, 5 and 7 also show the notion of temperature lagging CO₂ to be implausible.

Thus, the simplest explanation of observed changes in DIFF12 for atmospheric CO₂ is therefore that they are induced by changes in temperature, as illustrated by figures 2-8. Consequently, a substantial part of the atmospheric increase of CO₂ since January 1980 can be explained by associated changes in temperature, and presumably especially changes in ocean temperatures (Toggweiler 1999; Monnin et al. 2001; Goldberg 2008), as this is where we find both the strongest correlation to changes in CO₂ (Fig. 4, 6 and 8), and the longest time lag.

The maximum positive correlation between ocean temperatures and atmospheric CO₂ is within a range from 0.45 to 0.48, depending on which dataset is considered (HadSST2, NCDC or UAH). With a sample size of 361 (number of monthly DIFF12 values) these correlation coefficients are highly significant at the 0.05 level, and correspond to a goodness-of-fit ($r^2$) ranging from 0.20 to 0.23. This represents a fair degree of explanation, and far bigger than achieved by any other factor considered in the present analysis, but it also suggests that there are other factors
beyond ocean surface temperature which have influenced observed changes in atmospheric CO₂ since January 1980. Examples of such potential factors are changes in soil moisture, living biomass, volcanic eruptions, geological weathering processes, burning of fossil fuels, etc. The correlation between CO₂ released by anthropogenic sources and changes in atmospheric CO₂ is not stable (Fig. 15), and not able to explain much of the observed increase in atmospheric CO₂ since January 1980. A qualitatively identical conclusion may possibly be suggested for the effect of volcanic eruptions during the study period, but the character of the volcanic data available does not make it possible to carry out a comparable statistical analysis on this. Actually, on the time scale investigated, the net effect of a major volcanic eruption appears to be a reduction of the prevailing increase rate of atmospheric CO₂, probably an effect of ocean cooling induced from cloud effects.

Summing up, monthly data since January 1980 on atmospheric CO₂ and sea and air temperatures unambiguously demonstrate the overall global temperature change sequence of events to be 1) ocean surface, 2) surface air, 3) lower troposphere, and with changes in atmospheric CO₂ always lagging behind changes in any of these different temperature records.

A main control on atmospheric CO₂ appears to be the ocean surface temperature, and it remains a possibility that a significant part of the overall increase of atmospheric CO₂ since at least 1958 (start of Mauna Loa observations) simply reflects the gradual warming of the oceans, as a result of the prolonged period of high solar activity since 1920 (Solanki et al. 2004). Based on the GISP2 ice core proxy record from Greenland it has been pointed out that the present period of warming since 1850 to a high degree may be explained by a natural c. 1100 yr periodic temperature variation (Humlum et al. 2011).

The atmospheric CO₂ growth rate DIFF12 is seen to be at maximum about the same time when SST DIFF12 is on the falling limb from a previous peak, indicating ocean temperature to be approaching a temperature maximum (see, e.g. Fig. 2). As the rate of CO₂ net outgassing from the ocean then is affected by reduced solubility, this offers a simple physical explanation of the observed time lag. However, more work is needed to investigate the details of this. It is clear, however, that emission of CO₂ from the oceans and other natural sources plays a major role in observed changes of atmospheric CO₂. Independent of human emissions, this contribution to atmospheric CO₂ is not controllable and little predictable, although the coupling to ocean surface temperature demonstrated above may provide a starting basis for future prediction attempts.

An analysis of a Pole-to-Pole transect of atmospheric CO₂ records suggest that changes in atmospheric CO₂ are initiated slightly south of the Equator, and from there spreads towards the two Poles within a year or so (Fig. 13). This observation specifically points towards the oceans at or south of the Equator as an important source area for observed changes in atmospheric CO₂. The major release of anthropogenic CO₂ is taking place at mid-latitudes in the northern hemisphere (Fig. 12), but the north-south transect investigated show no indication of the main change signal in atmospheric CO₂ originating here.
In general, the level of atmospheric CO$_2$ is slightly higher in the northern hemisphere than in the southern hemisphere. This might be seen as due to anthropogenic CO$_2$ mainly being released in the northern hemisphere (Fig.12). However, had this being the explanation, the northern mid latitudes should had been the origin for changes in atmospheric CO$_2$, which is shown not to be the case by figure 13. Thus, presumably other processes are responsible for the north-south gradient in atmospheric CO$_2$, such as, e.g., respiration by land-based organisms (especially northern hemisphere) and ocean uptake (especially southern hemisphere).

