

Large enhancements in southern hemisphere satellite-observed trace gases due to the 2019/2020 Australian wildfires

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Key Points

- Satellite-retrieved carbon monoxide (CO) plumes from the Australian fires circumvent the Southern Hemisphere.
- Satellite-retrieved methanol (CH₃OH) shows downwind enhancement of CH₃OH:CO ratio suggesting in-plume secondary CH₃OH production as well as direct emission.

Key Words

Australian Fires; IASI; Carbon Monoxide; Methanol; Methane

Abstract:

The 2019/2020 Australian wildfires emitted large quantities of atmospheric pollutant gases and aerosols. Using state-of-the-art near-real-time satellite measurements of tropospheric composition, we present an analysis of several emitted trace gases and their long-range transport, and compare to the previous (2018/2019) fire season. Observations of carbon monoxide (CO) show that fire emissions were so intense that the distinct Australian fire plume managed to circumnavigate the Southern Hemisphere (SH) within a few weeks, with eastward propagation over the South Pacific,

South America, the South Atlantic, Africa and the Indian Ocean. Elevated atmospheric methane levels were also detected in January 2020 fire plumes over the Pacific, defined using CO as a plume tracer, even though sampling was restricted spatially by aerosols and clouds. Observations also show significant enhancements of methanol from the fires, where CH₃OH:CO enhancement ratios increased within the aged plume downwind over the South Pacific indicating secondary in-plume CH₃OH formation.

1. Introduction

Vegetation fires occur regularly in Australia between the months of August and December (Giglio et al., 2013; van der Werf et al., 2017). Burning activity predominantly occurs in northern Australia, but is widespread across the continent (Andela et al., 2017). Giglio et al., (2013) suggested that the majority of vegetation fires take place on savanna and shrubland, but in south-eastern Australia forest fires are most prevalent (Bradstock et al., 2012; van der Werf et al., 2010). Over recent decades, there have been large-scale decreases in Australian fire activity (Andela et al., 2017; Rabin et al., 2015). However, with present and future climate and land-use change, conditions in Australia are predicted to yield more frequent large-scale fire events (Pitman et al., 2007; Clarke et al., 2011; Di Virgilio et al., 2019). According to the Australian Bureau of Meteorology (2020), the 2019 summer was the warmest (1.52°C above the national average, 1961-1990) and driest (rainfall 40% lower than average) season on record. This provided suitable conditions for wildfires to ignite and spread.

The Australian wildfires of the 2019/2020 fire season, colloquially known as the “black summer”, represented some of the largest events in recent decades. The fires burned over 110,000 km² of bush, forest and parks (BBC, 2020). The majority of the fire activity occurred in south-eastern Australia (New South Wales and Victoria), which is predominantly eucalyptus forest and woodland (SOTE, 2016). The fires caused 33 deaths (BBC, 2020) and killed over approximately 1 billion animals (UoS, 2020). In comparison, the Black Saturday fires (February 2009) in Victoria burned approximately 4500 km² and killed 173 people (Siddaway and Petelina, 2011). Though the 2019/2020 fire death toll was lower, the burned area was much larger producing substantial quantities of smoke and pollutants.

Vegetation fires emit large quantities of smoke/aerosols and trace gases, which have important impacts on climate and the atmospheric radiation balance (Li et al., 2017, Rowlinson et al., 2019) and surface air quality (AQ, Bowman and Johnston, 2005; Haikerwal et al., 2016; Reisen et al., 2005; Kiely et al., 2019). The 2019/2020 Australian fires emitted approximately 250 million tonnes of carbon dioxide (CO₂), equivalent to nearly half the country’s annual anthropogenic emissions (Hope, 2020). This combination of fuel type, fire intensity and coverage, yielded large-scale fire plumes causing intense local pollution as well as long-range pollution transport. Wildfire-driven pyro-convection propagated vertically up into the stratosphere, reaching approximately 30 km (Ohneiser et al., 2020) over the South Pacific. In comparison, the Black Saturday fire plume reached 22 km (Siddaway and Petelina, 2011).

In this study, we use state-of-the-art satellite retrievals to provide detailed analysis of the spatial and temporal evolution of several trace gas distributions sourced from the 2019/2020 Australian fires. While available for the stratosphere, such capabilities were in their infancy when previous major burning events occurred in Australia (e.g. February 2009).

2. Observations

2.1 Fire Data Sets

We use two different satellite-derived fire activity datasets: fire radiative power (FRP) from the Global Fire Assimilation System (GFAS vn1.2; Kaiser et al., 2012) and burned area (BA) from the Fire INventory from NCAR (FINN near-real-time (NRT) vn1.0; Wiedinmyer et al., 2011). Both products are provided at a daily temporal resolution and are based on direct Moderate Resolution Imaging Spectroradiometer (MODIS) measurements (e.g. FRP and thermal anomalies). These quantities are merged with secondary information (including land surface type and emission factors) to derive top-down emissions for trace gases and aerosols (Kiely et al., 2019; Wooster et al., 2018).

