

1 **The ENSO signal in atmospheric composition fields:**
2 **Emission driven versus dynamically induced changes**

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13 **Abstract**

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

27

1

2 **1 Introduction**

3 The El Niño Southern Oscillation (ENSO) is the dominant mode of variability in the Tropics
4 (e.g. Allan et al. 1996). It does not only affect meteorological fields but has a large impact on
5 atmospheric composition too, for example on ozone (O₃), carbon monoxide (CO), nitrogen
6 oxides (NO_x) and aerosols (e.g. Logan et al. 2008; Ziemke and Chandra, 2003, Chandra et al.,
7 2002; Wang et al., 2004). As the result of an eastwards shift of the warm sea surface
8 temperatures (SST) and the large scale Walker circulation anomaly in the tropical Pacific
9 during El Niño years, downward motion is increased and convection and precipitation are
10 reduced over the Western Pacific and the Maritime Continent. During La Niña conditions the
11 opposite dynamical effects occur. Fire emissions over Indonesia show a large interannual
12 variability (IAV), with largest emissions during El Niño years (e.g. van der Werf et al., 2006,
13 Kaiser et al., 2012), when drought conditions and anthropogenic biomass burning lead to big
14 wild fires (Duncan et al., 2003; Lyon et al., 2004; Page et al., 2002) that emit large amounts of
15 trace gases and aerosols. During El Niño years tropospheric O₃ columns (TCO₃) are decreased
16 over the Central and Eastern Pacific and increased over the Western Pacific and Indonesia,
17 while CO concentrations and aerosols from biomass burning increase over Indonesia. Specific
18 humidity changes in the upper troposphere are anti-correlated with the changes in TCO₃ (e.g.
19 Chandra et al., 2007).

20 These atmospheric composition changes have been found in observations (Chandra et al.,
21 1998; Ziemke and Chandra 1999; Fujiwara et al., 1999; Chandra et al., 2007; Logan et al.,
22 2008) and were confirmed by modelling studies (Hauglustaine et al., 1999; Sudo and
23 Takashashi, 2001; Chandra et al., 2002; Doherty et al., 2006; Chandra et al., 2009; Nassar et
24 al., 2009) which also tried to quantify the relative importance of the dynamically induced and
25 the emission driven atmospheric composition changes. The reasons for the TCO₃ increase
26 over the Western Pacific and Indonesia during El Niño years are (i) changes in the vertical
27 transport that lead to enhanced downward transport of O₃ rich air from the upper troposphere
28 (and perhaps stratosphere) to the middle and lower troposphere, and reduced transport of O₃
29 poor air from the lower troposphere into the upper troposphere; (ii) a longer chemical lifetime
30 of O₃ because of reduced humidity which affects the concentrations of the hydroxyl radical
31 (OH) and hence the photochemical loss of tropospheric O₃; and (iii) enhanced photochemical

1 production of O₃ in the lower troposphere because of increased concentrations of O₃
2 precursors from biomass burning, such as NO_x, CO or Hydrocarbons. We refer to (i) and (ii)
3 as ‘dynamically induced changes’ and to (iii) as ‘emission driven changes’ throughout this
4 paper.

5 For El Niño events with large fires over Indonesia, such as in 1997 and 2006, the TCO₃
6 changes due to dynamics and due to increased emissions can be of similar magnitude (Sudo
7 and Takahashi, 2001; Chandra et al., 2002; Chandra et al., 2009), while for weaker events,
8 such as the 2004 El Niño, the dynamical impact dominates (Chandra et al., 2007).

9 The changes in CO are mainly emissions driven (Logan et al., 2008; Chandra et al., 2009;
10 Voulgarakis et al., 2010) and of smaller horizontal scale than the O₃ anomalies, but dynamical
11 interactions due to changes in water vapour (H₂O) and hence OH can also play a role. CO is
12 increased over the Western Pacific and the Maritime Continent during El Niño because of
13 increased emissions from fires and the increased chemical lifetime due to reduced OH. The
14 CO anomaly over Indonesia is usually gone by December (Logan et al., 2008; Chandra et al.,
15 2009), after the end of the biomass burning season, while the O₃ anomaly continues.

16 Large Indonesian wildfires can affect the air quality over South East Asia. Aouizerats et al.
17 (2015) investigated how the transport of biomass burning emissions from Sumatra affected
18 the air quality in Singapore. They found that 21% of the PM₁₀ loading in Singapore during
19 July to October 2006 was due to Sumatran fires, and that Sumatran fires were responsible for
20 about half of the days with PM₁₀ concentrations greater than 50 µgm⁻³ while the other half
21 was due to local anthropogenic pollution and contributions from smaller fires. The impact of
22 fire emissions on atmospheric aerosol concentrations is of particular interest because of the
23 potential feedbacks of fire-induced aerosols on climate. Several studies have looked at the
24 correlations between ENSO and aerosols or atmospheric haze produced by the Indonesian
25 fires. Wang et al. (2004) used visibility data over Sumatra as an indicator of biomass burning
26 and found that haze events were strongly correlated with El Niño during the 1973 to 2003
27 period. Tosca et al. (2010) used satellite data and modelling studies and found that the aerosol
28 optical depth (AOD) over Indonesia had a large IAV that was driven by wild fires during
29 periods of El Niño induced droughts. Their modelling study showed that the fire-emitted
30 aerosols could initiate a positive feedback loop. The aerosols acted to intensify drought over
31 the biomass burning regions. The aerosols also reduced land and sea surface temperatures,
32 and hence suppressed convection and precipitation in the area. Podgorny et al. (2003) looked

1 at the feedback between El Niño and the Indonesian biomass burning of 1997 and also found
2 that the haze from the fires reduced the solar radiation absorbed by the equatorial Indian
3 Ocean and increased the solar heating of the atmosphere, thus raising the possibility of
4 dynamical feedbacks of the smoke forcing on ENSO. Chung and Ramanathan (2003) carried
5 out modelling studies to assess the remote impact of changes in the South Asian haze and
6 found that fluctuations in the absorbing aerosol forcing could affect the interannual climate
7 variability in the Tropics (and Extratropics). It could remotely suppress convection in the
8 equatorial western Pacific and lead to an ocean-atmosphere response that was very similar to
9 El Niño like warming.

10 As part on the EU FP7 funded Monitoring Atmospheric Composition and Climate (MACC)
11 project (www.copernicus-atmosphere.eu) a 10-year reanalysis of atmospheric composition
12 (Inness et al. 2013) was constructed. This reanalysis provides fields of chemically active
13 gases, for example CO, O₃, and NO_x, as well as aerosols globally for both the troposphere
14 and the stratosphere for the years 2003 to 2012. It gives us the unprecedented possibility to
15 assess the impact of ENSO on atmospheric composition using an observationally constrained,
16 continuous, 3-dimensional atmospheric composition dataset with a resolution of about 80 km,
17 which is greater than the resolutions used in most previous modelling studies. In this paper we
18 show that the MACC reanalysis shows the ENSO induced anomalies in O₃, CO, NO_x and
19 aerosols described in earlier studies. We then use MACC's Composition Integrated
20 Forecasting System (C-IFS) model (Flemming et al. 2015) to quantify the relative impact of
21 the dynamics and the biomass burning emissions on the ENSO signal in the O₃, CO, NO_x and
22 smoke aerosol fields.

23 This paper is structured in the following way. Section 2 describes the MACC reanalysis and
24 the ENSO signal seen in MACC O₃, CO, NO_x and smoke aerosol fields. Section 3 describes
25 the additional C-IFS model runs that were carried out to quantify the relative impact of the
26 dynamics and biomass burning emission on the ENSO signal in the atmospheric composition
27 fields and their results, and Section 4 presents conclusions and outlook.

28

1 **2 ENSO signal in the MACC reanalysis**

2 **2.1 The MACC reanalysis**

3 The MACC data assimilation system provides analyses and forecasts of atmospheric
4 composition and was used to produce a reanalysis of atmospheric composition covering the
5 years 2003 to 2012, as described in Inness et al. (2013). O₃ retrievals from several instruments
6 (including the Ozone Monitoring Instrument (OMI), SCanning Imaging Absorption
7 spectroMeter for Atmospheric CHartographY (SCIAMACHY), Solar Backscatter Ultra-
8 Violet (SBUV/2), Microwave Limb Sounder (MLS), CO retrievals from Measurements of
9 Pollution in the Troposphere (MOPITT) and the Infrared Atmospheric Sounding
10 Interferometer (IASI), tropospheric NO₂ columns from SCIAMACHY, and AOD from the
11 Moderate Resolution Imaging Sctoradiometer (MODIS) were assimilated to constrain the
12 atmospheric compositions fields. For more information about the assimilated datasets and the
13 quality of the O₃, CO, and NO_x fields produced by the analysis see Inness et al. (2013). The
14 aerosol analysis in the MACC reanalysis is similar to that described in Benedetti et al. (2009)
15 and Morcrette et al. (2011) and is based on 4-dimensional variational assimilation of AOD
16 observations at 550 nm from the MODIS sensors, including a global adaptive bias correction.
17 Comparisons of multiyear averages of AOD over the period 2003–2010 from the MACC
18 reanalysis and from the Multi-angle Imaging SpectroRadiometer (MISR) sensors onboard the
19 Terra satellite indicate good qualitative agreement (not shown).

20 The anthropogenic emissions for the reactive gases for the MACC reanalysis were taken from
21 the MACCity inventory (Granier et al., 2011) which accounts for projected trends in the
22 emissions. For the aerosol fields they came from the EDGAR database (Dentener et al.,
23 2006). Monthly biomass burning emission for the years 2003 to 2008 from the GFED3.0
24 inventory (van der Werf et al., 2010) were scaled to daily resolution using MODIS active fire
25 observations. From 2009 to 2012 daily biomass burning emissions from MACC's GFAS,
26 Version 1.0 (Kaiser et al. 2012) were used. One advantage of the MACC reanalysis is that it
27 used daily fire emission, in contrast to several other studies that used monthly averages.
28 Biogenic emissions used in the MACC reanalysis were for 2003. They came from a recent
29 update (Barkley, 2010) of the Model of Emissions of Gases and Aerosols from Nature version
30 2 (MEGAN2; Guenther et al. 2006, <http://acd.ucar.edu/~guenther/MEGAN/MEGAN.htm>)
31 and were used as monthly surface flux fields without interannual variation.

1 The emissions are injected at the surface and distributed over the boundary layer by the
2 model's convection and vertical diffusion scheme. Despite the distribution being very
3 efficient, this is a limitation of the current system that will be addressed in future versions.
4 Experiments have been carried out with a new version that uses injection heights based on the
5 Plume Rise Model of Paugam et al. (2015). They show a significant impact on BC AOD for
6 single large fires; the impact at a global scale is smaller: BC AOD is increased by around 5%.
7 Most of the injection heights calculated with the Plume Rise Model lie within the boundary
8 layer and only a small fraction of smoke (often from particularly intense and well-studied
9 fires) is injected directly into the free troposphere. The largest smoke transport from the
10 boundary layer to the free troposphere occurs through larger-scale meteorological processes.
11 The lowering of the boundary layer height, when air is advected from land to sea, and strong
12 updrafts in frontal system have previously been identified as efficient smoke transport
13 mechanisms. Similarly, Veira et al. (2015) has studied the sensitivity of AOD in a global
14 climate model to different injection height parameterisations and the above-mentioned plume
15 rise model, with the conclusion that a simple parameterisation reproduces the average larger-
16 scale distribution sufficiently well.

17 The MACC models do not contain halogenated species, which would contribute to a small
18 additional loss term to O₃ and CO. Ocean emissions of volatile organic compounds (VOCs)
19 originate from climatological data from POET. Deposition on ocean surface depends on the
20 species solubility, which is negligible for O₃ and CO, but not for some of the VOCs. All these
21 aspects may contribute to overall biases in the model, but are not considered essential for the
22 signals investigated here.

23 Initial validation results from the MACC reanalysis are shown in Inness et al. (2013) and
24 Morcrette et al. (2011) and more detailed validation can be found in the MACC reanalysis
25 validation reports available from
26 http://atmosphere.copernicus.eu/services/aqac/global_verification/validation_reports/ .

27 **2.2 ENSO anomalies**

28 The MACC reanalysis was used to construct monthly composites of O₃, CO, NO_x fields at
29 500 hPa and of the smoke AOD, i.e. the sum of black carbon (BC) and organic matter (OM)
30 AOD, at 550 nm for El Niño and La Niña years for the months October, November and
31 December. The El Niño composite was constructed from the years 2004, 2006, 2009, the La

1 Niña composite from the years 2005, 2007, 2008, 2010, and 2011. Weak El Niño and La Niña
2 years were included in the composite calculation to increase the sample size. A recent
3 timeseries of the Multivariate ENSO index which was used to define the years used in our El
4 Niño and La Niña composites can be found on
5 <http://www.esrl.noaa.gov/psd/enso/mei/index.html>. Composites of vertical velocity and
6 specific humidity at 500 hPa were also calculated from the MACC reanalysis. SST and
7 precipitation composite fields, that were not available from the MACC reanalysis, were
8 constructed for the same years from the ERA Interim reanalysis (Dee et al. 2011), and
9 biomass burning composites were calculated from the GFAS v1.0 data set. The composites
10 were then used to calculate anomalies for the various fields by taking the difference between
11 the El Niño and La Niña composites for the months October, November and December.

