

# 1 Important role of stratospheric injection height for the distribution 2 and radiative forcing of smoke aerosol from the 2019/2020 Australian 3 wildfires

4  
5 Bernd Heinold<sup>1</sup>, Holger Baars<sup>1</sup>, Boris Barja<sup>2</sup>, Matthew Christensen<sup>3,4</sup>, Anne Kubin<sup>1</sup>, Kevin Ohneiser<sup>1</sup>,  
6 Kerstin Schepanski<sup>1,5</sup>, Nick Schutgens<sup>6</sup>, Fabian Senf<sup>1</sup>, Roland Schrödner<sup>1</sup>, Diego Villanueva<sup>1,7</sup>, and Ina  
7 Tegen<sup>1</sup>

8  
9 <sup>1</sup> Leibniz Institute for Tropospheric Research, Permoserstr. 15, 04318 Leipzig, Germany

10 <sup>2</sup> Department of Mathematics and Physics, University of Magallanes, Avenida Bulnes 01855 Punta Arenas, Chile

11 <sup>3</sup> Atmospheric, Oceanic and Planetary Physics, University of Oxford, Oxford OX1 3PU, United Kingdom

12 <sup>4</sup> Now at Atmospheric Sciences and Global Change Division, Pacific Northwest National Laboratory, Richland, WA, 99352,  
13 USA

14 <sup>5</sup> Now at Institute of Meteorology, Freie Universität Berlin, Carl-Heinrich-Becker-Weg 6-10, 12165 Berlin, Germany

15 <sup>6</sup> Department of Earth Science, Vrije Universiteit Amsterdam, 1081 HV Amsterdam, the Netherlands

16 <sup>7</sup> Now at Leipzig Institute for Meteorology, University of Leipzig, Germany

17 Correspondence to: Bernd Heinold ([bernd.heinold@tropos.de](mailto:bernd.heinold@tropos.de))

18  
19 **Abstract.** More than 1 Tg smoke aerosol was emitted into the atmosphere by the exceptional 2019-2020 Southeast Australian  
20 wildfires. Triggered by the extreme fire heat, several deep pyroconvective events carried the smoke directly into the  
21 stratosphere. Once there, smoke aerosol remained airborne considerably longer than in lower atmospheric layers. The thick  
22 plumes traveled eastward thereby being distributed across the high and mid-latitude Southern Hemisphere enhancing the  
23 atmospheric opacity. Due to the increased atmospheric lifetime of the smoke plume its radiative effect increased compared to  
24 smoke that remains lower altitudes. Global models describing aerosol-climate impacts lack adequate descriptions of the  
25 emission height of aerosols from intense wildfires. Here, we demonstrate by combination of aerosol-climate modeling and  
26 lidar observations the importance of the representation of those high-altitude fire smoke layers for estimating the atmospheric  
27 energy budget. Through observation-based input to the simulations, the Australian wildfire emissions by pyroconvection are  
28 explicitly prescribed to the lower stratosphere in different scenarios. Based on our simulations, the 2019-2020 Australian fires  
29 caused a significant top-of-atmosphere hemispheric instantaneous direct radiative forcing signal that reached a magnitude  
30 comparable to the radiative forcing induced by anthropogenic absorbing aerosol. Up to  $+0.50 \text{ W m}^{-2}$  instantaneous direct  
31 radiative forcing was modeled at top of the atmosphere, averaged for the Southern Hemisphere ( $+0.25 \text{ W m}^{-2}$  globally) for  
32 January to March 2020 under all-sky conditions. At the surface, on the other hand, an instantaneous solar radiative forcing of  
33 up to  $-0.81 \text{ W m}^{-2}$  was found for clear-sky conditions, depending on the model configuration *and subject to the model*  
34 *uncertainties in the smoke optical properties*. Since extreme wildfires are expected to occur more frequently in the rapidly  
35 changing climate, our findings suggest that high-altitude wildfire plumes must be adequately considered in climate projections  
36 in order to obtain reasonable estimates of atmospheric energy budget changes.

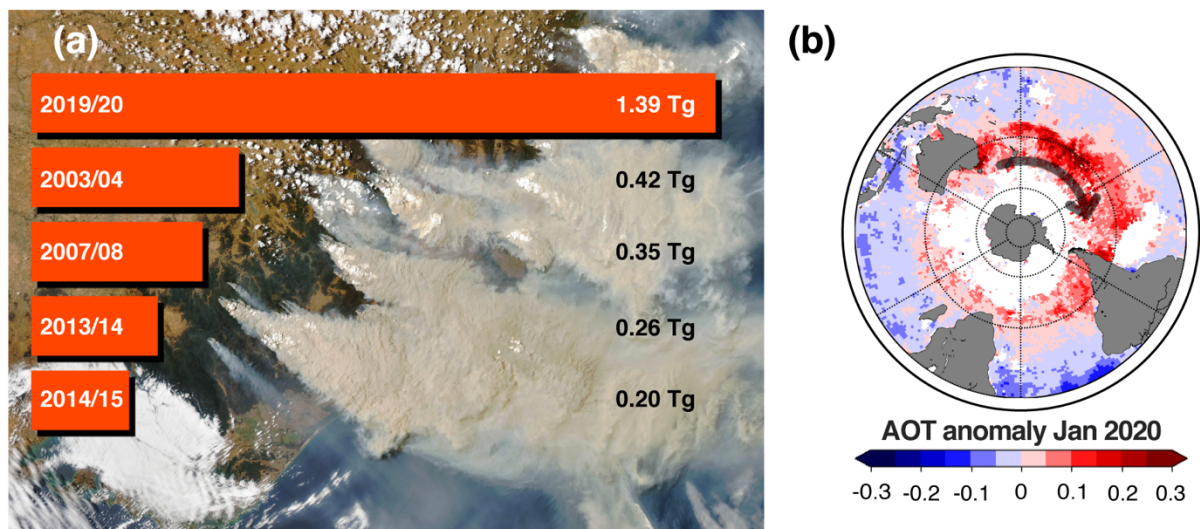
## 1 1 Introduction

2 During the record Australian 2019-2020 wildfire season, the aerosol load increased substantially over large parts of mid and  
3 high latitudes of the Southern Hemisphere due to the massive amounts of smoke aerosol injected into the stratosphere. The  
4 austral summer of 2019-2020 is remembered as Australia's Black Summer due to the unprecedented intensity and scale of  
5 wildfires. The devastating impact on local nature and life was particularly evident in the significant destruction of habitat for  
6 hundreds of endemic species (Ward et al., 2020; Wintle et al., 2020). In addition, the interactions of the fire plume with large-  
7 scale weather (Kablick et al., 2020; Khaykin et al., 2020) make the Black Summer fires also a distinct example for studying  
8 the climate impacts of stratospheric smoke injection.

9 Between September 2019 and January 2020, almost twice the area burnt compared to any previous record fire in Australia,  
10 emitting unprecedented amounts of smoke aerosol (Boer et al., 2020; Morgan et al., 2020) (Fig. 1a). Peaking between 29  
11 December 2019 and 4 January 2020, the fires caused a significant input of aerosol into the stratosphere. Several intense  
12 pyroconvective towers carried this aerosol directly up to 14–16 km height in the lower stratosphere (Kablick et al., 2020;  
13 Ohneiser et al., 2020; Boone et al., 2020). The mass of smoke emitted into the stratosphere by these fires has been estimated  
14 to range from 0.6 Tg (Khaykin et al., 2020) to 2.1 Tg (Hirsch and Koren, 2021). In a slightly later publication, Peterson et al.  
15 (2021) estimate the stratospheric injection of Australian smoke in a first phase of massive pyroconvective activity from 29 to  
16 31 December 2019 to amount 0.2 – 0.8 Tg and 0.1–0.3 Tg in the second phase on 4 January 2020. Within days, the smoke was  
17 distributed zonally across the southern mid and high latitudes, according to satellite measurements by NASA's Moderate  
18 Resolution Imaging Spectroradiometer (MODIS) (Hirsch and Koren, 2021), Cloud-Aerosol Lidar and Infrared Pathfinder  
19 Satellite Observations (CALIPSO) data (Kablick et al., 2020) as well as observations from the Stratospheric Aerosol and Gas  
20 Experiment (SAGE) III and TROPOspheric Monitoring Instrument (TROPOMI) (Khaykin et al., 2020). Furthermore, also  
21 ground-based lidar measurements at the southern tip of South America clearly showed the elevated smoke layer (Ohneiser et  
22 al., 2020). As a result, atmospheric opacity in the southern hemisphere was considerably enhanced. Between the latitudes  
23 20°S–60°S, the total column aerosol optical thickness (AOT) at 630 nm wavelength increased to 0.16 on average in January  
24 2020, a 51% deviation from the long-term mean, as shown for example by the observations of the Advanced Very High  
25 Resolution Radiometer (AVHRR) satellite instrument. Figure 1b shows the hemispheric dispersal of the Australian wildfire  
26 smoke in January 2020 in terms of the AVHRR AOT anomaly.

27 Satellite observations and global aerosol-climate model results show that this had significant effects on the radiation budget  
28 (Khaykin et al., 2020; Hirsch and Koren, 2021; Yu et al, 2021; *Sellitto et al., 2022*). For the stratospheric smoke from the  
29 Australian wildfires, Khaykin et al, (2020) found a cloud-free solar radiative forcing of about  $-1.0 \text{ W m}^{-2}$  at the top of the  
30 atmosphere (TOA) and  $-3.0 \text{ W m}^{-2}$  at the surface on average for the latitudes from 25°S to 60°S in February 2020, based on  
31 radiative transfer modeling using aerosol extinction profiles from the NASA Ozone Mapping and Profiler Suite Limb Profiler  
32 (OMPS-LP). *Building on the method by Khaykin et al. (2020), Sellitto et al. (2022) recently provided updated estimates and*  
33 *conducted sensitivity analyses to examine the effects of assumed aerosol optical properties. Their best estimate for the global-*  
34 *equivalent cloud-free solar radiative forcing in February 2020 is  $-0.35 \text{ W m}^{-2}$  at TOA and  $-0.94 \text{ W m}^{-2}$  at surface and reaches*  
35 *up to  $-2.0 \text{ W m}^{-2}$  and  $-4.5 \text{ W m}^{-2}$ , respectively, for the 25°S – 60°S latitude band, using optical properties characteristic for*  
36 *aged biomass burning aerosol. Assuming highly reflective underlying surfaces as approximation for clouds and highly*  
37 *absorbing particles, in contrast, they calculated a positive global average TOA forcing as high as  $+1.0 \text{ W m}^{-2}$ .* Hirsch and  
38 Koren (2021) derived an enhancement of outgoing solar radiation of  $1.1 \text{ W m}^{-2}$  in the latitude belt between 20°S and 60°S for  
39 January to March, from NASA's Clouds and the Earth's Radiant Energy System (CERES) satellite data. From model results  
40 and considering also the fast adjustment from stratospheric warming, Yu et al. (2021) obtained an estimate for global annual  
41 average clear-sky effective radiative forcing of  $-0.03 \text{ W m}^{-2}$  at TOA and  $-0.32 \text{ W m}^{-2}$  at the surface due to the smoke event.

1 Australia's Black Summer is among a recent series of extreme wildfires that has renewed scientific attention particularly to  
 2 wildfires with strong fire-induced convection and self-lifting. These include recent record fires in the Western United States  
 3 and Canada (2017, 2018), Siberia (2019, 2020) and the Eastern Mediterranean (2021). Triggered by the intense fire heat, the  
 4 pyroconvection can grow to pyrocumulonimbus (pyroCb) clouds which are the primary pathway of smoke injection into the  
 5 upper troposphere and lower stratosphere (Fromm et al., 2010; Fromm et al., 2019). Radiation-induced self-lifting has the  
 6 potential to cause the smoke plumes to continue rising (Boers et al., 2010). Also due to such events, biomass burning smoke  
 7 contributes considerably to the global aerosol composition, affecting the Earth's energy balance through aerosol-radiation and  
 8 tropospheric aerosol-cloud interactions (Bowman et al., 2009; Streets et al. 2009; Boucher et al., 2013). Such extreme wildfires  
 9 and associated deep pyroconvection can have similar effects as volcanic eruptions in terms of stratospheric aerosol injection  
 10 and radiative impact (Peterson et al., 2018). An important component is black carbon aerosol, which is among the strongest  
 11 warming short-lived radiative forcing agent (Bond et al., 2013; Lund et al., 2020; Thornhill et al., 2021). In addition, less-  
 12 absorbing organic carbon, ~~and~~ precursors for sulfate and other secondary aerosols are emitted. Depending on its optical  
 13 properties and the underlying surface reflectivity, the climate impact of biomass burning aerosol can vary regionally (Jiang et  
 14 al., 2016; Bellouin et al., 2020; Brown et al., 2021) and also strongly depends on the altitude of the aerosol layer (Ban-Weiss  
 15 et al., 2012). During strong pyroCb events, radiative effects can be enhanced due to long stratospheric lifetime of aerosol.  
 16 While the high-altitude injection of wildfire plumes is yet insufficiently represented in aerosol-climate models (Paugam et al.,  
 17 2016), the recent series of extreme wildfires and their potentially increased occurrence with climate change (Jolly et al., 2015;  
 18 Abazoglou et al., 2019; Dowdy et al., 2019; Kirchmeier-Young et al., 2019) call for greater attention in global climate  
 19 modeling.  
 20



21  
 22 **Figure 1.** (a) Biomass burning plumes in the Canberra region in Southeast Australia as seen from NASA's Aqua satellite on  
 23 4 January 2020 (<https://aqua.nasa.gov>). Overlaid is a ranking of carbon aerosol emissions accumulated for the annual Southeast  
 24 Australian bushfire seasons (September to March) based on Global Fire Assimilation System (GFAS) data (Kaiser et al., 2012).  
 25 (b) Anomaly in monthly mean aerosol optical thickness (AOT) at 630 nm for January 2020 (by remapping to  $1^\circ \times 1^\circ$ , pixels  
 26 with less than 300 valid retrievals at  $0.1^\circ$  original resolution are excluded, see further details in Sect. 2.3) compared to the  
 27 long-term January mean (1982 to 2019), as observed by NOAA's AVHRR instrument (Zhao et al., 2017) (missing data shown  
 28 in white, continents in gray).

