RESEARCH ARTICLE
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Key Points:
• Smoke drives strong diurnal responses in convection, cloud cover, and precipitation that cannot be observed by polar-orbiting satellites.
• The radiative effect of smoke is sensitive to plume longevity and the response of high-altitude clouds that exert a strong positive effect.
• Simulations suggest the localized suppression of convection scales with the optical depth of smoke over the Amazon rainforest.

Abstract
Absorbing aerosol from biomass burning impacts the hydrological cycle and radiation fluxes both directly and indirectly via modifications to convective processes and cloud development. Using the ICOsahedral Non-hydrostatic modelling framework in a regional configuration with 1,500 m convection-permitting resolution, we isolate the response of the Amazonian atmosphere to biomass burning smoke via enhanced cloud droplet number concentrations $N_d$ (aerosol-cloud interactions; ACI) and changes to radiative fluxes (aerosol-radiation interactions; ARI) over a period of 8 days. We decompose ARI into contributions from reduced shortwave radiation and localized heating of the smoke. We show ARI influences the formation and development of convective cells: surface cooling below the smoke drives suppression of convection that increases with smoke optical depth, while the elevated heating promotes initial suppression and subsequent intensification of convection overnight; a corresponding diurnal response (repeating temporal response day-after-day) from high precipitation rates is shown. Enhanced $N_d$ (ACI) perturbs the bulk cloud properties and suppresses low-to-moderate precipitation rates. Both ACI and ARI result in enhanced high-altitude ice clouds that have a strong positive longwave radiative effect. Changes to low-cloud coverage (ARI) and albedo (ACI) drive an overall negative shortwave radiative effect, that slowly increases in magnitude due to a moistening of the boundary layer. The overall net radiative effect is dominated by the enhanced high-altitude clouds, and is sensitive to the plume longevity. The considerable diurnal responses that we simulate cannot be observed by polar orbiting satellites widely used in previous work, highlighting the potential of geostationary satellites to observe large-scale impacts of aerosols on clouds.

Plain Language Summary
There remain important uncertainties on how smoke from forest and grassland fires impacts the past, present, and future climates. In this study, we use a detailed model of the atmosphere over the Amazon rainforest to understand and quantify the processes by which smoke influences clouds and rain via two pathways: the first driven by changes to the absorption of solar radiation through the smoky atmosphere, and the second driven by an increase in the number of cloud droplets due to smoke particles. We find that the diurnal cycle of convection that drives much of the Amazon rainfall is greatly affected by changes to the radiation with less activity during the day and increasing activity overnight. The more numerous cloud droplets make the clouds brighter and help suppress rainfall rates. Changes to both radiation and cloud droplet number result in more extensive and thicker ice-phase clouds that exert a warming effect on the climate; the longer the smoke plume persists for, the stronger the warming effect. Our findings highlight important processes that are not sufficiently represented in global climate models, and also highlight a need to use time-resolved geostationary satellite observations of the region to capture more of the diurnal cycle.

1. Introduction
Airborne aerosol particles, such as sea salt, mineral dust, or carbonaceous material, impact our climate via interactions with solar and terrestrial radiation known as aerosol-radiation interactions (ARI) and via their ability to act as cloud condensation nuclei (CCN) and subsequent follow-up effects (known as aerosol-cloud interactions; ACI). Wildfires occur annually in many tropical regions across the globe (e.g., the Amazon, central Africa, Borneo) and often cover vast areas with optically thick layers of aerosol that can persist for weeks to months. These plumes can transport the carbonaceous biomass burning aerosol (BBA) high into the atmosphere and far from the source (Holanda et al., 2020), demonstrating the potential for widespread
non-local impacts. BBA influences the environment via both ARI and ACI. The strongly absorbing optical properties of BBA result in an instantaneous localized heating of the plume and a reduction in surface solar radiation (Feingold et al., 2005). BBA also acts as CCN and therefore have the potential to enhance the cloud droplet number concentration ($N_d$). The potential for BBA to have far-reaching and prolonged impacts on radiation and cloud cover, coupled with remaining uncertainties in BBA optical and physical properties, has resulted in BBA being identified as a key source of uncertainty in future projections of the climate (Bond et al., 2013; Boucher et al., 2013).