12 Figure 1 shows the warm SST anomaly over the Central Pacific associated with El Niño
13 conditions and the resulting precipitation changes for October, November and December from
14 ERA Interim. Precipitation is increased over the central Pacific and reduced over the Western
15 Pacific, Maritime Continent, northern Australia and part of the Indian Ocean. Figure 2 shows
16 that the increased precipitation over the Central Pacific and the reduced precipitation over the
17 Maritime continent are collocated with increased ascent and increased descent at 500 hPa,
18 respectively. At the same time, specific humidity at 500 hPa shows a positive anomaly in the
19 area of increased ascent and precipitation over the Central Pacific and a negative anomaly
20 over the Maritime continent. Cloud cover shows a similar signal to humidity, with a negative
21 anomaly over the Maritime continent and a positive anomaly over the central Pacific (not
22 shown).

23 The increased biomass burning emissions related to the lack of rainfall over Indonesia and
24 Northern Australia can be seen in the FRP anomalies shown in Figure 3. Increased fire
25 activity can be seen over Indonesia in October and November, but has stopped by December
26 after the end of the fire season, while a weaker biomass burning anomaly continues in
27 Northern Australia into December. Over Brazil decreased fire activity can be seen in October.

28 Figure 4 shows the IAV of the biomass burning emissions for CO from GFAS v1.0 for the 10
29 years covered by the MACC reanalysis for the area around Indonesia (10°N and 10°S, 90°E
30 and 130°E) and illustrates that the emissions are higher during the El Niño years 2004, 2006,
31 2009 than at other times. The largest values are seen for 2006. Table 1 shows the average CO,
32 NO_x and fire aerosol emissions from GFAS v1.0 during October, November, December for

1 the El Niño years and La Niña years. The average CO emissions during El Niño years during
2 October, November, and December are a factor of about 9, 12, and 2 larger, respectively, than
3 during La Niña years. For NO_x fire emissions the factors are 6, 7, 2 and for smoke aerosols 8,
4 10, 2, respectively. The values in Table 1 are slightly smaller than the values in Chandra et al.
5 (2009) who list CO fire emissions over Indonesia as 44.2 Tg/month, 6.3 Tg/month and 0.2
6 Tg/months and NO₂ fire emissions of 0.76, 0.11, and 0.0 for October, November, and
7 December 2006, respectively.

8 Figures 5 to 8 illustrate the impact that the dynamical and emission related changes have on
9 the atmospheric composition fields, by showing the anomalies calculated from the MACC
10 reanalysis at 500 hPa for O₃, CO, NO_x, and smoke AOD, respectively. O₃ shows positive
11 anomalies over the Western Pacific, Indonesia, Northern Australia and the Eastern Indian
12 Ocean, and negative anomalies over the Central and Eastern Pacific (Figure 5). These
13 anomalies are quite large scale and continue into December after the end of the biomass
14 burning anomaly over Indonesia. The O₃ anomalies agree well with those described in other
15 studies based on MLS and Total Ozone Mapping Spectrometer (TOMS) data (Chandra et al.
16 2009) as well as Tropospheric Emission Spectrometer (TES) data (Logan et al. 2008). The
17 negative O₃ anomaly over Africa and the western Indian Ocean in December was also noted
18 by Nassar et al. (2009) in TES data and modelling studies with the Goddard Earth Observing
19 System 3-D chemical transport model (GEOS-Chem), and we also note a negative O₃
20 anomaly over Brazil in October. The O₃ anomalies are a combination of biomass burning
21 changes and meteorological changes as a consequence of changes in SSTs and the resulting
22 eastward shift of the Walker circulation (Sudo and Takahashi 2001; Chandra et al. 2002;
23 Chandra et al. 2009). The O₃ decrease over the Central and East Pacific is due to enhanced
24 upward transport of O₃ poor air from the boundary layer into the middle and upper
25 troposphere, and a shorter O₃ lifetime and larger photochemical loss due to increased H₂O
26 (and hence OH) concentrations.

27 During October and November both dynamical and emission driven effects contribute, and
28 modelling studies (e.g. Sudo and Takahashi 2001; Chandra et al. 2002; Chandra et al. 2009)
29 have shown that emissions and dynamical changes can contribute equally for El Niño years
30 with strong biomass burning. The O₃ changes in December are due to the dynamical changes
31 after the end of the fire season over Indonesia. This agrees with what was seen by Logan et al.
32 (2008), Chandra et al. (2009) and Nassar et al (2009).

1 Figure 6 shows the CO anomalies at 500 hPa calculated from the MACC reanalysis. These
2 anomalies are more confined to the areas of the biomass burning anomalies (see Figure 3)
3 than the O₃ anomalies (Figure 5). The strongest positive anomaly is found over the Maritime
4 Continent during October and is linked to increased emissions from enhanced biomass
5 burning under drought conditions. A negative CO anomaly extends from South America over
6 the Southern Atlantic to Africa in October and is related to lower fire emissions over Brazil.
7 Similar anomalies were described by Logan et al. (2008), Nassar et al. (2009) and Chandra et
8 al. (2009). The CO anomalies during November are weaker than the October ones, and by
9 December the anomalies have all but disappeared. This confirms that the CO anomalies are
10 mainly emission driven and are not affected much by the dynamical changes that cause the O₃
11 anomalies after the end of the biomass burning season in December.

12 The NO_x anomalies (Figure 7) clearly show the impact of the increased emissions from
13 biomass burning over Indonesia during October and November, but also seem to indicate some
14 large scale response. For example, the negative anomaly over the Eastern Pacific is colocated
15 with the negative O₃ anomaly here (Figure 5) and could indicate enhanced upward transport
16 of NO_x poor air to these levels. Also the negative NO_x anomaly over the Pacific in
17 December is again co-located with a (larger) negative O₃ anomaly.

18 Figure 8 shows the anomaly of smoke AOD at 550 nm calculated from the MACC reanalysis.
19 The largest positive anomaly is found over Indonesia in October and November,
20 corresponding to increased aerosol concentrations from biomass burning emissions. The
21 negative aerosol anomaly over South America in October is related to the reduced fire activity
22 seen in Figure 3. By December the anomalies have disappeared. Similar AOD anomaly
23 patterns over Indonesia were seen by Tosca et al. (2010) when comparing El Niño and La
24 Niña years for August to October for the period 2000 and 2006 from the MISR and MODIS
25 data.

26 The 3-dimensional nature of the MACC reanalysis allows us to look at the vertical
27 distribution of the anomalies in the troposphere. Figures 9 to 12 show height versus longitude
28 cross-sections of O₃, CO, NO_x and smoke AOD anomalies averaged over the latitude range
29 from 0 to 12°S. The O₃ cross section in Figure 9 illustrates that the largest positive O₃
30 anomalies in October and November are located in the lower troposphere and are likely to be
31 the result of enhanced O₃ production due to increased concentrations of O₃ precursors from
32 enhanced fire emissions. However, the positive and negative anomalies extend into the upper

1 troposphere, and some of the anomalies (for example the negative anomaly over the Central
2 Pacific) are clearly not connected to the surface but seem to originate in the middle or upper
3 troposphere. These anomalies continue into December after the end of the fire season, and are
4 likely to be a result of the dynamically induced changes mentioned above.

5 Figure 10 shows that CO anomalies are largest in the lower troposphere but can extend
6 throughout the troposphere over Indonesia and South America. There is a clear connection to
7 increased CO emissions over Indonesia and decreased emissions over South America due to
8 changes in biomass burning. By December the anomalies have all but gone and show that
9 there is no dynamically induced anomaly, unlike in O₃. Now a small positive anomaly is
10 found over South America.

11 Figure 11 shows cross sections of NO_x anomalies calculated from the reanalysis. The largest
12 anomalies are located in the lower troposphere and are again clearly connected to changes in
13 the fire emissions. A large positive anomaly is found over Indonesia and negative anomalies
14 over South America in October and Africa in October and November. In these areas there are
15 reduced fire activities and increased precipitation during El Niño years. Positive NO_x
16 anomalies are found in the upper troposphere and could be a result of increased NO_x
17 production from lightning over South America where there is positive precipitation anomaly
18 pointing to increased convection. The flash rates in the lightning NO parametrization are 5-10
19 larger over land than over ocean which might explain why no signal is seen over the Central
20 Pacific. The positive NO_x anomalies around 100°E in October and November are collocated
21 with high O₃ values in the lower troposphere (Figure 9) pointing to enhanced O₃ production
22 due to enhanced NO_x concentrations from biomass burning. In December, when NO_x does
23 not show such a positive anomaly any more, O₃ concentrations in the lower troposphere are
24 lower and the maximum of the O₃ anomaly is located above 700 hPa.

25 Figure 12 depicts cross sections of smoke AOD and shows that, as for CO and NO_x, there is a
26 clear connection to increased emissions over Indonesia in October and reduced emissions over
27 South America. By December the positive anomaly over Indonesia is much reduced and
28 confined to the lower troposphere. Enhanced AOD concentrations can be seen in November
29 over South America. In the lower troposphere there is a negative aerosol anomaly over the
30 Central Pacific that is not seen in the other atmospheric composition fields. This anomaly is
31 likely to be the result of the increased precipitation in this area during El Niño conditions (see

1 Figure 1) which leads to increased wet deposition and removal of aerosols, while not
2 removing the gas-phase species in the same way.

3

4 **3 Quantifying the relative importance of dynamically and emission driven** 5 **changes on the atmospheric composition fields**

6 **3.1 Experiment setup**

7 To quantify the relative impact of increased biomass burning emissions and dynamically
8 induced changes on the atmospheric composition fields during El Niño conditions two
9 experiments are run for the years 2005 and 2006: one with normal and one with
10 climatological GFAS v1.0 fire emissions. 2006 was an El Niño year, and 2005 is used to
11 represent normal to weak La Niña conditions. The additional experiments use the most recent
12 version of the MACC system, the C-IFS model (Flemming et al. 2015; Inness et al. 2015).
13 This model is different to the one used in the MACC reanalysis (Inness et al. 2013) because it
14 has chemistry routines included directly in ECMWF's Integrated Forecasting System (IFS). A
15 basic initial validation of CIFS-fields can be found in Flemming et al. (2015) and Inness et al.
16 (2015) and more detailed validation of C-IFS can be found in the validation reports available
17 from <http://www.copernicus-atmosphere.eu/>.

18 The chemistry scheme implemented in the C-IFS model version used for these experiments is
19 an extended, modified version of the Carbon Bond Mechanism 5 (Yarwood et al., 2005)
20 chemical mechanism as originally implemented in the Tracer Model 5 (TM5) CTM (Huijnen
21 et al., 2010, Williams et al., 2013; Huijnen et al., 2014). This is a tropospheric chemistry
22 scheme with 54 species and 126 reactions. For O₃ a simple stratospheric parameterisation
23 based on Cariolle and Teyssèdre (2007) has been added. Monthly mean dry deposition
24 velocities are currently based on climatological fields from MOCAGE (Michou et al., 2004).
25 The module for wet deposition is based on the Harvard wet deposition scheme (Jacob et al.,
26 2000 and Liu et al., 2001). The output of the IFS convection scheme is used to calculate NO
27 emissions from lightning. They are parameterised using estimates of the flash rate density, the
28 flash energy release and the vertical emission profile. Estimates of the flash rate density are
29 based on parameters of the convection scheme and calculated using convective precipitation
30 as input parameter (Meijer et al 2001). Documentation of the technical implementation of C-
31 IFS and more details about the model can be found in Flemming et al. (2015). In the present

1 study, the C-IFS aerosol fields are not used in the radiation scheme, where an aerosol
2 climatology based on Tegen et al. (1997) is used instead. Also, heterogeneous chemistry on
3 aerosols is not included.

4 The anthropogenic emissions used in the C-IFS runs come from the MACCity emission data
5 base (Granier et al., 2011). Biogenic emissions are taken from the POET database for the year
6 2000 (Granier et al. 2005; Olivier et al. 2003), with isoprene emissions from MEGAN2.1,
7 again for the year 2000 (Guenther et al., 2006). Biomass burning emissions for the runs are
8 either taken from GFAS v1.0 (Kaiser et al. 2012) or from a GFAS v1.0 climatology. This
9 daily climatology was constructed using the GFAS v1.0 dataset from 2000 to 2014 (Kaiser et
10 al. 2012, Remy and Kaiser, 2014). Biomass burning emissions for each day of the year were
11 defined as the average of the emissions of the same day of the year for the 15 years of the
12 dataset.

