



The ENSO signal in  
atmospheric  
composition fields

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# The ENSO signal in atmospheric composition fields: emission driven vs. dynamically induced changes

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## Abstract

The El Niño Southern Oscillation (ENSO) does not only affect meteorological fields but also has a large impact on atmospheric composition. Atmospheric composition fields from the Monitoring Atmospheric Composition and Climate (MACC) reanalysis are used to identify the ENSO signal in tropospheric ozone, carbon monoxide, nitrogen oxide and smoke aerosols, concentrating on the months October to December. During El Niño years all these fields have increased concentrations over maritime South East Asia in October. The MACC Composition Integrated Forecasting System (C-IFS) model is used to quantify the relative magnitude of dynamically induced and emission driven changes in the atmospheric composition fields. While changes in tropospheric ozone are a combination of dynamically induced and emission driven changes, the changes in carbon monoxide, nitrogen oxide and smoke aerosols are almost entirely emission driven in the MACC model. The ozone changes continue into December, i.e. after the end of the Indonesian fire season while changes in the other fields are confined to the fire season.

## 1 Introduction

The El Niño Southern Oscillation (ENSO) is the dominant mode of variability in the Tropics (e.g. Allan et al., 1996). It does not only affect meteorological fields but has a large impact on atmospheric composition too, for example on ozone ( $O_3$ ), carbon monoxide (CO), nitrogen oxides ( $NO_x$ ) and aerosols (e.g. Logan et al., 2008; Ziemke and Chandra, 2003; Chandra et al., 2002; Wang et al., 2004). As the result of an eastwards shift of the warm sea surface temperatures (SST) and the large scale Walker circulation anomaly in the tropical Pacific during El Niño years, downward motion is increased and convection and precipitation are reduced over the western Pacific and the maritime continent. During La Niña conditions the opposite dynamical effects occur. Fire emissions over Indonesia show a large interannual variability (IAV), with largest

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emissions during El Niño years (e.g. van der Werf et al., 2006; Kaiser et al., 2012), when drought conditions and anthropogenic biomass burning lead to big wild fires (Duncan et al., 2003; Lyon et al., 2004; Page et al., 2002) that emit large amounts of trace gases and aerosols. During El Niño years tropospheric  $O_3$  columns ( $TCO_3$ ) are decreased over the Central and eastern Pacific and increased over the western Pacific and Indonesia, while CO concentrations and aerosols from biomass burning increase over Indonesia. Specific humidity changes in the upper troposphere are anti-correlated with the changes in  $TCO_3$  (e.g. Chandra et al., 2007).

These atmospheric composition changes have been found in observations (Chandra et al., 1998, 2007; Ziemke and Chandra, 1999; Fujiwara et al., 1999; Logan et al., 2008) and were confirmed by modelling studies (Hauglustaine et al., 1999; Sudo and Takahashi, 2001; Chandra et al., 2002, 2009; Doherty et al., 2006; Nassar et al., 2009) which also tried to quantify the relative importance of the dynamically induced and the emission driven atmospheric composition changes. The reasons for the  $TCO_3$  increase over the western Pacific and Indonesia during El Niño years are (i) changes in the vertical transport that lead to enhanced downward transport of  $O_3$  rich air from the upper troposphere (and perhaps stratosphere) to the middle and lower troposphere, and reduced transport of  $O_3$  poor air from the lower troposphere into the upper troposphere, (ii) a longer chemical lifetime of  $O_3$  because of reduced humidity which affects the concentrations of the hydroxyl radical (OH) and hence the photochemical loss of tropospheric  $O_3$ ; and (iii) enhanced photochemical production of  $O_3$  in the lower troposphere because of increased concentrations of  $O_3$  precursors from biomass burning, such as  $NO_x$ , CO or Hydrocarbons. We refer to (i) and (ii) as “dynamically induced changes” and to (iii) as “emission driven changes” throughout this paper.

For El Niño events with large fires over Indonesia, such as in 1997 and 2006, the  $TCO_3$  changes due to dynamics and due to increased emissions can be of similar magnitude (Sudo and Takahashi, 2001; Chandra et al., 2002, 2009), while for weaker events, such as the 2004 El Niño, the dynamical impact dominates (Chandra et al., 2007).

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The changes in CO are mainly emissions driven (Logan et al., 2008; Chandra et al., 2009; Voulgarakis et al., 2010) and of smaller horizontal scale than the O<sub>3</sub> anomalies, but dynamical interactions due to changes in water vapour (H<sub>2</sub>O) and hence OH can also play a role. CO is increased over the western Pacific and the maritime continent during El Niño because of increased emissions from fires and the increased chemical lifetime due to reduced OH. The CO anomaly over Indonesia is usually gone by December (Logan et al., 2008; Chandra et al., 2009), after the end of the biomass burning season, while the O<sub>3</sub> anomaly continues.

Large Indonesian wildfires can affect the air quality over South East Asia. Aouizerats et al. (2015) investigated how the transport of biomass burning emissions from Sumatra affected the air quality in Singapore. They found that 21 % of the PM<sub>10</sub> loading in Singapore during July to October 2006 was due to Sumatran fires, and that Sumatran fires were responsible for about half of the days with PM<sub>10</sub> concentrations greater than 50 µg m<sup>-3</sup> while the other half was due to local anthropogenic pollution and contributions from smaller fires. The impact of fire emissions on atmospheric aerosol concentrations is of particular interest because of the potential feedbacks of fire-induced aerosols on climate. Several studies have looked at the correlations between ENSO and aerosols or atmospheric haze produced by the Indonesian fires. Wang et al. (2004) used visibility data over Sumatra as an indicator of biomass burning and found that haze events were strongly correlated with El Niño during the 1973 to 2003 period. Tosca et al. (2010) used satellite data and modelling studies and found that the aerosol optical depth (AOD) over Indonesia had a large IAV that was driven by wild fires during periods of El Niño induced droughts. Their modelling study showed that the fire-emitted aerosols could initiate a positive feedback loop. The aerosols acted to intensify drought over the biomass burning regions. The aerosols also reduced land and sea surface temperatures, and hence suppressed convection and precipitation in the area. Podgorny et al. (2003) looked at the feedback between El Niño and the Indonesian biomass burning of 1997 and also found that the haze from the fires reduced the solar radiation absorbed by the equatorial Indian Ocean and increased the solar heating of

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the atmosphere, thus raising the possibility of dynamical feedbacks of the smoke forcing on ENSO. Chung and Ramanathan (2003) carried out modelling studies to assess the remote impact of changes in the South Asian haze and found that fluctuations in the absorbing aerosol forcing could affect the interannual climate variability in the Tropics (and Extratropics). It could remotely suppress convection in the equatorial western Pacific and lead to an ocean–atmosphere response that was very similar to El Niño like warming.

As part on the EU FP7 funded Monitoring Atmospheric Composition and Climate (MACC) project ([www.copernicus-atmosphere.eu](http://www.copernicus-atmosphere.eu)) a 10 year reanalysis of atmospheric composition (Inness et al., 2013) was constructed. This reanalysis provides fields of chemically active gases, for example CO, O<sub>3</sub>, and NO<sub>x</sub>, as well as aerosols globally for both the troposphere and the stratosphere for the years 2003 to 2012. It gives us the unprecedented possibility to assess the impact of ENSO on atmospheric composition using an observationally constrained, continuous, 3 dimensional atmospheric composition dataset with a resolution of about 80 km, which is greater than the resolutions used in most previous modelling studies. In this paper we show that the MACC reanalysis shows the ENSO induced anomalies in O<sub>3</sub>, CO, NO<sub>x</sub> and aerosols described in earlier studies. We then use MACC's Composition Integrated Forecasting System (C-IFS) model (Flemming et al., 2015) to quantify the relative impact of the dynamics and the biomass burning emissions on the ENSO signal in the O<sub>3</sub>, CO, NO<sub>x</sub> and smoke aerosol fields.

This paper is structured in the following way. Section 2 describes the MACC reanalysis and the ENSO signal seen in MACC O<sub>3</sub>, CO, NO<sub>x</sub> and smoke aerosol fields. Section 3 describes the additional C-IFS model runs that were carried out to quantify the relative impact of the dynamics and biomass burning emission on the ENSO signal in the atmospheric composition fields and their results, and Sect. 4 presents conclusions and outlook.