The modern relation between temperature and CO$_2$ is qualitatively identical to that demonstrated by ice cores for the Quaternary glacial-interglacial transitions (Mudelsee 2001; Caillon et al. 2003), although the modern time lag between temperature and CO$_2$ is considerably shorter. However, this is presumably reflecting the much coarser time resolution provided by ice cores, displaying only changes on a multi-decadal scale. This is partly due to sampling resolution, partly due to gas diffusion within the ice that averages out any surface temperature variability shorter than a few decades (Severinghaus et al. 1998).

12 Conclusions

There exist a clear phase relationship between changes of atmospheric CO$_2$ and the different global temperature records, whether representing sea surface temperature, surface air temperature, or lower troposphere temperature, with changes in the amount of atmospheric CO$_2$ always lagging behind corresponding changes in temperature.

1. The overall global temperature change sequence of events appears to be from 1) the ocean surface to 2) the land surface to 3) the lower troposphere.
2. Changes in global atmospheric CO$_2$ are lagging about 11-12 months behind changes in global sea surface temperature.
3. Changes in global atmospheric CO$_2$ are lagging 9.5-10 months behind changes in global air surface temperature.
4. Changes in global atmospheric CO$_2$ are lagging about 9 months behind changes in global lower troposphere temperature.
5. Changes in ocean temperatures appear to explain a substantial part of the observed changes in atmospheric CO$_2$ since January 1980.
6. CO$_2$ released from anthropogenic sources apparently have little influence on the observed changes in atmospheric CO$_2$, and changes in atmospheric CO$_2$ are not tracking changes in human emissions.
7. On the time scale investigated, the overriding effect of large volcanic eruptions appears to be a reduction of atmospheric CO$_2$, presumably due to the dominance of associated cooling effects from clouds associated with volcanic gases/aerosols and volcanic debris.

8. Since at least 1980 changes in global temperature, and presumably especially southern ocean temperature, appears to represent a major control on changes in atmospheric CO$_2$.

Acknowledgements

The present study could not have been carried out without the existence of open-access international databases on temperature and atmospheric CO$_2$. We thank one unknown referee for pointing out the variable correlation between changes in the release of CO$_2$ from anthropogene sources and changes in atmospheric CO$_2$ on a decadal time scale. In addition, we are grateful for informal discussions in Oslo with Drs. O.H. Ellestad, O. Engvold, and P. Brekke. Also Drs. T.V. Segalstad and H. Yndestad has for long time been important sources of information and inspiration.
References


Shackleton, N.J. 2000. The 100,000 year ice-age cycle identified and found to lag temperature, carbon dioxide and orbital eccentricity. *Science* 289, 1897-1902.

Shakun, J.D., Clark, P.U., He, F., Marcott, S.A., Mix, A.C., Liu, Z., Otto-Bliesner, B., Schmittner, A. and Bard, E. 2012. Global warming preceded by increasing carbon dioxide concentrations during the last deglaciation. *Nature* 484, 49-55, doi:10.1038/nature10915.


Web references

CO₂ globally averaged marine surface data (accessed March 11, 2012):
ftp://ftp.cmdl.noaa.gov/ccg/co2/trends/co2_mm_gl.txt

HadCRUT3 monthly global surface air temperature (accessed March 11, 2012):

HadSST2 monthly global sea surface temperature (accessed March 11, 2012):
http://www.cru.uea.ac.uk/cru/data/temperature/hadsst2gl.txt

UAH monthly global lower troposphere temperature (accessed March 11, 2012):
http://vortex.nsstc.uah.edu/data/msu/t2lt/uahncdc.lt

Goddard Institute for Space Studies, GISS (accessed March 11, 2012):
http://data.giss.nasa.gov/gistemp/tabledata_v3/GLB.Ts+dSST.txt

National Climatic Data Center, NCDC (accessed March 11, 2012):

Individual station atmospheric CO₂ records from Earth System Research Laboratory, NOAA

Carbon Dioxide Information Analysis Center, CDIAC (accessed March 11, 2012):

Global Volcanism Program, GWP (accessed March 11, 2012):
http://www.volcano.si.edu/world/largeeruptions.cfm
Table 1. Volcanic Explosivity Index (VEI) eruption size classes based on Newhall and Self (1962) and Simkin and Siebert (1994).