2.2 Trace Gas Data Sets

In this study we use trace gas retrievals from the Infrared Atmospheric Sounding Interferometer (IASI). IASI is a Michelson interferometer which observes the spectral range 645 to 2760 cm^{-1} with spectral sampling of 0.25 cm^{-1} (Illingworth et al., 2011). It measures simultaneously in four fields of view (FOV, each circular at nadir with a diameter of 12 km) which are scanned across track to sample a 2200 km-wide swath (Clerbaux et al., 2009). IASI is one of a suite of nadir-sounders flying on Eumetsat's MetOp-A, -B and -C satellites in sun-synchronous polar orbits with equator crossing times of 9.30 (day) and 21.30 (night). Here we use CO, methanol (CH_3OH) and methane (CH_4) data from MetOp-B produced by NRT processing systems developed by the Rutherford Appleton Laboratory (RAL). CO profiles are co-retrieved with column amounts of CH_3OH , other trace gases and dust in an extended version of RAL's Infrared-Microwave-Sounding (IMS) scheme, which is described in the supplementary material (**SM-2**). IMS was developed originally to retrieve temperature, water vapour, ozone, surface spectral emissivity and cloud jointly from co-located measurements by IASI, the Microwave Humidity Sounder (MHS) and the Advanced Microwave Sounding Unit (AMSU-A) on MetOp. CH_4 data are retrieved by an improved version of the IASI scheme reported by RAL Space (2015), which is detailed by Siddans et al. (2017). Data are available from the Centre of Environmental Data Analysis (CEDA, Siddans et al., 2020). CO and CH_3OH have been quality filtered for a geometric cloud fraction of 0.5 or less (0.1 or less for CH_4 given the greater sensitivity to interference from cloud/aerosol) and a cost value of 1000.0 or less (120 or less for CH_4 plus a convergence flag equally 1.0). For CO and CH_3OH , we experimented with a stricter geometric cloud fraction threshold of <0.2 . We found this had negligible impact on the scientific results, but did reduce the spatial coverage, making the fire signals noisier. Hence we used the <0.5 cloud fraction threshold. We have also investigated more localised enhancements in tropospheric column nitrogen dioxide (TCNO_2) from the Tropospheric Monitoring Instrument (TROPOMI) on-board ESA's Sentinel 5 – Precursor (S5P) satellite, which is discussed in the SM (see **SM-1**).

3. Results

3.1 Fire Activity

During the 2019/2020 fire season (November-December-January, NDJ), satellite observations detected substantially larger fire activity on the Australian south-eastern coastline around highly populated regions such as Sydney compared to the average of the previous 10 seasons (NDJ 2009-2019 climatology). GFAS FRP suggests that on average (NDJ climatology) (**Figure 1a**) there was limited fire activity over south-eastern Australia (i.e. 10-20 mW/m^2). The peak activity was more widespread across the north-western territories with FRP typically between 20 and 30 mW/m^2 . However, in NDJ 2019/2020 the entire Australian south-eastern coastline experienced large-scale fires with intensities well above 50 mW/m^2 (**Figure 1b**). This is supported by the FINN BA (**Figure 1c & d**) with fire events peaking above 10 km^2 widespread across the south-east coast. In contrast, the

fire events in the NDJ climatology predominantly occurred over the north-western territories and northern coastlines, with lower BA between 3.0 and 7.0 km². **Figure 1e** shows the time-series of the total daily Australian FRP and BA for the climatological (median daily totals, 2009-2019) and 2019/2020 fire seasons. Between 1st November and 31st January, the climatological FRP has a small range (i.e. 25th-75th percentiles) between approximately 0 and 10 GW. Climatological BA ranges between approximately near-zero and 1700 km². However, the 2019/2020 FRP (BA) is typically between 20 (100) and 150 (1700) GW (km²), but with peaks in late December (FRP=~320 GW, BA=~2200 km²) and early January (FRP=~580 GW, BA=~3000 km²). The peak FRP and BA values sit well outside the variability of the climatologies, highlighting the extreme fire activity experienced in the 2019/2020 fire season.

3.2 Carbon Monoxide

The 2019/2020 Australian fire season produced extensive quantities of emitted CO, as observed by IASI. **Figure 2c** shows a large total-column CO (TCCO) plume originating over south-eastern Australia and propagating across the entire Pacific, reaching South America. Here, the TCCO ranges between approximately 16-21 ×10¹⁷ molecules/cm², peaking over the Australian coastline and midway between continents in the NDJ period 2019/2020. In the previous burning season (NDJ 2018/2019), TCCO was considerably lower and ranged between 10-13 ×10¹⁷ molecules/cm², with no obvious fire signal over Australia (**Figure 2a**). **Figure 2e** shows the difference between fire seasons to be widespread and large (1.5-7.5 ×10¹⁷ molecules/cm²) across the Pacific. Peak inter-year differences of over 5.0 ×10¹⁷ molecules/cm² occur over south-eastern Australia and the mid-Pacific, as the plume propagates eastwards. The inter-year differences are significant over large areas (99% confidence level based on the student t-test and where absolute mean differences are greater than 1.0 ×10¹⁷ molecules/cm²) as shown by the green polygon-outlined regions.