## 2 ENSO signal in the MACC reanalysis

### 2.1 The MACC reanalysis

The MACC data assimilation system provides analyses and forecasts of atmospheric composition and was used to produce a reanalysis of atmospheric composition covering the years 2003 to 2012, as described in Inness et al. (2013). O<sub>3</sub> retrievals from several instruments (including the Ozone Monitoring Instrument (OMI), SCanning Imaging Absorption spectroMETER for Atmospheric CHartographyY (SCIAMACHY), Solar Backscatter Ultra-Violet (SBUV/2), Microwave Limb Sounder (MLS)), CO retrievals from Measurements of Pollution in the Troposphere (MOPITT) and the Infrared Atmospheric Sounding Interferometer (IASI), tropospheric NO<sub>2</sub> columns from SCIAMACHY, and AOD from the Moderate Resolution Imaging Spectroradiometer (MODIS) were assimilated to constrain the atmospheric compositions fields. For more information about the assimilated datasets and the quality of the O<sub>3</sub>, CO, and NO<sub>x</sub> fields produced by the analysis see Inness et al. (2013). The aerosol analysis in the MACC reanalysis is similar to that described in Benedetti et al. (2009) and Morcrette et al. (2011) and is based on 4 dimensional variational assimilation of AOD observations at 550 nm from the MODIS sensors, including a global adaptive bias correction. Comparisons of multi-year averages of AOD over the period 2003–2010 from the MACC reanalysis and from the Multi-angle Imaging SpectroRadiometer (MISR) sensors onboard the Terra satellite indicate good qualitative agreement (not shown).

The anthropogenic emissions for the reactive gases for the MACC reanalysis were taken from the MACCcity inventory (Granier et al., 2011) which accounts for projected trends in the emissions. For the aerosol fields they came from the EDGAR database (Dentener et al., 2006). Monthly biomass burning emission for the years 2003 to 2008 from the GFED3.0 inventory (van der Werf et al., 2010) were scaled to daily resolution using MODIS active fire observations. From 2009 to 2012 daily biomass burning emissions from MACC's GFAS, Version 1.0 (Kaiser et al., 2012) were used. One advantage of the MACC reanalysis is that it used daily fire emission, in contrast to several other

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studies that used monthly averages. Biogenic emissions used in the MACC reanalysis were for 2003. They came from a recent update (Barkley, 2010) of the Model of Emissions of Gases and Aerosols from Nature version 2 (MEGAN2; Guenther et al., 2006, <http://acd.ucar.edu/~guenther/MEGAN/MEGAN.htm>) and were used as monthly surface flux fields without interannual variation.

Initial validation results from the MACC reanalysis are shown in Inness et al. (2013) and Morcrette et al. (2011) and more detailed validation can be found in the MACC reanalysis validation reports available from [http://atmosphere.copernicus.eu/services/aqac/global\\_verification/validation\\_reports/](http://atmosphere.copernicus.eu/services/aqac/global_verification/validation_reports/).

## 2.2 ENSO anomalies

The MACC reanalysis was used to construct monthly composites of  $O_3$ ,  $CO$ ,  $NO_x$  fields at 500 hPa and of the smoke AOD, i.e. the sum of black carbon (BC) and organic matter (OM) AOD, at 550 nm for El Niño and La Niña years for the months October, November and December. The El Niño composite was constructed from the years 2004, 2006, 2009, the La Niña composite from the years 2005, 2007, 2008, 2010, and 2011. Weak El Niño and La Niña years were included in the composite calculation to increase the sample size. A recent timeseries of the Multivariate ENSO index which was used to define the years used in our El Niño and La Niña composites can be found on <http://www.esrl.noaa.gov/psd/enso/mei/index.html>. SST and precipitation composite fields, that were not available from the MACC reanalysis, were constructed for the same years from the ERA Interim reanalysis (Dee et al., 2011), and biomass burning composites were calculated from the GFAS v1.0 data set. The composites were then used to calculate anomalies for the various fields by taking the difference between the El Niño and La Niña composites for the months October, November and December.

Figure 1 shows the warm SST anomaly over the central Pacific associated with El Niño conditions and the resulting precipitation changes for October, November and December from ERA Interim. Precipitation is increased over the central Pacific and reduced over the western Pacific, maritime continent, northern Australia and part of the

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Indian Ocean. The increased biomass burning emissions related to the lack of rainfall over Indonesia and Northern Australia can be seen in the FRP anomalies shown in Fig. 2. Increased fire activity can be seen over Indonesia in October and November, but has stopped by December after the end of the fire season, while a weaker biomass burning anomaly continues in Northern Australia into December. Over Brazil decreased fire activity can be seen in October.

Figure 3 shows the IAV of the biomass burning emissions for CO from GFAS v1.0 for the 10 years covered by the MACC reanalysis for the area around Indonesia ( $10^{\circ}$  N and  $10^{\circ}$  S,  $90$  and  $130^{\circ}$  E) and illustrates that the emissions are higher during the El Niño years 2004, 2006, 2009 than at other times. The largest values are seen for 2006. Table 1 shows the average CO,  $\text{NO}_x$  and fire aerosol emissions from GFAS v1.0 during October, November, December for the El Niño years and La Niña years. The average CO emissions during El Niño years during October, November, and December are a factor of about 9, 12, and 2 larger, respectively, than during La Niña years. For  $\text{NO}_x$  fire emissions the factors are 6, 7, 2 and for smoke aerosols 8, 10, 2, respectively. The values in Table 1 are slightly smaller than the values in Chandra et al. (2009) who list CO fire emissions over Indonesia as  $44.2$ ,  $6.3$  and  $0.2 \text{ Tg months}^{-1}$  and  $\text{NO}_2$  fire emissions of  $0.76$ ,  $0.11$ , and  $0.0$  for October, November, and December 2006, respectively.

Figures 4 to 7 illustrate the impact that the dynamical and emission related changes have on the atmospheric composition fields, by showing the anomalies calculated from the MACC reanalysis at 500 hPa for  $\text{O}_3$ , CO,  $\text{NO}_x$ , and smoke AOD, respectively.  $\text{O}_3$  shows positive anomalies over the western Pacific, Indonesia, northern Australia and the eastern Indian Ocean, and negative anomalies over the central and eastern Pacific (Fig. 4). These anomalies are quite large scale and continue into December after the end of the biomass burning anomaly over Indonesia. The  $\text{O}_3$  anomalies agree well with those described in other studies based on MLS and Total Ozone Mapping Spectrometer (TOMS) data (Chandra et al., 2009) as well as Tropospheric Emission Spectrometer (TES) data (Logan et al., 2008). The negative  $\text{O}_3$  anomaly over Africa and the western



Indian Ocean in December was also noted by Nassar et al. (2009) in TES data and modelling studies with the Goddard Earth Observing System 3-D chemical transport model (GEOS-Chem), and we also note a negative  $O_3$  anomaly over Brazil in October. The  $O_3$  anomalies are a combination of biomass burning changes and meteorological changes as a consequence of changes in SSTs and the resulting eastward shift of the Walker circulation (Sudo and Takahashi, 2001; Chandra et al., 2002, 2009). The  $O_3$  decrease over the central and east Pacific is due to enhanced upward transport of  $O_3$  poor air from the boundary layer into the middle and upper troposphere, and a shorter  $O_3$  lifetime and larger photochemical loss due to increased  $H_2O$  (and hence OH) concentrations.

During October and November both dynamical and emission driven effects contribute, and modelling studies (e.g. Sudo and Takahashi, 2001; Chandra et al., 2002, 2009) have shown that emissions and dynamical changes can contribute equally for El Niño years with strong biomass burning. The  $O_3$  changes in December are due to the dynamical changes after the end of the fire season over Indonesia. This agrees with what was seen by Logan et al. (2008), Chandra et al. (2009) and Nassar et al. (2009).

Figure 5 shows the CO anomalies at 500 hPa calculated from the MACC reanalysis. These anomalies are more confined to the areas of the biomass burning anomalies (see Fig. 2) than the  $O_3$  anomalies (Fig. 4). The strongest positive anomaly is found over the maritime continent during October and is linked to increased emissions from enhanced biomass burning under drought conditions. A negative CO anomaly extends from South America over the southern Atlantic to Africa in October and is related to lower fire emissions over Brazil. Similar anomalies were described by Logan et al. (2008), Nassar et al. (2009) and Chandra et al. (2009). The CO anomalies during November are weaker than the October ones, and by December the anomalies have all but disappeared. This confirms that the CO anomalies are mainly emission driven and are not affected much by the dynamical changes that cause the  $O_3$  anomalies after the end of the biomass burning season in December.