13 The differences between the GFAS v1.0 and climatological GFAS emissions for the area
14 between 10°N, 10°S, 90°E, 130°E are shown in Figure 13. The figure illustrates that 2006 was
15 a year with exceptionally large biomass burning emissions over Indonesia during the biomass
16 burning season (as already seen in Figure 5), while in 2005 emissions were slightly below
17 average.

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

- 22 i. the overall impact of changes to the atmospheric composition fields due to El Niño
23 related dynamically and emission induced changes by comparing BASE for the years
24 2006 and 2005 (BASE06 minus BASE05),
- 25 ii. changes of atmospheric composition due to differences in the biomass burning
26 emissions under El Niño conditions by comparing BASE and CLIM for 2006
27 (BASE06 minus CLIM06),
- 28 iii. the impact of the El Niño induced dynamical changes on atmospheric composition and
29 O₃ production by comparing CLIM for the years 2006 and 2005 (CLIM06 minus
30 CLIM05).

31

1 **3.2 Results of the C-IFS experiments**

2 Figure 14 shows timeseries of the tropospheric CO, O₃ and NO₂ burdens from the BASE and
3 CLIM experiments averaged over the area between 10°N, 10°S, 90°E, 130°E. Between
4 September and November 2006 the GFAS v1.0 fire emissions used in BASE lead to an
5 increased CO burden, which reaches values up to 21 Tg, almost double the values seen in
6 CLIM (around 11 Tg). In 2005 the tropospheric CO burden in both experiments is similar to
7 the CLIM values of 2006 (around 10-12 Tg). Tropospheric O₃ burdens show a smaller
8 increase (about 8%) in 2006 from about 7.4 Tg in CLIM to 8 Tg in BASE. The 2006 O₃
9 burdens in BASE are increased by about 30% relative to 2005, when the tropospheric O₃
10 burden is about 6 Tg in both experiments. It should be noted that the tropospheric O₃ mass
11 shows considerable intra seasonal fluctuations. The tropospheric NO₂ burden in BASE is
12 increased by about 20-30% compared to CLIM in September 2006 as a result of the increased
13 fire emissions. During 2005 NO₂ burdens from BASE and CLIM are of similar magnitude.

14 The top panels of Figure 15 show the overall impact of changes to the tropospheric O₃ column
15 due to dynamically and emission driven changes, by comparing BASE06 and BASE05 for
16 October and December. The patterns are very similar to the ones seen in the MACC
17 reanalysis composite O₃ anomalies at 500 hPa (Figure 5). The combined effect of dynamically
18 induced and emission driven changes leads to an increase of TCO₃ by over 50% in a large
19 area surrounding Indonesia and to a reduction of 10-30% over large parts of the Central
20 Pacific. TCO₃ values are also reduced by more than 30% over Brazil. TCO₃ changes due to
21 changes in the fire emissions alone (middle panels of Figure 15) can only explain part of the
22 observed O₃ increase over Indonesia (which is consistent with the small differences between
23 BASE and CLIM seen in Figure 14 for September to November 2006) and a small decrease
24 over Brazil, and can not explain the reduction of O₃ over the Pacific. The dynamically
25 induced changes in October (Figure 15, bottom left) show a similar pattern to the overall
26 differences between El Niño and normal conditions. This illustrates that while emission
27 driven changes can explain about half of the total TCO₃ changes in a small area surrounding
28 Indonesia, the TCO₃ increase outside this region and the negative O₃ anomaly over the
29 Pacific is unrelated to changes in the fire emissions. This is also confirmed by the December
30 plots, when no fire related anomaly is seen any more (Figure 15, middle right). The
31 dynamically driven O₃ anomalies persist into December and can explain most of the TCO₃

1 anomaly (Figure 15, bottom right). Over Indonesia the O₃ maxima are now located around
2 10°N and 10°S, and over the Pacific they are slightly smaller scale than in October.

3 The importance of the dynamically driven ozone changes was also highlighted by Lin et al.
4 (2014 and 2015). Despite large El Nino enhancements to wildfire activity in equatorial Asia,
5 the model sensitivity experiments in Lin et al. (2014) indicated that wildfire emissions are not
6 the main driver of ENSO-related ozone variability observed at Mauna Loa, Hawaii. The
7 dynamically induced eastward extension and equatorward shift of the subtropical jet stream
8 during El Nino plays a key role on observed interannual variability of springtime lower
9 tropospheric ozone at Mauna Loa. These shifts enhance long range transport of Asian ozone
10 and CO pollution towards the eastern North Pacific in winter and spring during El Nino. Lin
11 et al. (2015) demonstrated a connection between springtime western US ozone air quality and
12 jet characteristics associated with strong La Nina winters. They showed more frequent late
13 spring deep stratospheric ozone intrusions when the polar jet stream meanders southward over
14 the western United States as occurs following strong La Nina winters.

15 The TCO₃ changes seen in the bottom panel of Figure 15 are anti-correlated with changes in
16 specific humidity (Figure 16, top panels) pointing to an enhanced O₃ lifetime over Indonesia
17 due to reduced humidity and hence OH concentrations. Furthermore, there is enhanced ascent
18 over the Central Pacific and enhanced descent over Indonesia (Figure 16, bottom panels) so
19 that increased upward transport of clean O₃ poor air over the Pacific and increased downward
20 transport from the upper troposphere/stratosphere in the Indonesian region will also affect the
21 tropospheric O₃ columns. In October the peak in specific humidity is located south of the
22 ozone enhancement. This agrees with Nassar et al. (2009) who showed that the equatorial
23 component of the October ozone anomaly was related to fire emissions, while the southern
24 component of the ozone anomaly was due to other factors. It should be noted that the positive
25 specific humidity anomalies over the Arabian peninsula and over Australia in October do not
26 correspond to decreased ozone values, while the ones over southern Africa, South America
27 and the Central Pacific do. The reason for this is that relative anomalies are shown and that
28 the absolute humidity values over the Arabian peninsula and Australia are much lower than in
29 the other areas, so that the absolute humidity changes between 2006 and 2005 are actually
30 relatively small. This all suggests that the correlation of O₃ to specific humidity is strongest in
31 tropical regions with large variability in water vapour, combined with low NO_x conditions.

1 Figure 17 shows that the TCCO anomalies over Indonesia are almost entirely emission driven,
2 in contrast to the TCO₃ anomalies seen in Figure 15. Using GFAS v1.0 emissions rather than
3 climatological GFAS emissions can explain most of the TCCO anomaly over Indonesia in
4 October, apart from two small positive dynamically induced anomalies to the east and west of
5 the maritime continent. By December, after the end of the fire season in Indonesia, the TCCO
6 anomalies have almost gone.

7 As for CO, the NO₂ and smoke aerosol anomalies are entirely emission driven (not shown).
8 For both these fields, no anomalies are seen when comparing CLIM06 and CLIM05, and the
9 anomalies seen when comparing BASE06 and BASE05 are gone by December. It is possible
10 that we would also see some dynamically induced changes in the smoke aerosols if aerosols
11 were interactive with the radiation scheme in the model runs. However, without this feedback
12 the smoke aerosol anomalies are entirely emission driven.

13 Figure 18 shows O₃-CO correlations for October 2005 and 2006 from the BASE and CLIM
14 experiments. We focus only on October as the month with the largest anomaly in the fire
15 emissions. Such correlations have been used in several studies (e.g., Kim et al., 2013,
16 Voulgarakis et al., 2011 and references therein) going back to Fishman and Seiler (1983) to
17 identify regions of photochemically produced O₃ (positive correlations) and O₃ from other
18 sources (e.g. downward transport from stratosphere) as well as O₃ loss due to chemistry or
19 deposition (negative correlations). In 2005, free tropospheric O₃-CO correlations (Figure 18 a
20 and c) show a similar distribution across the Maritime Continent with relatively weak ($r < 0.7$)
21 negative correlations extending from the Indian Ocean south of Indonesia to East and West
22 Malaysia and the Philippines, and positive correlations over northern and eastern Indonesia.
23 Slight differences in the distribution of the O₃-CO correlations between Figure 18 (a) and (c)
24 reflect differences in the fire emissions between the BASE and CLIM experiments. Larger
25 differences are seen in the distribution of free tropospheric O₃-CO correlations between
26 BASE and CLIM for the El Nino year of 2006 (Figure 18 b and d). The increased fire activity
27 in 2006 (Figures 4 and 13) gives rise to larger positive ($r > 0.7$) O₃-CO correlations extending
28 across most of Indonesia and Malaysia in the BASE experiment (Figure 18b) and reflect
29 enhanced O₃ photochemistry associated with increased emissions of O₃ precursors from the
30 fires. This agrees well with the area of increased O₃ concentrations due to fires seen in Figure
31 16 when comparing BASE06 and CLIM06. In contrast, O₃-CO correlations in the CLIM
32 experiment (Figure 18d) are generally negative over much of Indonesia and Malaysia and

1 reflect the influence of transport across the region and the lack of enhanced O₃ production
2 when the climatological fire emissions are used.

3

4 **4 Conclusions and outlook**

5 In this paper O₃, CO, NO₂ and smoke aerosol fields from the MACC reanalysis (Inness et al.,
6 2013) were used to identify the ENSO signal in tropical atmospheric composition fields,
7 concentrating on the months September to December. The MACC atmospheric composition
8 fields show a clear ENSO related anomaly signal with increased O₃, CO, NO₂ and smoke
9 aerosols over the Maritime Continent during El Niño years. O₃ also shows larger scale
10 changes with decreased tropospheric columns over the Central and Eastern Pacific and
11 increased columns over the Western Pacific and the Maritime continent that continue after the
12 end of the Indonesian fire season.

13 Two simulations were carried out with the C-IFS model to quantify to what extent the ENSO
14 signal seen in the atmospheric composition fields was due to changes in the biomass burning
15 emissions or due to dynamically induced changes, e.g. related to changes in the vertical
16 transport of O₃ from the lower troposphere and the stratosphere, and to changes of the
17 photolysis of O₃ due to changes to OH. While the CO, NO₂ and smoke aerosol changes were
18 almost entirely driven by changes in biomass burning emissions due to increased wild fires
19 over the Maritime Continent during El Niño related drought conditions, changes in
20 tropospheric O₃ were largely dynamically induced and only to a small part driven by changes
21 in the emissions. The emission driven O₃ changes were confined to the area surrounding
22 Indonesia, where enhanced photochemical O₃ production occurs under El Niño conditions
23 because of increased biomass burning activities, while the larger-scale O₃ anomalies were
24 dynamically induced.

25 Comparing simulations with daily GFAS v1.0 emissions for the years 2005 and 2006 and a
26 daily GFAS v1.0 climatology of the period 2000 to 2014 showed that tropospheric CO was
27 almost doubled in September 2006 relative to September 2005 due to increased fire emissions,
28 NO₂ was increased by 20-30 % and O₃ by about 8%. For tropospheric O₃, dynamically
29 induced changes dominated the differences between 2006 and 2005. The fire induced O₃
30 anomaly was smaller in magnitude and horizontal extent than the dynamically induced
31 changes which affected much of the Tropics. In 2006, tropospheric O₃ was increased by more
32 than 50% over the Maritime Continent and Indian Ocean compared to 2005, and decreased by

1 between 20-30 % over large parts of the Tropical Pacific when the same climatological fire
2 emissions were used in both years. Only in a small area over Indonesia was the O₃ increase
3 due to fires of similar magnitude to the dynamically induced changes. A future study will look
4 in more detail at the chemistry budgets and chemical processes that cause the changes in the
5 atmospheric composition fields.

6 The results from this paper show that the MACC system is able to model changes in
7 atmospheric composition fields found under El Niño and La Niña conditions. After a more
8 thorough validation of the MACC atmospheric fields against observations, it could be
9 interesting to investigate the ocean-atmosphere response to ENSO induced changes in
10 atmospheric composition in a further study. A first step would be to include the aerosol direct
11 and indirect effects through the cloud microphysics in the radiation scheme of the IFS and to
12 look at the feedback of fire-induced aerosols on climate. We would expect a positive
13 feedback, i.e. reduced convection due to increased atmospheric stability, as carbonaceous
14 aerosols usually absorb (and thus re-emit) a significant amount of solar radiation in the mid
15 troposphere, and increased aerosol concentrations also lead to reduced land and sea surface
16 temperatures. Their presence should therefore act to reduce convection and precipitation over
17 the Maritime Continent. Including the aerosols in the radiation scheme will also affect the
18 chemical fields through changes in the UV radiation and hence photolysis rates. A second
19 step could see the coupling of the chemistry and aerosol fields by including heterogeneous
20 chemistry on aerosols. In a final step it can be envisaged to fully couple the MACC system
21 with ECMWF's ocean model to investigate how the forcing from ENSO induced changes to
22 atmospheric composition fields can feedback on the ENSO dynamics.