During the 2018/2019 season, fire plumes of limited extent were detected, with moderate CO outflow from the east coast in November 2018 peaking at approximately 19.0 ×10¹⁷ molecules/cm² (**Figures 2b & d**). In NDJ 2019/2020, there is large and frequent CO outflow throughout the entire season. TCCO peaks at more than 30.0 ×10¹⁷ molecules/cm² and persists across the 20°-40°S band. **Figure 2f** shows that the inter-year difference exceeds 15.0 ×10¹⁷ molecules/cm² and the larger quantities of CO propagating out into the Pacific (**Figure 2c**). CO outflow from the peak fire activity (**Figure 1e**) in late December 2019/early January 2020 went on to circumnavigate the entire SH (**Figure 3**). Between 27th December 2019 and 9th January 2020, the fire emissions led to large quantities of CO (>25 ×10¹⁷ molecules/cm²), which formed large-scale plumes propagating towards South America. Pyroconvection during this period uplifted plumes to altitudes where CO and other trace gases could more easily be detected by IASI (i.e. colder temperatures and above clouds). In the following fortnight, 10th – 23rd January 2020, these plumes reached South America (peak TCCO >20.0×10¹⁷ molecules/cm²) and started propagating into the South Atlantic (15-18×10¹⁷ molecules/cm²). A week later, the TCCO plume reached southern Africa (no local fire sources apparent there, unlike e.g. 15th-28th November 2019) with values still between 15-18×10¹⁷ molecules/cm². The TCCO quantities were also enhanced to 14-16×10¹⁷ molecules/cm² over the Indian Ocean, reaching the Australian west coastline and fully encircling the SH. Over south-eastern Australia (black box in **Figure S1c**), the 2019/2020 fires (NDJ) emitted 9.06 Tg of CO (1.73 times larger than the 2018/2019 annual total anthropogenic Australian CO emissions; NPI, 2020) in comparison to 0.33 Tg in NDJ 2018/2019, based on FINN emissions.

3.3. Methanol and Methane

Focusing on the 3rd – 16th of January 2020, when the fire plumes were most spatially extensive (see TCCO in **Figure 3, Figure 4a**), CH₃OH and CH₄ both showed substantial increases from the fires (**Figures 4b & c**). Total column CH₃OH (TCCCH₃OH), relative to the 3rd – 16th January 2019, show peak enhancements of over 10×10¹⁵ molecules/cm² co-locating with TCCO enhancements of over 15×10¹⁷ molecules/cm² (**Figure 4a**). Though CH₃OH infrared absorption features are much weaker than those of CO and low background CH₃OH abundances (e.g. over the ocean) are difficult to detect in individual soundings (i.e. large estimated errors; see **SM-3, Figure S6**), robust signals are detected in large sources such as the Australian fire plume (i.e. >15×10¹⁵ molecules/cm², **Figure 5b, SM-3, Figure S7**).

CH₄, on spatial and temporal scales observable by satellite fluctuates by only a few % of its global mean value, unlike the order of magnitude of variability in CO and methanol. To retrieve CH₄ perturbations at the % level requires careful handling of cloud and, in the case of the Australian wildfire plumes, also smoke (and dust). As shown in **SM-4 and Figure SM8**, IASI detects daily CO plumes, but the corresponding CH₄ distribution is less well sampled due to stringent filtering necessary for cloud. However, the column average CH₄ (CACH₄) anomaly for the 3rd-16th January 2020, with reference to the de-trended and de-seasonalised multi-annual mean for January (2007-17, MetOp-A & B offline version) (**Figure 4c**) is positive over much of the region (e.g. ~0.01 ppmv), and some features (e.g. 0.015-0.03 ppmv) cohere with prominent structure (e.g. over the central South Pacific) in the CO and CH₃OH plumes (see **Figure 4a & b**). Therefore, we use CO as an in-plume tracer to quantify the much lower amplitude CH₄ signals from the fires. **Figure 4d** shows the time evolution of in-plume column average CO (CACO) and CACH₄ averaged over 150°E-90°W, 50-20°S (black box in **Figure S8**). The in-plume thresholds for CACO and CACH₄ were set at 0.07 ppmv and 1.75 ppmv. Both thresholds had to be met, along with stringent cloud screening and other quality control for CH₄, in order for the pair of CO and CH₄ soundings at a given location to be sampled.

In the first two weeks of January 2020, the in-plume domain-averaged daily CACO ranges between 0.09 and 0.12 ppmv, and the standard deviation indicates large spatial variability. In the second two weeks of January, the spatially-averaged CACO is substantially lower, ranging between 0.07 and 0.08 ppmv, with much smaller spatial variability, consistent with mixing with neighbouring CO-poor airmasses in the east and less intense incoming plumes in the west. In contrast, for CACH₄ the spatially averaged value drops only slightly and spatial variability is unchanged in the second two weeks, attributable to mixing with high CH₄ tropical airmasses in the east. The peak average CACH₄ values (1.78-1.79 ppmv) occur in the first few weeks and then 27th-29th January. Temporal correlation of the spatially averaged CO and CH₄ time series is 0.63 for the whole month and 0.86 between 1st and 20th January. The daily spatial correlation in this period is between 0.2 and 0.5, indicative of a substantial fire-induced component of CH₄. In the latter period of January (21st-31st) the correlation drops below 0.0 indicating the fire-induced component to be less significant for CH₄ in comparison to other processes. Overall, the relationship between CO and CH₄ in data produced from IASI on MetOp-B demonstrates that the plumes emitted from the Australian fires contained CH₄ as well as CO. Emission factors of CH₄ from vegetation fires have been shown to be similar in magnitude to those of methanol on a mass emitted per mass of dry matter consumed basis (Agaki et al., 2011). It is worth noting that given the long CH₄ lifetime against hydroxyl radical (OH) oxidation (~9 years (McNorton et al., 2016)), any enhancement in methane due to OH suppression in the plume on the transport timescale of a few days would be minimal (~0.1% over 3 days under complete suppression of OH oxidation).