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The  $\text{NO}_x$  anomalies (Fig. 6) clearly show the impact of the increased emissions from biomass burning over Indonesia during October and November, but also seem to indicate some large scale response. For example, the negative anomaly over the eastern Pacific is colocated with the negative  $\text{O}_3$  anomaly here (Fig. 4) and could indicate enhanced upward transport of  $\text{NO}_x$  poor air to these levels. Also the negative  $\text{NO}_x$  anomaly over the Pacific in December is again co-located with a (larger) negative  $\text{O}_3$  anomaly.

Figure 7 shows the anomaly of smoke AOD at 550 nm calculated from the MACC reanalysis. The largest positive anomaly is found over Indonesia in October and November, corresponding to increased aerosol concentrations from biomass burning emissions. The negative aerosol anomaly over South America in October is related to the reduced fire activity seen in Fig. 2. By December the anomalies have disappeared. Similar AOD anomaly patterns over Indonesia were seen by Tosca et al. (2010) when comparing EL Niño and La Niña years for August to October for the period 2000 and 2006 from the MISR and MODIS data.

The 3 dimensional nature of the MACC reanalysis allows us to look at the vertical distribution of the anomalies in the troposphere. Figures 8 to 11 show height vs. longitude cross-sections of  $\text{O}_3$ , CO,  $\text{NO}_x$  and smoke AOD anomalies averaged over the latitude range from 0 to  $12^\circ$  S. The  $\text{O}_3$  cross section in Fig. 8 illustrates that the largest positive  $\text{O}_3$  anomalies in October and November are located in the lower troposphere and are likely to be the result of enhanced  $\text{O}_3$  production due to increased concentrations of  $\text{O}_3$  precursors from enhanced fire emissions. However, the positive and negative anomalies extend into the upper the troposphere, and some of the anomalies (for example the negative anomaly over the central Pacific) are clearly not connected to the surface but seem to originate in the middle or upper troposphere. These anomalies continue into December after the end of the fire season, and are likely to be a result of the dynamically induced changes mentioned above.

Figure 9 shows that CO anomalies are largest in the lower troposphere but can extend throughout the troposphere over Indonesia and South America. There is a clear

connection to increased CO emissions over Indonesia and decreased emissions over South America due to changes in biomass burning. By December the anomalies have all but gone and show that there is no dynamically induced anomaly, unlike in O<sub>3</sub>. Now a positive anomaly is found over Africa.

Figure 10 shows cross sections of NO<sub>x</sub> anomalies calculated from the reanalysis. The largest anomalies are located in the lower troposphere and are again clearly connected to changes in the fire emissions. A large positive anomaly is found over Indonesia and negative anomalies over South America in October and Africa in October and November. In these areas there are reduced fire activities and increased precipitation during El Niño years. Positive NO<sub>x</sub> anomalies are found in the upper troposphere and could be a result of increased NO<sub>x</sub> production from lightning over South America where there is positive precipitation anomaly pointing to increased convection. The flash rates in the lightning NO parametrization are 5–10 larger over land than over ocean which might explain why no signal is seen over the Central Pacific.

Figure 11 depicts cross sections of smoke AOD and shows that, as for CO and NO<sub>x</sub>, there is a clear connection to increased emissions over Indonesia in October and reduced emissions over South America. By December the positive anomaly over Indonesia is much reduced and confined to the lower troposphere. Enhanced AOD concentrations can be seen in November over South America.

### 3 Quantifying the relative importance of dynamically and emission driven changes on the atmospheric composition fields

#### 3.1 Experiment setup

To quantify the relative impact of increased biomass burning emissions and dynamically induced changes on the atmospheric composition fields during El Niño conditions two experiments are run for the years 2005 and 2006: one with normal and one with climatological GFAS v1.0 fire emissions. 2006 was an El Niño year, and 2005 is used

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to represent normal to weak La Niña conditions. The additional experiments use the most recent version of the MACC system, the C-IFS model (Flemming et al., 2015). This model is different to the one used in the MACC reanalysis (Inness et al., 2013) because it has chemistry routines included directly in ECMWF's Integrated Forecasting System (IFS).

The chemistry scheme implemented in the C-IFS model version used for these experiments is an extended, modified version of the Carbon Bond Mechanism 5 (Yarwood et al., 2005) chemical mechanism as originally implemented in the Tracer Model 5 (TM5) CTM (Huijnen et al., 2010, 2014; Williams et al., 2013). This is a tropospheric chemistry scheme with 54 species and 126 reactions. For O<sub>3</sub> a simple stratospheric parameterisation based on Cariolle and Teyssèdre (2007) has been added. Monthly mean dry deposition velocities are currently based on climatological fields from MOCAGE (Michou et al., 2004). The module for wet deposition is based on the Harvard wet deposition scheme (Jacob et al., 2000; Liu et al., 2001). The output of the IFS convection scheme is used to calculate NO emissions from lightning. They are parameterised using estimates of the flash rate density, the flash energy release and the vertical emission profile. Estimates of the flash rate density are based on parameters of the convection scheme and calculated using convective precipitation as input parameter (Meijer et al., 2001). Documentation of the technical implementation of C-IFS and more details about the model can be found in Flemming et al. (2015). In the present study, the C-IFS aerosol fields are not used in the radiation scheme, where an aerosol climatology based on Tegen et al. (1997) is used instead. Also, heterogeneous chemistry on aerosols is not included.

The anthropogenic emissions used in the C-IFS runs come from the MACCity emission data base (Granier et al., 2011). Biogenic emissions are taken from the POET database for the year 2000 (Granier et al., 2005; Olivier et al., 2003), with isoprene emissions from MEGAN2.1, again for the year 2000 (Guenther et al., 2006). Biomass burning emissions for the runs are either taken from GFAS v1.0 (Kaiser et al., 2012) or from a GFAS v1.0 climatology. This daily climatology was constructed using the

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GFAS v1.0 dataset from 2000 to 2014 (Kaiser et al., 2012; Remy and Kaiser, 2014). Biomass burning emissions for each day of the year were defined as the average of the emissions of the same day of the year for the 15 years of the dataset.

The differences between the GFAS v1.0 and climatological GFAS emissions for the area between 10° N, 10° S, 90° E, 130° E are shown in Fig. 12. The figure illustrates that 2006 was a year with exceptionally large biomass burning emissions over Indonesia during the biomass burning season (as already seen in Fig. 4), while in 2005 emissions were slightly below average.

The experiments are started on 1 January 2005 and run until the end of 2006. The first experiment (BASE) uses daily GFAS v1.0 emissions, while the second experiment (CLIM) uses the climatological GFAS data set described above. We look at fields from these experiments for October and December 2005 and 2006 to determine

- i. the overall impact of changes to the atmospheric composition fields due to El Niño related dynamically and emission induced changes by comparing BASE for the years 2006 and 2005 (BASE06 minus BASE05),
- ii. changes of atmospheric composition due to differences in the biomass burning emissions under El Niño conditions by comparing BASE and CLIM for 2006 (BASE06 minus CLIM06),
- iii. the impact of the El Niño induced dynamical changes on atmospheric composition and O<sub>3</sub> production by comparing CLIM for the years 2006 and 2005 (CLIM06 minus CLIM05).

### 3.2 Results of the C-IFS experiments

Figure 13 shows timeseries of the tropospheric CO, O<sub>3</sub> and NO<sub>2</sub> burdens from the BASE and CLIM experiments averaged over the area between 10° N, 10° S, 90° E, 130° E. Between September and November 2006 the GFAS v1.0 fire emissions used in BASE lead to an increased CO burden, which reaches values up to 21 Tg, almost

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double the values seen in CLIM (around 11 Tg). In 2005 the tropospheric CO burden in both experiments is similar to the CLIM values of 2006 (around 10–12 Tg). Tropospheric O<sub>3</sub> burdens show a smaller increase (about 8 %) in 2006 from about 7.4 Tg in CLIM to 8 Tg in BASE. The 2006 O<sub>3</sub> burdens in BASE are increased by about 30 % relative to 2005, when the tropospheric O<sub>3</sub> burden is about 6 Tg in both experiments. It should be noted that the tropospheric O<sub>3</sub> mass shows considerable intra seasonal fluctuations. The tropospheric NO<sub>2</sub> burden in BASE is increased by about 20–30 % compared to CLIM in September 2006 as a result of the increased fire emissions. During 2005 NO<sub>2</sub> burdens from BASE and CLIM are of similar magnitude.