<table>
<thead>
<tr>
<th>VEI class</th>
<th>0</th>
<th>1</th>
<th>2</th>
<th>3</th>
<th>4</th>
<th>5</th>
<th>6</th>
<th>7</th>
<th>8</th>
</tr>
</thead>
<tbody>
<tr>
<td>Description</td>
<td>Non-</td>
<td>Small</td>
<td>Moderate</td>
<td>Moderate-</td>
<td>Large</td>
<td>Very</td>
<td>Very</td>
<td>Very</td>
<td>Very</td>
</tr>
<tr>
<td></td>
<td>explosive</td>
<td></td>
<td></td>
<td>large</td>
<td></td>
<td>large</td>
<td>large</td>
<td>large</td>
<td>large</td>
</tr>
<tr>
<td>Tephra (m$^3$)</td>
<td>$&lt;1 \times 10^4$</td>
<td>$&lt;1 \times 10^6$</td>
<td>$&lt;1 \times 10^7$</td>
<td>$&lt;1 \times 10^8$</td>
<td>$&lt;1 \times 10^9$</td>
<td>$&lt;1 \times 10^{10}$</td>
<td>$&lt;1 \times 10^{11}$</td>
<td>$&lt;1 \times 10^{12}$</td>
<td>$&gt;1 \times 10^{12}$</td>
</tr>
<tr>
<td>Cloud</td>
<td>$&lt;0.1$</td>
<td>0.1-1</td>
<td>1-5</td>
<td>3-15</td>
<td>10-25</td>
<td>$&gt;25$</td>
<td>$&gt;25$</td>
<td>$&gt;25$</td>
<td>$&gt;25$</td>
</tr>
<tr>
<td>height (km)</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
</tbody>
</table>
Figure captions

Figure 1. Monthly global atmospheric CO$_2$ (NOAA; green), monthly global sea surface temperature (HadSST2; blue stippled) and monthly global surface air temperature (HadCRUT3; red), since January 1980. Last month shown is December 2011.

Figure 2. 12-month change of global atmospheric CO$_2$ concentration (NOAA; green), global sea surface temperature (HadSST2; blue) and global surface air temperature (HadCRUT3; red dotted). The upper panel shows unfiltered monthly values (e.g. January 2000 minus January 1999), while the lower panel shows filtered values (DIFF12, the difference between the average of the last 12 month and the average for the previous 12 months for each data series).

Figure 3. 12-month change of global atmospheric CO$_2$ concentration (NOAA; green), global sea surface temperature (HadSST2; blue) and global surface air temperature (GISS; red dotted). All graphs are showing monthly values of DIFF12, the difference between the average of the last 12 month and the average for the previous 12 months for each data series.

Figure 4. Correlation coefficients between DIFF12 monthly surface air temperatures (a: GISS; b: HadCRUT3), sea surface temperature (c: HadSST2) and global atmospheric CO$_2$, for different monthly lags of CO$_2$. The maximum positive correlation is found for CO$_2$ lagging 9.5 months behind GISS, 10 months behind HadCRUT3 and 11 months behind HadSST2. Numbers in parentheses show the maximum positive correlation coefficient and the associated time lag of CO$_2$ in months. The grey vertical arrows indicate no lag.

Figure 5. 12-month change of global atmospheric CO$_2$ concentration (NOAA; green), change in global surface air temperature (NCDC; blue), land surface air temperature (NCDC; yellow) and ocean surface air temperature (NCDC; red dotted). All graphs are showing monthly values of DIFF12, the difference between the average of the last 12 month and the average for the previous 12 months for each data series.