3.4. Enhancement Ratios

We investigate how the TCCH₃OH:TCCO enhancement ratio changes as the fire plume traverses the Pacific across the four regional boxes defined in **Figure 5a** between the 1st and 17th January 2020 (i.e. period of large fire plume). Box 1 covers the primary fire region and Boxes 2-4 cover the downwind outflow. In all cases, we see TCCO-TCCH₃OH correlation values (R in **Figure 5c-f**) above 0.63, peaking at 0.92 in Box 3, indicative of common origins. The in-plume TCCO and TCCH₃OH values are defined based on a subjectively chosen threshold (TCCO = 18.0×10^{17} molecules/cm² and TCCH₃OH = 5.0×10^{15} molecules/cm², where both criteria have to be met for the two species to be co-sampled), but sensitivity analysis of these thresholds (**Figure SM5, Table S1**), and the time period in which the plume(s) are sampled, show that our results are robust and relatively insensitive to our choices of these parameters.

In Box 1, the TCCH₃OH:TCCO enhancement ratio (M), based on a simple linear least-squares fit, is $0.0036 \pm 5.98\%$, which is similar to the FINN CH₃OH:CO fire emissions ratio ($0.0031 \pm 0.04\%$). This suggests that the satellite observations of atmospheric enhancements close to the fires are consistent with freshly emitted fire pollution. As the plume propagates eastwards over the Pacific, there is an increase in the TCCH₃OH:TCCO enhancement ratio. In Boxes 2, 3 and 4 M is $0.0059 \pm 3.13\%$, $0.0091 \pm 1.28\%$ and $0.0081 \pm 1.94\%$, respectively. When using the full range of the IASI TCCH₃OH retrievals (i.e. the retrieval values \pm the random errors) and recalculating the enhancement ratios, we find that M is perturbed by approximately 10%, which is relatively small, and the regional ratios follow the same tendency. This increase of TCCH₃OH with distance, and therefore time, from the fires is suggestive of in-plume chemical production of CH₃OH. This is similar to Holzinger et al., (2005) and Coheur et al., (2009), who used aircraft and satellite data respectively, to investigate fire plume CH₃OH:CO enhancement ratios suggesting signs of secondary CH₃OH production when studying southern European fire plumes. The magnitude of the CH₃OH/CO enhancement ratios derived here are consistent with other studies (e.g. Yokelson et al., 1999; Christian et al., 2003; Holzinger et al., 2004; Singh et al., 2004; Karl et al., 2007). Our results imply that IASI detected such secondary formation of CH₃OH within the fire plume, but with a robust enhancement across the large portion of the Pacific, on a scale previously not discussed in the literature, to the best of our knowledge. Mixing with background air would likely dilute the plume counteracting some of the downwind increase in CH₃OH enhancement relative to CO. This suggests that the in-plume production of CH₃OH is likely larger than that suggested by the observed increase in the CH₃OH:CO ratio alone. Such large-scale enhancements in CH₃OH may have an important influence on the CH₃OH budget, impact the oxidative capacity of the remote atmosphere, and potentially the CH₄ lifetime (Read et al., 2012). Such secondary methanol production could be driven by the self-reaction of methylperoxy (CH₃O₂), the reaction of CH₃O₂ with higher order peroxy (RO₂) radicals (Jacob et al., 2005) and OH (Müller et al., 2016), or possibly by less well-established oxidation of organics in the fire plume (Holzinger et al., 2005).

4. Conclusions

The 2019/2020 Australian fires (“black summer”) constituted some of the largest regional wildfires in recent decades and produced large quantities of smoke, aerosols and trace gases. Peak fire activity occurred on eucalyptus forest vegetation in south-eastern Australia during December and January. Data from the MetOp-B satellite produced by RAL’s NRT processing system show that carbon monoxide (CO) emitted from the fires circumnavigated the entire Southern Hemisphere. Compared with the 2018/2019 fire season, CO levels from the fire-plumes were substantially (and

significantly, 99% confidence level) larger by approximately 30-70% over the South Pacific in the November-December-January average.

Methanol (CH_3OH), which is difficult to detect in normal circumstances due to its weak absorption signature, was in sufficient abundance to retrieve in-plume column amounts with a good precision. Satellite-observed enhancements in total column CH_3OH relative total column CO show a substantial increase downwind from the fires, over the Pacific Ocean. This is strongly suggestive of CH_3OH production within the plume, on a scale not previously reported, as far as we are aware, with potentially important implications for the methanol budget and oxidative capacity of the remote atmosphere. Elevated levels of CH_4 were also detected in association with the fire-plumes during peak activity in early-mid January 2020, even though the CH_4 emission rate is considerably lower than for CO and perturbations from uniform mixing less than 2%.