The top panels of Fig. 14 show the overall impact of changes to the tropospheric O<sub>3</sub> column due to dynamically and emission driven changes, by comparing BASE06 and BASE05 for October and December. The patterns are very similar to the ones seen in the MACC reanalysis composite O<sub>3</sub> anomalies at 500 hPa (Fig. 4). The combined effect of dynamically induced and emission driven changes leads to an increase of TCO<sub>3</sub> by over 50 % in a large area surrounding Indonesia and to a reduction of 10–30 % over large parts of the central Pacific. TCO<sub>3</sub> values are also reduced by more than 30 % over Brazil. TCO<sub>3</sub> changes due to changes in the fire emissions alone (middle panels of Fig. 14) can only explain part of the observed O<sub>3</sub> increase over Indonesia (which is consistent with the small differences between BASE and CLIM seen in Fig. 13 for September to November 2006) and a small decrease over Brazil, and can not explain the reduction of O<sub>3</sub> over the Pacific. The dynamically induced changes in October (Fig. 14, bottom left) show a similar pattern to the overall differences between El Niño and normal conditions. This illustrates that while emission driven changes can explain about half of the total TCO<sub>3</sub> changes in a small area surrounding Indonesia, the TCO<sub>3</sub> increase outside this region and the negative O<sub>3</sub> anomaly over the Pacific is unrelated to changes in the fire emissions. This is also confirmed by the December plots, when no fire related anomaly is seen any more (Fig. 14, middle right). The dynamically driven O<sub>3</sub> anomalies persist into December and can explain most of the TCO<sub>3</sub> anomaly (Fig. 14,

bottom right). Over Indonesia the  $O_3$  maxima are now located around  $10^\circ$  N and  $10^\circ$  S, and over the Pacific they are slightly smaller scale than in October.

The  $TCO_3$  changes seen in the bottom panel of Fig. 14 are anti-correlated with changes in specific humidity (Fig. 15, top panels) pointing to an enhanced  $O_3$  lifetime over Indonesia due to reduced humidity and hence OH concentrations. Furthermore, there is enhanced ascent over the central Pacific and enhanced descent over Indonesia (Fig. 15, bottom panels) so that increased upward transport of clean  $O_3$  poor air over the Pacific and increased downward transport from the upper troposphere/stratosphere in the Indonesian region will also affect the tropospheric  $O_3$  columns.

Figure 16 shows that the TCCO anomalies over Indonesia are almost entirely emission driven, in contrast to the  $TCO_3$  anomalies seen in Fig. 14. Using GFAS v1.0 emissions rather than climatological GFAS emissions can explain most of the TCCO anomaly over Indonesia in October, apart from two small positive dynamically induced anomalies to the east and west of the maritime continent. By December, after the end of the fire season in Indonesia, the TCCO anomalies have almost gone.

As for CO, the  $NO_2$  and smoke aerosol anomalies are entirely emission driven (not shown). For both these fields, no anomalies are seen when comparing CLIM06 and CLIM05, and the anomalies seen when comparing BASE06 and BASE05 are gone by December. It is possible that we would also see some dynamically induced changes in the smoke aerosols if aerosols were interactive with the radiation scheme in the model runs. However, without this feedback the smoke aerosol anomalies are entirely emission driven.

Figure 17 shows  $O_3$ -CO correlations for October 2005 and 2006 from the BASE and CLIM experiments. We focus only on October as the month with the largest anomaly in the fire emissions. Such correlations have been used in several studies (e.g., Kim et al., 2013; Voulgarakis et al., 2011, and references therein) going back to Fishman and Seiler (1983) to identify regions of photochemically produced  $O_3$  (positive correlations) and  $O_3$  from other sources (e.g. downward transport from stratosphere) as well as  $O_3$  loss due to chemistry or deposition (negative correlations). In 2005, free tro-

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pospheric O<sub>3</sub>-CO correlations (Fig. 17a and c) show a similar distribution across the maritime continent with relatively weak ( $r < 0.7$ ) negative correlations extending from the Indian Ocean south of Indonesia to East and West Malaysia and the Philippines, and positive correlations over northern and eastern Indonesia. Slight differences in the distribution of the O<sub>3</sub>-CO correlations between Fig. 17a and c reflect differences in the fire emissions between the BASE and CLIM experiments. Larger differences are seen in the distribution of free tropospheric O<sub>3</sub>-CO correlations between BASE and CLIM for the El Niño year of 2006 (Fig. 17b and d). The increased fire activity in 2006 (Figs. 4 and 12) gives rise to larger positive ( $r > 0.7$ ) O<sub>3</sub>-CO correlations extending across most of Indonesia and Malaysia in the BASE experiment (Fig. 17b) and reflect enhanced O<sub>3</sub> photochemistry associated with increased emissions of O<sub>3</sub> precursors from the fires. This agrees well with the area of increased O<sub>3</sub> concentrations due to fires seen in Fig. 16 when comparing BASE06 and CLIM06. In contrast, O<sub>3</sub>-CO correlations in the CLIM experiment (Fig. 17d) are generally negative over much of Indonesia and Malaysia and reflect the influence of transport across the region and the lack of enhanced O<sub>3</sub> production when the climatological fire emissions are used.

## 4 Conclusions and outlook

In this paper O<sub>3</sub>, CO, NO<sub>2</sub> and smoke aerosol fields from the MACC reanalysis (Inness et al., 2013) were used to identify the ENSO signal in tropical atmospheric composition fields, concentrating on the months September to December. The MACC atmospheric composition fields show a clear ENSO related anomaly signal with increased O<sub>3</sub>, CO, NO<sub>2</sub> and smoke aerosols over the maritime continent during El Niño years. O<sub>3</sub> also shows larger scale changes with decreased tropospheric columns over the Central and Eastern Pacific and increased columns over the western Pacific and the maritime continent that continue after the end of the Indonesian fire season.

Two simulations were carried out with the C-IFS model to quantify to what extent the ENSO signal seen in the atmospheric composition fields was due to changes in



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the biomass burning emissions or due to dynamically induced changes, e.g. related to changes in the vertical transport of  $O_3$  from the lower troposphere and the stratosphere, and to changes of the photolysis of  $O_3$  due to changes to OH. While the CO,  $NO_2$  and smoke aerosol changes were almost entirely driven by changes in biomass burning emissions due to increased wild fires over the maritime continent during El Niño related drought conditions, changes in tropospheric  $O_3$  were largely dynamically induced and only to a small part driven by changes in the emissions. The emission driven  $O_3$  changes were confined to the area surrounding Indonesia, where enhanced photochemical  $O_3$  production occurs under El Niño conditions because of increased biomass burning activities, while the larger-scale  $O_3$  anomalies were dynamically induced.

Comparing simulations with daily GFAS v1.0 emissions for the years 2005 and 2006 and a daily GFAS v1.0 climatology of the period 2000 to 2014 showed that tropospheric CO was almost doubled in September 2006 relative to September 2005 due to increased fire emissions,  $NO_2$  was increased by 20–30 % and  $O_3$  by about 8 %. For tropospheric  $O_3$ , dynamically induced changes dominated the differences between 2006 and 2005. The fire induced  $O_3$  anomaly was smaller in magnitude and horizontal extent than the dynamically induced changes which affected much of the Tropics. In 2006, tropospheric  $O_3$  was increased by more than 50 % over the maritime continent and Indian Ocean compared to 2005, and decreased by between 20–30 % over large parts of the Tropical Pacific when the same climatological fire emissions were used in both years. Only in a small area over Indonesia was the  $O_3$  increase due to fires of similar magnitude to the dynamically induced changes. A future study will look in more detail at the chemistry budgets and chemical processes that cause the changes in the atmospheric composition fields.

The results from this paper show that the MACC system is able to successfully model the ENSO signal in atmospheric composition fields and could therefore be used in further studies to investigate the ocean–atmosphere response to ENSO induced changes in atmospheric composition. A first step would be to include the aerosol direct and indirect effects through the cloud microphysics in the radiation scheme of the IFS and to

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look at the feedback of fire-induced aerosols on climate. We would expect a positive feedback, i.e. reduced convection due to increased atmospheric stability, as carbonaceous aerosols usually absorb (and thus re-emit) a significant amount of solar radiation in the mid troposphere, and increased aerosol concentrations also lead to reduced land and sea surface temperatures. Their presence should therefore act to reduce convection and precipitation over the maritime continent. Including the aerosols in the radiation scheme will also affect the chemical fields through changes in the UV radiation and hence photolysis rates. A second step could see the coupling of the chemistry and aerosol fields by including heterogeneous chemistry on aerosols. In a final step it can be envisaged to fully couple the MACC system with ECMWF's ocean model to investigate how the forcing from ENSO induced changes to atmospheric composition fields can feedback on the ENSO dynamics.