29  
 30 In order to clarify the role of smoke injection of wildfire pyroconvection in the aerosol-climate modeling context, here we use  
 31 the aerosol-climate model ECHAM6.3-HAM2.3 (Zhang et al., 2012; Tegen et al., 2019). Specifically, we aim to show the  
 32 importance of considering these extreme fire events in determining the global energy budget, while they are not adequately

1 reflected in today's climate simulations. The fire emission fluxes in the model are prescribed from the Global Fire Assimilation  
2 System (GFAS; Kaiser et al., 2012), and the injection height of Australian fire smoke is set to the tropopause level for the  
3 known pyroCb events and varied accordingly in sensitivity experiments. The modeled transport patterns are evaluated with  
4 active and passive ground-based and spaceborne remote sensing, providing the basis for analyzing the radiative impact of the  
5 carbonaceous smoke aerosol. Finally, we discuss implications and perspectives for climate models to address the extreme  
6 wildfires and their effects in a changing climate.

## 7 **2 Observations and modeling**

8 The analysis of the 2019-2020 Australian fire season in this study is based on global aerosol-climate simulations. An important  
9 part of the modeling is concerned with finding a configuration that best represents the pyroconvective fires. Since the typical  
10 horizontal resolution of global climate models is too coarse to explicitly resolve convection, observed pyroCb events are  
11 explicitly prescribed and the injection height of the wildfire plume is varied in terms of sensitivity experiments. Their results  
12 are compared to the original settings for biomass burning emissions as well as evaluated with ground-based and spaceborne  
13 remote sensing observations to show how realistically these can be represented if the injection heights for pyroCbs are  
14 considered accordingly. The model results are then used to investigate the impact of pyroconvective smoke injection on plume  
15 transport and radiative effects for January to March 2020.

### 16 **2.1 AERONET sun photometer measurements**

17 Information on column aerosol properties including aerosol optical thickness (AOT) at specific wavelengths and corresponding  
18 information on effective aerosol size are available from quality-controlled measurements by the global sunphotometer network  
19 AEROSOL ROBOTIC NETWORK (AERONET (Holben et al., 1998; Giles et al., 2019); <http://aeronet.gsfc.nasa.gov>). These data are  
20 widely used for aerosol studies including evaluation of aerosol model results. In this study we use level 1.5 or, where available,  
21 level 2 cloud-screened, 6-hour averages of AOT measurements. AERONET AOT values at 550 nm are extrapolated from the  
22 measured values at 500 nm making use of the Angstrom exponent for each observation, which in turn is computed from the  
23 ratio of observed AOT values at 500 nm and 675 nm, respectively. AOT measurements are compared to model results by  
24 linearly interpolating model values to the times and locations of the measurements of the respective AERONET stations: Punta  
25 Arenas, Chile (53.14°S, 70.89°W), Amsterdam Island (37.80°S, 77.57°E), Marambio (64.24°S, 56.63°W), Vechernaya Hill  
26 (67.66°S, 46.16°E) and South Pole (90.00°S, 70.30°E).

### 27 **2.2 Ground-based lidar remote sensing**

28 The lidar observations at Punta Arenas (53.14°S, 70.89°W; 9 m above sea level), Chile, were conducted in the framework of  
29 the long-term DACAPO-PESO campaign (Dynamics, Aerosol, Cloud And Precipitation Observations in the Pristine  
30 Environment of the Southern Ocean; <https://dacapo.tropos.de>). Main goal of DACAPO-PESO is the investigation of aerosol-  
31 cloud interaction processes in rather pristine, unpolluted marine conditions (Radenz et al., 2021). The Polly instrument  
32 (PORTABLE Lidar SYstem; Engelmann et al., 2016) was operated at the University of Magallanes (UMAG) at Punta Arenas  
33 from November 2018 until October 2021. The lidar has 13 channels and continuously measures elastic and Raman backscatter  
34 signals at the laser wavelengths of 355, 532, and 1064 nm and respective Raman backscattering wavelengths of 387 and 607  
35 nm for nitrogen Raman scattering and 407 nm for water vapor Raman scattering (Baars et al., 2016; Baars et al., 2019). At the  
36 laser wavelengths of 355 nm and 532 nm, particle extinction coefficients, the respective extinction-to-backscatter ratio (i.e.  
37 lidar ratio), and the linear depolarization ratio are determined. Moreover, vertical profiles of the particle backscatter coefficient  
38 can be derived at these wavelengths and, additionally, at 1064 nm. The mixing ratio of water vapor to dry air is obtained from

1 measurements in the UV. Auxiliary meteorological data, in particular temperature and pressure profiles, are required in the  
2 lidar data analysis in order to calculate and correct for atmospheric molecular backscatter and extinction. To this end, GDAS1  
3 (Global Data Assimilation System 1) temperature and pressure profiles with 1° horizontal resolution from the National  
4 Weather Service's National Center for Environmental Prediction (GDAS et al., 2020) were used. *In addition, values for single*  
5 *scattering albedo at 355 nm and 532 nm were retrieved from the Polly multi-wavelength backscatter and extinction profiles*  
6 *by using the Veselovskii et al. (2002) lidar data inversion (Ohneiser et al., 2022).*

## 7 **2.3 Spaceborne remote sensing**

### 8 **2.3.1 AVHRR aerosol optical thickness**

9 Observations with the Advanced Very High Resolution Radiometer (AVHRR) onboard the National Oceanic and Atmospheric  
10 Administration (NOAA) operational satellites are available for almost four decades. For the present study, we use version 3  
11 of the AVHRR AOT product (Zhao et al., 2017). It provides daily mean AOT at 630 nm for cloud-free pixels over none-glint  
12 water surfaces with a horizontal resolution of 0.1°. The uncertainty of a single AOT retrieval is 0.2. Because of clouds  
13 obscuring the view, there are no valid retrievals and consequently no daily averages for every pixel on every day. For the  
14 illustration of the hemispherical spread of Australian wildfire smoke in Fig. 1b, showing the observed January 2020 AOT  
15 anomaly, the original 0.1°x0.1° AVHRR data is compiled onto a grid with a spatial resolution of 1°x1° to account for sufficient  
16 samples in the temporal mean.

17 The long-term mean for January from 1982 to 2019 is calculated by averaging in space and time over all pixels in the 1°x1°  
18 cells. The averaging for January 2020 is performed analogously. However, due to the sparseness of the observations in a single  
19 month, 1°x1° averaging boxes with less than 300 valid retrievals in January 2020 (i.e. approx. 10% of the 100 potentially  
20 available data points per are not considered here. In particular, at high latitudes, data coverage is sparse due to the low angle  
21 of the sun and high cloud cover.

### 22 **2.3.2 CALIOP space-based lidar observations**

23 Lidar observations from the Cloud-Aerosol Lidar with Orthogonal Polarization (CALIOP, Winker et al., 2013) instrument are  
24 used to retrieve the extinction coefficient at 532 nm and 1064 nm. We use the level 2 version 4 aerosol profile product which  
25 is averaged over 5-km horizontal segments along the near nadir-view ground track (05kmAPro product). The cloud-aerosol  
26 discrimination (CAD) score is used to include only those columns in which at least one aerosol retrieval was successfully  
27 performed, using a threshold of < -20 CAD scores. This level of quality screening is the same as that described in Winker et  
28 al. (2013). However, despite the use of the highest quality data, CALIOP is known to frequently fail to detect thin aerosol  
29 layers in the upper troposphere *and lower stratosphere (Watson-Parris et al., 2018; Liu et al., 2019)*. The CALIOP level 2  
30 aerosol classification selection algorithm defines six aerosol types: clean marine, dust, polluted continental, clean continental,  
31 polluted dust, and smoke which is based on the extinction-to-backscatter ratio (i.e., lidar ratio). Comparisons of the CALIOP  
32 backscatter with airborne measurements using a High Spectral Resolution Lidar (HSRL), conducted during the ObserVations  
33 of Aerosols above CLouds and their intEractionS (ORACLES) campaign independently demonstrated the lack of detection of  
34 these aerosol types using the CALIOP lidar, and as such, have carried out the necessary steps to account for these biases as  
35 discussed in detail in Watson-Parris et al. (2018). As a result of its low sensitivity the mean fraction of aerosol detected by  
36 CALIOP is globally up to 44% lower than the aerosol-climate model ECHAM-HAM (Watson-Parris et al., 2018). Despite this  
37 bias, the substantial increase in aerosols resulting from the Australian wildfires is evidently detected by CALIOP. While  
38 sampling and detection biases occur on individual profiles the trends in the extinction profiles offer valuable constraints for  
39 the ECHAM-HAM model.

## 1 2.4 Aerosol-climate simulations

### 2 2.4.1 Model description and setup

3 The simulations for this study were made with the global state-of-the-art aerosol-climate model ECHAM6.3-HAM2.3 (Tegen  
4 et al., 2019). This model uses the aerosol microphysics model M7 (Vignati et al., 2004) to predict the evolution of black carbon  
5 (BC), organic carbon (OC), sulphate, sea salt and mineral dust. The mass and number concentrations of the aerosols are  
6 influenced by emission, loss processes, particle microphysics and atmospheric chemistry. The *aerosol* particles can interact  
7 with radiation and clouds. *Atmospheric radiative transfer in ECHAM6.3 is computed using the PSrad/RRTMG (Rapid  
8 Radiative Transfer Model for GCMs; Iacono et al., 2008; Pincus and Stevens, 2013) radiation package, which considers 16  
9 shortwave and 14 longwave wavelength bands, respectively, as well as relevant atmospheric compounds. Aerosol optical  
10 properties are dynamically computed in the model depending on chemical composition and particle size. For each aerosol  
11 mode, the Mie-scattering parameter and the volume-averaged refractive indices are derived, assuming internal mixing, and  
12 are mapped to pre-calculated values of extinction cross section, single scattering albedo, and the asymmetry parameter from  
13 a look-up table. For analyzing the instantaneous direct aerosol radiative forcing, a double call to the radiation scheme exists  
14 to exclude atmospheric adjustments in the diagnostic.*

15 Anthropogenic and biomass burning emissions of aerosols are prescribed. Daily data from the Global Fire Assimilation System  
16 (GFAS; Kaiser et al., 2012) based on fire radiative power observations by the MODIS instruments aboard NASA's Terra and  
17 Aqua satellites are used for the biomass burning related aerosol emissions of BC, OC, sulphate and dimethyl sulphide. In its  
18 original version, 75% of the biomass burning aerosol mass is injected in the planetary boundary layer (PBL), 17% in the first  
19 layer above the PBL and 8% in the second layer above the PBL. A more suitable representation of the smoke injection height  
20 for this extreme wildfire event is explored in a series of sensitivity experiments. In these simulations, Australian smoke aerosol  
21 is directly emitted into the tropopause region in different configurations for known days of pyroconvective activity, as  
22 described in the next Section 2.4.2.

23 The ECHAM6.3-HAM2.3 simulations were performed for the time period November 2019 to March 2020, using T63  
24 horizontal resolution (approximately  $1.875^\circ \times 1.875^\circ$ ). In the vertical, the model is set up with 47 levels with increasing layer  
25 thickness from the ground to 0.01 hPa ( $\sim 80$  km). The vertical resolution ranges from approximately 70 m at surface to 500 m  
26 at 2.5 km and 1100 m at 15 km height and coarsens accordingly thereabove. The dynamics in all simulations was nudged  
27 towards ECMWF ERA5 reanalysis data (Hersbach et al., 2020). Sea surface temperatures and sea-ice concentrations were  
28 prescribed as lower boundary conditions using Atmospheric Model Intercomparison Project (AMIP) data (Giorgetta et al.,  
29 2012). Concentrations of long-lived greenhouse gases were specified following the Representative Concentration Pathway  
30 (RCP) 4.5 scenario. Output was written every 6 hours. The simulated aerosol properties include AOT at 550 nm and vertical  
31 profiles of aerosol extinction at 532 nm wavelength from the online lidar simulator, implemented specifically for comparisons  
32 with CALIOP and ground-based lidar measurements.

### 34 2.4.2 Sensitivity experiments on wildfire smoke injection

35 Wildfire injection heights are usually parameterized in coarser-scale models by schemes of various complexity (Paugam et al.,  
36 2016), but these do not necessarily represent the deep pyroconvection that is observed during very intense wildfires (Remy et  
37 al., 2017; Haarig et al., 2018; Ohneiser et al., 2020). In order to reconstruct the elevated smoke injection due to pyroconvection  
38 and to explore the impacts on plume transport and climate radiative effects of the 2019-2020 Australian fire plume, we adapted  
39 the high-altitude smoke injection height by pyroconvection for the days 29 – 31 December 2019 and 4 January 2020 (pyroCb  
40 days) in the model, on which strong pyroconvective activities were reported in the Southeastern Australian region affected by  
41 the fire (Kablick et al., 2020). Since no direct information was available on the actual pyroconvective injection heights, these

1 were varied in the model in sensitivity experiments and verified with the range of above-mentioned remote sensing  
 2 observations, in particular with the lidar measurements over Punta Arenas in Chile.  
 3 The Australian fire emissions in the model, based on the GFAS inventory, are 0.6 Tg and 0.2 Tg for the two pyroconvective  
 4 phases 29 – 31 December 2019 and 4 January 2020, respectively, with a black carbon to total carbon (BC/(BC+OC)) ratio in  
 5 the fire emissions of about 0.06 – 0.07. These emission values agree well with the previously mentioned estimates by Peterson  
 6 et al. (2021) *and are kept unchanged throughout the sensitivity experiments*. For the four pyroCb days (29 – 31 December  
 7 2019 and 4 January 2020), the smoke injection from Southeastern Australia was set to the model layers above and below the  
 8 tropopause as in the scenarios listed in the following: ‘*TP+1*’: 100% smoke injection into the model layer above the tropopause;  
 9 ‘*TP*’: 100% smoke injection into the model layer containing the tropopause, ‘*TP-1*’, 100% smoke injection into the model  
 10 layer below the tropopause; ‘*TPI\_8020*’: as *TP+1* but only 80% of the emitted smoke injected above *the tropopause* and 20%  
 11 distributed *between tropopause level and surface below the tropopause*; ‘*TPI\_5050*’: as *TP+1*, but only 50% of the emitted  
 12 smoke injected above and 50% distributed below the tropopause; and ‘*14km*’: smoke injection into 14 km height as suggested  
 13 from the spaceborne CALIOP lidar measurements showing smoke plumes in the lower stratosphere up to 17 km near the  
 14 Australian continent (Hirsch and Koren, 2021). It is important to note that a larger height range is directly affected *because of*  
 15 *due to* the model layer thickness at these heights.