With future climate and land-use change it is expected that wildfires are going to become more frequent and intense. Therefore, Earth observation (EO), as presented here, is going to be a vital resource to help monitor and understand future wildfire events globally. These EO capabilities will improve with the planned launches of advanced infrared and shortwave spectrometers such as IASI Next Generation and Sentinel 5 on the MetOp Second Generation in polar orbit (ESA, 2020) and the Infrared Sounder and Sentinel 4 on Meteosat Third Generation in geostationary orbit.

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Figures

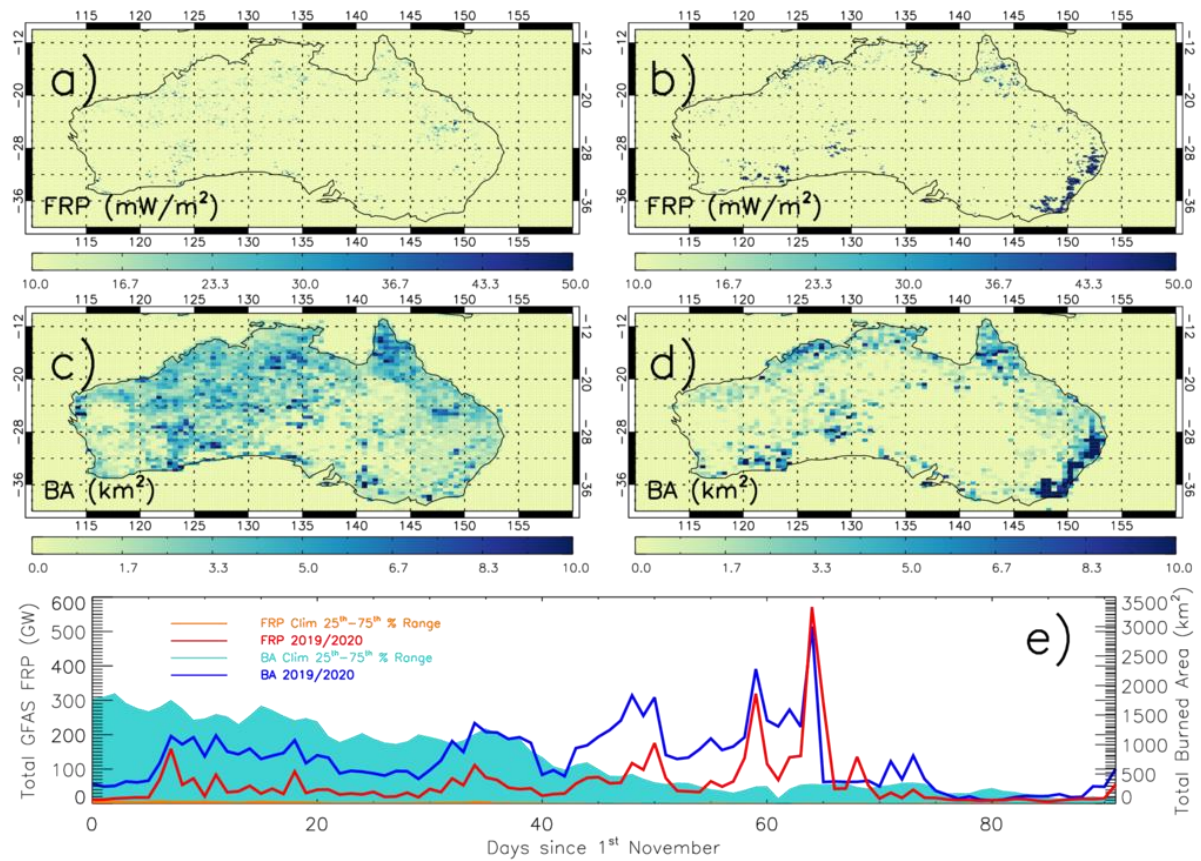


Figure 1: Global Fire Assimilation (GFAS) fire radiative power (FRP, mW/m^2) for a) November-December-January (NDJ) climatology (2009-2019) and b) NDJ 2019/2020. Panels c) and d) show Fire INventory from NCAR (FINN) burned area (km^2) for c) NDJ climatology and d) NDJ 2019/2020. Panel e) shows daily time series of accumulated FRP (GW, blue solid line) and BA (km^2 , red line) across Australian for NDJ 2019/2020. The orange and light blue shading represent the 25th-75th percentile spread in the climatology for FRP and BA, respectively.