MACC atmospheric composition data are freely available from [www.copernicus-atmosphere.eu](http://www.copernicus-atmosphere.eu).

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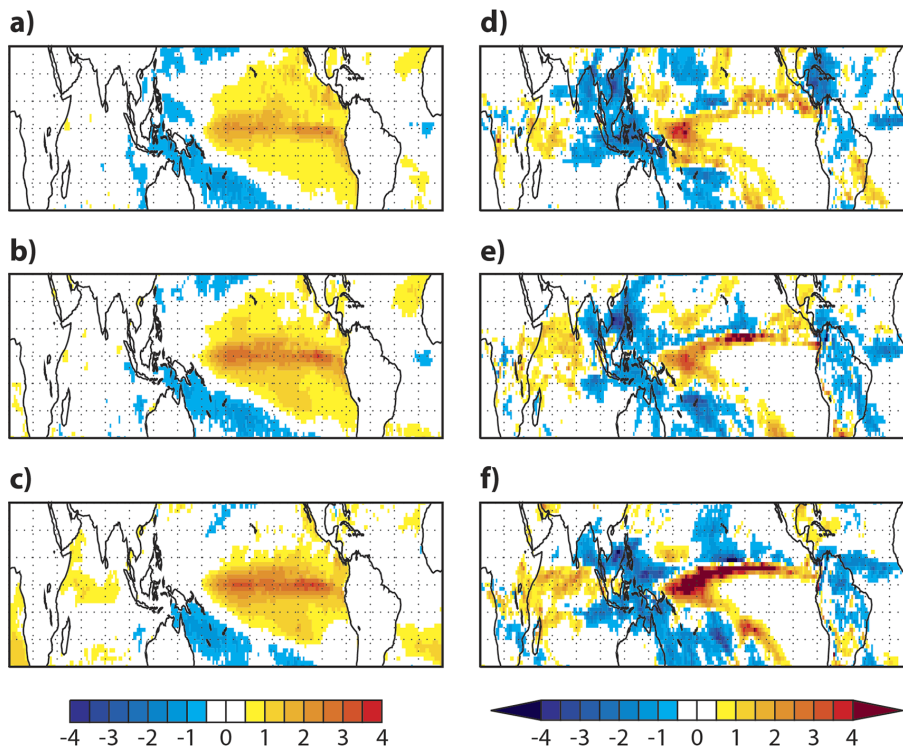
**Table 1.** Biomass burning emissions in Tg per month for CO, NO<sub>x</sub>, and the smoke aerosols (sum of organic matter and black carbon) from GFAS v1.0 for the region 10° N–10° S, 90–130° E averaged over the El Niño years (2004, 2006, 2009) and the La Niña years (2005, 2007, 2008, 2010, 2011), as well as the ratio of the El Niño/La Niña values.

	CO Oct	CO Nov	CO Dec	NO <sub>x</sub> Oct	NO <sub>x</sub> Nov	NO <sub>x</sub> Dec	OM + BC Oct	OM + BC Nov	OM + BC Dec
El Niño	13.9	3.0	$2.4 \times 10^{-1}$	$8.4 \times 10^{-2}$	$1.9 \times 10^{-2}$	$2.8 \times 10^{-3}$	$6.3 \times 10^{-1}$	$1.4 \times 10^{-1}$	$1.3 \times 10^{-2}$
La Niña	1.5	$2.5 \times 10^{-1}$	$1.4 \times 10^{-1}$	$1.4 \times 10^{-2}$	$3.0 \times 10^{-3}$	$1.5 \times 10^{-3}$	$7.3 \times 10^{-2}$	$1.4 \times 10^{-2}$	$7.3 \times 10^{-3}$
El Niño/La Niña	9.5	11.8	1.7	6.1	6.3	1.9	8.6	10.0	1.8



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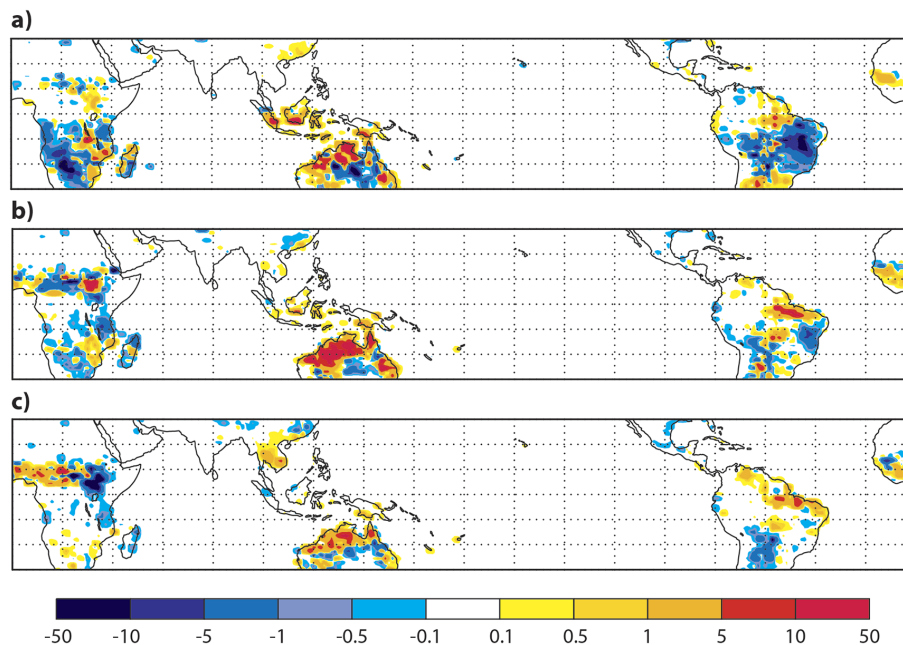
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**Figure 1.** Left panels: SST anomaly in K calculated from ERA Interim as the difference of El Niño composite minus La Niña composite for October (a), November (b) and December (c). Right panels: precipitation anomaly in  $\text{mm day}^{-1}$  calculated from ERA Interim as the difference of El Niño composite minus La Niña composite for October (d), November (e) and December (f). Red colours indicate positive values, blue colours negative values.

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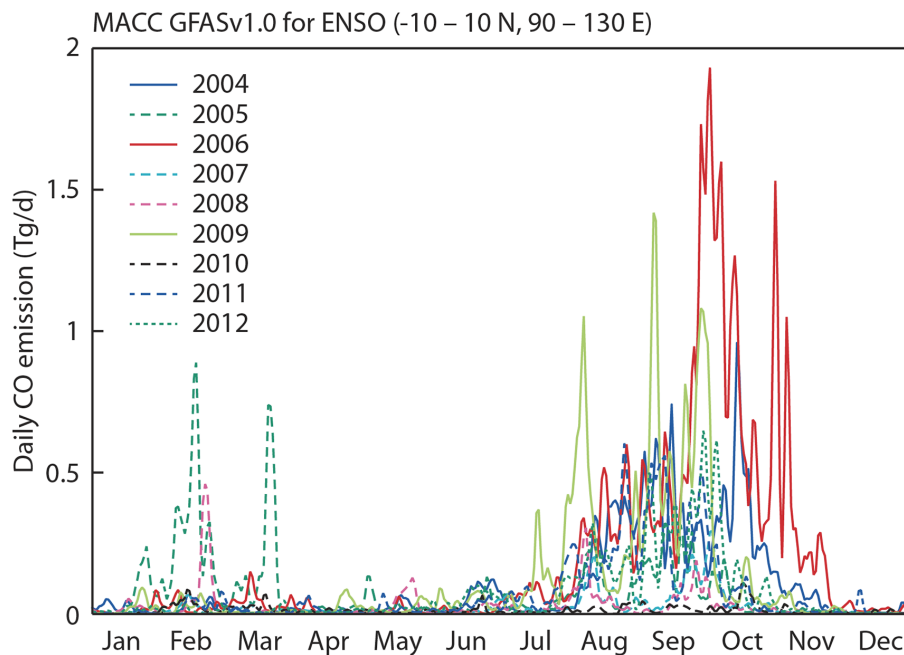
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**Figure 2.** Biomass burning (fire radiative power areal density) anomaly in  $\text{mW m}^{-2}$  calculated from the GFAS v1.0 dataset as the difference of El Niño composite minus La Niña composite for October (a), November (b) and December (c). Red colours indicate positive values, blue colours negative values.