Figure 6. Correlation coefficients between DIFF12 monthly NCDC surface air temperatures (a: land; b: ocean; c: global) and global atmospheric CO$_2$, for different monthly lags of CO$_2$. The maximum positive correlation is found for CO$_2$ lagging 9.5 months behind land surface air temperature, 12 months behind ocean surface air temperature and 10 months behind the global surface air temperature. Numbers in parentheses show the maximum positive correlation coefficient and the associated time lag of CO$_2$ in months. The grey vertical arrows indicate no lag.

Figure 7. 12-month change of global atmospheric CO$_2$ concentration (NOAA; green), change in lower troposphere temperature over oceans in the Northern Hemisphere (UAH; blue), in the Southern Hemisphere (UAH; yellow) and in the Tropics (UAH; red dotted). All graphs are
showing monthly values of DIFF12, the difference between the average of the last 12 month and the average for the previous 12 months for each data series.

Figure 8. Correlation coefficients between DIFF12 monthly lower troposphere temperatures (UAH) and global atmospheric CO$_2$, for different monthly lags of CO$_2$ in relation to temperature. The maximum positive correlation is found for CO$_2$ lagging 9 months after lower global troposphere temperatures, 10 months after lower troposphere temperatures over oceans, and 8 months after lower troposphere temperatures over land. Numbers in parentheses show the maximum positive correlation coefficient and the associated time lag of CO$_2$ in months. The grey vertical arrows indicate no lag.

Figure 9. 12-month change of global atmospheric CO$_2$ concentration (NOAA; green), change in northern hemisphere surface air temperature (NCDC; red dotted), and southern hemisphere air temperature (NCDC; blue). All graphs are showing monthly values of DIFF12, the difference between the average of the last 12 month and the average for the previous 12 months for each data series.

Figure 10. Correlation coefficients between DIFF12 monthly surface air temperatures in the northern (a) and the southern hemisphere (b) and global atmospheric CO$_2$, for different monthly lags of CO$_2$. The maximum positive correlation is found for CO$_2$ lagging 9.5 months behind northern hemisphere temperature, and 11 months behind southern hemisphere temperature. Numbers in parentheses show the maximum positive correlation coefficient and the associated time lag of CO$_2$ in months. The grey vertical arrows indicate no lag.

Figure 11. 12-month change of global atmospheric CO$_2$ concentration (NOAA; green), and the change in release of CO$_2$ from burning of fossil fuels (CDIAC; red). Both graphs are showing monthly values of DIFF12, the difference between the average of the last 12 month and the average for the previous 12 months for each data series.

Figure 12. Latitudinal 1990 summary of CO$_2$ emissions from anthropogenic sources. Data source: Carbon Dioxide Information Analysis Center (CDIAC).

Figure 13. 12-month change of global atmospheric CO$_2$ concentration for Alert (NOAA; blue), Mauna Loa (NOAA; green), Ascension Island (NOAA; yellow), and the south Pole (NOAA; purple), and the change in release of anthropogenic CO$_2$ (CDIAC; red). All graphs show monthly values of DIFF12, the difference between the average of the last 12 month and the average for the previous 12 months for each data series.

Figure 14. Correlation coefficients between DIFF12 change in release of anthropogenic CO$_2$ (CDIAC) and DIFF12 change in atmospheric CO$_2$ at different stations, for different lags of CO$_2$ in
relation to CO₂ from fossil fuel. The maximum negative correlation is found for changes in atmospheric CO₂ ranges from 4 to 19 months after changes in CO₂ release from fossil fuels. Numbers in parentheses show the maximum absolute correlation coefficient and the associated time lag of CO₂ in months. The grey vertical arrows indicate no lag (the time of release).

Figure 15. Correlation coefficients between DIFF12 change in release of anthropogenic CO₂ (CDIAC) and global atmospheric CO₂, for different lags of atmospheric CO₂ in relation to CO₂ from fossil fuel. Numbers in parentheses show the maximum positive correlation coefficient and the associated time lag of CO₂ in months. The grey vertical arrows indicate no lag (the time of release).