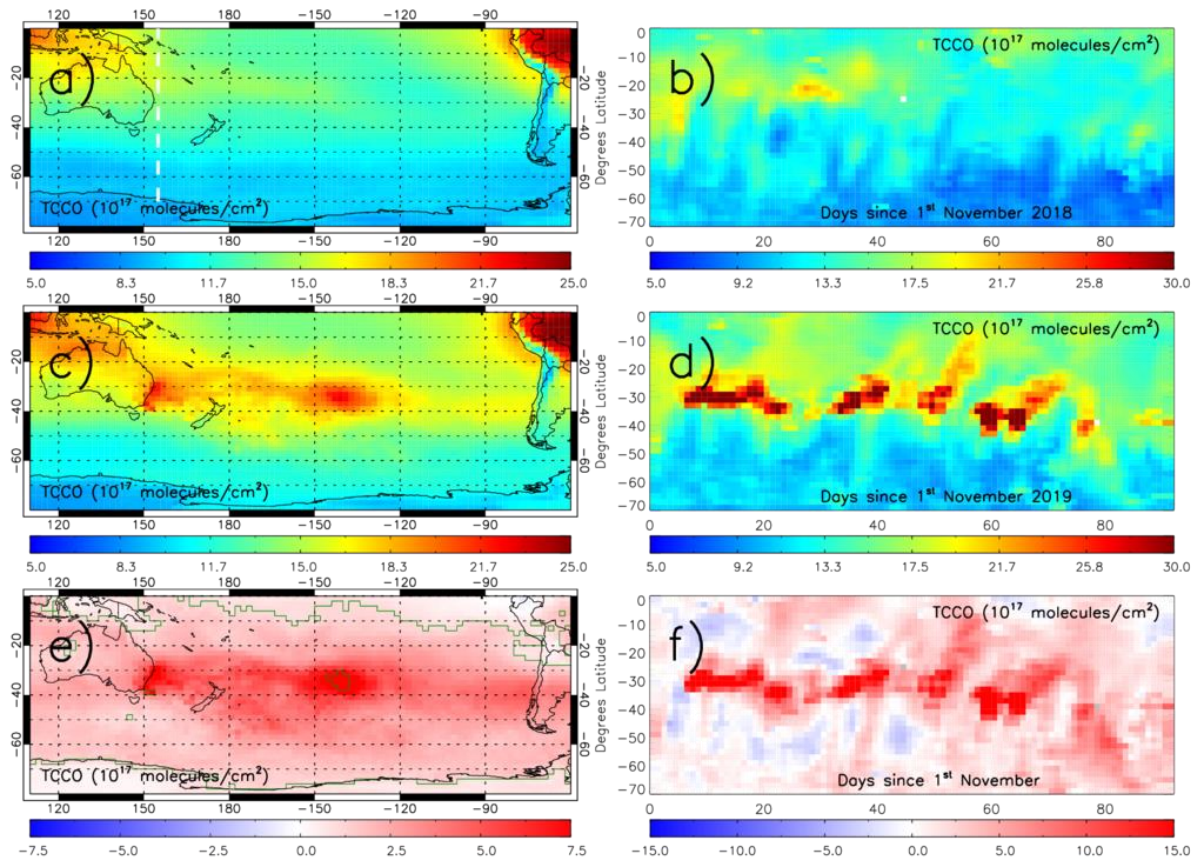


Figure 2: Infrared Atmospheric Sounding Interferometer (IASI) NDJ total-column carbon monoxide (TCCO, 10^{17} molecules/cm²) for a) 2018/2019, c) 2019/2020 and e) 2019/2020-2018/2019 difference. Green polygon-outlined regions in panel e) represent statistically significant differences between the fire seasons at the 99% confidence level (CL, based on the Student t-Test) and where absolute differences are greater than 1.0×10^{17} molecules/cm². Panels b), d) and f) represent Hovmöller diagrams of IASI TCCO from November – January at 155°E, between 70°S-0°S (white dashed line in panel a)), for 2018/2019, 2019/2020 and 2019/2020-2018/2019 difference, respectively.