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**Figure 3.** Timeseries of daily CO emissions in  $\text{Tg}(10^{12} \text{g})\text{day}^{-1}$  from GFASv1.0 for the region  $10^\circ \text{N}$ – $10^\circ \text{S}$ ,  $90$ – $130^\circ \text{E}$  for the years 2003 to 2012.

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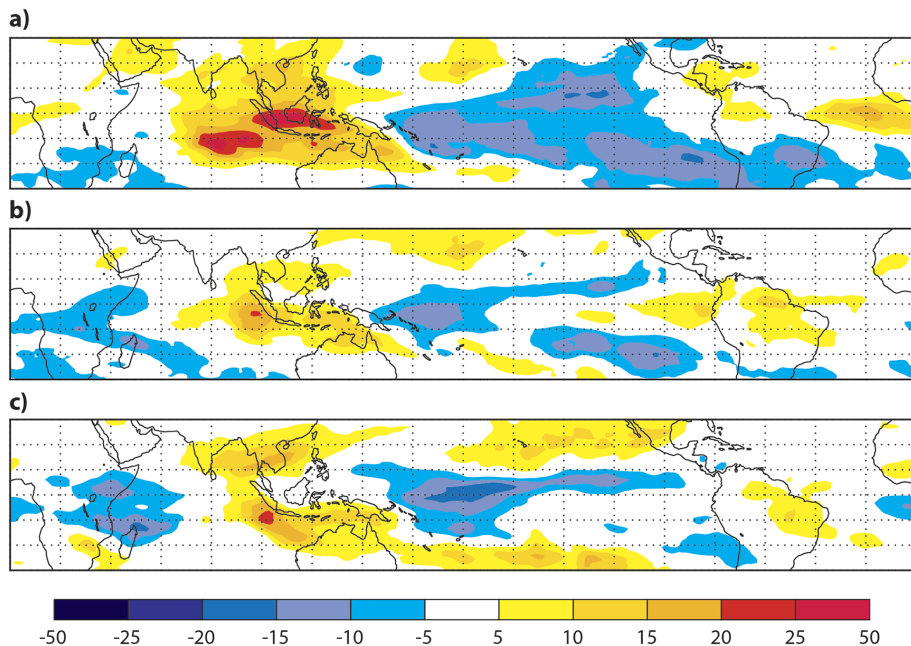
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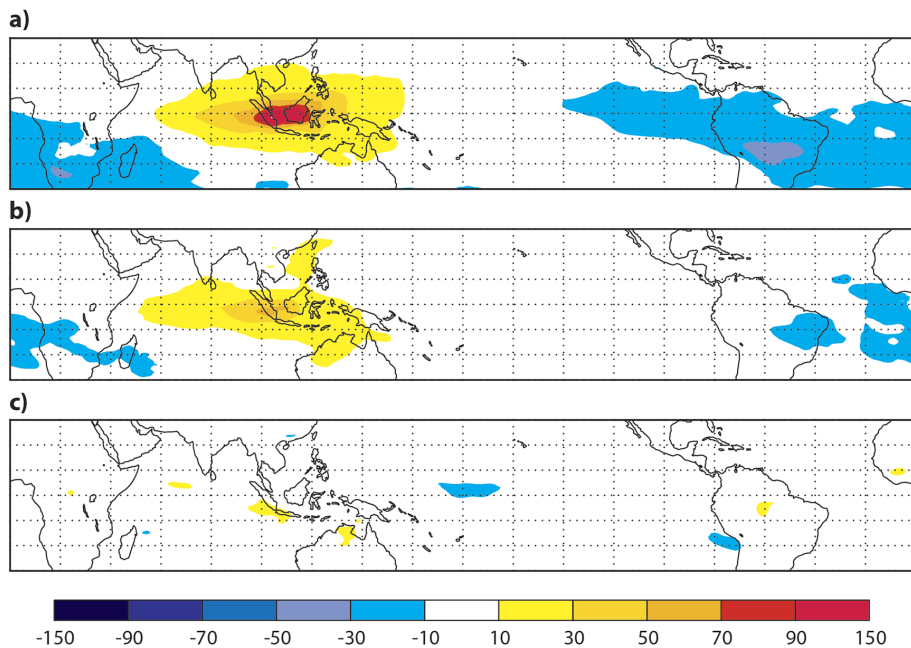


**Figure 4.** O<sub>3</sub> anomaly at 500 hPa in ppb calculated from the MACC reanalysis as the difference of El Niño composite minus La Niña composite for October (a), November (b) and December (c). Red colours indicate positive values, blue colours negative values.

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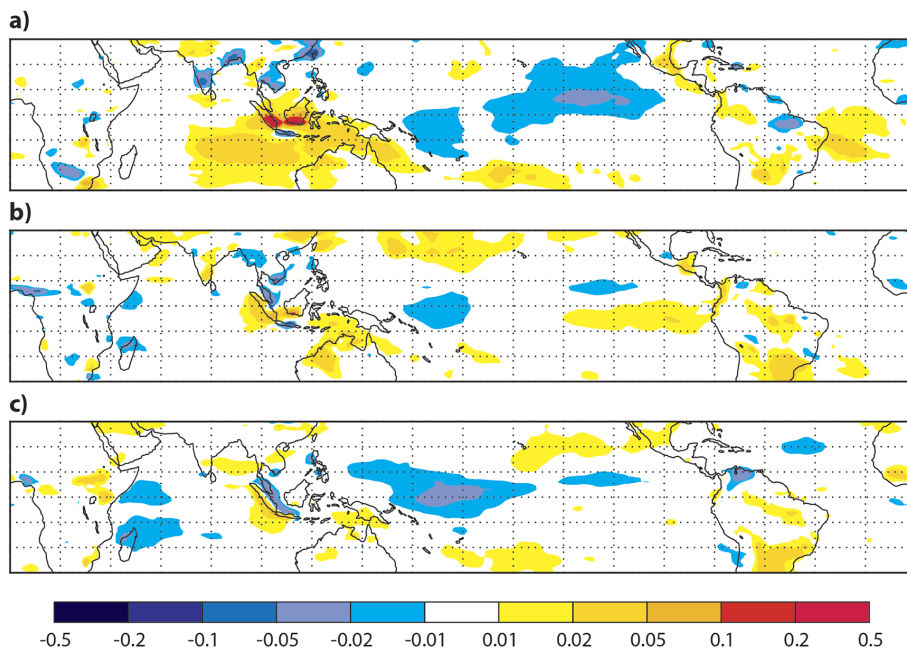


**Figure 5.** Like Fig. 4 but for CO anomaly at 500 hPa in ppb.

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**Figure 6.** Like Fig. 4 but for  $\text{NO}_x$  anomaly in ppb.

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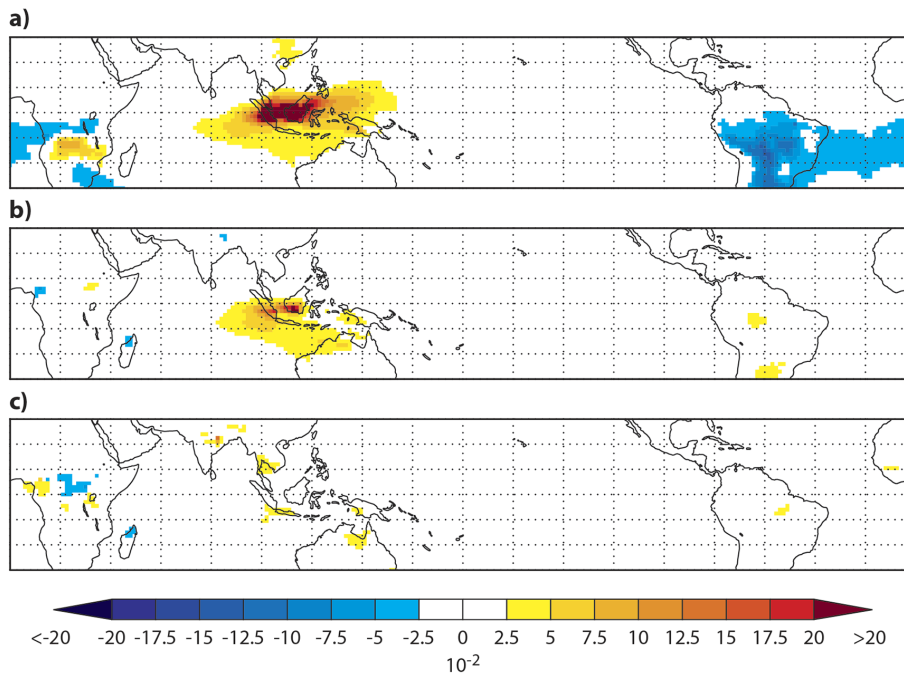
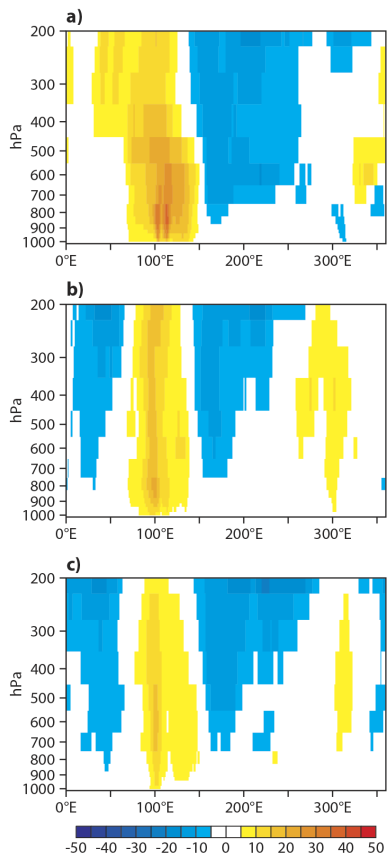