16 In addition, a reference simulation using the original model configuration with 75% wildfire emissions within the planetary  
 17 boundary layer and 25% into the two model layers above was carried out, which hereafter is referred to as *BASE* case. To  
 18 estimate the input of fire aerosol to the stratosphere from the pyroconvective fires, a model run was also performed in which  
 19 the Southeastern Australian wildfire emissions were set to zero for the pyroCb days while they were treated as in the original  
 20 setup for all other days (referred to as case *NoEmiss*). Further experiments include model runs without interactive aerosol-  
 21 radiation interaction for the *BASE*, *TP+1*, and *TPI\_8020* case scenarios in order to test the hypothesis that self-lifting due to  
 22 radiative heating has significantly influenced the smoke plume evolution. The different model experiments are summarized in  
 23 Table 1.  
 24

25 **Table 1.** Overview of scenarios simulated with ECHAM6.3-HAM2.3 using different assumptions for the emission height of  
 26 the emitted biomass burning aerosol over Southeastern Australia.

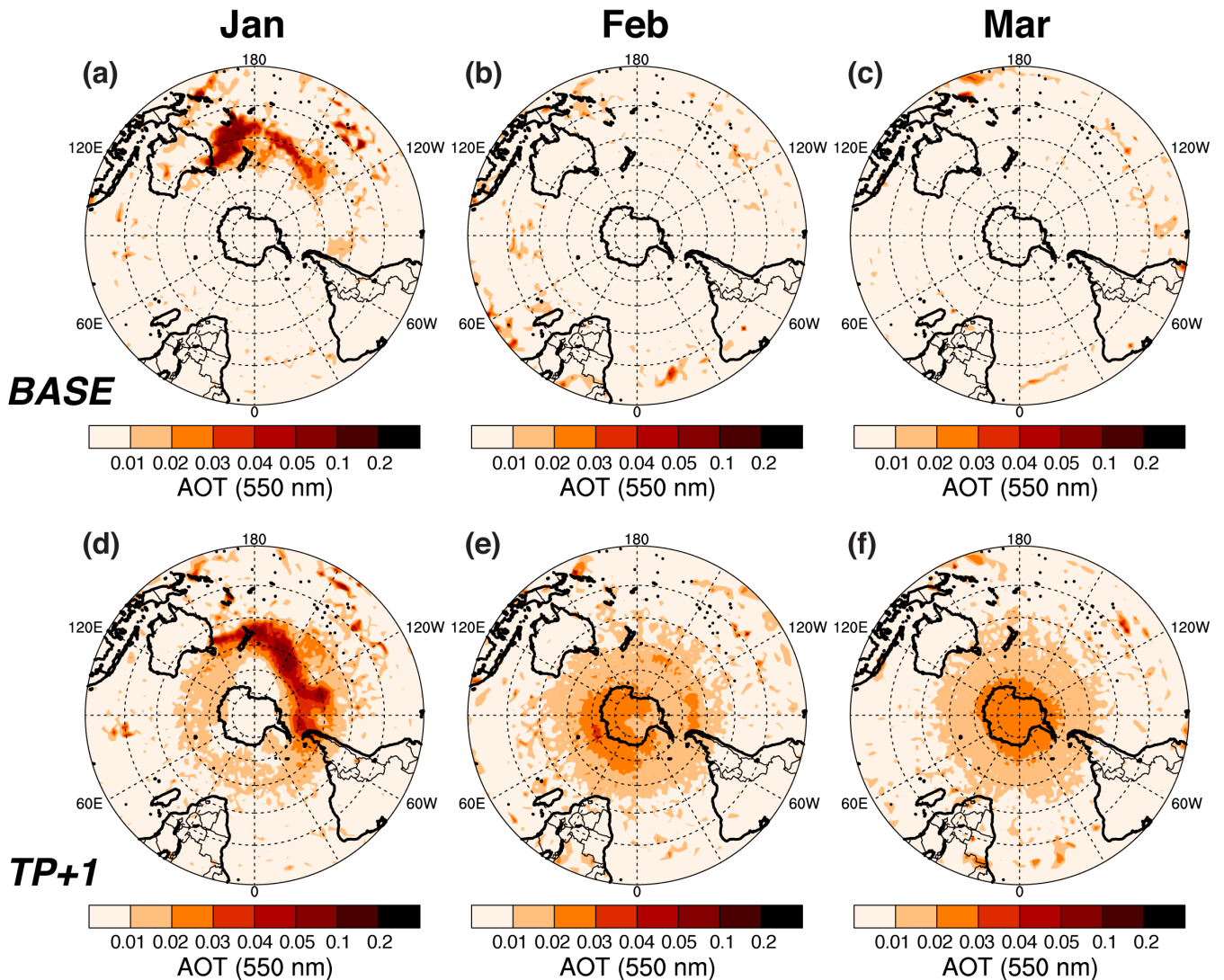
Scenario	Description
<i>BASE</i>	Standard emission height as prescribed in the ECHAM-HAM model for wildfires (75% in PBL, 17% in the first layer and 8% in the second layer above PBL)
<i>NoEmiss</i>	Wildfire smoke emission set to zero for the pyroCb days 29 – 31 December 2019 and 4 January 2020 in Southeastern Australia
<i>TP+1</i>	Wildfire smoke emission from Southeastern Australia injected into the model layer above the tropopause for the pyroCb days
<i>TP</i>	As <i>TP+1</i> , but smoke injection into the model layer containing the tropopause
<i>TP-1</i>	As <i>TP+1</i> , but smoke injection into the model layer below the tropopause
<i>TPI_8020</i>	As <i>TP+1</i> , but only 80% of the emitted smoke injected above <i>the tropopause</i> and 20% distributed <i>between tropopause level and surface below the tropopause below the tropopause</i>
<i>TPI_5050</i>	As <i>TP+1</i> , but only 50% of the emitted smoke injected above <i>the tropopause</i> and 50% distributed <i>between tropopause level and surface below the tropopause below the tropopause</i>

1

2 **3 Results**3 **3.1 Smoke transport simulation and model evaluation**

4 According to satellite observations, the 2019-2020 Australian wildfire plume considerably increased the AOT of the usually  
 5 pristine Southern Hemisphere (Hirsch and Koren, 2021). The average AOT at 630 nm wavelength derived from the AVHRR  
 6 satellite instrument between 20°S–60°S was increased to 0.16 for January 2020, which implies a 51% offset from the long-  
 7 term mean. Ground-based observations *also* show *significantly increased AOT values after a major peak after mid-January*.  
 8 *For* example, at the Punta Arenas station, the January to March 2020 average 550-nm AOT was 0.10, which is more than a  
 9 factor 2 increase compared to the year 2019. One year later, in January 2021, the observed 500-nm AOT over Punta Arenas  
 10 was still increased with a monthly mean of 0.06 (50% increase relative to the monthly mean AOT for January 2019).

11



12

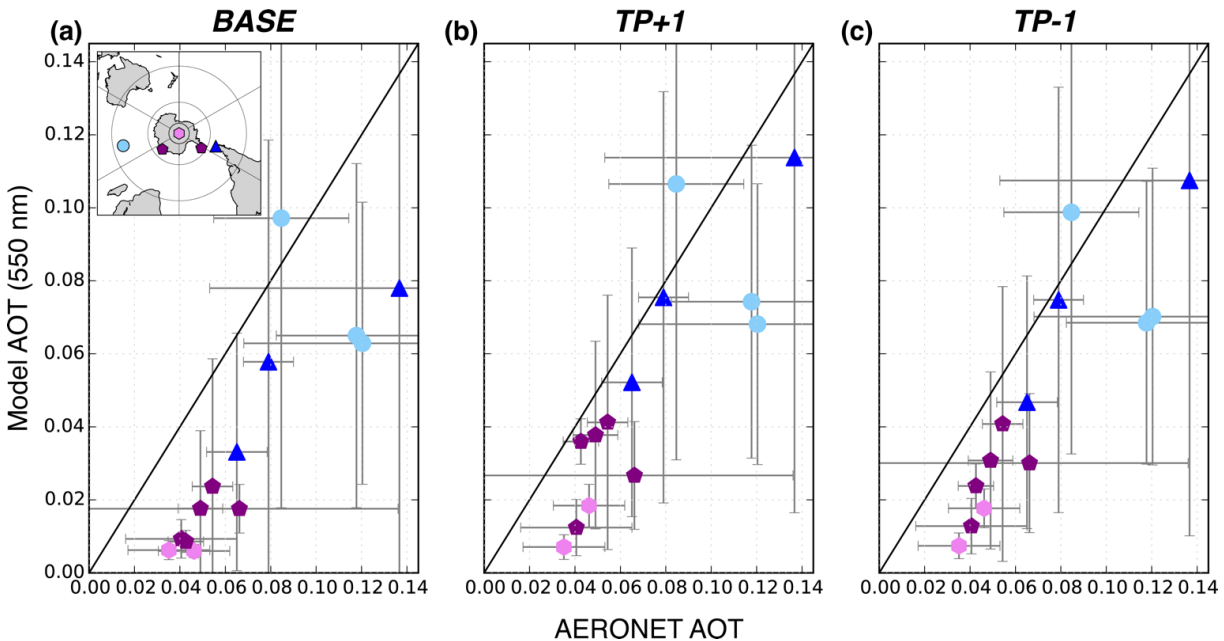
13 **Figure 2.** Monthly mean simulated AOT differences for January to March 2020 for the cases *BASE - NoEmiss* (top panels)  
 14 showing the contribution of smoke AOT for the case when no smoke injection by pyroconvection is prescribed in the model,



1 and *TP+1 - NoEmiss* (bottom panels), showing the effect on AOT of smoke injection into the model layer above the tropopause  
 2 for the pyroCb days 29 – 31 December 2019 and 4 January, 2020 in Southeastern Australia.

3  
 4 The dispersal of this smoke plume is reproduced using the global aerosol-climate model ECHAM6.3-HAM2.3 with the  
 5 pyroconvective injection heights prescribed. The comparison of the modeled AOT of the *BASE* and *TP+1* experiments,  
 6 respectively, and that of the *NoEmiss* experiment (see Fig. 2) provide an insight into the AOT distribution caused by the  
 7 pyroCb events on 29 – 31 December 2019 and 4 January 2020 and illustrate the role of the smoke injection height. While *the*  
 8 monthly mean smoke AOT *is simulated in absolute values* as high as 0.26 and 0.22 for January just downwind of the fire  
 9 region in Southeast Australia for the *BASE* and *TP+1* experiments, respectively, the results of the *BASE* experiment do not  
 10 show increased smoke AOT eastward of 120°W in January and none in the later months (Fig. 2a–c). In contrast, the results of  
 11 the *TP+1* model experiment in which the smoke aerosol was injected into the model layer above the tropopause for the four  
 12 Southeastern Australian pyroCb days show persistently enhanced smoke AOT south of 30°S with AOT differences between  
 13 *TP+1* and *NoEmiss* of 0.01 to 0.03 until at least March 2020 (Fig. 2d–f). Also, a southward transport of the stratospheric smoke  
 14 during the three months leading to maximum smoke AOT anomaly above Antarctica in March is evident. Similar smoke  
 15 transport to Antarctica was reported by Jumelet et al. (2020) for the earlier major Australian fires in 2009. The effect of the  
 16 stratospheric transport of the smoke plume on simulated monthly mean AOT from the Australian wildfires is shown in Fig.  
 17 2d–f. For the simulations that consider an explicit prescription of the aerosol injection into the upper troposphere or lower  
 18 stratosphere, the model shows significantly enhanced AOT in large parts of the Southern Hemisphere. These model results  
 19 indicate, in agreement with observations (Khaykin et al., 2020; Ohneiser et al., 2020), that elevated levels of wildfire smoke  
 20 were sustained over several months and markedly impacted the radiative conditions in the Southern Hemisphere.

21



22  
 23 **Figure 3.** Scatter plots of observed versus simulated monthly mean 550-nm AOT at southern mid and high-latitude  
 24 AERONET stations for January to March 2020. The error bars represent the standard deviation based on daily values.  
 25 Compared are model results for the cases (a) *BASE*, (b) *TP+1*, and (c) *TP-1*. The stations are color-coded respectively: Punta  
 26 Arenas, Chile (53.14°S, 70.89°W), blue triangle; Amsterdam Island (37.80°S, 77.57°E), light blue circles; Antarctic Stations  
 27 Marambio (64.24°S, 56.63°W) and Vechernaya Hill (67.66°S, 46.16°E), purple pentagons; South Pole (90.00°S), light  
 28 purple hexagons.