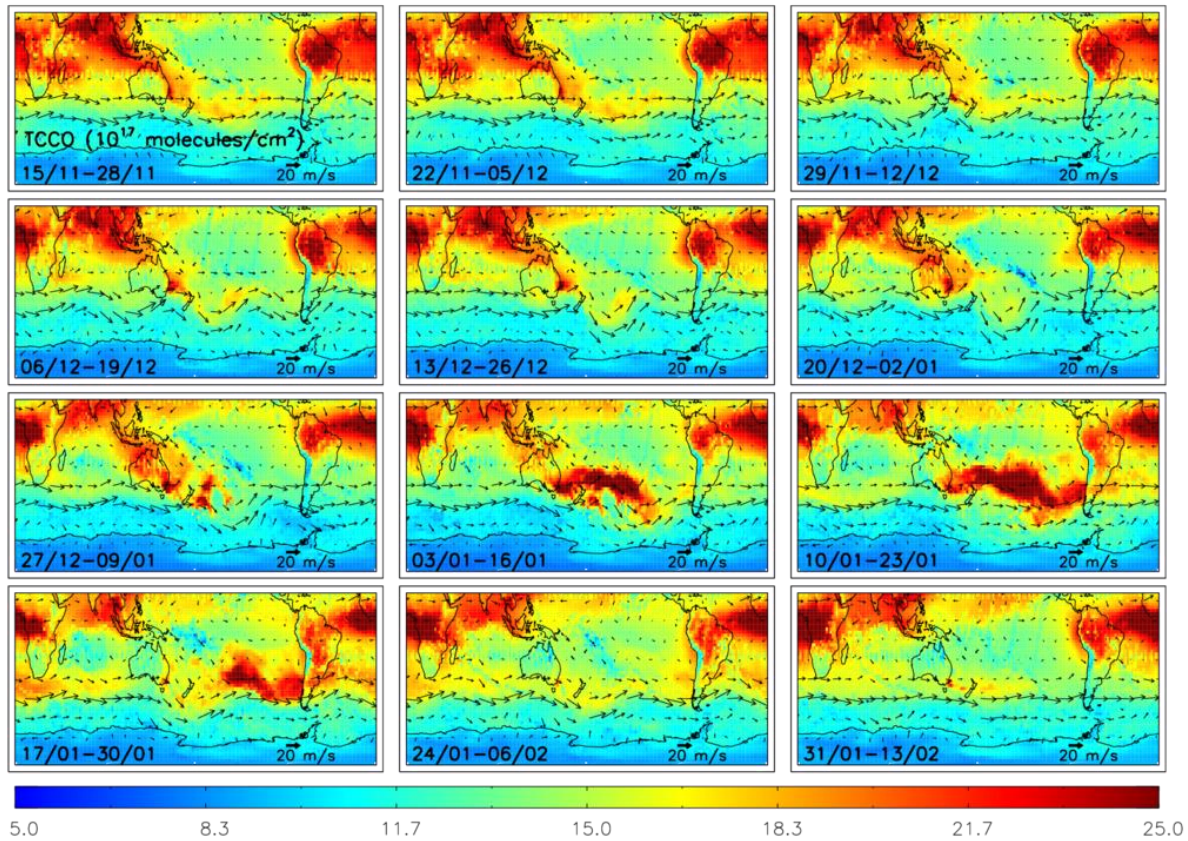


Figure 3: Temporal evolution of the IASI mean TCCO (10^{17} molecules/ cm^2) between the 15th November 2019 and 13th February 2020. Each panel represents a 2-week average with a weekly step between the first day of each map. The arrows show 500 hPa winds from the National Centers for Environmental Prediction (NCEP) reanalysis.