The Amazon region is the world’s largest forested region, representing a vast store of carbon and playing an important role in the world’s hydrological cycle (Lovejoy & Nobre, 2019). Wildfires in the Amazon forest and the surrounding savannah coincide with the dry season (August to October) and may become more numerous under drought conditions (Aragão et al., 2007; Ten Hoeve et al., 2012). Future climate projections predict an increase in the frequency and intensity of drought events (Stocker et al., 2013), and a lengthening of the dry season (Boisier et al., 2015), both of which may enhance fire activity. Additionally, in the past decades anthropogenic activities, mainly from large-scale clearing and burning for agriculture, have greatly impacted the number and scale of fires in the region (Morton et al., 2008) and have resulted in an increase in savannah grassland at the expense of forest cover (Nobre et al., 2016). A recent upward trend in deforestation (de Oliveira et al., 2020) highlights both recent increases in fire activity, but also substantial variation in decadal deforestation trends.

Observations made at field sites across the Amazon suggest precipitation is suppressed by the presence of BBA alongside intensification in certain conditions. Andreae et al. (2004) report evidence of smoke reducing the size of cloud droplets, delaying and elevating the altitude for the onset of precipitation, and increasing the vertical extent of clouds. An overall suppression of the warm rain process was observed using a ground-based radar at Manaus (Gonçalves et al., 2015), though some intensification of convection was observed in more unstable atmospheres. Camponogara et al. (2014) and Bevan et al. (2008) also observed correlations between aerosol optical depth (AOD) and the suppression of precipitation at several ground sites across the Amazon. The spatial extent of the Amazon and associated BBA plumes necessitates the use of spaceborne retrievals to examine BBA impacts on the larger-scale environment. Kaufman and Nakajima (1993) found that increases in AOD coincided with warmer cloud tops (lower altitude) pointing toward a suppression of convection instead of invigoration; this conclusion was also reported by Jiang et al. (2018). Koren et al. (2008) reported that at low loadings (below AOD of ~0.3) low cloud cover increases with AOD, but decreases at higher loadings and Ten Hoeve et al. (2011) reported a similar relationship for cloud optical depth. These relationships may however occur in part due to confounding factors such as meteorology (Gryspeerdt et al., 2017; Yu et al., 2007). The previously listed satellite-based studies have a local overpass time of ~1330 local solar time (LST), which poses the question as to whether these observations are representative of the full diurnal response of the cloud to BBA.

Modeling studies have been performed on a range of resolutions and domain sizes over the region and highlight some key impacts to the large-scale environment. BBA reduces surface fluxes (Zhang et al., 2008) that can directly lead to reduced cloudiness (Feingold et al., 2005) and promote a stabilization of the boundary layer (BL) via surface cooling and heating aloft (Liu et al., 2020; Zhang et al., 2009). The changes to the thermodynamic structure of the BL can have widespread impacts on circulation and moisture transport (Lee et al., 2014; Yu et al., 2007; Zhang et al., 2008, 2009), the suppression of convection at the surface (Liu et al., 2020; Martins et al., 2009; Wu et al., 2011), and potentially enhance convection aloft (Wu et al., 2011). Generally, these studies suggest that cumulative precipitation in the dry season decreases due to smoke presence (Liu et al., 2020; Martins et al., 2009; Thornhill et al., 2018; Wu et al., 2011) but peak intensities may increase (Martins et al., 2009). There is less agreement as to how liquid water path (LWP), ice water path (IWP), and cloudiness responds, which may be due to the differing mechanisms by which ARI and ACI impact cloud formation and evolution. One such mechanism, referred to as CCN-induced cloud invigoration, links an increase in CCN with an increase in buoyancy and convective cloud updraft speed, which may occur via enhanced condensation of cloud droplets or enhanced flux of condensate to the freezing level (e.g., Fan et al., 2018; Khain et al., 2005; Lebo, 2018; Rosenfeld et al., 2008). The lack of consensus among modeling studies has implications for the overall radiative effect of BBA. Generally, an increase in low-cloud cover cools the climate, whereas an increase in high-cloud cover warms the climate. Without a
In this modeling study, we aim to isolate the response of the clouds, precipitation, and larger-scale environment to BBA from microphysical effects (ACI) and radiative effects (ARI). We then further decompose ARI to focus on the roles of surface cooling (due to a reduction in downwelling shortwave radiation) and in situ heating of the BBA layer. We use a large domain (3,000 × 2,000 km) with high spatial resolution (1.5 km) that resolves the diurnal cycle of convection and permits the full evolution of mesoscale convective systems across the Amazon. We also use a semi-idealized treatment of the BBA plume, grounded by observations, to reduce the uncertainty that arises from the aerosol distribution and evolution.