Figure 7. Like Fig. 4 but for smoke AOD (BC + OM). AOD is unitless.

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**Figure 8.** Vertical cross section of  $O_3$  anomalies in ppb plotted against longitude and averaged between  $0^\circ$  and  $12^\circ$  S calculated from the MACC reanalysis as the difference of El Niño composite minus La Niña composite for October (a), November (b) and December (c). Red colours indicate positive values, blue colours negative values.



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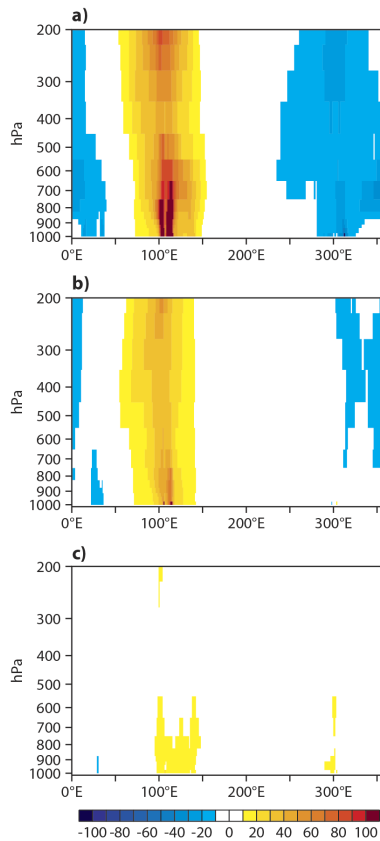
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**Figure 9.** Like Fig. 8 but for CO in ppb.

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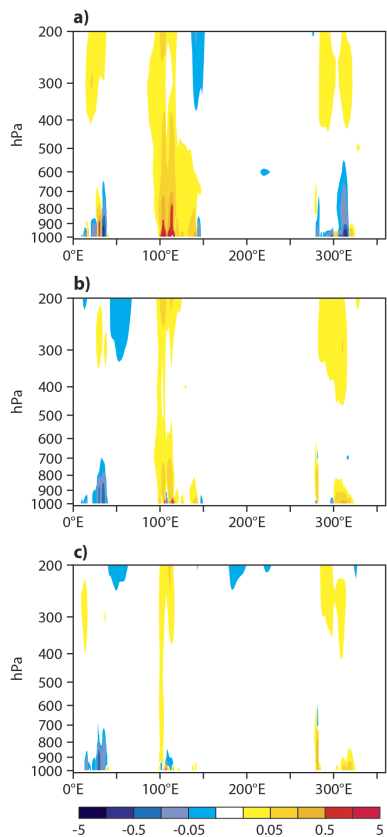
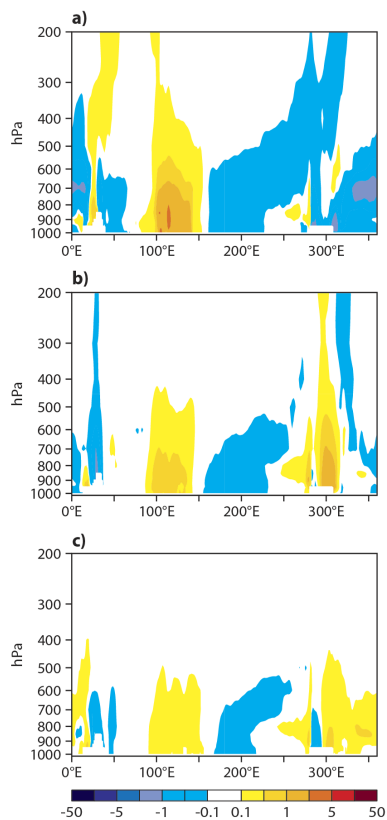


Figure 10. Like Fig. 8 but for  $\text{NO}_x$  in ppb.

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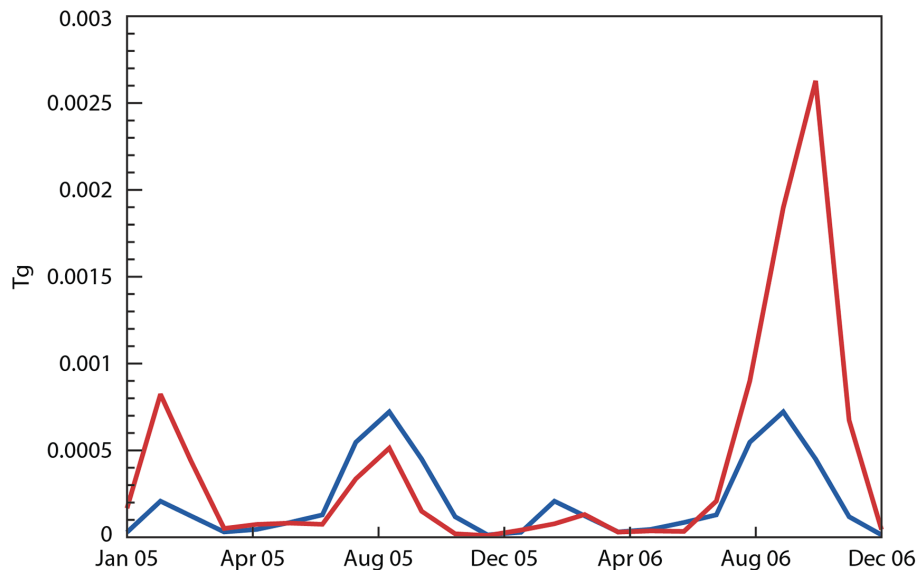
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**Figure 11.** Like Fig. 8 but for smoke aerosol in ppb.

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**Figure 12.** Timeseries of CO biomass burning emissions in Tg averaged over the region between 10° N, 10° S, 90° E, 130° E from GFAS v1.0 (red) and climatological GFAS v1.0 data set (blue.)