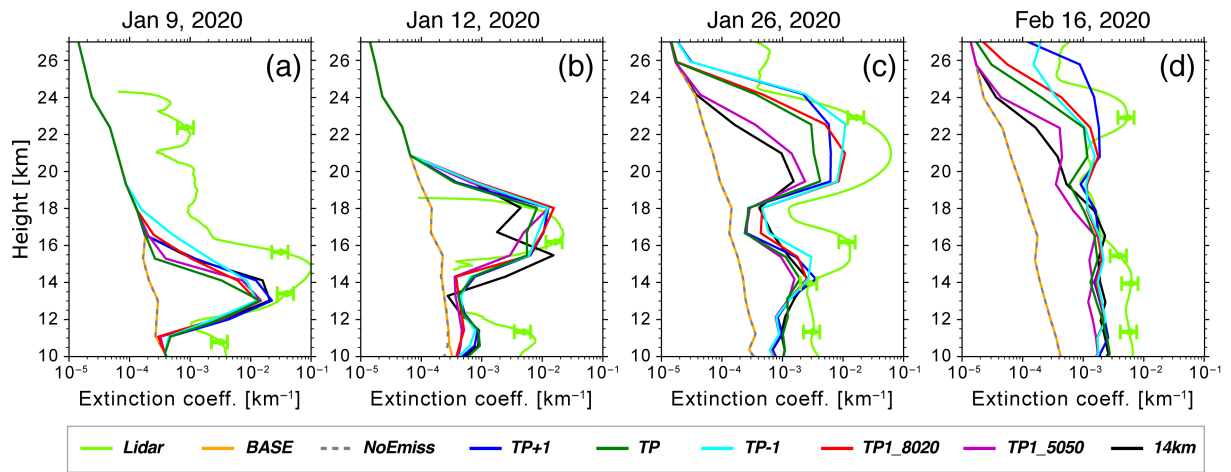
29

1 To evaluate the representation of smoke emission height during the pyroCb days, the model results for the different sensitivity  
2 cases representing different injection heights are compared to sunphotometer measurements of mid- and high-latitude  
3 AERONET stations in the Southern Hemisphere (Holben et al., 1998) for the months January to March 2020 (Fig. 3), and with  
4 ground-based lidar measurement from the PollyXT instrument at Punta Arenas for several days January 2020 (Fig. 4).  
5 Particularly for the AERONET stations located in Antarctica the observed AOT was enhanced in early 2020 compared to  
6 previous years. The agreement of model results with AOT measured at five AERONET stations is clearly better for the cases  
7 *TP+1* and *TP-1* with prescribed fire injection heights compared to the *BASE* case (Fig. 3). Considering the overall very low  
8 levels of AOT at the Southern Hemisphere sites, the alternative injection heights lead to a substantial improvement with up to  
9 65% higher modeled AOT values, e.g. at Punta Arenas. Still, all model results show a negative bias compared to the  
10 observations, indicating that the modeled effects of the smoke will underestimate the actual load and potentially the solar  
11 absorption of stratospheric smoke (see also Sect. 3.2). In the *BASE* case the bias is on average about 30% larger than for the  
12 other cases representing smoke injection into the upper troposphere and lower stratosphere, and the correlation is also slightly  
13 lower, at least compared to the *TP+1*, *TP*, *TP-1* and *TP1\_8020* cases (Table 2). The results for *TP+1*, *TP*, *TP-1*, *TP1\_8020*  
14 and *14km* agree similarly well with the observations, with less agreement for the *TP1\_5050* case. The two cases *BASE* and  
15 *TP1\_5050* therefore represent the observations worst, while no clear best fit is apparent for the other setups. The  
16 underestimation of the fire aerosol loading in all configurations can be caused by a still too low source strength in the GFAS  
17 data *compared to, e.g., Peterson et al. (2021) or is partly due to missing secondary aerosol formed in the plume, which is not*  
18 *considered by the model.*

19  
20 **Table 2.** Statistical key figures for the comparison of measured and simulated AOTs for the different model cases at the  
21 AERONET sun photometer stations Punta Arenas, Chile (53.14°S, 70.89°W), Amsterdam Island (37.80°S, 77.57°E), Antarctic  
22 Stations Marambio (64.24°S, 56.63°W), Vechernaya Hill (67.66°S, 46.16°E), and South Pole (90.00°S). The numbers in bold  
23 denote the case with the best match for the respective statistical variable, the number in brackets the case with least agreement  
24 (excluding case *BASE*). From top: Normalized Root Mean Square error (NRMS; normalized by mean), bias, Pearson  
25 correlation coefficient (R), and probability of correlation (p-value).

	<i>BASE</i>	<i>TP+1</i>	<i>TP</i>	<i>TP-1</i>	<i>TP1_8020</i>	<i>TP1_5050</i>	<i>14km</i>
NRMS	0.61	<b>0.43</b>	0.44	0.45	0.47	(0.51)	0.47
Bias	-0.035	<b>-0.021</b>	<b>-0.021</b>	-0.024	-0.025	(-0.028)	-0.025
Correlation R	0.84	0.84	0.86	<b>0.87</b>	0.86	0.84	0.84
p-value of Correlation	3 x 10 <sup>-4</sup>	3 x 10 <sup>-4</sup>	1 x 10 <sup>-4</sup>	1 x 10 <sup>-4</sup>	2 x 10 <sup>-4</sup>	3 x 10 <sup>-4</sup>	3 x 10 <sup>-4</sup>

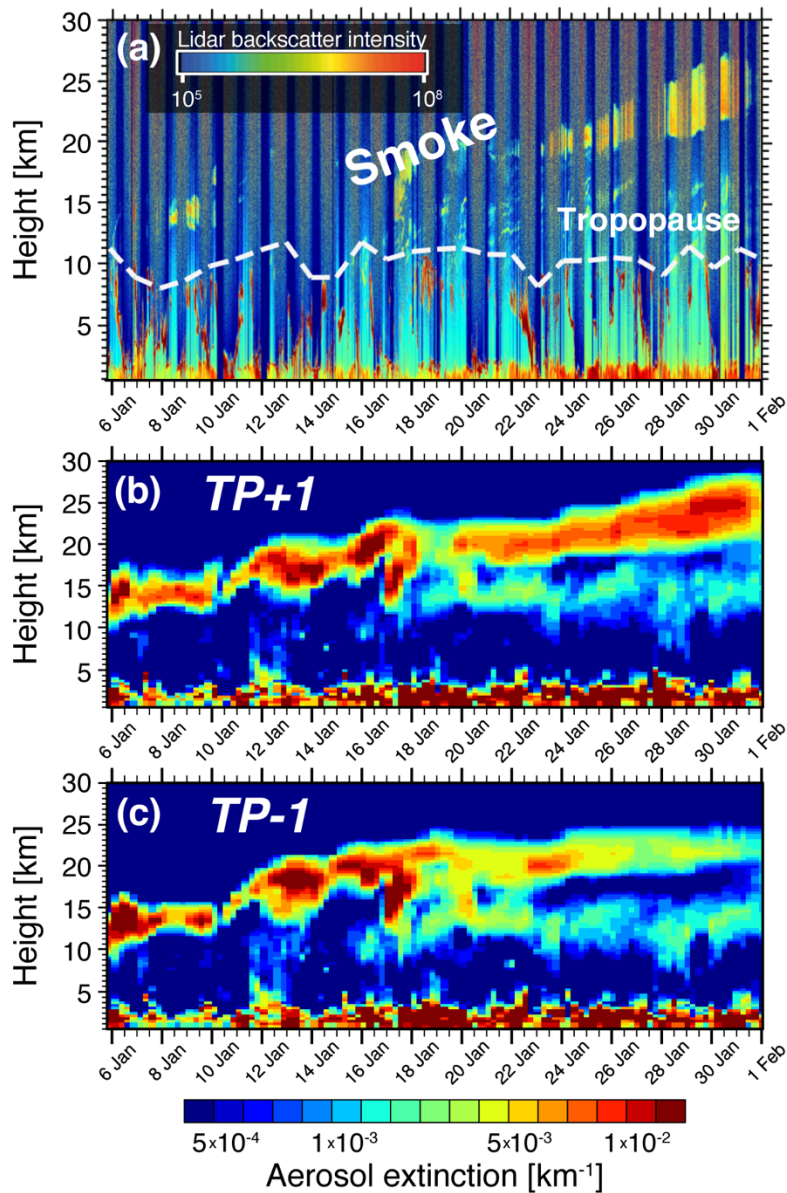
26  
27  
28 The Australian wildfire smoke was observed in early 2020 above Punta Arenas with a ground-based lidar. Pronounced smoke  
29 layers arrived first on 8 January and were clearly above the local tropopause (Fig. 5). The altitude of the observed smoke  
30 plumes steadily increased and reached top heights of 26–27 km at the end of January. For four observations in January and  
31 February 2020 (Fig. 4), the exceptionally thick smoke plume is also evident in the measured extinction coefficients. This  
32 remarkable wildfire smoke layering in terms of structure and magnitude shown in Fig. 4 can only be captured by the model  
33 with stratospheric Australian fire injection heights. Although the grid-cell to point-measurement comparison remains uncertain  
34 in detail, again a tendency for an underestimation of the stratospheric smoke is apparent in Fig. 4. This could be due to an  
35 underestimation of the fire emissions or partly due to missing secondary smoke aerosol, which is not included in the model,  
36 as already suspected from the AOT comparison.



1  
 2 **Figure 4.** Comparison of modeled and observed profiles of aerosol extinction coefficients at the field site in Punta Arenas for  
 3 9, 12, and 26 January and 16 February 2020. Error bars indicate the estimated lidar measurement uncertainties of 30%, values  
 4 below the lidar detection limit are omitted from the graph. See Table 1 for details of the different simulations.

5  
 6 When using the model with original injection height (*BASE* case), none of the structures in the stratosphere can be simulated,  
 7 *just like in the NoEmiss case*. This gives *ing* the clear evidence that the deep pyroconvection in the wildfire hotspots in Southeast  
 8 Australia did emit smoke well above the usually assumed injection heights (Remy et al., 2017; Val Martin et al., 2018). The  
 9 model results also indicate the role of absorptive aerosol heating for the vertical transport of the smoke layer. In the lidar  
 10 profiles, a continuous rise of the smoke layer is visible, with plume center heights increasing from 15 km to 23 km (Figs. 4,  
 11 5). The radiatively-induced self-lifting of smoke can only be reproduced if aerosol-radiation interactions are considered in the  
 12 simulations that finally lead to a considerable absorptive heating and associated buoyancy production (see discussion below,  
 13 *including Fig. 6*).

14

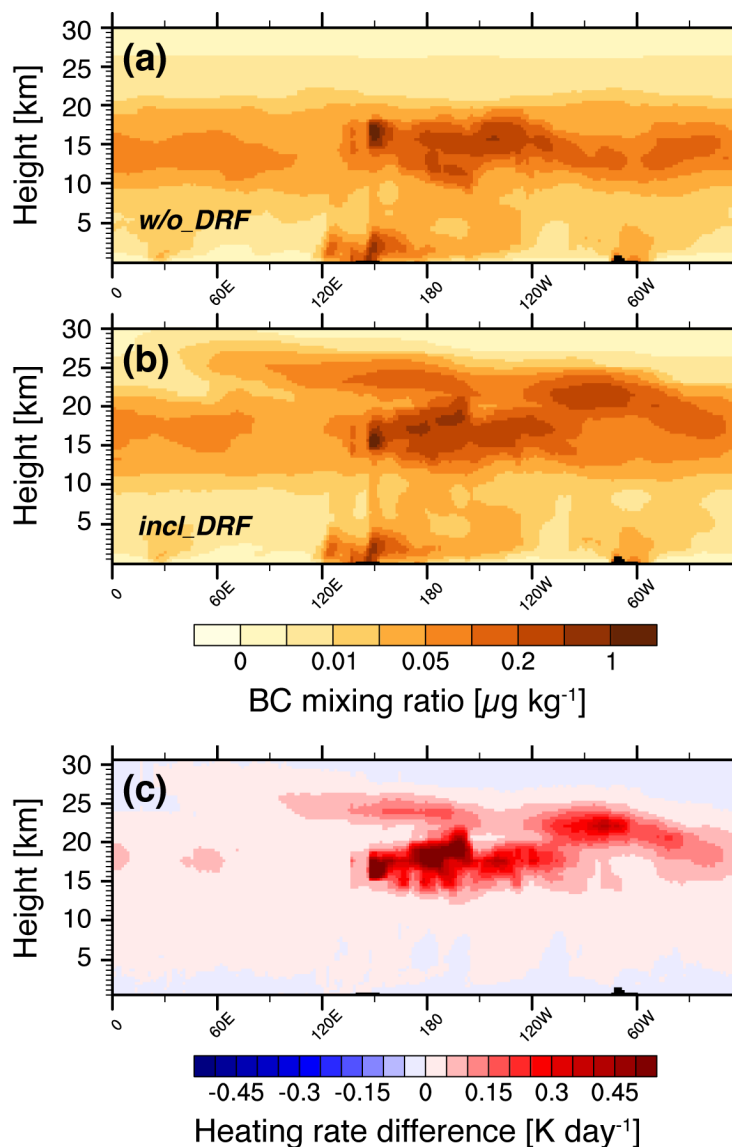


1  
 2 **Figure 5.** Comparison of the pattern of the temporal evolution of stratospheric smoke observed by lidar measurements and  
 3 model results at Punta Arenas, Chile for January 2020. (a) Time-height curtain plot of aerosol attenuated backscatter coefficient  
 4 from the PollyXT lidar at Punta Arenas in southern Chile (53.14°S, 70.89°W). (b, c) Simulated aerosol extinction for the model  
 5 results for the cases *TP+1* and *TP-1*, respectively.

6  
 7 Figure 5 qualitatively compares the development of the smoke extinction profile for the cases *TP+1* and *TP-1* with the aerosol  
 8 backscatter measurements at Punta Arenas, where the rise of the smoke plume center to 24 km by 31 January is particularly  
 9 well matched for the *TP+1* case (cf. Fig. 5a and 5b). For the other model scenarios with prescribed pyroCb smoke injection,  
 10 the plume is lifted to lower heights of 20–21 km by the end of January. But even for the case *TP-1* for which the smoke was  
 11 injected below the tropopause the smoke has lifted into the stratosphere in the model (Fig. 5c). This result underlines the  
 12 importance of a correct representation of fire injection heights, especially for intense wildfires, which is essential to realistically  
 13 assess the radiative effects of smoke plumes.