The results will highlight the important diurnal response (repeating temporal pattern of the response day-after-day) of clouds and precipitation and the processes that drive them, while allowing us to identify important gaps in our knowledge and help understand conflicting observations and results in other modeling studies. The remainder of the paper is set out as follows. In Section 2, we briefly describe the atmospheric model that is being used for this study along with the setup of the domain, then describe the representation of aerosol effects in the model, and outline the experiments being performed. In Section 3, we present the main results of our study starting with a stepwise focus on the domain response in each experiment, followed by the top-of-atmosphere (TOA) radiative effects, and then we examine the relationship between BBA optical depth and the suppression of convection. Finally, in Section 4, we put our results into a broader context and provide a comparison with other studies, before concluding the study.

**2. Methodology**

**2.1. Model Description and Setup**

We use the ICOsahedral Non-hydrostatic (ICON) modelling framework in a regional, limited-area, configuration; a full description and evaluation of the model is provided by Zängl et al. (2015). This setup has been used in previous studies (Dagan et al., 2020; Klocke et al., 2017) to study clouds and ACI. The model domain, covering an area of $\sim 6 \times 10^6$ km$^2$, is centered on the Amazon between longitudes 68°W–4°W and latitudes 15°S–3°N (Figure 1) on a triangular grid with a horizontal resolution of $\sim 1,500$ m, which is able to represent convection without the need for sub-grid scale parameterizations of mid-level and deep convection (Klocke et al., 2017). We use the terrain-following Smooth Level Vertical SLEVE coordinate system (Leuenberger et al., 2010) in the vertical, with 75 levels extending up to an altitude of 30 km. Radiation is represented interactively using the Rapid Radiative Transfer Model for GCMs (RRTM-G) that operates on 16 longwave bands and 14 shortwave bands (Iacono et al., 2008; Mlawer et al., 1997). Cloud microphysical processes are represented using the double-moment microphysics scheme of Seifert and Beheng (2006) that predicts the mass and number evolution of five hydrometeor species; microphysics and radiation schemes are coupled via the cloud droplet effective radius that is calculated in the former. In these experiments, the $N_d$ at each grid point is prescribed using a relationship between AOD and $N_d$ (see Section 2.2 for details). The exchange of heat, moisture, and momentum fluxes at the surface are interactively represented using the TERRA interactive surface scheme (Schrodin & Heise, 2002).

ERA5 reanalysis from the Copernicus Climate Change Service (C3S, 2017) is used for initialization of the model and for boundary conditions that are supplied every 6 h throughout the simulation and linearly interpolated between the time intervals. This frequency is sufficient given the scale of the domain and characteristic timescale for changes to the large-scale environment which is greater than 6 h (Dagan et al., 2018). All simulations are run for 8 days with a model time step of 12 s; 3D fields are output every 3 h, and 2D fields

![Figure 1. September climatology of 850 hPa winds (arrows; scale shown above the plot) and total cloud fraction (contours; labeled as percentages) from the ERA-Interim reanalysis data set (2000–2018) over the Amazon rainforest. The green box shows the boundaries of the model domain used in this study. The pink and cyan points show the locations of the Alta Floresta and Rio Branco AERONET sites, respectively.](image-url)
output hourly. The simulations are initialized at 0000 UTC on August 31, 2007 and run until 0000 UTC September 8, 2007. This date coincides with a prolonged period of elevated AOD over the Amazon during the dry season, shown in the Aerosol Robotic Network (AERONET; Holben et al., 2001) measurements of AOD in Figure 2.