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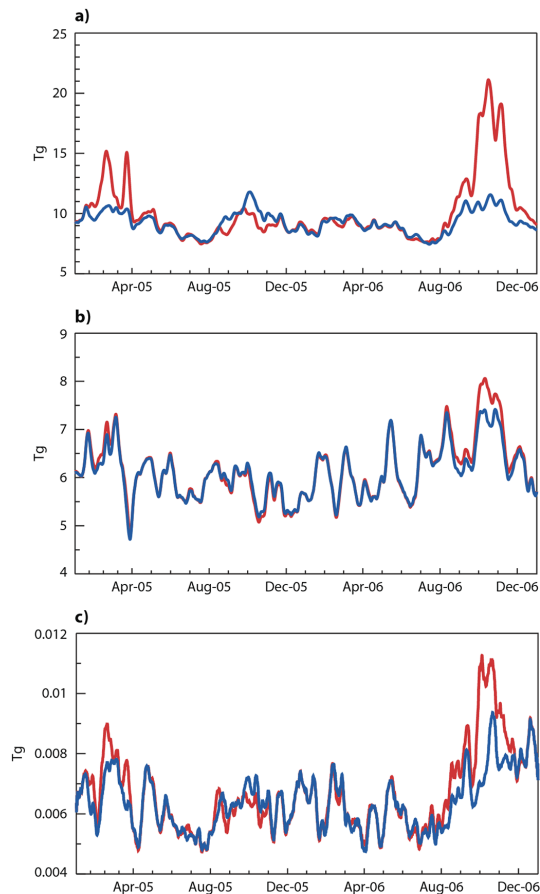
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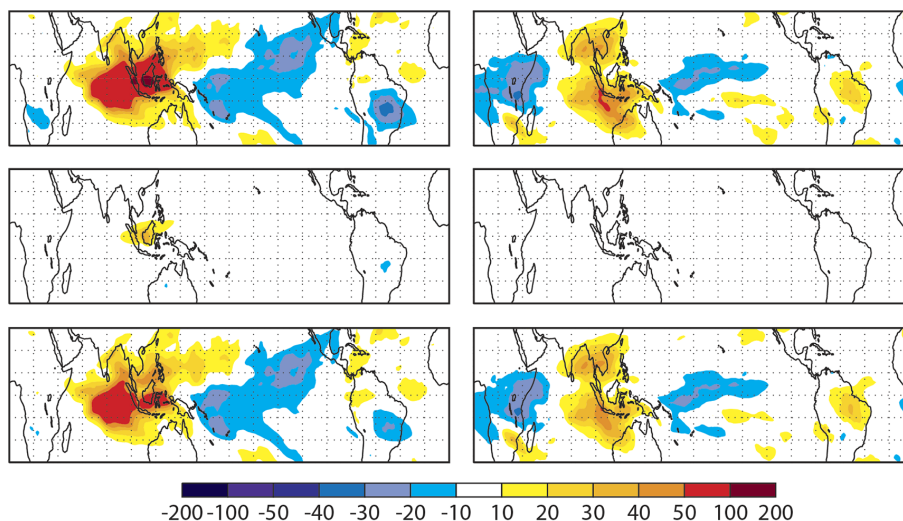
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**Figure 13.** Timeseries of the tropospheric CO (a), O<sub>3</sub> (b), NO<sub>2</sub> (c) burden in Tg from BASE (red) and CLIM (blue) for 2005 and 2006 averaged over the area between 10° N, 10° S, 90° E, 130° E.

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**Figure 14.**  $\text{TCO}_3$  differences in % for October (left) and December (right) from the experiments BASE06 – BASE05 (top), BASE06 – CLIM06 (middle) and CLIM06 – CLIM05 (bottom). The top panels show the overall differences of  $\text{TCO}_3$  due to the combined effects of El Niño related dynamical changes and changes in the fires emissions between El Niño and normal conditions. The middle panels show the impact of changes to the fire emissions under El Niño conditions, and the bottom panels show the impact of the El Niño induced dynamical changes on  $\text{TCO}_3$  when climatological fire emissions are used for both years. Red colours indicate positive values, blue colours negative values.

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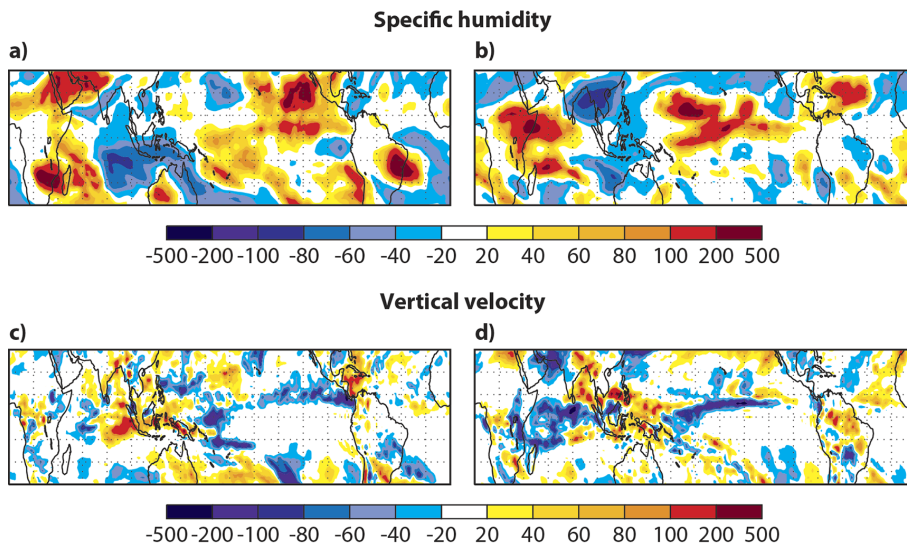
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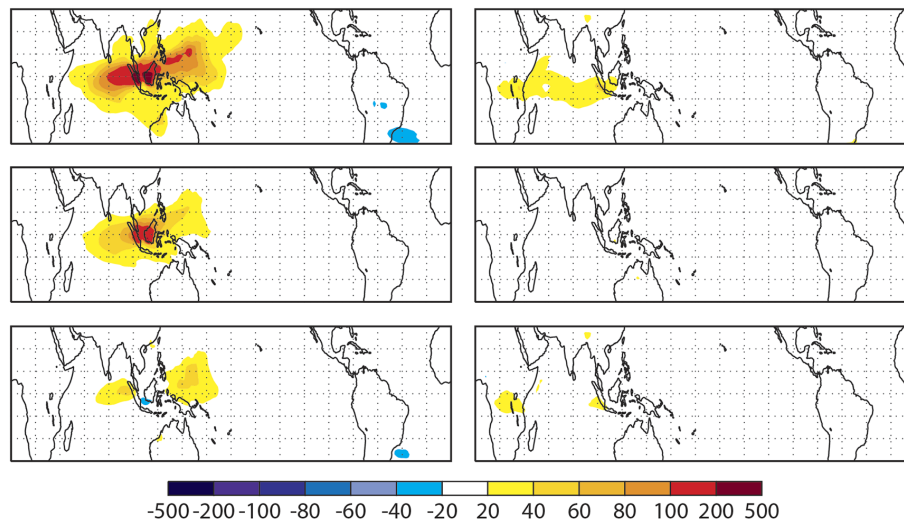


**Figure 15.** Top panels: specific humidity differences at 500 hPa in % for October (a) and December (b) from the experiments BASE06 minus BASE05. Blue colours show reduced specific humidity, red colours increased values. Bottom panels: differences of vertical velocity in  $\text{mm s}^{-1}$  for October (c) and December (d) from the experiments BASE06 minus BASE05. Blue colours show increased ascent, red colours increased descent.

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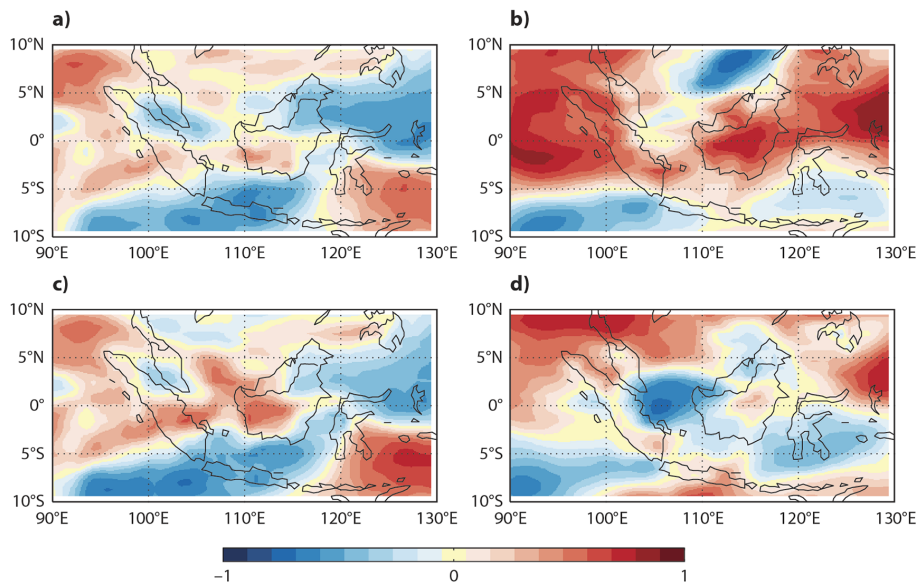
**Figure 16.** TCCO differences in % for October (left) and December (right) from the experiments BASE06 – BASE05 (top), BASE06 – CLIM06 (middle) and CLIM06 – CLIM05 (bottom). Red colours indicate positive values, blue colours negative values.

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**Figure 17.** October O<sub>3</sub>-CO correlations calculated for free tropospheric (approx. 750–350 hPa) column abundances over the maritime continent from the BASE (top) and CLIM (bottom) experiments for 2005 (left) and 2006 (right). Red colours indicated positive correlations, blue colours negative ones.

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