14 The role of the self-lifting of the smoke caused by the radiative heating of the absorbing BC aerosol in the smoke is also  
 15 illustrated by the vertical distribution of modeled BC mixing ratios (shown for case *TP1\_8020* in Fig. 6) averaged for January  
 16 2020 at 35°S latitude where the fires occurred. The BC mixing ratios for a model simulation where the aerosol is not interacting  
 17 with radiation and thus do not heat the smoke containing atmospheric layers the smoke BC remains below 20 km height, while  
 18 ascending to 24 km for radiatively interactive aerosol in the model (Fig. 6a, b). The monthly heating rate increase caused by

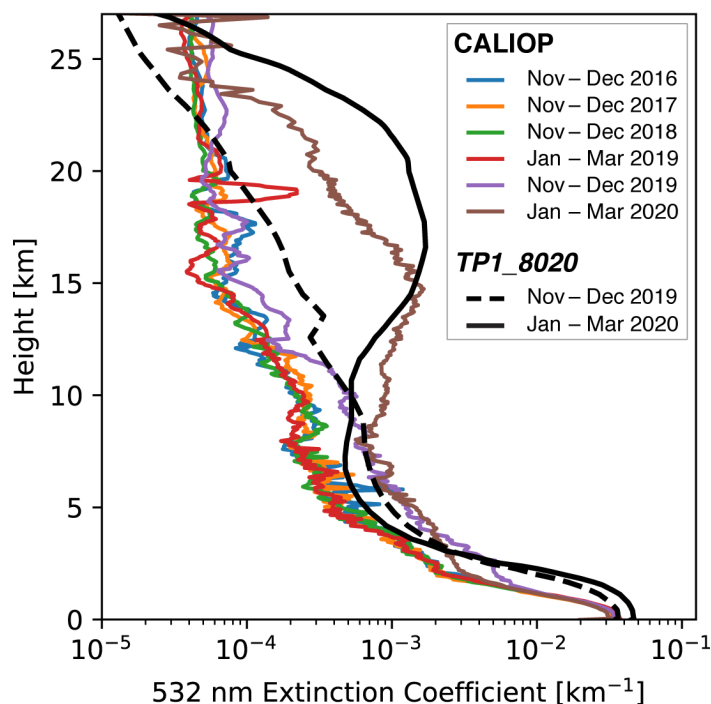
1 the wildfire smoke leading to the self-lifting of the smoke plume is computed as the difference between the *TPI\_8020* and the  
2 *NoEmiss* scenarios (Fig. 6 c). This heating rate reached monthly mean shortwave values up to  $1.7 \text{ K day}^{-1}$  in January 2020.  
3



4  
5 **Figure 6.** Longitude-height distributions of modeled black carbon mixing ratios at latitude  $35^\circ\text{S}$  on average for January 2020  
6 for the case *TPI\_8020* (a) without (*w/o\_DRF*) and (b) with *direct aerosol radiative forcing aerosol-radiation interaction*  
7 (*incl\_DRF*). (c) Change in monthly mean *shortwave* heating rate caused by the absorption of solar radiation by the Australian  
8 wildfire smoke from the pyroCb days, computed as the difference between the cases *TPI\_8020* and *NoEmiss*.  
9

10 Evidence that the 2019-2020 Australian wildfires caused significantly increased upper tropospheric/lower stratospheric aerosol  
11 loading across the Southern Hemisphere is also shown by the CALIOP satellite lidar observations in Figure 7. The extinction  
12 profiles averaged over the latitudes  $30^\circ\text{S}$ – $60^\circ\text{S}$  and the region between Australia and South America ( $145^\circ\text{E}$ – $70^\circ\text{W}$ ) are  
13 considerably enhanced up to 12 km altitude for November to December 2019 compared to those of the previous years 2016 to  
14 2018. In the period from January to March 2020, the extinction is again massively increased in the altitude range from 8 km  
15 to 24 km with a peak at an altitude of 15 km. Interesting to note is also the fact that the CALIOP profile for January to March  
16 2020 (brown line in Fig. 7) comprises about 50 times as many retrievals in the upper troposphere and lower stratosphere as  
17 the other averaging periods. This difference in sampling with far more CALIOP aerosol detections is clearly a response to the  
18 Australian wildfires. Comparing the model results with the CALIOP observations, it can be seen that our approach of  
19 prescribing pyroconvective smoke injection also reproduces well the vertical distribution of Australian wildfire aerosol across

1 the southern mid-latitudes between Australia and South America, as shown in example of scenario *TP1\_8020*. Discrepancies  
 2 at altitudes above 15 km are likely related to the CALIOP sampling bias discussed in Sect. 2.3 while between 4 km and 12 km  
 3 the model underestimates the fire aerosol likely because of the smoke injection mainly in the tropopause region, which is also  
 4 to be seen in Fig. 4. In the boundary layer, a slight overestimation occurs again in the model. Overall, these misrepresentations  
 5 are probably due to the comparatively simple approach to simulate the deep pyroconvective smoke injection.  
 6



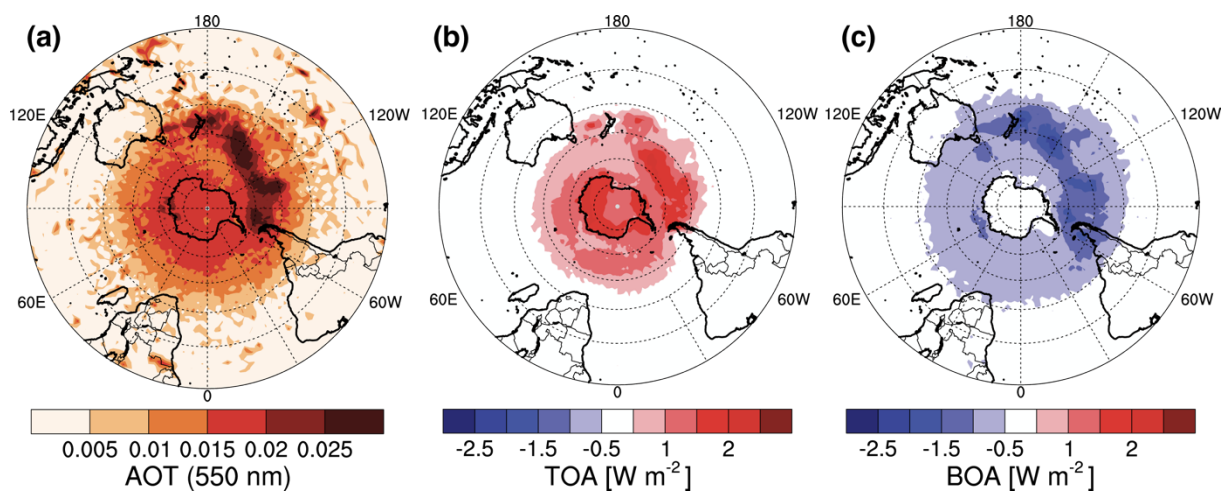
7  
 8 **Figure 7.** Mean vertical profile of the 532-nm retrieved extinction coefficient from CALIOP (colored lines) for several selected  
 9 periods within 2016 – 2020 and the 2019-2020 results from the ECHAM6.3-HAM2.3 model for the *TP1\_8020* case, averaged  
 10 over the area covering the latitudes and longitudes between 30°S and 60°S and 145°E and 70°W.

### 11 3.2 Estimates of direct radiative perturbation

12 Regionally varying climate forcing agents such as aerosols substantially modulate the greenhouse forcing by anthropogenic  
 13 aerosol particles and gases. We find that the individual extreme Australian fire event caused a significant direct instantaneous  
 14 shortwave radiative forcing signal as shown in Fig. 8. The instantaneous direct aerosol radiative forcing in the model is  
 15 calculated by calling the radiation scheme twice in each simulation in order to diagnose the radiative forcing without affecting  
 16 the atmospheric conditions such as dynamics, moisture fields and clouds. The instantaneous shortwave forcing due to the  
 17 elevated smoke layers ranged up to  $+0.50 \text{ W m}^{-2}$  at TOA averaged for the Southern Hemisphere for January to March 2020  
 18 under all-sky conditions for the scenario *TP+1* (Table 3). This would correspond to a global-average TOA shortwave radiative  
 19 forcing of  $+0.25 \text{ W m}^{-2}$ . In Table 3, the range of forcing estimates is given for all considered model scenarios except the clearly  
 20 unrealistic cases *BASE* and *TP1\_5050*. This instantaneous forcing by the singular fire event is of similar magnitude as the  
 21 latest multi-model estimate of the global-average instantaneous forcing for all anthropogenic black carbon with  $+0.28$  ( $0.13$ –  
 22  $0.37$ )  $\text{W m}^{-2}$  (Thornhill et al., 2021). Previous studies, in contrast, found negative TOA radiative forcing estimates  
 23 of  $-1 \text{ W m}^{-2}$  (Khaykin et al., 2020) for this event but only considered clear-sky situations in which the smoke aerosols appear  
 24 brighter over the dark ocean surface due to the dominant scattering effect (Bellouin et al., 2020). However, the elevated  
 25 Australian smoke layers that contain absorbing black carbon were located above clouds and to a large extent also above the  
 26 strongly reflecting snow and ice cover of the Antarctic. Over such bright surfaces, the balance between aerosol scattering and

1 absorption is shifted and smoke aerosol darkens the scene seen from TOA. At surface (bottom of atmosphere, BOA), the clear-  
 2 sky instantaneous solar radiative forcing was estimated to ranging from  $-0.68$  to  $-0.81$   $\text{W m}^{-2}$  for the different model scenarios.  
 3 This corresponds to the short-term surface dimming caused by a large volcanic eruption (Andersson et al., 2015; Schmidt et  
 4 al., 2018). On the other hand, according to the model, the smoke-containing air layer itself experienced significant absorptive  
 5 heating with maximum shortwave heating rates of  $1.7$   $\text{K day}^{-1}$  on average in January 2020 for the TP+1 case. While the  
 6 effective TOA radiative forcing is expected to be low due to stratospheric adjustment to the instantaneous forcing, these heating  
 7 rate changes may have the potential to trigger responses in the atmospheric dynamics. Khaykin et al. (2020) actually showed  
 8 that a self-sustained 1000-km anticyclonic vortex formed as a result, which traveled through the stratosphere for weeks,  
 9 accompanied by a local ozone reduction. *Such a vortex is also seen in the 50-hPa wind fields of the ECHAM6.3-HAM2.3*  
 10 *simulations (not shown here). The analysis of the atmospheric dynamic effects, however, is the subject of a follow-up study.*

11



12

13 **Figure 8.** AOT and estimates of shortwave radiative forcing of the 2019-2020 Australian wildfire smoke plume in the Southern  
 14 Hemisphere. Model results of (a) AOT and (b, c) instantaneous shortwave radiative forcing of the elevated smoke aerosol  
 15 layer, averaged over the months January to March 2020. All values are differences between model ECHAM6.3-HAM2.3  
 16 results with Australian wildfire smoke injection for the scenario *TP+1* and *NoEmiss*. The instantaneous radiative flux  
 17 differences are shown for all-sky conditions at top (TOA; b) and bottom of the atmosphere (BOA; c).

18 Major uncertainties in the model estimates of aerosol radiative forcing are due to uncertainties in AOT, particle angular  
 19 scattering properties (asymmetry parameter) and in particular aerosol absorption that is characterized by its single scattering  
 20 albedo (SSA). A recent comprehensive analysis of aircraft data indicates that model parameterizations may generally  
 21 overestimate absorption by biomass burning aerosol due to an insufficient representation of the mixing state for fire aerosol  
 22 (Brown et al., 2021). In our model, at the height of maximum extinction of the smoke plume, the ratio of black to total carbon  
 23 (BC/(BC+OC) mixing ratio) is approximately 0.05 – 0.08, corresponding to a particle SSA between 0.82–0.85 at 550 nm. This  
 24 is within the range of other aerosol models (Bellouin et al., 2020; Brown et al., 2021). Accordingly, the model may be biased  
 25 toward too strong positive forcing. On the other hand, inversion results of multispectral lidar observations in the Northern  
 26 Hemisphere from the strong 2017 Canadian fires yield an SSA of 0.80 for the stratospheric smoke (Haarig et al., 2018). For  
 27 the 2019-2020 Australian fires, the lidar inversion method of Veselovskii et al. (2002) was applied to the Polly  
 28 multiwavelength backscatter and extinction observations on 26 January 2020 to obtain values for the single scattering albedo  
 29 *by Ohneiser et al. (2022). They show an SSA of 0.79 with an uncertainty range of 0.05 for the smoke-filled vortex on 26*  
 30 *January above Punta Arenas in Chile, which is considered also representative for the mid and high latitudes in the Southern*  
 31 *Hemisphere (Ohneiser et al., 2022) other smoke measurements. A steady decline in the lidar depolarization ratio at Punta*