### 2.2. Representation of Aerosol Effects

In these experiments, we simulate a persistent plume of prescribed aerosol. The horizontal and vertical distribution of aerosol is represented using the Max Planck Institute Aerosol Climatology version 2 - Simple Plume model (MACv2-SP; Stevens et al., 2017). MACv2-SP allows the user to prescribe a number of idealized radiatively active plumes each with a prescribed location, spatial structure, vertical profile, annual cycle, angstrom exponent, AOD at plume center, single scattering albedo (SSA), and asymmetry parameter (three previous parameters given at 550 nm). In our simulations, we isolate the “South America” plume and prescribe parameters as set in Stevens et al. (2017) except for the AOD and SSA (see Table 1 in Stevens et al., 2017). The resulting vertical and horizontal distribution of aerosol is shown in Figures 3a and 3b. The plume location and horizontal extent are in good agreement with a climatology presented by Malavelle et al. (2019); the plume center is at 62°W 10°S and decreases toward the north and east, with background AOD values <0.1 at the Atlantic coast. MACv2-SP places much of the BBA plume within the lower 2 or

### Table 1

**Details of the Experiments Being Performed in This Study**

| Experiment name | Details |
|-----------------|---------|
| CONTROL         | Background AOD (0.06) and $N_d$ (100 cm$^{-3}$) distribution |
| MPHYS           | Perturb $N_d$ distribution only: isolate microphysical effect of BBA on $N_d$ |
| RAD             | Perturb AOD distribution only: isolate full radiative effect of BBA |
| RAD-scat        | As RAD experiment but with SSA of 1.0: isolate radiative effect of BBA at surface only |
| TOTAL           | Perturb both AOD and $N_d$ distribution: full radiative and microphysical effects from BBA |

*Note. All simulations are performed for 8 days with the same initial conditions and lateral boundary conditions.*
3 km of the atmosphere with a peak at ∼0.5 km altitude (Figure 3a) which is in agreement with observations (Andreae et al., 2004; Gonzalez-Alonso et al., 2019; Reddington et al., 2019).

Two setups using the MACv2-SP model were used to represent the radiative effect of aerosol. The first is the standard representation of BBA, with a maximum AOD of 1.5 and an SSA of 0.87, which results in localized heating of the aerosol layer and a reduced surface flux of shortwave (SW) radiation. This peak AOD is within sustained magnitudes observed during this period (Figure 2a; see also Schafer et al., 2008) and only represents a small region of the domain (Figure 3b). The second setup is designed to only represent the reduced downward surface SW flux and associated surface flux changes. Increasing the SSA to 1.0 removes the localized heating but simultaneously enhances the SW diffuse component reaching the surface; the result is a weaker reduction in surface SW flux when compared to the first setup. In order to achieve a consistent SW reduction at the surface, the AOD was increased by a factor of 2.55 (determined using SOCRATES: Suite of Community Radiation Codes based on Edwards & Slingo, 1996). This results in an unrealistically high AOD but for this experiment, we are only concerned with the effect that the changes to the surface fluxes have on the evolution of the domain.

The MACv2-SP model includes a parameterized ACI effect that represents the enhancement of \( N_d \) due to increased anthropogenic aerosol. The parameterization presented by Stevens et al. (2017), and shown in Figure 3c (dashed line), is based on a mean climatology of \( N_d \) and AOD retrieved by MODIS over the ocean. As will be shown below, observations of \( N_d \) from campaigns throughout the Amazon routinely report \( N_d \) exceeding 100 cm\(^{-3}\) and indeed this maximum, as reflected by the parameterization, may be better suited as a minimum background value for the region. Therefore, the current MACv2-SP parameterization, which
may well be representative of large-scale perturbations over marine environments, is not suitable for our region of interest.