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

25

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1 **5 Tables**

2

3 Table 1: Biomass burning emissions in Tg per month for CO, NOx, and the smoke aerosols
 4 (sum of organic matter and black carbon) from GFAS v1.0 for the region 10°N - 10°S, 90°E -
 5 130°E averaged over the El Niño years (2004, 2006, 2009) and the La Niña years (2005,
 6 2007, 2008, 2010, 2011), as well as the ratio of the El Niño/ La Niña values.

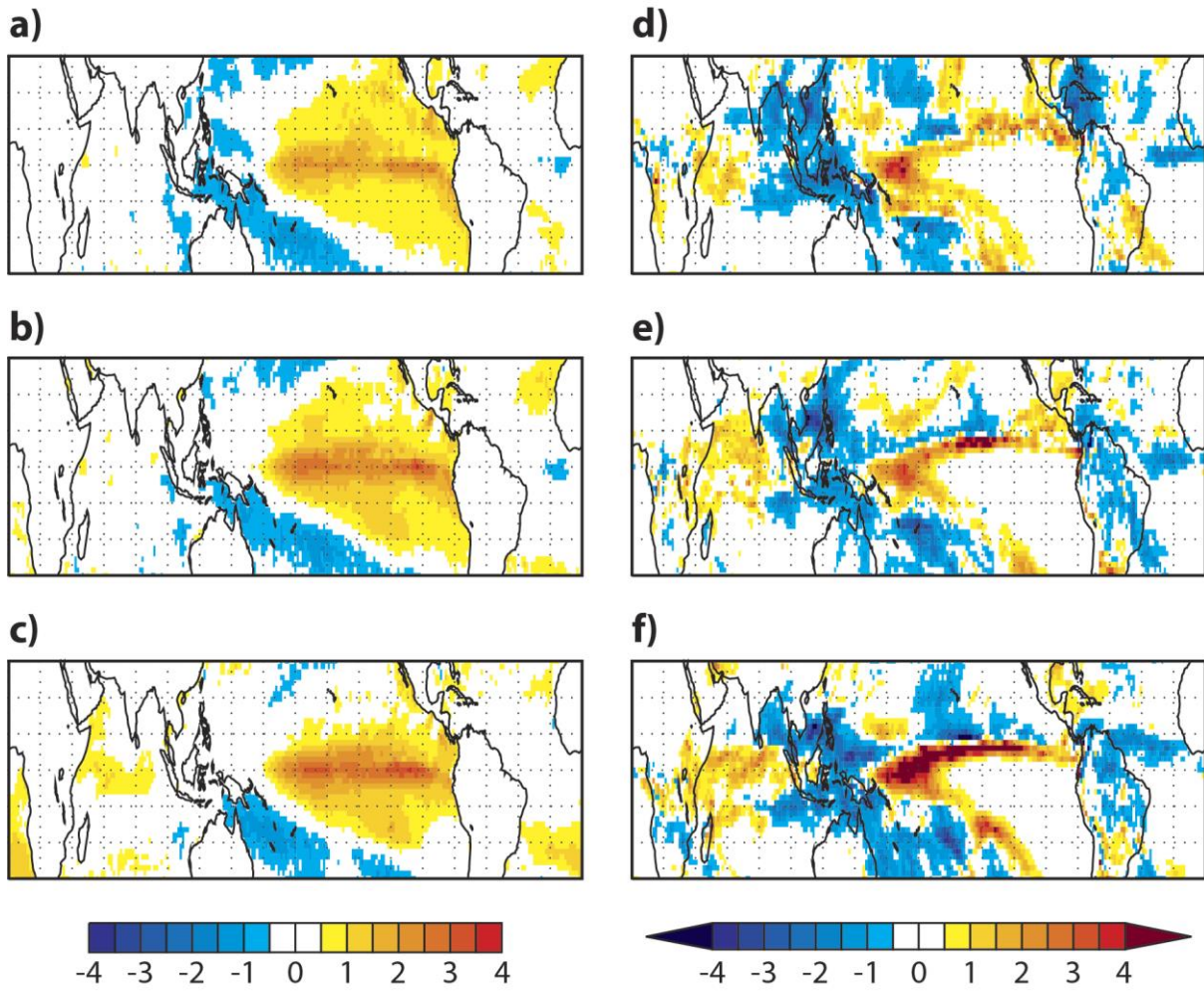
	CO	CO	CO	NOx	NOx	NOx	OM+BC	OM+BC	OM+BC
	Oct	Nov	Dec	Oct	Nov	Dec	Oct	Nov	Dec
El Niño	13.9	3.0	2.4 10 ⁻¹	8.4 10 ⁻²	1.9 10 ⁻²	2.8 10 ⁻³	6.3 10 ⁻¹	1.4 10 ⁻¹	1.3 10 ⁻²
La Niña	1.5	2.5 10 ⁻¹	1.4 10 ⁻¹	1.4 10 ⁻²	3.0 10 ⁻³	1.5 10 ⁻³	7.3 10 ⁻²	1.4 10 ⁻²	7.3 10 ⁻³
El Niño/ La Niña	9.5	11.8	1.7	6.1	6.3	1.9	8.6	10.0	1.8

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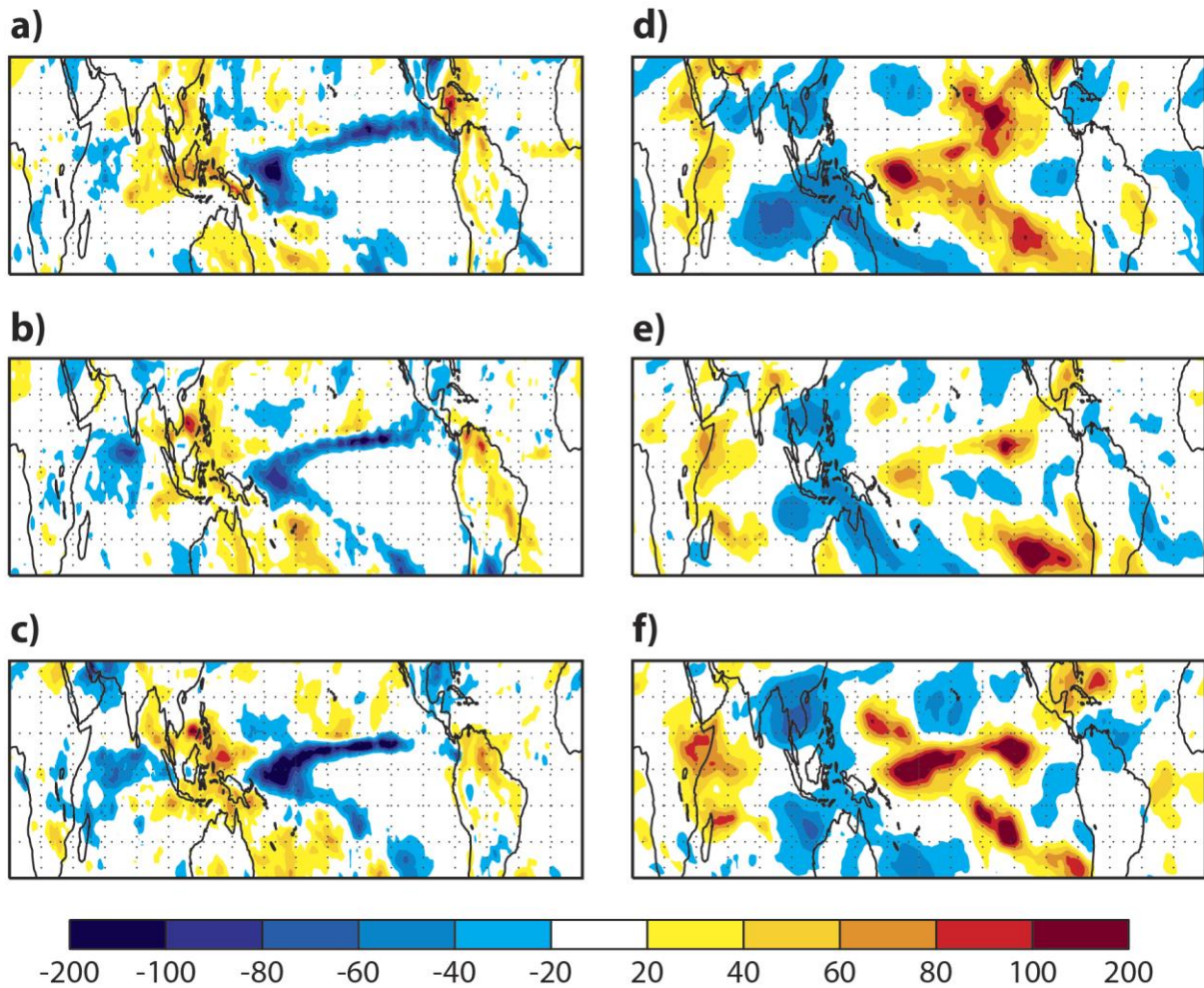
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2 **6 Figures**



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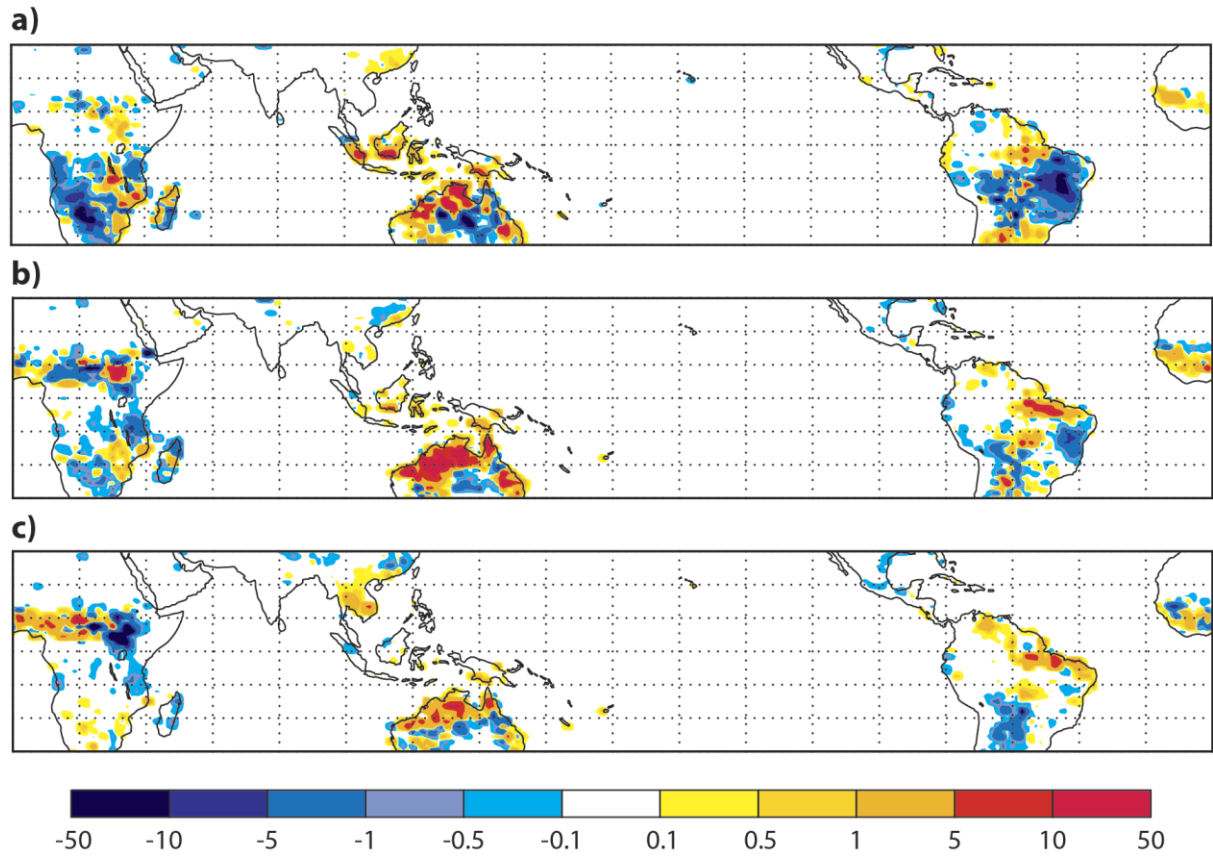
4 Figure 1: Left panels: SST anomaly in K calculated from ERA Interim as the difference of El
5 Niño composite minus La Niña composite for October (a), November (b) and December (c).
6 Right panels: Precipitation anomaly in mm/day calculated from ERA Interim as the difference
7 of El Niño composite minus La Niña composite for October (d), November (e) and December
8 (f). Red colours indicate positive values, blue colours negative values.