Figure 16. 12-month change of global atmospheric CO₂ concentration (NOAA; green), and the timing of major (VEI>3) volcanic eruptions (GWP). The CO₂ graph is showing monthly values of DIFF12, the difference between the average of the last 12 month and the average for the previous 12 months for each data series. The DIFF12 CO₂ values have been shifted 12 month forward to correspond correctly with the time axis used for plotting the volcanic eruptions.

Figure 17. Fourier frequency spectra (Best Exact N composite algorithm; SeaSolve 2003) for DIFF12’s for changes in a) atmospheric CO₂, b) HadCRUT3, c) HadSST2 and d) CO₂ released by use of fossil fuels. The DIFF12 atmospheric CO₂ record is dominated by highly significant periods of about 3.8 and 2.5 yr length, while especially a period of about 3.7 yr is dominant in both DIFF12 temperature records. The DIFF12 record for CO₂ from fossil fuels is characterised by a different type of spectrum, and with peaks of generally less significance than characterising the other records. The grey tone indicates increasing amplitude within each record. The stippled lines indicate peak-based critical limit significance levels.
Figures

Figure 1. Monthly global atmospheric CO\(_2\) (NOAA; green), monthly global sea surface temperature (HadSST2; blue stippled) and monthly global surface air temperature (HadCRUT3; red), since January 1980. Last month shown is December 2011.

Figure 2. 12-month change of global atmospheric CO\(_2\) concentration (NOAA; green), global sea surface temperature (HadSST2; blue) and global surface air temperature (HadCRUT3; red dotted). The upper panel shows unfiltered monthly values (e.g. January 2000 minus January 1999), while the lower panel shows filtered values (DIFF12, the difference between the average of the last 12 month and the average for the previous 12 months for each data series).
Figure 3. 12-month change of global atmospheric CO$_2$ concentration (NOAA; green), global sea surface temperature (HadSST2; blue) and global surface air temperature (GISS; red dotted). All graphs are showing monthly values of DIFF12, the difference between the average of the last 12 month and the average for the previous 12 months for each data series.

Figure 4. Correlation coefficients between DIFF12 monthly surface air temperatures (a: GISS; b: HadCRUT3), sea surface temperature (c: HadSST2) and global atmospheric CO$_2$, for different monthly lags of CO$_2$. The maximum positive correlation is found for CO$_2$ lagging 9.5 months behind GISS, 10 months behind HadCRUT3 and 11 months behind HadSST2. Numbers in parentheses show the maximum positive correlation coefficient and the associated time lag of CO$_2$ in months. The grey vertical arrows indicate no lag.
Figure 5. 12-month change of global atmospheric CO$_2$ concentration (NOAA; green), change in global surface air temperature (NCDC; blue), land surface air temperature (NCDC; yellow) and ocean surface air temperature (NCDC; red dotted). All graphs are showing monthly values of DIFF12, the difference between the average of the last 12 month and the average for the previous 12 months for each data series.

Figure 6. Correlation coefficients between DIFF12 monthly NCDC surface air temperatures (a: land; b: ocean; c: global) and global atmospheric CO$_2$, for different monthly lags of CO$_2$. The maximum positive correlation is found for CO$_2$ lagging 9.5 months behind land surface air temperature, 12 months behind ocean surface air temperature and 10 months behind the global surface air temperature. Numbers in parentheses show the maximum positive correlation coefficient and the associated time lag of CO$_2$ in months. The grey vertical arrows indicate no lag.
Figure 7. 12-month change of global atmospheric CO$_2$ concentration (NOAA; green), change in lower troposphere temperature over oceans in the Northern Hemisphere (UAH; blue), in the Southern Hemisphere (UAH; yellow) and in the Tropics (UAH; red dotted). All graphs are showing monthly values of DIFF12, the difference between the average of the last 12 month and the average for the previous 12 months for each data series.