1 *Arenas indicates an aging of the fire plume by mid-February, while high lidar ratios at 532 nm were observed with values*  
2 *ranging from 75 to 115 sr (mean 91 sr) well beyond that time, further indicating strongly absorbing smoke particles and low*  
3 *SSA values (see Fig. 8 in Ohmeiser et al., 2022). The lidar simulator of the model, in comparison, provides slightly lower lidar*  
4 *ratios at 532 nm between 70 and 100 sr for the stratospheric smoke layer. ~~This, in turn, would suggest that the model~~*  
5 *~~underestimates absorption by the 2019–2020 Australian stratospheric smoke, which is supported by the comparison to~~*  
6 *~~observed extinction profiles (Fig. 4).~~ Regarding the asymmetry parameter, it is difficult to make an evaluation because the*  
7 *exact morphology of the smoke particles is not known. In the model, the asymmetry parameter for the Australian smoke is*  
8 *about 0.6 at 550 nm, which is a typical value for wildfire aerosol (e.g., Reid et al., 2005). ~~In conclusion, together with the low~~*  
9 *~~bias of the modeled smoke AOT, we argue that our results illustrate a conservative estimate for the positive TOA forcing of~~*  
10 *~~this outstanding biomass burning event. In terms of single scattering albedo and lidar ratio, this would imply that the model~~*  
11 *~~underestimates absorption by Australian stratospheric smoke in 2019–2020, whereas it tends to overestimate it in other,~~*  
12 *~~weaker vegetation fires according to Brown et al. (2021). The comparison to the ground-based remote sensing retrievals~~*  
13 *~~provides the best available confirmation for the optical properties of the 2019–2020 Australian fire smoke during this~~*  
14 *~~outstanding biomass burning event in the model. However, we cannot rule out the possibility that away from the core plume,~~*  
15 *~~more reflective secondary organic particles made a larger contribution to the stratospheric fire aerosol. In addition, dilution~~*  
16 *~~impacts on smoke aging with regard to particle size and mixing state could have influenced the smoke optical properties more~~*  
17 *~~strongly toward less absorption (Hodshire et al., 2021; Sellitto et al., 2022). The resulting effects of this spatial and temporal~~*  
18 *~~variability and evolution in the optical properties of the Australian wildfire plume would not or only insufficiently be~~*  
19 *~~represented by the model.~~*

20

21 **Table 3.** Shortwave instantaneous direct radiative forcing ( $\text{W m}^{-2}$ ) of the elevated smoke plume during the 2019–2020  
22 Australian wildfires. The estimates are calculated differences between the instantaneous shortwave irradiances of the model  
23 results including stratospheric smoke injection and the case *NoEmiss* without smoke emission from Southeastern Australia for  
24 the pyroCb days, averaged over the Southern Hemisphere. Ranges are given for the different configurations *TP+1*, *TP1\_8020*,  
25 *TP*, *TP-1*, and *14km* (see Table 1). Shown are the differences for all-sky and clear-sky conditions at top and bottom of  
26 atmosphere (TOA, BOA) averaged for the months January to March (JFM) 2020.

	TOA all sky	TOA clear sky	BOA all sky	BOA clear sky
<b>January 2020</b>	+0.45 – +0.56	-0.02 – -0.05	-0.54 – -0.61	-0.86 – -0.97
<b>February 2020</b>	+0.40 – +0.57	+0.003 – +0.07	-0.42 – -0.51	-0.67 – -0.84
<b>March 2020</b>	+0.25 – +0.37	-0.01 – +0.07	-0.28 – -0.38	-0.46 – -0.63
<b>JFM Average</b>	+0.37 – +0.50	-0.02 – +0.02	-0.42 – -0.50	-0.68 – -0.81

#### 27 4 Implications and perspectives

28 In order to determine the impact of biomass burning aerosol on the global energy budget, accurately estimating emission fluxes  
29 and their injection height in the atmosphere is essential. State-of-the-science global aerosol-atmosphere models generally  
30 consider biomass burning aerosol, but still show uncertainties in the spatio-temporal distribution. In particular, large emission  
31 events like Australia's Black Summer wildfires of 2019–2020 are underrepresented.



1 A key uncertainty is the vertical injection of fire smoke into the atmosphere that may ultimately cause misrepresentation of  
2 the plume evolution. The results of this study show that using fire emission data from the GFAS dataset and injecting the  
3 smoke into the tropopause region for pyroCb events gives results that are reasonable, although still somewhat underestimated  
4 in the present study.

5 The substantial increase in stratospheric AOT in the Southern Hemisphere, and thus the perturbation of the radiative balance,  
6 from the southeastern Australian wildfire smoke from just four days of pyroconvection events is remarkable. The local sub-  
7 grid scale nature of fire plume rising challenges the representation in models beyond the 1-km scale, but especially in global  
8 models that do not resolve convection (Paugam et al., 2016; Veira et al., 2015). In these coarse models, the vertical distribution  
9 of fire emissions is based on climatological profiles (Val Martin et al., 2018) or prescribed by injection heights estimated from  
10 satellite retrievals of fire radiative power (Remy et al., 2017). While this is appropriate for the majority of vegetation fires, the  
11 vertical transport during deep pyroconvective events with potentially far-reaching effects is most likely underestimated due to  
12 the obstruction of satellite observations by dense pyroCb clouds (Remy et al., 2017). Adequate plume-rise parameterizations  
13 exist particularly for mesoscale chemistry-transport models, but have not found their way into climate modeling on a wider  
14 scale yet (Paugam et al., 2016; Val Martin et al., 2018; Veira et al., 2015).

15 Consequently, while aerosol-climate models have been shown to overestimate the radiative forcing biomass burning aerosol  
16 in general (Brown et al., 2020), they likely underestimate the wildfire aerosol impacts on the energy balance for pyro-  
17 convective fires, as the vertical location of the smoke is fundamental to its radiative impact. To solve this, adjustments are  
18 needed in the representation of biomass burning injection. By implementing a more realistic emission scenario based on  
19 aerosol-profiling observations but still using the emission fluxes from the standard GFAS database, we enhance the ability of  
20 our model to capture the extreme 2019-2020 Australian pyroCb event and can thus showcase the potential of global aerosol-  
21 climate models to realistically reproduce the spatio-temporal evolution of smoke plumes of intense wildfires. This further  
22 allows for an improved estimate on aerosol impacts on radiation and clouds. Ultimately, these improvements are essential to  
23 any estimate on the Earth's energy balance and climate state. In this respect, it is particularly important to make climate models  
24 capable of dealing with exceptional outliers of wildfires, which are anticipated to increase in frequency and severity worldwide  
25 in response to anthropogenic climate warming (Abatzoglou et al, 2019; Jolly et al., 2015; Wotton et al, 2017). The increased  
26 risk of serious wildfires is related to extreme heat and drier conditions, as well as record-low snow cover in boreal regions  
27 (Box et al., 2019; Dowdy et al., 2019). More frequent and intense fire weather extremes will also increase the likelihood of  
28 deep pyroconvection (Dowdy et al., 2019). *Along with this, of course, there continues to be the challenge and need to develop*  
29 *better model representations of smoke aging and secondary aerosol formation, and the associated evolution of particle optical*  
30 *properties in the lifecycle of wildfire plumes.* In essence, biomass burning emissions are an important source of aerosol  
31 particles, and individual wildfires are shown to have more widespread effects than previously assumed. An as-accurate-as-  
32 possible description, therefore, is key to successfully estimate aerosol climate effects, and future climate projections will  
33 clearly benefit from an improved aerosol representation in Earth system models.

34  
35  
36  
37  
38  
39  
40  
41  
42  
43

1 *Code availability.* The ECHAM-HAMMOZ code is maintained and made available to the scientific community under  
2 <https://redmine.hammoz.ethz.ch>. The availability is regulated under the HAMMOZ Software Licence Agreement that can be  
3 downloaded from [https://redmine.hammoz.ethz.ch/attachments/download/291/License\\_ECHAM-HAMMOZ\\_June2012.pdf](https://redmine.hammoz.ethz.ch/attachments/download/291/License_ECHAM-HAMMOZ_June2012.pdf).  
4 *Data availability.* The ECHAM6.3-HAM2.3 model output, on which the figures are based, and the analyzed lidar profiles are  
5 accessible at Zenodo from <https://zenodo.org/deposit/5571545> (Heinold et al., 2021) The ground-based Polly lidar data time  
6 series is visualized at [polly.tropos.de](https://polly.tropos.de) and will become publicly available via the data portal of the European Aerosols, Clouds  
7 and Trace gases Research Infrastructure (ACTRIS) when its implementation phase is completed. Until then, raw data is  
8 available on request via [polly@tropos.de](mailto:polly@tropos.de). AERONET data can be obtained with the Aerosol Robotic Network download tool  
9 [https://aeronet.gsfc.nasa.gov/cgi-bin/webtool\\_opera\\_v2\\_new](https://aeronet.gsfc.nasa.gov/cgi-bin/webtool_opera_v2_new). The Aerosol Optical Thickness CDR used in this study was  
10 acquired from NOAA's National Climatic Data Center (<http://www.ncdc.noaa.gov>).

11 *Author contribution.* BH and IT conceived the idea and led the study. BH and AK performed the model development and ran  
12 the simulations. NS provided the LIDAR simulator for ECHAM6.3-HAM2.3. IT, FS, KS, AK and BH focused on the analysis  
13 and interpretation of model results. RS analyzed NASA's AVHRR AOT climatology data, and DV researched the Earth history  
14 context. KO, HB and BB performed the lidar analysis and interpretation. BH, IT, KS, FS and RS wrote the paper with  
15 contributions from all co-authors. All authors participated in the revision and editing of the paper.

16 *Competing interests.* The authors declare no competing interests.

17 *Acknowledgements.* The authors thank Johannes Quaas and Albert Ansmann for sharing their comments on an earlier version  
18 of the manuscript. We also would like to thank the whole DACAPO-PESO team for ensuring the continuous operation of all  
19 the different instruments at Punta Arenas. Special thanks to Patric Seifert, Martin Radenz, Cristofer Jimenez and Ronny  
20 Engelmann for taking care of the lidar, measurements and ensuring high-quality data operation at the cost of spending many  
21 lifetime hours in an airplane. Igor Veselovskii is acknowledged for computing lidar inversion products. The AVHRR Aerosol  
22 Optical Thickness Climate Data Record (CDR) was originally developed by Xuepeng Zhao and colleagues for NOAA's CDR  
23 Program. We thank P. Goloub, V. E. Cachorro Revilla, P. Seifert, J. Butler, B. Holben, and A. Chaikovsky for their efforts in  
24 establishing and maintaining the AERONET sites Amsterdam Island, Marambio, Punta Arenas, South Pole, and Vechernaya  
25 Hill. The ECHAM-HAMMOZ model is developed by a consortium composed of ETH Zürich, Max Planck Institute for  
26 Meteorology, Forschungszentrum Jülich, the University of Oxford, the Finnish Meteorological Institute and the Leibniz  
27 Institute for Tropospheric Research (TROPOS) and managed by the Center for Climate Systems Modelling (C2SM) at ETH  
28 Zürich. We are grateful for computing time from the Deutsches Klimarechenzentrum (DKRZ). Computing resources at DKRZ  
29 were granted under project number bb1004. We thank Pasquale Sellitto and an anonymous reviewer for their helpful feedback.

30

## 31 **References**

32 Abatzoglou, J. T., Williams, A. P., and Barbero, R.: Global emergence of anthropogenic climate change in fire weather indices,  
33 *Geophys. Res. Lett.*, 46, 326–336, <https://doi.org/10.1029/2018GL080959>, 2019.

34 Andersson, S., Martinsson, B., Vernier, J.-P., Friberg, J., Brenninkmeijer, C. A. M., Hermann, M., van Velthoven, P. F. J., and  
35 Zahn, A.: Significant radiative impact of volcanic aerosol in the lowermost stratosphere. *Nat. Commun.*, 6, 7692,  
36 <https://doi.org/10.1038/ncomms8692>, 2015.

37 Ban-Weiss, G. A., Cao, L., Bala, G., and Caldeira, K.: Dependence of climate forcing and response on the altitude of black  
38 carbon aerosols, *Clim. Dyn.*, 38, 897–911, <https://doi.org/10.1007/s00382-011-1052-y>, 2012.

1 Baars, H., Kanitz, T., Engelmann, R., Althausen, D., Heese, B., Komppula, M., Preißler, J., Tesche, M., Ansmann, A.,  
2 Wandinger, U., Lim, J.-H., Ahn, J. Y., Stachlewska, I. S., Amiridis, V., Marinou, E., Seifert, P., Hofer, J., Skupin, A.,  
3 Schneider, F., Bohlmann, S., Foth, A., Bley, S., Pfüller, A., Giannakaki, E., Lihavainen, H., Viisanen, Y., Hooda, R. K.,  
4 Pereira, S. N., Bortoli, D., Wagner, F., Mattis, I., Janicka, L., Markowicz, K. M., Achtert, P., Artaxo, P., Pauliquevis, T.,  
5 Souza, R. A. F., Sharma, V. P., van Zyl, P. G., Beukes, J. P., Sun, J., Rohwer, E. G., Deng, R., Mamouri, R.-E., and Zamorano,  
6 F.: An overview of the first decade of PollyNET: an emerging network of automated Raman-polarization lidars for continuous  
7 aerosol profiling, *Atmos. Chem. Phys.*, 16, 5111–5137, <https://doi.org/10.5194/acp-16-5111-2016>, 2016.

8 Baars, H., Ansmann, A., Ohneiser, K., Haarig, M., Engelmann, R., Althausen, D., Hanssen, I., Gausa, M., Pietruczuk, A.,  
9 Szkop, A., Stachlewska, I. S., Wang, D., Reichardt, J., Skupin, A., Mattis, I., Trickl, T., Vogelmann, H., Navas-Guzmán, F.,  
10 Haefele, A., Acheson, K., Ruth, A. A., Tatarov, B., Müller, D., Hu, Q., Podvin, T., Goloub, P., Veselovskii, I., Pietras, C.,  
11 Haeffelin, M., Fréville, P., Sicard, M., Comerón, A., Fernández García, A. J., Molero Menéndez, F., Córdoba-Jabonero, C.,  
12 Guerrero-Rascado, J. L., Alados-Arboledas, L., Bortoli, D., Costa, M. J., Dionisi, D., Liberti, G. L., Wang, X., Sannino, A.,  
13 Papagiannopoulos, N., Boselli, A., Mona, L., D'Amico, G., Romano, S., Perrone, M. R., Belegante, L., Nicolae, D., Grigorov,  
14 I., Gialitaki, A., Amiridis, V., Soupiona, O., Papayannis, A., Mamouri, R.-E., Nisantzi, A., Heese, B., Hofer, J., Schechner, Y.  
15 Y., Wandinger, U., and Pappalardo, G.: The unprecedented 2017–2018 stratospheric smoke event: decay phase and aerosol  
16 properties observed with the EARLINET, *Atmos. Chem. Phys.*, 19, 15183–15198, [https://doi.org/10.5194/acp-19-15183-](https://doi.org/10.5194/acp-19-15183-2019)  
17 2019, 2019.