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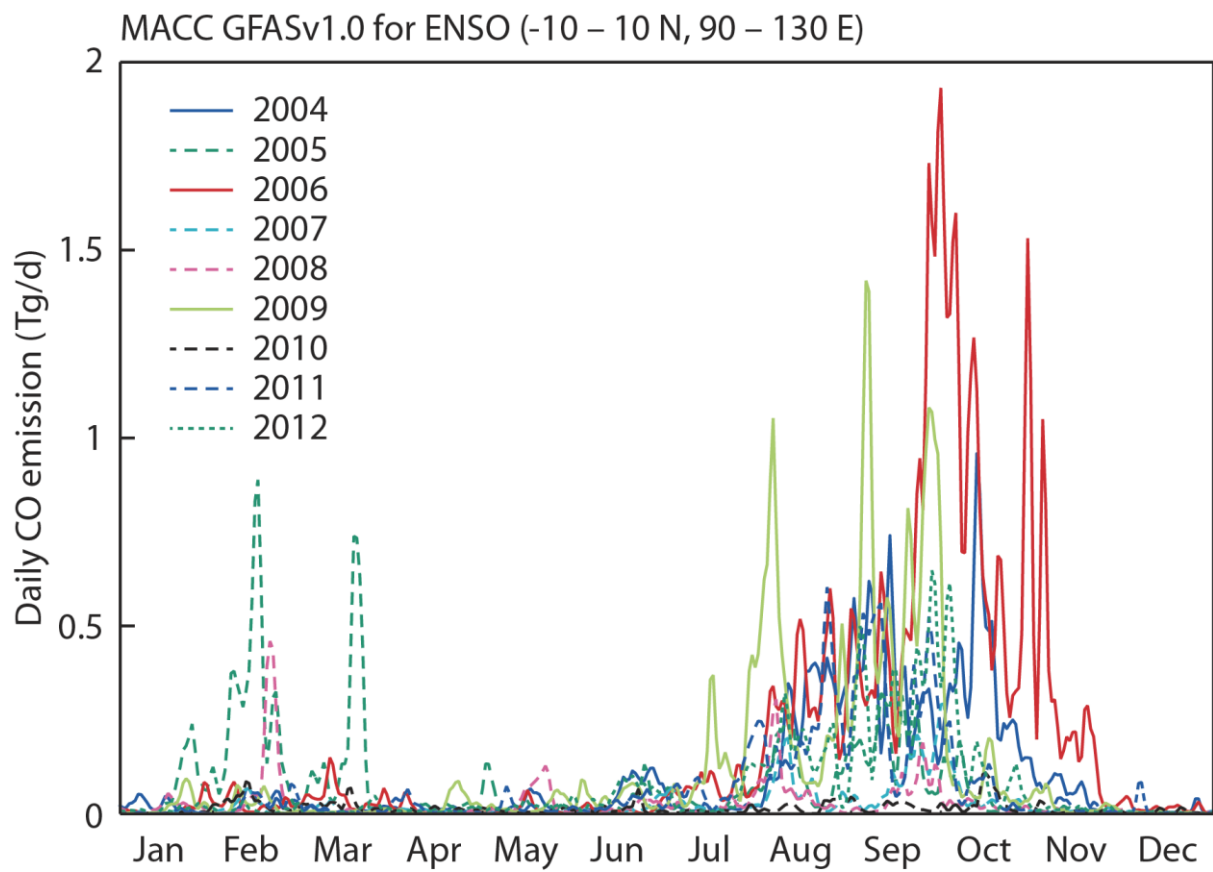
Figure 2: Left panels: Anomaly of vertical velocity at 500 hPa in mm/s calculated from the MACC reanalysis as the difference of El Niño composite minus La Niña composite for October (a), November (b) and December (c). Blue colours show increased ascent, red colours increased descent. Right panels: Specific humidity anomaly at 500 hPa in % calculated from the MACC reanalysis as the difference of El Niño composite minus La Niña composite for October (d), November (e) and December (f). Blue colours show reduced specific humidity, red colours increased values.

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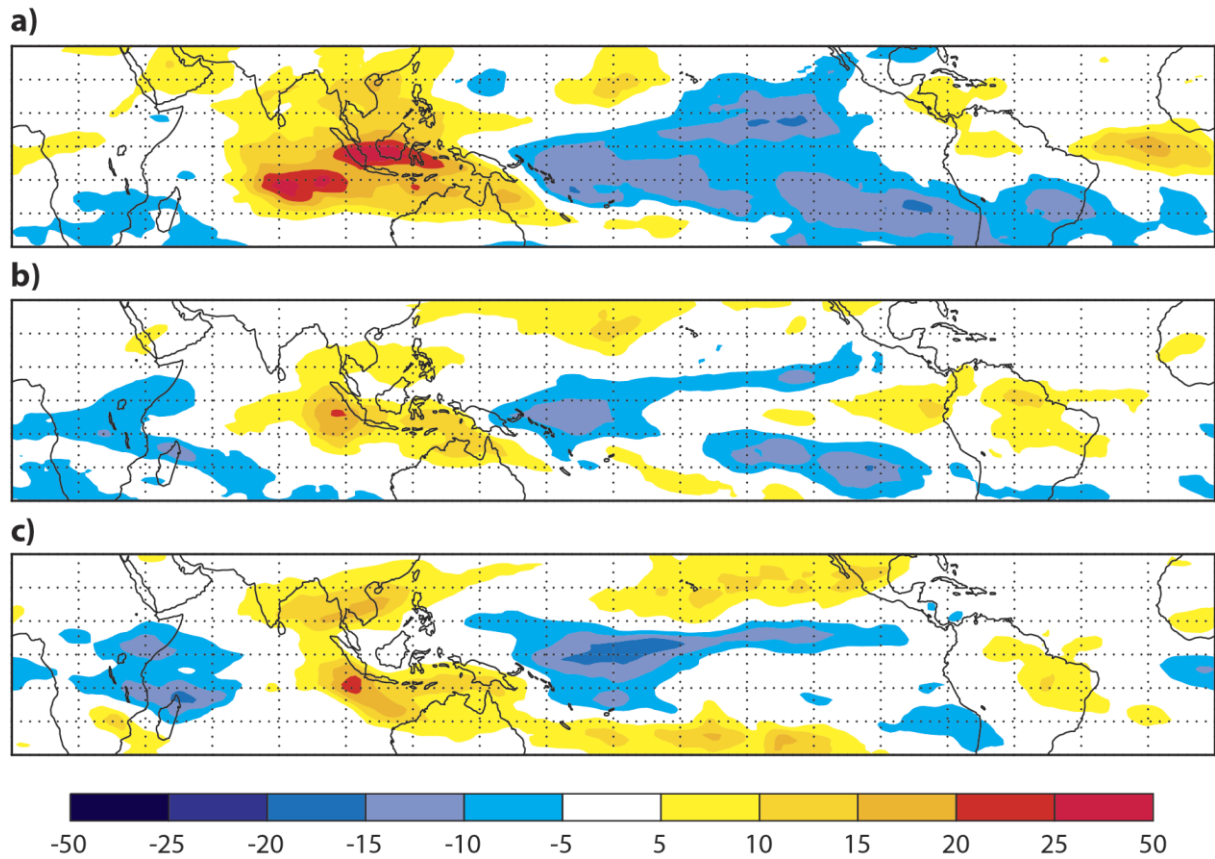
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Figure 3: Biomass burning (fire radiative power areal density) anomaly in mW/m^2 calculated from the GFAS v1.0 dataset as the difference of El Niño composite minus La Niña composite for October (a), November (b) and December (c). Red colours indicate positive values, blue colours negative values.



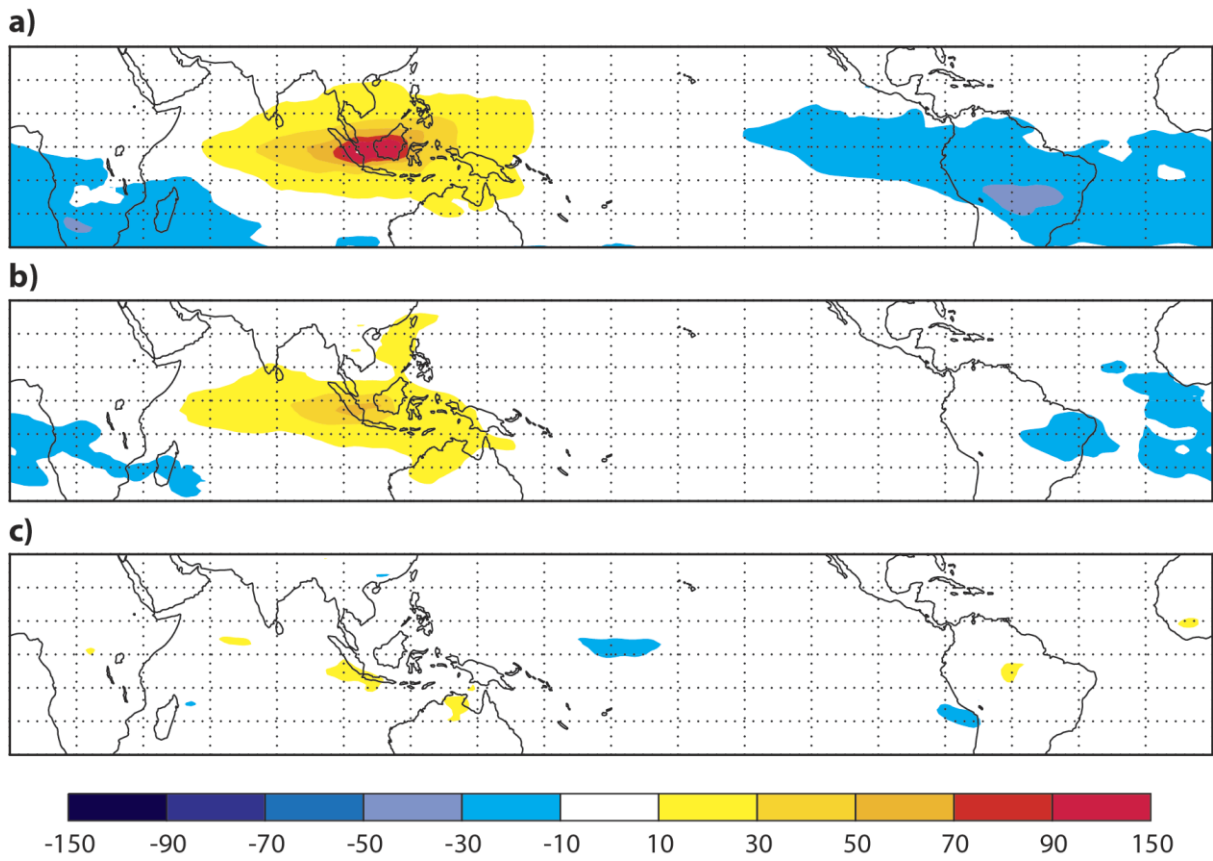
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2 Figure 4: Timeseries of daily CO emissions in Tg (10^{12} g) per day from GFASv1.0 for the
3 region $10^{\circ}\text{N} - 10^{\circ}\text{S}$, $90^{\circ}\text{E} - 130^{\circ}\text{E}$ for the years 2003 to 2012.
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3 Figure 5: O₃ anomaly at 500 hPa in ppb calculated from the MACC reanalysis as the
4 difference of El Niño composite minus La Niña composite for October (a), November (b) and
5 December (c). Red colours indicate positive values, blue colours negative values.

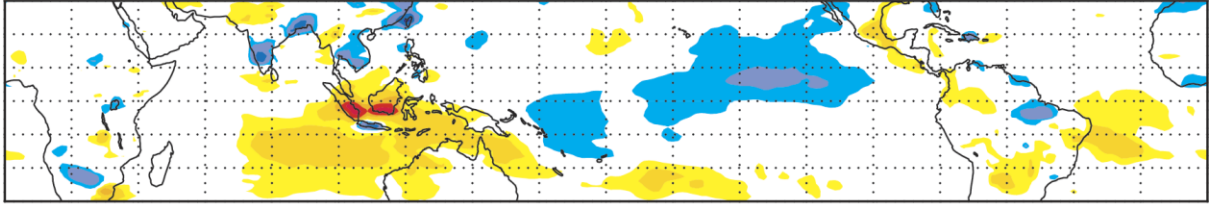


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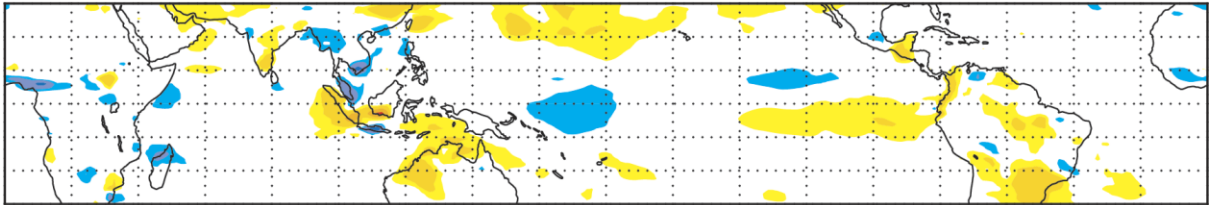
Figure 6: Like Figure 5 but for CO anomaly at 500 hPa in ppb.