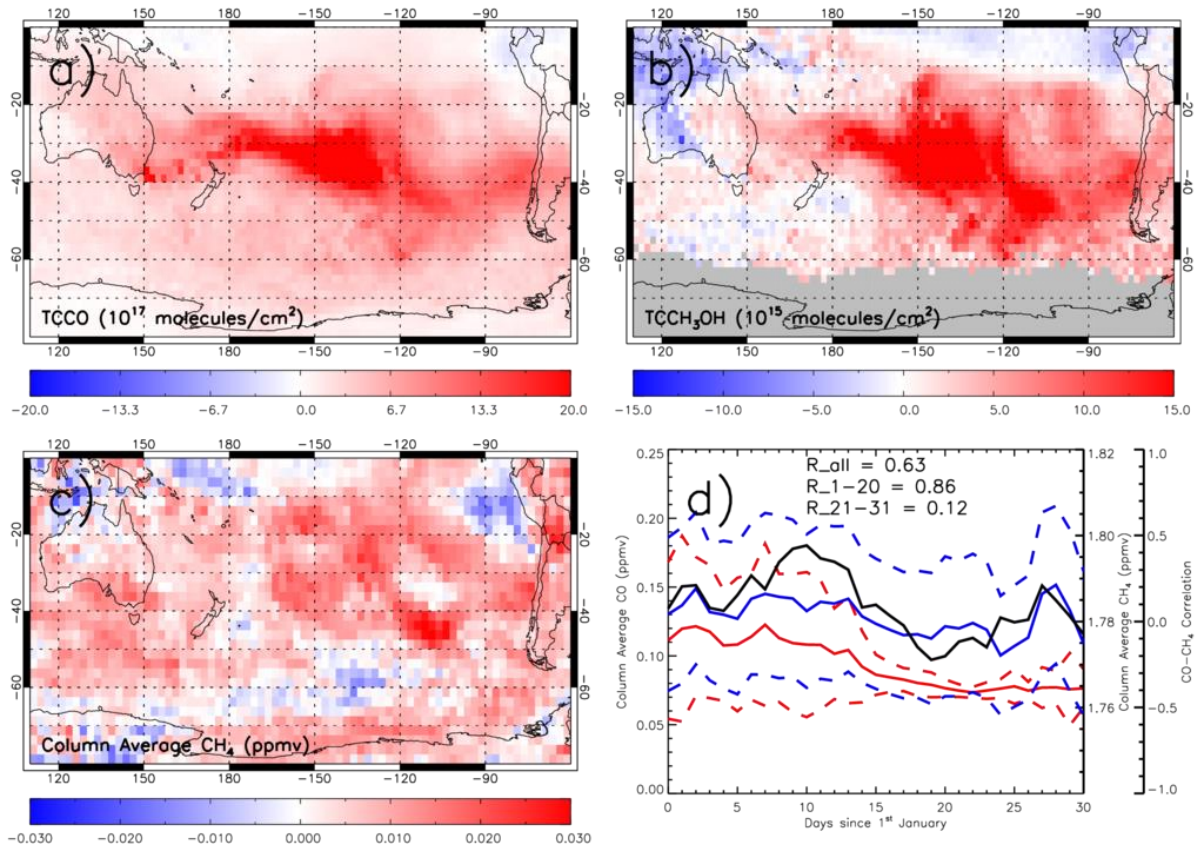


Figure 4: a) IASI two-week (3rd–16th January) 2020 - 2019 difference a) TCCO (10^{17} molecules/cm²) and b) total column methanol (TCCH₃OH, 10^{15} molecules/cm²). Grey regions represent missing satellite data (i.e. average values with error terms $>15.0 \times 10^{15}$ molecules/cm²). c) Two-week (3rd–16th January 2020) column average methane (CH₄, ppmv) anomaly with respect to the de-seasonalised and de-trended multi-annual mean for January (2007-2017). d) Daily time series of spatially averaged in-plume (150°E-90°W, 50-20°S; black box in Figure S4) IASI-observed CO (red) and CH₄ (blue) column average mixing ratios for January 2020. In-plume data are defined where CO and CH₄ values are both larger than the corresponding thresholds of 0.07 ppmv and 1.75 ppmv. Dashed lines represent the uncertainty range (average \pm standard deviation). The black line represents daily spatial correlations between in-plume CO and CH₄. R_{all}, R₁₋₂₀ and R₂₁₋₃₁ are the CO-CH₄ time series correlations for all of January, 1st-20th January and 21st-31st January, respectively.

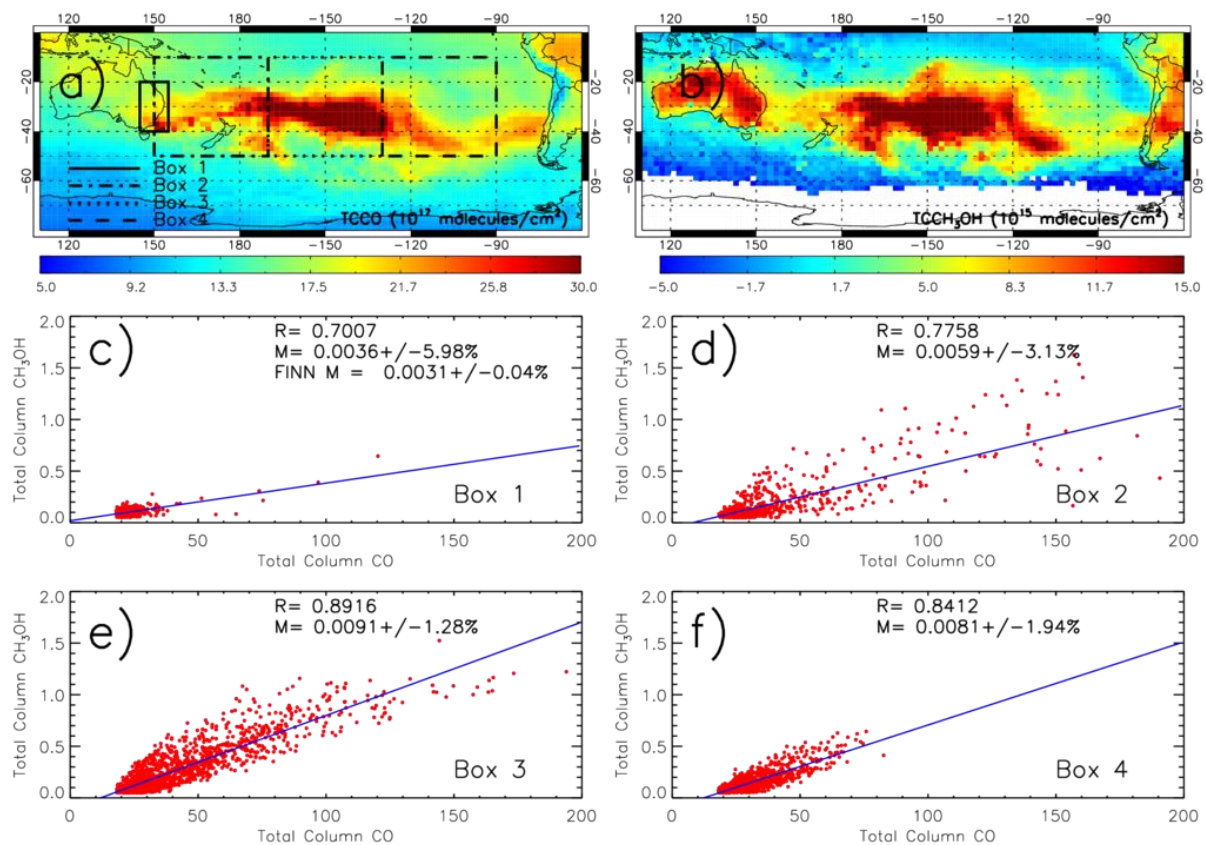


Figure 5: a) TCCO and b) TCCH₃OH (10^{17} molecules/cm²) for 1st-17th January 2020. White regions represent missing satellite data (i.e. average values with error terms $>15.0 \times 10^{15}$ molecules/cm²). Panels c) – f) show scatter plots of TCCH₃OH versus TCCO within Boxes 1-4 outlined in panel a) with values of correlation (R) and gradient (enhancement ratio, M) indicated.