18 Bellouin, N., Quaas, J., Gryspeerdt, E., Kinne, S., Stier, P., Watson-Parris, D., Boucher, O., Carslaw, K. S., Christensen, M.,  
19 Daniau, A.-L., Dufresne, J.-L., Feingold, G., Fiedler, S., Forster, P. M., Gettelman, A., Haywood, J. M., Lohmann, U.,  
20 Malavelle, F., Mauritsen, T., McCoy, D. T., Myhre, G., Mülmenstädt, J., Neubauer, D., Possner, A., Rugenstein, M., Sato, Y.,  
21 Schulz, M., Schwartz, S. E., Sourdeval, O., Storelvmo, T., Toll, V., Winker, D., and Stevens, B.: Bounding global aerosol  
22 radiative forcing of climate change, *Rev. of Geophys.*, 58, e2019RG000660, <https://doi.org/10.1029/2019RG000660>, 2020.

23 Boer, M. M., Resco de Dios, V., and Bradstock, R. A.: Unprecedented burn area of Australian mega forest fires, *Nat. Clim.*  
24 *Chang.*, 10, 171–172, <https://doi.org/10.1038/s41558-020-0716->, 2020.

25 Boers, R., de Laat, A. T., Stein Zweers, D. C., and Dirksen, R. J.: Lifting potential of solar-heated aerosol layers, *Geophys.*  
26 *Res. Lett.*, 37, L24802, doi:10.1029/2010GL045171, 2010.

27 Boone, C. D., Bernath, P. F., and Fromm, M. D.: Pyrocumulonimbus stratospheric plume injections measured by the ACE-  
28 FTS, *Geophys. Res. Lett.*, 47(15), 1-9, e2020GL088442, <https://doi.org/10.1029/2020GL088442>, 2020.

29 Boucher, O., Randall, D., Artaxo, P., Bretherton, C., Feingold, G., Forster, P. M., Kerminen, V.-M., Kondo, Y., Liao, H.,  
30 Lohmann, U., Rasch, P., Satheesh, S. K., Sherwood, S., Stevens, B. and Zhang, X. Y.: Clouds and aerosols. In *Climate Change*  
31 *2013: The Physical Science Basis. Contribution of Working Group I to the Fifth Assessment Report of the Intergovernmental*  
32 *Panel on Climate Change.* Stocker, T. F., Qin, D., Plattner, G.-K., Tignor, M., Allen, S. K., Doschung, J., Nauels, A., Xia, Y.,  
33 Bex, V., and Midgley, P. M. Eds. Cambridge University Press, pp. 571-657, doi:10.1017/CBO9781107415324.016, 2013.

34 Bond, T. C., Doherty, S. J., Fahey, D. W., Forster, P. M., Berntsen, T., DeAngelo, B. J., Flanner, M. G., Ghan, S., Kärcher,  
35 B., Koch, D., Kinne, S., Kondo, Y., Quinn, P. K., Saroim, M. C., Schultz, M. G., Schulz, M., Venkataraman, C., Zhang, H.,  
36 Zhang, S., Bellouin, N., Guttikunda, S. K., Hopke, P. K., Jacobson, M. Z., Kaiser, J. W., Klimont, Z., Lohmann, U., Schwarz,  
37 J. P., Shindell, D., Storelvmo, T., Warren, S. G., and Zender, C. S.: Bounding the role of black carbon in the climate system:  
38 A scientific assessment, *J. Geophys. Res. Atmos.*, 118, 5380–5552, <http://dx.doi.org/10.1002/jgrd.50171>, 2013.

1 Bowman, D. M. J. S., Balch, J. K., Artaxo, P., Bond, W. J., Carlson, J. M., Cochrane, M. A., D'Antonio, C. M., DeFries, R.  
2 S., Doyle, J. C., Harrison, S. P., Johnston, F. H., Keeley, J. E., Krawchuk, M. A., Kull, C. A., Marston, J. B., Moritz, M. A.,  
3 Prentice, I. C., Roos, C. I., Scott, A. C., Swetnam, T. W., van der Werf, G. R., and Pyne, S. J.: Fire in the Earth system,  
4 *Science*, 324, 481–4, <https://doi.org/10.1126/science.1163886>, 2009.

5 Box, J. E., Colgan, W. T., Christensen, T. R., Schmidt, N. M., Lund, M., Parmentier, F.-J. W., Brown, R., Bhatt, U. S.,  
6 Euskirchen, E. S., Romanovsky, V. E., Walsh, J. E., Overland, J. E., Wang, M., Corell, R. W., Meier, W. N., Wouters, B.,  
7 Mernild, S., Mård, ., Pawlak, J., and Olsen, M. S.: Key indicators of Arctic climate change: 1971–2017, *Environ. Res. Lett.*,  
8 14, 045010, <http://dx.doi.org/10.1088/1748-9326/aafc1b>, 2019.

9 Brown, H., Liu, X., Pokhrel, R., Murphy, S., Lu, Z., Saleh, R., Mielonen, T., Kokkola, H., Bergman, T., Myhre, G., Skeie, R.  
10 B., Watson-Paris, D., Stier, P., Johnson, B., Bellouin, N., Schulz, M., Vakkari, V., Beukes, J. P., van Zyl, P. G., Liu, S., and  
11 Chand, D.: Biomass burning aerosols in most climate models are too absorbing, *Nat. Commun.*, 12, 277,  
12 <https://doi.org/10.1038/s41467-020-20482-9>, 2021.

13 Dowdy, A. J., Ye, H., Pepler, A., Thatcher, M., Osbrough, S. L., Evans, J. P., Di Virgilio, G., and McCarthy, N.: Future  
14 changes in extreme weather and pyroconvection risk factors for Australian wildfires, *Sci. Rep.* 9, 10073,  
15 <https://doi.org/10.1038/s41598-019-46362-x>, 2019.

16 Engelmann, R., Kanitz, T., Baars, H., Heese, B., Althausen, D., Skupin, A., Wandinger, U., Komppula, M., Stachlewska, I.  
17 S., Amiridis, V., Marinou, E., Mattis, I., Linné, H., and Ansmann, A.: The automated multiwavelength Raman polarization  
18 and water-vapor lidar PollyXT: the neXT generation, *Atmos. Meas. Tech.*, 9, 1767–1784, [https://doi.org/10.5194/amt-9-1767-](https://doi.org/10.5194/amt-9-1767-2016)  
19 2016, 2016.

20 Fromm, M., Lindsey, D. T., Servranckx, R., Yue, G., Trickl, T., Sica, R., Doucet, P., and Godin-Beekmann, S.: The Untold  
21 Story of Pyrocumulonimbus, *Bull. Amer. Meteor. Soc.*, 91(9), 1193-1210, <https://doi.org/10.1175/2010BAMS3004.1>, 2010.

22 Fromm, M., Peterson, D., and Di Girolamo, L.: The Primary Convective Pathway for Observed Wildfire Emissions in the  
23 Upper Troposphere and Lower Stratosphere: A Targeted Reinterpretation, *J. Geophys. Res.* 124, 13254–13272,  
24 <https://doi.org/10.1029/2019JD031006>, 2019.

25 GDAS: Global Data Assimilation System, meteorological database, available at: <https://www.ready.noaa.gov/gdas1.php>, last  
26 access: December 21, 2020.

27 Giles, D. M., Sinyuk, A., Sorokin, M. G., Schafer, J. S., Smirnov, A., Slutsker, I., Eck, T. F., Holben, B. N., Lewis, J. R.,  
28 Campbell, J. R., Welton, E. J., Korkin, S. V., and Lyapustin, A. I.: Advancements in the Aerosol Robotic Network  
29 (AERONET) Version 3 database – automated near-real-time quality control algorithm with improved cloud screening for  
30 Sun photometer aerosol optical depth (AOD) measurements, *Atmos. Meas. Tech.*, 12, 169–209, [https://doi.org/10.5194/amt-](https://doi.org/10.5194/amt-12-169-2019)  
31 12-169-2019, 2019.

32 Giorgetta, M., Jungclaus, J., Reick, C., Legutke, S., Brovkin, V., Crueger, T., Esch, M., Fieg, K., Glushak, K., Gayler, V.,  
33 Haak, H., Hollweg, H.-D., Kinne, S., Kornbluh, L., Matei, D., Mauritsen, T., Mikolajewicz, U., Müller, W., Notz, D., Raddatz,  
34 T., Rast, S., Roeckner, E., Salzmann, M., Schmidt, H., Schnur, R., Segschneider, J., Six, K., Stockhause, M., Wegner, J.,  
35 Widmann, H., Wieners, K.-H., Claussen, M., Marotzke, J., and Stevens, B.: CMIP5 simulations of the Max Planck Institute  
36 for Meteorology (MPI-M) based on the MPI-ESM-LR model: The amip experiment, served by ESGF, World Data Center for  
37 Climate (WDCC) at DKRZ, <https://doi.org/10.1594/WDCC/CMIP5.MXELam>, 2012.

1 Haarig, M., Ansmann, A., Baars, H., Jimenez, C., Veselovskii, I., Engelmann, R., and Althausen, D.: Depolarization and lidar  
2 ratios at 355, 532, and 1064 nm and microphysical properties of aged tropospheric and stratospheric Canadian wildfire smoke,  
3 *Atmos. Chem. Phys.*, 18, 11847–11861, <https://doi.org/10.5194/acp-18-11847-2018>, 2018.

4 Heinold, B., Baars, H., Barja, B., Kubin, A., Ohneiser, K., and Tegen, I.: Data for paper publication 'Important role of  
5 stratospheric injection height for the distribution and radiative forcing of smoke aerosol from the 2019/2020 Australian  
6 wildfires' (v2.0) [data set], <https://doi.org/10.5281/zenodo.5571545>, 2021.

7 Hersbach, H., Bell, B., Berrisford, P., Hirahara, S., Horányi, ., Muñoz-Sabater, ., Nicolas, J., Peubey, C., Radu, R., Schepers,  
8 D., Simmomons, A., Soci, C., Abdalla, S., Abellan, X., Balsamo, G., Bechtold, P., Biavati, G., Bidlot, J., Bonavita, M., De  
9 Chiara, G., Dahlgren, P., Dee, D., Diamantakis, M., Dragani, R., Flemming, J., Forbes, R., Fuentes, M., Geer, A., Haimberger,  
10 L., Healy, S., Hogan, R. J., Hólm, E., Janisková, M., Keeley, S., Laloyaux, P., Lopez, P., Lupu, C., Radnoti, G., de Rosnay,  
11 P., Rozum, I., Vamborg, F., Villaume, S., and Thépaut, J.-N: The ERA5 global reanalysis, *Q. J. R. Meteorol Soc.*, 146, 1999-  
12 2049, <https://doi.org/10.1002/qj.3803>, 2020.

13 Hirsch, E. and Koren, I.: Record-breaking aerosol levels explained by smoke injection into the stratosphere, *Science*, 371,  
14 Issue 6535, 1269-1274, doi:10.1126/science.abe1415, 2021.

15 *Hodshire, A. L., Ramnarine, E., Akherati, A., Alvarado, M. L., Farmer, D. K., Jathar, S. H., Kreidenweis, S. M., Lonsdale, C.*  
16 *R., Onasch, T. B., Springston, S. R., Wang, J., Wang, Y., Kleinman, L. I., Sedlacek III, A. J., and Pierce, J. R.: Dilution impacts*  
17 *on smoke aging: evidence in Biomass Burning Observation Project (BBOP) data, Atmos. Chem. Phys.*, 21, 6839–6855,  
18 <https://doi.org/10.5194/acp-21-6839-2021>, 2021.

19 Holben, B. N., Eck, T. F., Slutsker, I., Tanré, D., Buis, J. P., Setzer, A., Vermote, E., Reagan, J. A., Kaufman, Y. J., Nakajima,  
20 T., Lavenue, F., Jankowiak, I., and Smirnov, A.: AERONET—A Federated Instrument Network and Data Archive for Aerosol  
21 Characterization. *Remote Sens. Environ.*, 66, 1-16, [https://doi.org/10.1016/S0034-4257\(98\)00031-5](https://doi.org/10.1016/S0034-4257(98)00031-5), 1998.

22 Jiang, Y., Lu, Z., Liu, X., Qian, Y., Zhang, K., Wang, Y., and Yang, X.-Q.: Impacts of global open-fire aerosols on direct  
23 radiative, cloud and surface-albedo effects simulated with CAM5, *Atmos. Chem. Phys.*, 16, 14805–14824,  
24 <https://doi.org/10.5194/acp-16-14805-2016>, 2016.

25 Jolly, W. et al.: Climate-induced variations in global wildfire danger from 1979 to 2013, *Nat. Commun.*, 6, 7537, 2015.

26 Jumelet, J., Klekociuk, A. R., Alexander, S. P., Bekki, S., Hauchecorne, A., Vernier, J. P., Fromm, M., and Keckhut, P.:  
27 Detection of aerosols in Antarctica from long-range transport of the 2009 Australian wildfires, *J. Geophys. Res.-Atmos.*, 125,  
28 e2020JD032542, <https://doi.org/10.1029/2020JD032542>, 2020.

29 Kablick, G. P. III, Allen, D. R., Fromm, M. D., and Nedoluha, G. E.: Australian pyroCb smoke generates synoptic-scale  
30 stratospheric anticyclones, *Geophys. Res. Lett.*, 47, e2020GL088101, <https://doi.org/10.1029/2020GL088101>, 2020.