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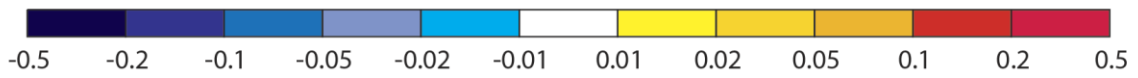
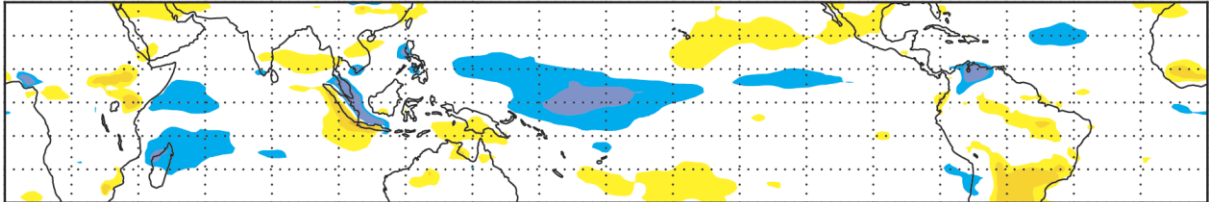
a)



b)



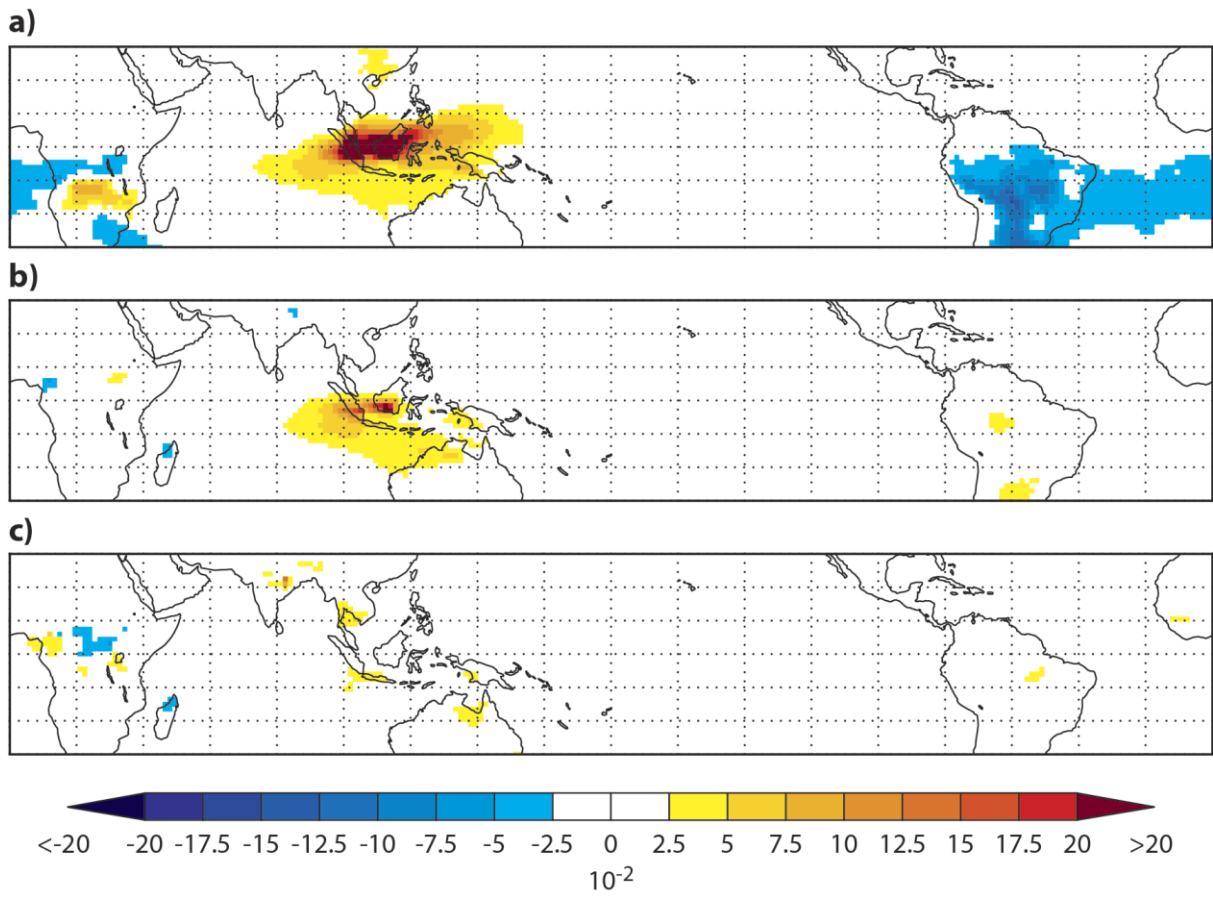
c)



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3 Figure 7: Like Figure 5 but for NO_x anomaly in ppb.

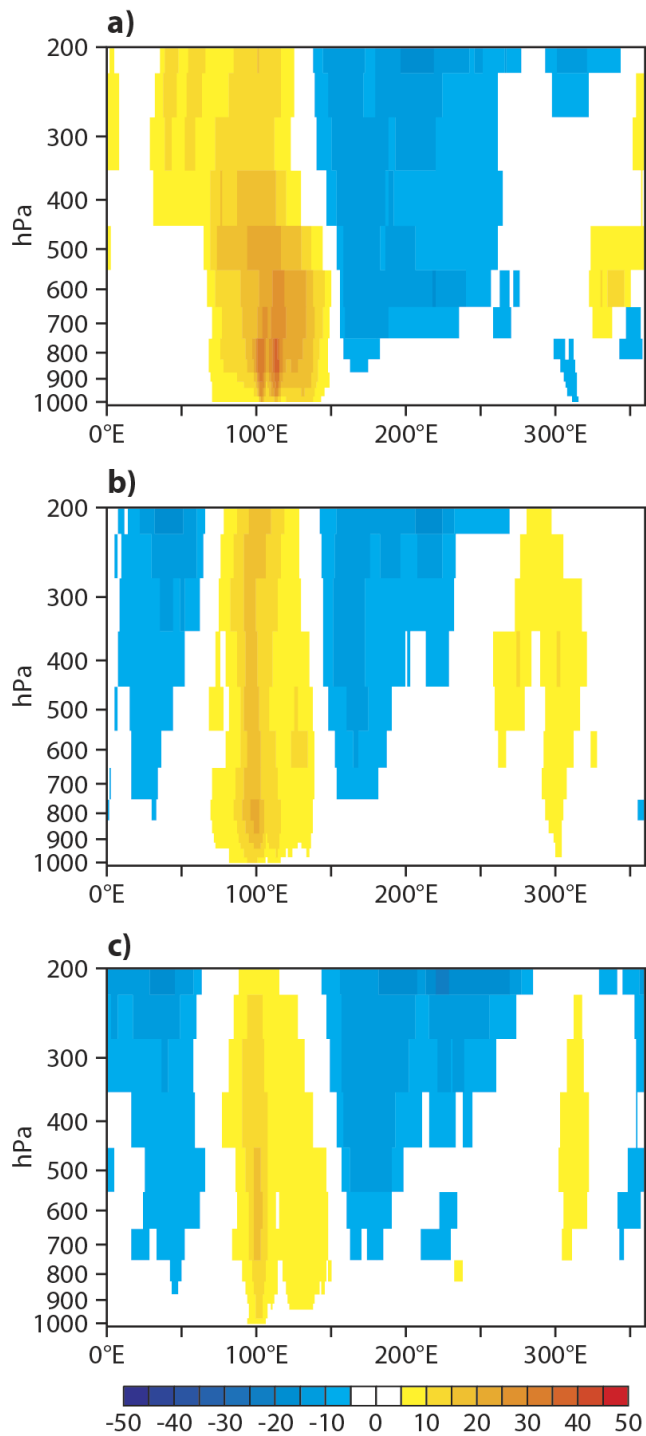
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2 Figure 8: Like Figure 5 but for smoke AOD (BC+OM). AOD is unitless.

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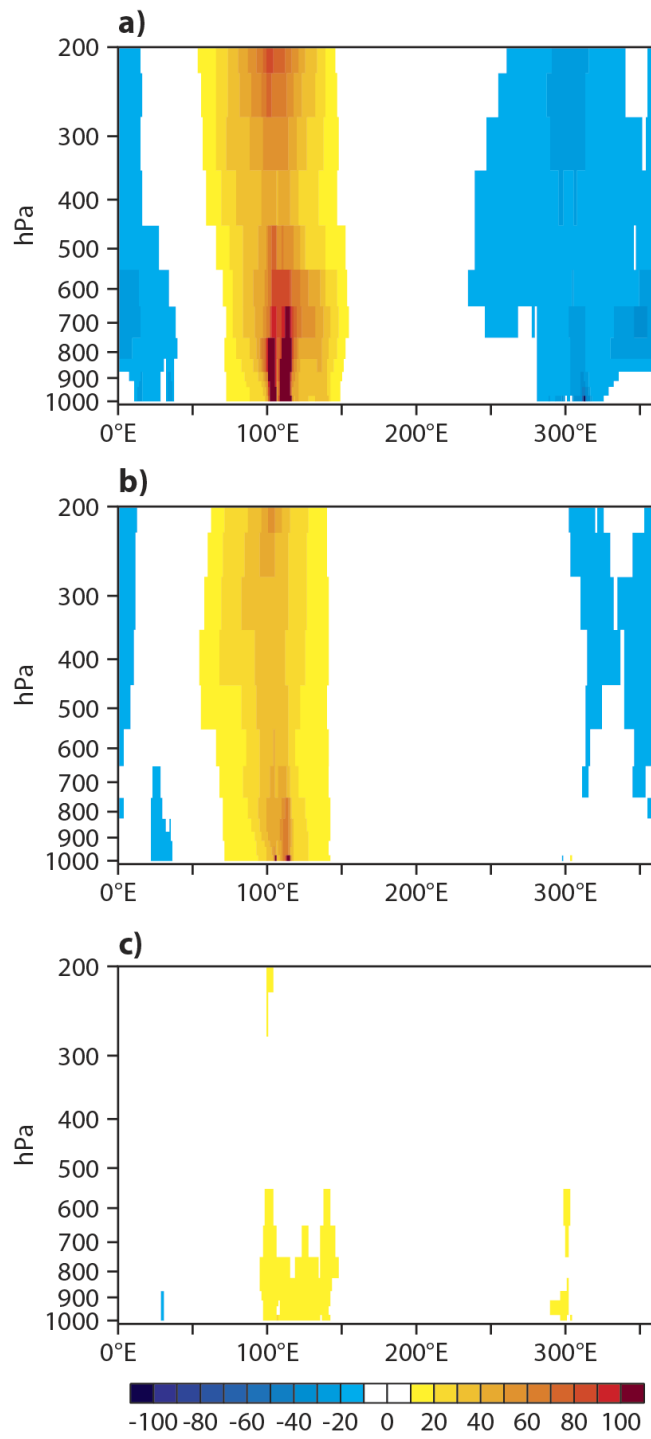


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2 Figure 9: Vertical cross section of O₃ anomalies in ppb plotted against longitude and averaged
 3 between 0° and 12°S calculated from the MACC reanalysis as the difference of El Niño
 4 composite minus La Niña composite for October (a), November (b) and December (c). Red
 5 colours indicate positive values, blue colours negative values.