Figure 8. Correlation coefficients between DIFF12 monthly lower troposphere temperatures (UAH) and global atmospheric CO$_2$, for different monthly lags of CO$_2$ in relation to temperature. The maximum positive correlation is found for CO$_2$ lagging 9 months after lower global troposphere temperatures, 10 months after lower troposphere temperatures over oceans, and 8 months after lower troposphere temperatures over land. Numbers in parentheses show the maximum positive correlation coefficient and the associated time lag of CO$_2$ in months. The grey vertical arrows indicate no lag.
Figure 9. 12-month change of global atmospheric CO$_2$ concentration (NOAA; green), change in northern hemisphere surface air temperature (NCDC; red dotted), and southern hemisphere air temperature (NCDC; blue). All graphs are showing monthly values of DIFF12, the difference between the average of the last 12 month and the average for the previous 12 months for each data series.

Figure 10. Correlation coefficients between DIFF12 monthly surface air temperatures in the northern (a) and the southern hemisphere (b) and global atmospheric CO$_2$, for different monthly lags of CO$_2$. The maximum positive correlation is found for CO$_2$ lagging 9.5 months behind northern hemisphere temperature, and 11 months behind southern hemisphere temperature. Numbers in parentheses show the maximum positive correlation coefficient and the associated time lag of CO$_2$ in months. The grey vertical arrows indicate no lag.
Figure 11. 12-month change of global atmospheric CO\textsubscript{2} concentration (NOAA; green), and the change in release of CO\textsubscript{2} from burning of fossil fuels (CDIAC; red). Both graphs are showing monthly values of DIFF12, the difference between the average of the last 12 month and the average for the previous 12 months for each data series.

Figure 12. Latitudinal 1990 distribution of CO\textsubscript{2} emissions from anthropogene sources. Data source: Carbon Dioxide Information Analysis Center (CDIAC).
Figure 13. 12-month change of global atmospheric CO$_2$ concentration for Alert (NOAA; blue), Mauna Loa (NOAA; green), Ascension Island (NOAA; yellow), and the south Pole (NOAA; purple), and the change in release of anthropogenic CO$_2$ (CDIAC; red). All graphs show monthly values of DIFF12, the difference between the average of the last 12 months and the average for the previous 12 months for each data series.

Figure 14. Correlation coefficients between DIFF12 change in release of CO$_2$ from burning of fossil fuels (CDIAC) and DIFF12 change in atmospheric CO$_2$ at different stations, for different lags of CO$_2$ in relation to anthropogenic CO$_2$. The maximum negative correlation is found for changes in atmospheric CO$_2$ ranges from 4 to 19 months after changes in CO$_2$ release from fossil fuels, depending on latitude. Numbers in parentheses show the maximum absolute correlation coefficient and the associated time lag of CO$_2$ in months. The grey vertical arrows indicate no lag (the time of release).
Figure 15. Correlation coefficients between DIFF12 change in release of CO₂ from burning of fossil fuels (CDIAC) and global atmospheric CO₂, for different lags of atmospheric CO₂ in relation to anthropogene CO₂. Numbers in parentheses show the maximum positive correlation coefficient and the associated time lag of CO₂ in months. The grey vertical arrows indicate no lag (the time of release).

Figure 16. 12-month change of global atmospheric CO₂ concentration (NOAA; green), and the timing of major (VEI>3) volcanic eruptions (GWP). The CO₂ graph is showing monthly values of DIFF12, the difference between the average of the last 12 month and the average for the previous 12 months for each data series. The DIFF12 CO₂ values have been shifted 12 month forward to correspond correctly with the time axis used for plotting the volcanic eruptions.
Figure 17. Fourier frequency spectra (Best Exact N composite algorithm; SeaSolve 2003) for DIFF12’s for changes in a) atmospheric CO$_2$, b) HadCRUT3, c) HadSST2 and d) CO$_2$ released by use of fossil fuels. The DIFF12 atmospheric CO$_2$ record is dominated by highly significant periods of about 3.8 and 2.5 yr length, while especially a period of about 3.7 yr is dominant in both DIFF12 temperature records. The DIFF12 record for CO$_2$ from fossil fuels is characterised by a different type of spectrum, and with peaks of generally less significance than characterising the other records. The grey tone indicates increasing amplitude within each record. The stippled lines indicate peak-based critical limit significance levels.