31 Kaiser, J. W., Heil, A., Andreae, M. O., Benedetti, A., Chubarova, N., Jones, L., Morcrette, J.-J., Razinger, M., Schultz, M.  
32 G., Suttie, M., and van der Werf, G. R.: Biomass burning emissions estimated with a global fire assimilation system based on  
33 observed fire radiative power, *Biogeosciences*, 9, 527–554, <https://doi.org/10.5194/bg-9-527-2012>, 2012.

- 1 Khaykin, S., Legras, B., Bucci, S., Sellitto, P., Isaksen, I., Tencé, J., Bekki, S., Bourassa, A., Rieger, L., Zawada, D., Jumelet,  
2 J., and Godin-Beekmann, S.: The 2019/20 Australian wildfires generated a persistent smoke-charged vortex rising up to 35 km  
3 altitude, *Commun. Earth Environ.*, 1, 22, <https://doi.org/10.1038/s43247-020-00022-5>, 2020.
- 4 Kirchmeier-Young, M. C., Gillett, N. P., Zwiers, F. W., Cannon, A. J., and Anslow, F. S.: Attribution of the influence of  
5 human-induced climate change on an extreme fire season, *Earth Future*, 7, 2–10, <https://doi.org/10.1029/2018EF001050>, 2019.
- 6 *Liu, Z., Kar, J., Zeng, S., Tackett, J., Vaughan, M., Avery, M., Pelon, J., Getzewich, B., Lee, K.-P., Magill, B., Omar, A., Lucker,*  
7 *P., Trepte, C., and Winker, D.: Discriminating between clouds and aerosols in the CALIOP version 4.1 data products, *Atmos.*  
8 *Meas. Tech.*, 12, 703–734, <https://doi.org/10.5194/amt-12-703-2019>, 2019.*
- 9 Lund, M. T., Aamaas, B., Stjern, C. W., Klimont, Z., Berntsen, T. K., and Samset, B. H.: A continued role of short-lived  
10 climate forcings under the Shared Socioeconomic Pathways, *Earth Syst. Dynam.*, 11, 977–993, [https://doi.org/10.5194/esd-11-](https://doi.org/10.5194/esd-11-977-2020)  
11 [977-2020](https://doi.org/10.5194/esd-11-977-2020), 2020.
- 12 Morgan, G. W., Tolhurst, K. G., Poynter, M. W., Cooper, N., McGuffog, T., Ryan, R., Wouters, M. A., Stephens, N., Black,  
13 P., Sheehan, D., Leeson, P., Whight, S., and Davey, S. M.: Prescribed burning in south-eastern Australia: History and future  
14 directions, *Aust. For.*, 83, 4-28, doi:10.1080/00049158.2020.1739883, 2020.
- 15 Ohneiser, K., Ansmann, A., Baars, H., Seifert, P., Barja, B., Jimenez, C., Radenz, M., Teisseire, A., Floutsi, A., Haarig, M.,  
16 Foth, A., Chudnovsky, A., Engelmann, R., Zamorano, F., Bühl, J., and Wandinger, U.: Smoke of extreme Australian bushfires  
17 observed in the stratosphere over Punta Arenas, Chile, in January 2020: optical thickness, lidar ratios, and depolarization ratios  
18 at 355 and 532 nm, *Atmos. Chem. Phys.*, 20, 8003–8015, <https://doi.org/10.5194/acp-20-8003-2020>, 2020.
- 19 Ohneiser, K., Ansmann, A., Kaifler, B., Chudnovsky, A., Barja, B., Knopf, D. A., Kaifler, N., Baars, H., Seifert, P., Villanueva,  
20 D., Jimenez, C., Radenz, M., Engelmann, R., Veselovskii, I., and Zamorano, F.: Australian wildfire smoke in the stratosphere:  
21 the decay phase in 2020/21 and impact on ozone depletion, *Atmos. Chem. Phys. Discuss.* [preprint],  
22 <https://doi.org/10.5194/acp-2021-1097>, in review, 2022.
- 23 Paugam, R., Wooster, M., Freitas, S., and Val Martin, M.: A review of approaches to estimate wildfire plume injection height  
24 within large-scale atmospheric chemical transport models, *Atmos. Chem. Phys.*, 16, 907–925, [https://doi.org/10.5194/acp-16-](https://doi.org/10.5194/acp-16-907-2016)  
25 [907-2016](https://doi.org/10.5194/acp-16-907-2016), 2016.
- 26 Peterson, D. A., Campbell, J. R., Hyer, E. J., Fromm, M. D., Kablick, G. P., Cossuth, J. H., and DeLand, M. T.: Wildfire-  
27 driven thunderstorms cause a volcano-like stratospheric injection of smoke, *npj Clim. Atmos. Sci.*, 1, 30,  
28 <https://doi.org/10.1038/s41612-018-0039-3>, 2018.
- 29 Peterson, D. A., Fromm, M. D., McRae, R. H. D. et al.: Australia’s Black Summer pyrocumulonimbus super outbreak reveals  
30 potential for increasingly extreme stratospheric smoke events, *npj Clim. Atmos. Sci.*, 4, 38, [https://doi.org/10.1038/s41612-](https://doi.org/10.1038/s41612-021-00192-9)  
31 [021-00192-9](https://doi.org/10.1038/s41612-021-00192-9), 2021.
- 32 Radenz, M., Bühl, J., Seifert, P., Baars, H., Engelmann, R., Barja González, B., Mamouri, R.-E., Zamorano, F., and Ansmann,  
33 A.: Hemispheric contrasts in ice formation in stratiform mixed-phase clouds: disentangling the role of aerosol and dynamics  
34 with ground-based remote sensing, *Atmos. Chem. Phys.*, 21, 17969–17994, <https://doi.org/10.5194/acp-21-17969-2021>, 2021.

- 1 Reid, J. S., Eck, T. F., Christopher, S. A., Koppmann, R., Dubovik, O., Eleuterio, D. P., Holben, B. N., Reid, E. A., and Zhang,  
2 J.: A review of biomass burning emissions part III: intensive optical properties of biomass burning particles, *Atmos. Chem.*  
3 *Phys.*, 5, 827–849, <https://doi.org/10.5194/acp-5-827-2005>, 2005.
- 4 Rémy, S., Veira, A., Paugam, R., Sofiev, M., Kaiser, J. W., Marenco, F., Burton, S. P., Benedetti, A., Engelen, R. J., Ferrare,  
5 R., and Hair, J. W.: Two global data sets of daily fire emission injection heights since 2003, *Atmos. Chem. Phys.*, 17, 2921–  
6 2942, <https://doi.org/10.5194/acp-17-2921-2017>, 2017.
- 7 Schmidt, A., Mills, M. J., Ghan, S., Gregory, J. M., Allan, R. P., Andrews, T., Bardeen, C. G., Conley, A., Forster, P. M.,  
8 Gettelman, A., Portmann, R. W., Solomon, S., and Toon, O. B.: Volcanic Radiative Forcing From 1979 to 2015, *J. Geophys.*  
9 *Res.-Atmos.*, 123, 12491–12508, <https://doi.org/10.1029/2018JD028776>, 2018.
- 10 *Sellitto, P., Belhadji, R., Kloss, C., and Legras, B.: Radiative impacts of the Australian bushfires 2019–2020 – Part 1: Large-*  
11 *scale radiative forcing, EGUsphere [preprint], <https://doi.org/10.5194/egusphere-2022-42>, 2022.*
- 12 Streets, D. G., Yan, F., Chin, M., Diehl, T., Mahowald, N., Schultz, M., Wild, M., Wu, Y., and Yu, C.: Anthropogenic and  
13 natural contributions to regional trends in aerosol optical depth, 1980–2006. *J. Geophys. Res.*, 114, D00D18,  
14 <https://doi.org/10.1029/2008JD011624>, 2009.
- 15 Tegen, I., Neubauer, D., Ferrachat, S., Siegenthaler-Le Drian, C., Bey, I., Schutgens, N., Stier, P., Watson-Parris, D., Stanelle,  
16 T., Schmidt, H., Rast, S., Kokkola, H., Schultz, M., Schroeder, S., Daskalakis, N., Barthel, S., Heinold, B., and Lohmann, U.:  
17 The global aerosol–climate model ECHAM6.3–HAM2.3 – Part 1: Aerosol evaluation, *Geosci. Model Dev.*, 12, 1643–1677,  
18 <https://doi.org/10.5194/gmd-12-1643-2019>, 2019.
- 19 Thornhill, G. D., Collins, W. J., Kramer, R. J., Olivié, D., Skeie, R. B., O'Connor, F. M., Abraham, N. L., Checa-Garcia, R.,  
20 Bauer, S. E., Deushi, M., Emmons, L. K., Forster, P. M., Horowitz, L. W., Johnson, B., Keeble, J., Lamarque, J.-F., Michou,  
21 M., Mills, M. J., Mulcahy, J. P., Myhre, G., Nabat, P., Naik, V., Oshima, N., Schulz, M., Smith, C. J., Takemura, T., Tilmes,  
22 S., Wu, T., Zeng, G., and Zhang, J.: Effective radiative forcing from emissions of reactive gases and aerosols – a multi-model  
23 comparison, *Atmos. Chem. Phys.*, 21, 853–874, <https://doi.org/10.5194/acp-21-853-2021>, 2021.
- 24 Veira, A., Kloster, S., Wilkenskjaeld, S., and Remy, S.: Fire emission heights in the climate system – Part 1: Global plume  
25 height patterns simulated by ECHAM6–HAM2, *Atmos. Chem. Phys.*, 15, 7155–7171, [https://doi.org/10.5194/acp-15-7155-](https://doi.org/10.5194/acp-15-7155-2015)  
26 2015, 2015.
- 27 Val Martin M., Kahn R. A., and Tosca M. G.: A Global Analysis of Wildfire Smoke Injection Heights Derived from Space-  
28 Based Multi-Angle Imaging, *Remote Sens.*, 10, 1609, <https://doi.org/10.3390/rs10101609>, 2018.
- 29 Veselovskii, I., Kolgotin, A., Griaznov, V., Müller, D., Wandinger, U., and Whiteman, D. N.: Inversion with regularization  
30 for the retrieval of tropospheric aerosol parameters from multiwavelength lidar sounding, *Appl. Opt.* 41, 3685–3699, doi:  
31 10.1364/AO.41.003685, 2002.
- 32 Vignati, E., Wilson, J., and Stier, P.: M7: An efficient size resolved aerosol microphysics module for large scale aerosol  
33 transport, *J. Geophys. Res.*, 109, D22202, <https://doi.org/10.1029/2003JD004485>, 2004.
- 34 Ward, M., Tulloch, A. I. T., Radford, J. Q., Williams, B. A., Reside, A. E., Macdonald, S. L., Mayfield, H. J., Maron, M.,  
35 Possingham, H. P., Vine, S. J., O'Connor, J. L., Massingham, E. J., Greenville, A. C., Woinarski, J. C. Z., Garnett, S. T.,

- 1 Lintermans, M., Scheele, B. C., Carwardine, J., Nimmo, D. G., Lindenmayer, D. B., Kooyman, R. M., Simmonds, J. S., Sonter,  
2 L. J., and Watson, J. E. M.: Impact of 2019–2020 mega-fires on Australian fauna habitat, *Nat. Ecol. Evol.*, 4, 1321–1326,  
3 <https://doi.org/10.1038/s41559-020-1251-1>, 2020.
- 4 Watson-Parris, D., Schutgens, N., Winker, D., Burton, S. P., Ferrare, R. A., and Stier, P.: On the limits of CALIOP for  
5 constraining modeled free tropospheric aerosol, *Geophys. Res. Lett.*, 45, 9260–9266, <https://doi.org/10.1029/2018GL078195>,  
6 2018.
- 7 Winker, D. M., Tackett, J. L., Getzewich, B. J., Liu, Z., Vaughan, M. A., and Rogers, R. R.: The global 3-D distribution of  
8 tropospheric aerosols as characterized by CALIOP, *Atmos. Chem. Phys.*, 13, 3345–3361, <https://doi.org/10.5194/acp-13-3345-2013>, 2013.
- 10 Wintle, B. A., Legge, S., and Woinarski, J. C. Z.: After the Megafires: What Next for Australian Wildlife?, *Trends Ecol. Evol.*,  
11 35, 753–757, <https://doi.org/10.1016/j.tree.2020.06.009>, 2020.
- 12 Wotton, B. M., Flannigan, M. D., and Marshall, G. A.: Potential climate change impacts on fire intensity and key wildfire  
13 suppression thresholds in Canada, *Environ. Res. Lett.*, 12, 095003, <http://dx.doi.org/10.1088/1748-9326/aa7e6e>, 2017.
- 14 Yu, P., Davis, S. M., Toon, O. B., Portmann, R. W., Bardeen, C. G., Barnes, J. E., Telg, H., Maloney, C., and Rosenlof, K. H.:  
15 Persistent stratospheric warming due to 2019–2020 Australian wildfire smoke, *Geophys. Res. Lett.*, 48, e2021GL092609,  
16 <https://doi.org/10.1029/2021GL092609>, 2021.
- 17 Zhang, K., O'Donnell, D., Kazil, J., Stier, P., Kinne, S., Lohmann, U., Ferrachat, S., Croft, B., Quaas, J., Wan, H., Rast, S.,  
18 and Feichter, J.: The global aerosol-climate model ECHAM-HAM, version 2: sensitivity to improvements in process  
19 representations, *Atmos. Chem. Phys.*, 12, 8911–8949, <https://doi.org/10.5194/acp-12-8911-2012>, 2012.
- 20 Zhao, X., and NOAA CDR Program: NOAA Climate Data Record (CDR) of AVHRR Daily and Monthly Aerosol Optical  
21 Thickness (AOT) over Global Oceans, Version 3.0, published 2016-12-01, NOAA National Centers for Environmental  
22 Information, doi:10.7289/V5BZ642P, last access 2021-01-27, 2017.