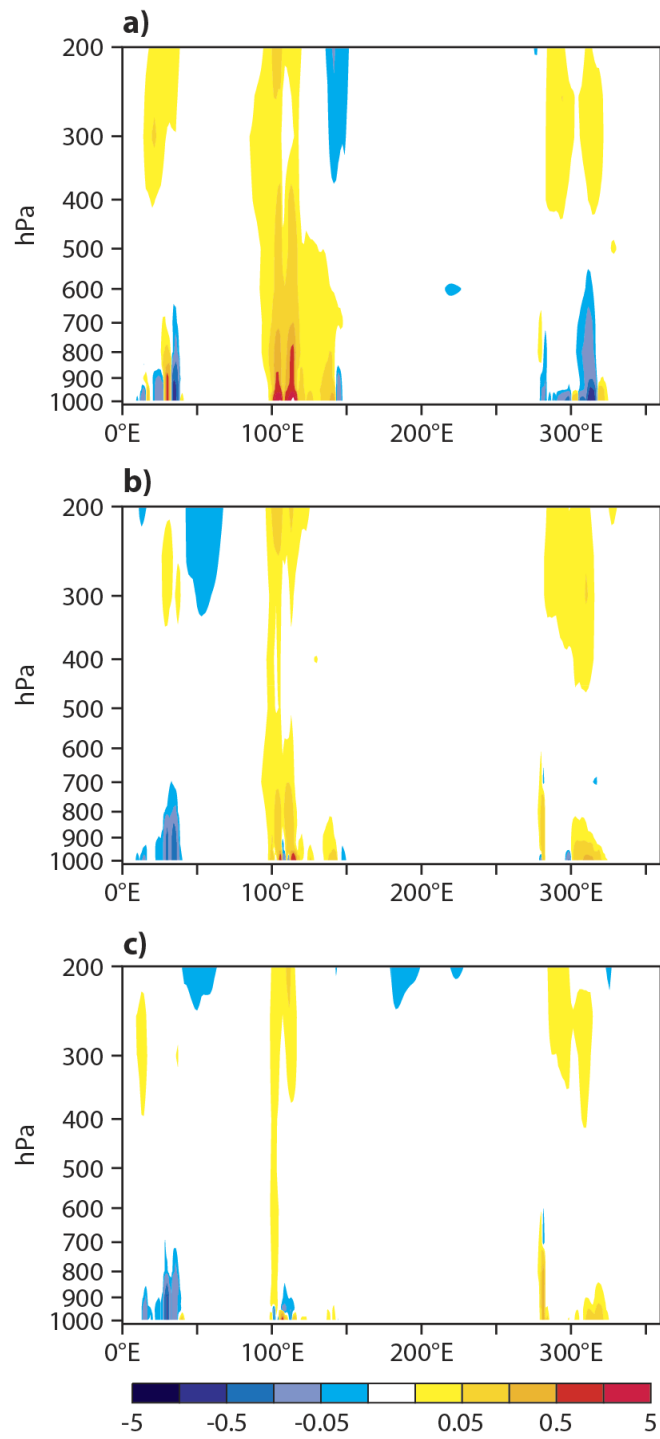
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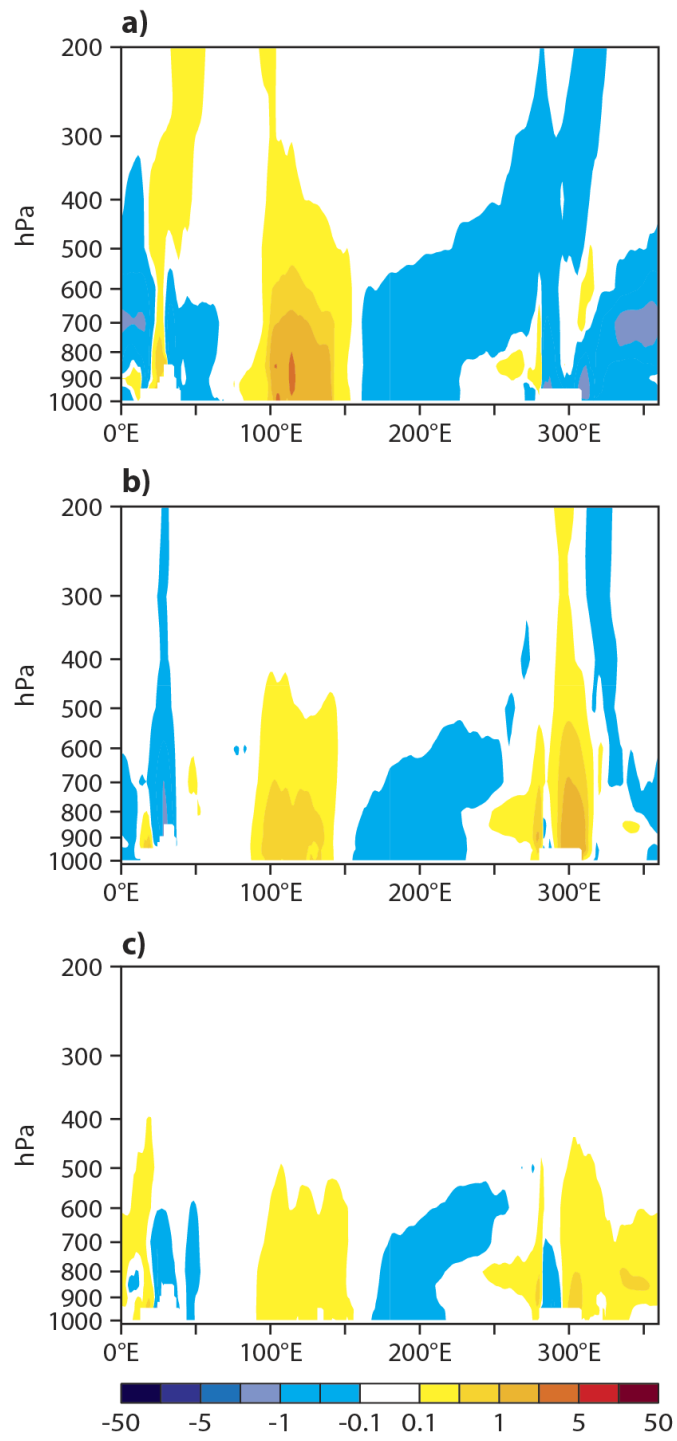
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2 Figure 10: Like Figure 9 but for CO in ppb.



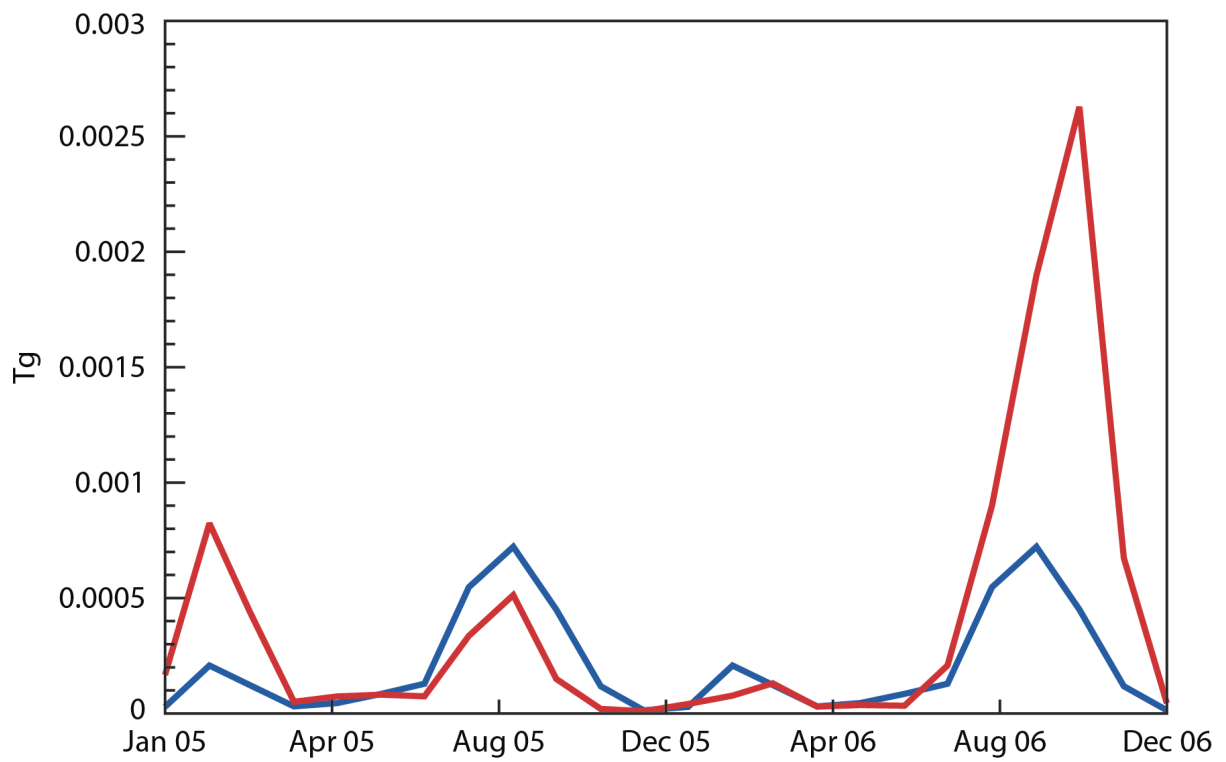
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2 Figure 11: Like Figure 9 but for NO_x in ppb.

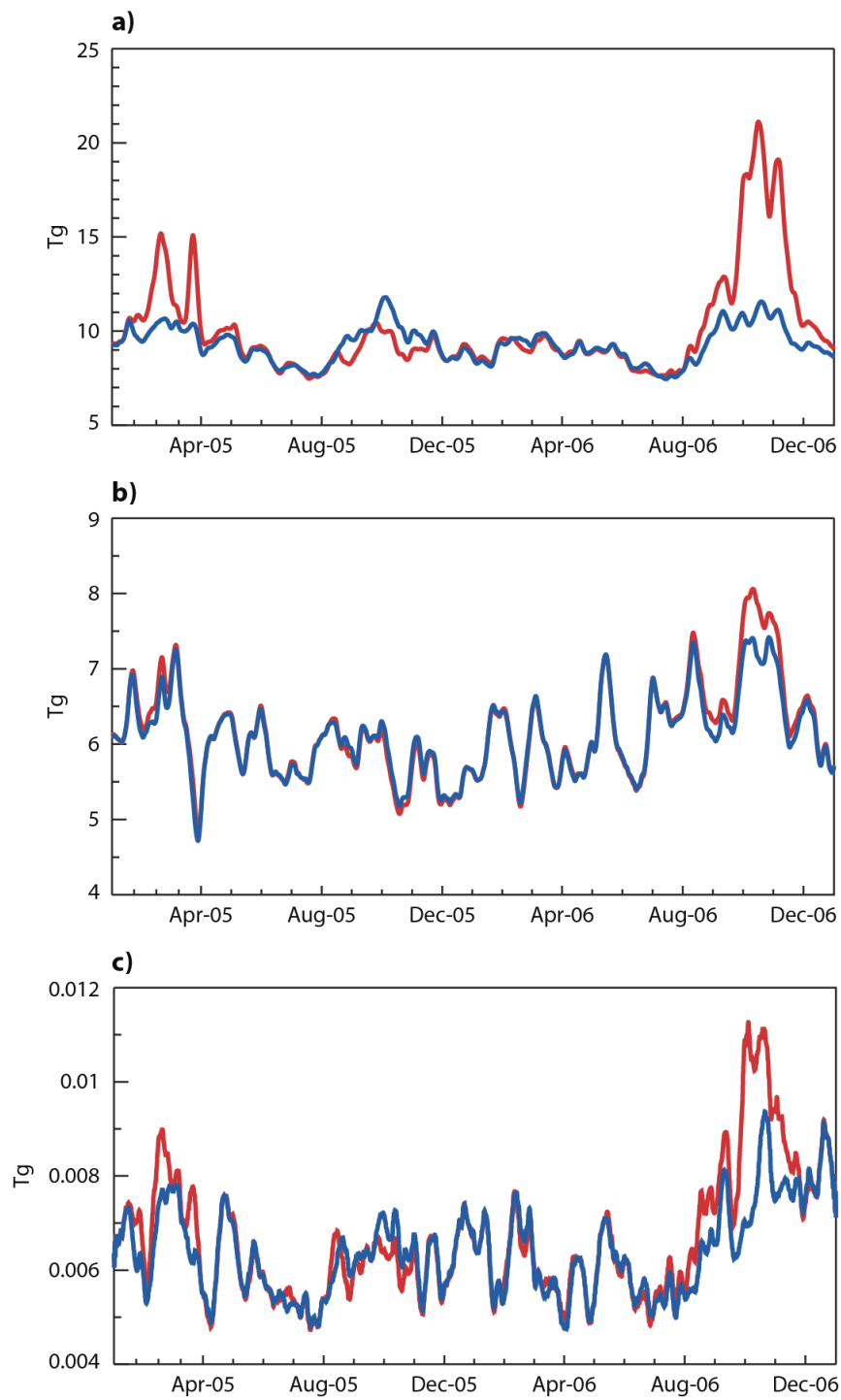


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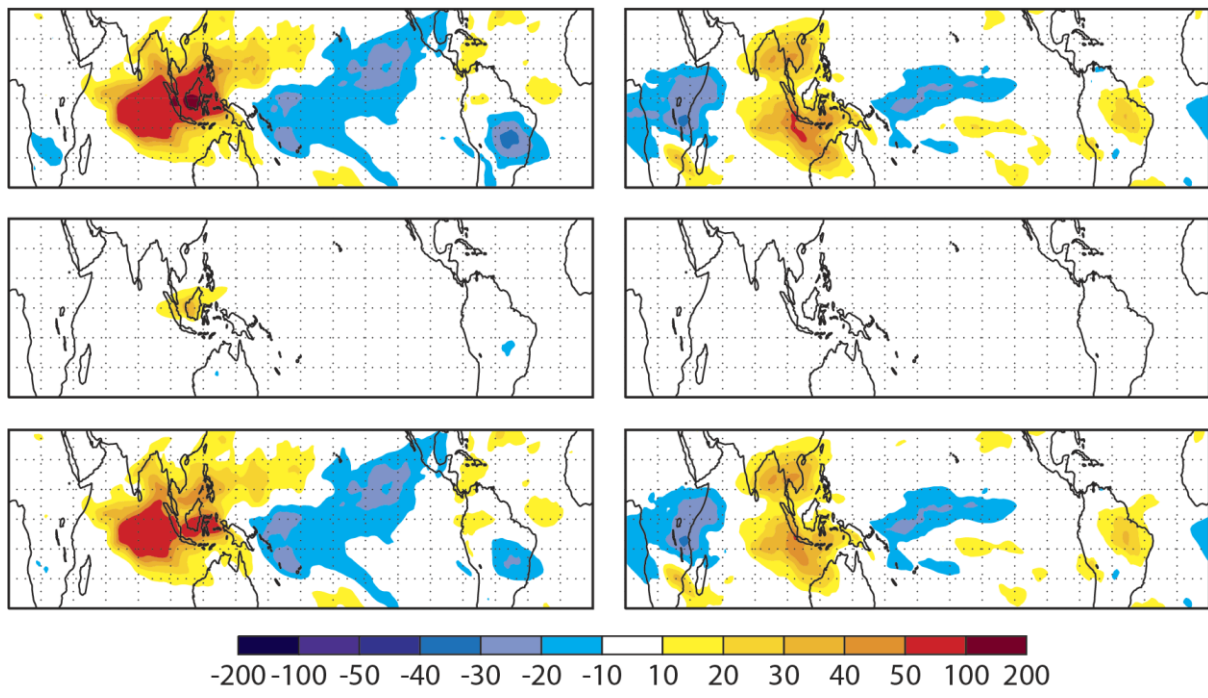
2 Figure 12: Like Figure 9 but for smoke aerosol in ppb.



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 2 Figure 13: Timeseries of CO biomass burning emissions in Tg averaged over the region
 3 between 10°N, 10°S, 90°E, 130°E from GFAS v1.0 (red) and climatological GFAS v1.0 data
 4 set (blue.)



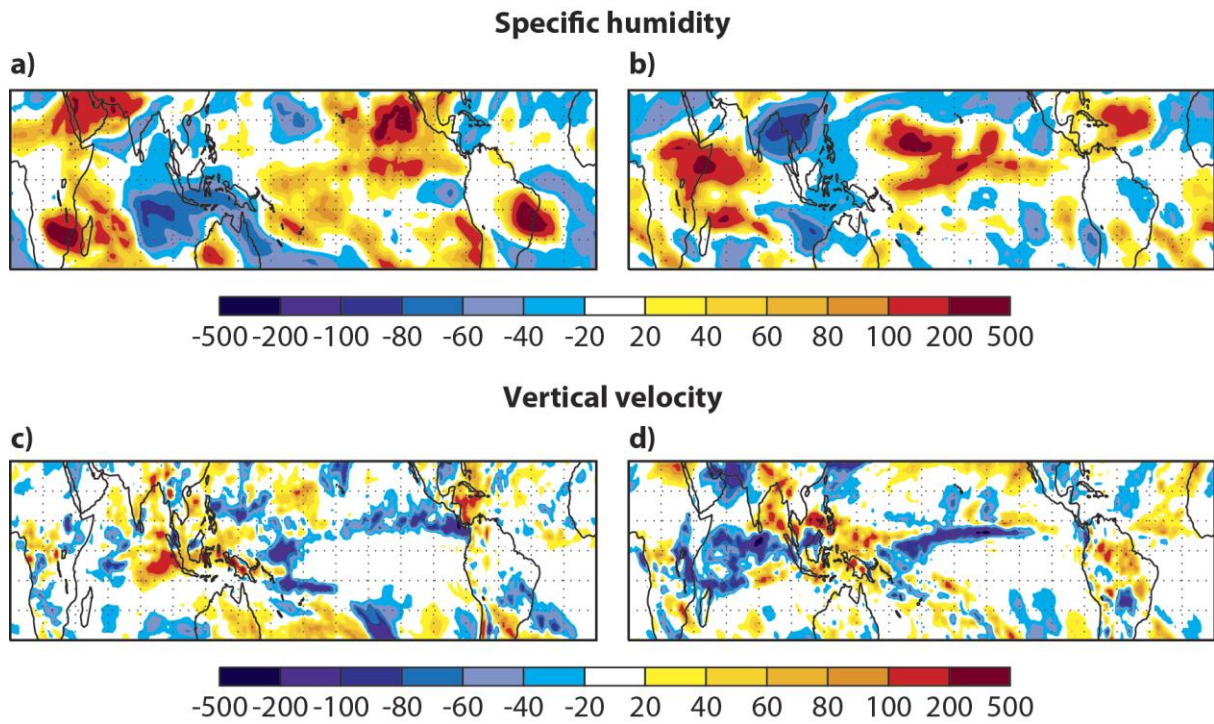
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 2 Figure 14: Timeseries of the tropospheric CO (a), O₃ (b), NO₂ (c) burden in Tg from BASE
 3 (red) and CLIM (blue) for 2005 and 2006 averaged over the area between 10°N, 10°S, 90°E,
 4 130°E.



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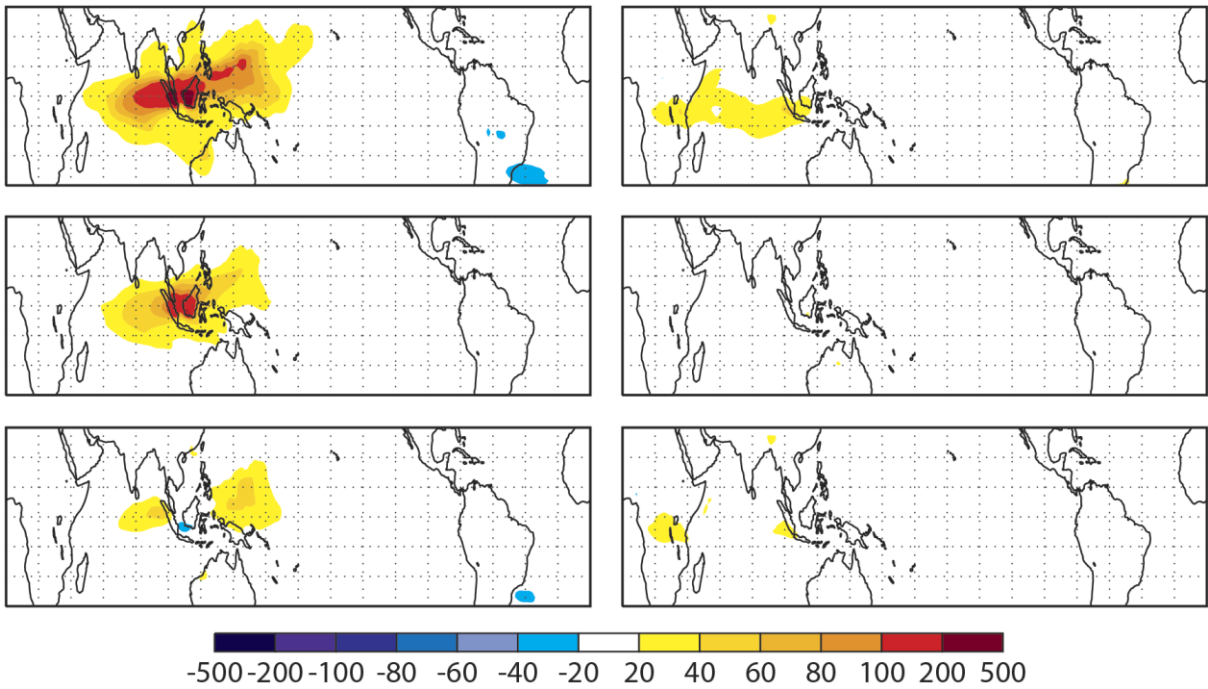
2 Figure 15: TCO3 differences in % for October (left) and December (right) from the
 3 experiments BASE06 - BASE05 (top), BASE06 - CLIM06 (middle) and CLIM06 - CLIM05
 4 (bottom). The top panels show the overall differences of TCO3 due to the combined effects of
 5 El Niño related dynamical changes and changes in the fires emissions between El Niño and
 6 normal conditions. The middle panels show the impact of changes to the fire emissions under
 7 El Niño conditions, and the bottom panels show the impact of the El Niño induced dynamical
 8 changes on TCO3 when climatological fire emissions are used for both years. Red colours
 9 indicate positive values, blue colours negative values.

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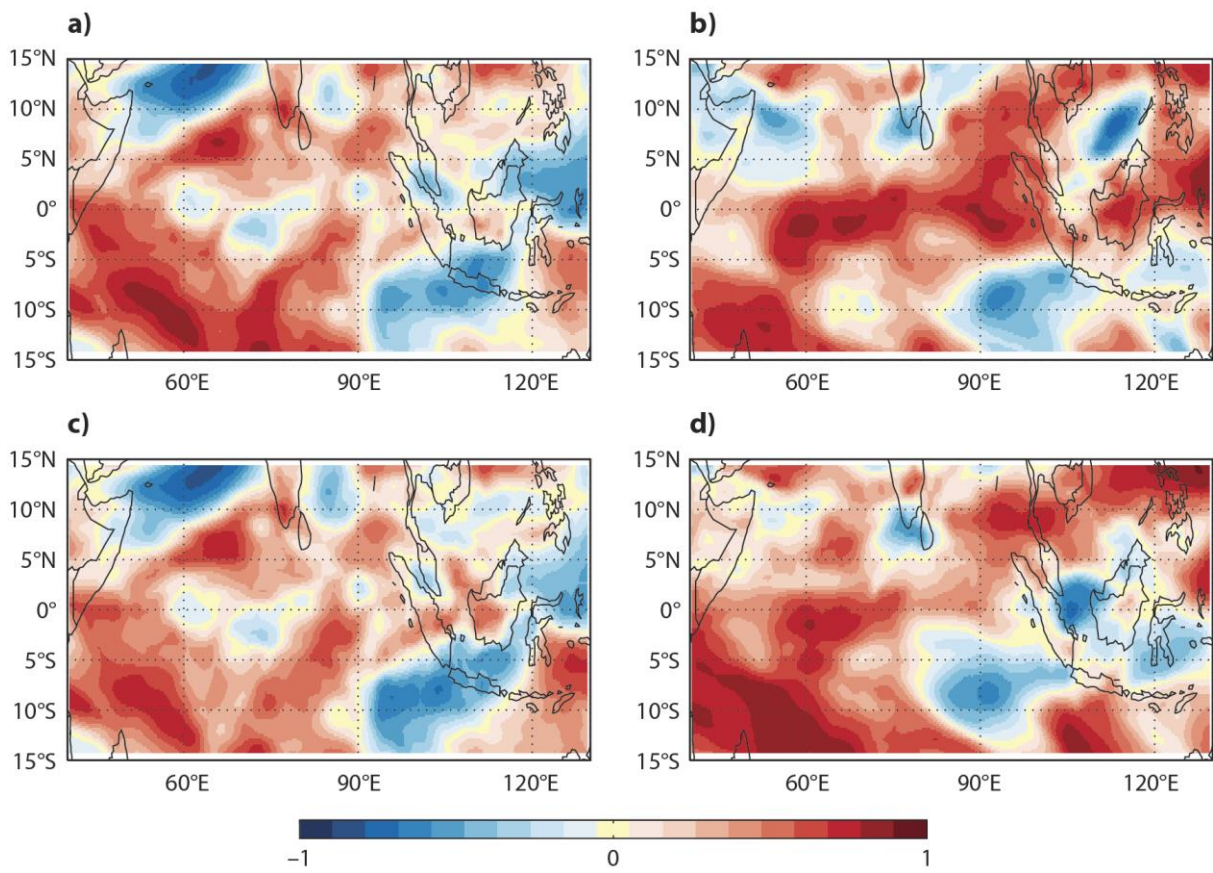
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 2 Figure 16: Top panels: Specific humidity differences at 500 hPa in % for October (a) and
 3 December (b) from the experiments BASE06 minus BASE05. Blue colours show reduced
 4 specific humidity, red colours increased values. Bottom panels: Differences of vertical
 5 velocity in mm/s for October (c) and December (d) from the experiments BASE06 minus
 6 BASE05. Blue colours show increased ascent, red colours increased descent.

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 2 Figure 17: TCCO differences in % for October (left) and December (right) from the
 3 experiments BASE06 - BASE05 (top), BASE06 - CLIM06 (middle) and CLIM06 - CLIM05
 4 (bottom). Red colours indicate positive values, blue colours negative values.
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3 Figure 18: October O₃-CO correlations calculated for free tropospheric (approx. 750-350 hPa)
4 column abundances over the Maritime Continent from the BASE (top) and CLIM (bottom)
5 experiments for 2005 (left) and 2006 (right). Red colours indicated positive correlations, blue
6 colours negative ones.

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