In our simulations, we represent the enhancement of \( N_d \) due to BBA using a relationship between AOD (total column optical depth measured in clear skies) and \( N_d \) based on in situ measurements made throughout the troposphere during three campaigns over the Amazon. We stress that we are not presenting a new parameterization, but rather producing a realistic relationship that allows us to explore the impact of enhanced \( N_d \) to the simulated domain in an idealized setup. The goAMAZON campaign (Martin et al., 2016) took place at Manaus and included localized flights during the dry season. Mean values of \( N_d \) for each flight (10th–90th percentile of data above 1 cm\(^{-3}\)) were determined using in situ measurements of \( N_d \) from the CDP and 2DS probes (https://www.arm.gov/research/campaigns/amf2014goamazon), and corresponding values of AOD taken from the Manaus AERONET station. The ACRIDICON-CHUVA campaign (Wendisch et al., 2016) included flights that extended over much of the Amazon. Median \( N_d \) measured during the flights are presented by Polonik et al. (2020) and used here. AOD measurements were taken from AERONET stations either at Manaus or Alta Floresta, depending on availability and proximity to the flight. The SAMBBA campaign (details in Brito et al., 2014 and Darbyshire et al., 2019) was based at Porto Velho with flights extending over much of the Amazon. In situ measurements from the CDP probe (http://data.ceda.ac.uk/badc/sambba/data) were used to determine flight-mean values of \( N_d \) (10th–90th percentile of data above 1 cm\(^{-3}\)). Due to reservations on the quality of data, we only use the maximum SAMBBA value of \( \sim 1,000 \text{ cm}^{-3} \) (flight b742) to inform our relationship; this value is within the upper range as observed by Polonik et al. (2020). This maximum \( N_d \) is combined with a maximum AOD of 2.0, which lies within sustained AOD measurements shown in Figure 2 and other observations (Schafer et al., 2008).

The data from the campaigns are plotted in Figure 3c along with a logarithmic function (following Stevens et al., 2017) that describes the combined data sets:

\[
N_d = 410 \ln(5\tau + 1),
\]  

where \( N_d \) is in cm\(^{-3}\) and \( \tau \) is the dimensionless AOD. The new relationship displays much more sensitivity of \( N_d \) to AOD than the MACv2-SP parameterization. Applying the new relationship to the horizontal AOD distribution from the MACv2-SP model (Figure 3b) provides a horizontal distribution of prescribed \( N_d \) (Figure 3d). We couple this to the cloud microphysics scheme via prescribing the liquid cloud droplet number at all cloudy grid points (note that the \( N_d \) distribution is constant in the vertical and not applied to the ice particle distribution). An enhancement of \( N_d \) in the microphysics scheme will reduce the average radius of the cloud droplet distribution, which may suppress autoconversion and warm-phase precipitation. This idealized treatment of aerosol and \( N_d \) sacrifices some realism in that the distribution of both would naturally evolve both spatially and temporally, but allows us to reduce the uncertainty that the complexity of aerosol microphysics processes and aerosol activation would add to understanding aerosol-cloud–radiation interactions (Gryepeedt et al., 2017; Rothenberg et al. 2018; Simpson et al., 2014). Additionally, we routinely see sustained periods (weeks to months) of enhanced BBA over the Amazon (see Figure 2) which would provide a continuous source of aerosol and CCN, so our simulations could be seen as a proxy for these prolonged periods of BBA.

2.3. Experiments

We perform five simulations using the same initial meteorological and lateral boundary conditions. A list of experiments is shown in Table 1. In CONTROL, we use background distributions of AOD and \( N_d \) to produce a control simulation to compare with the remaining experiments. Based on the lowest values measured during field campaigns in the region we use a background \( N_d \) of 100 cm\(^{-3}\) which equates to a background AOD (using Equation 1) of 0.06; the background \( N_d \) and AOD are included in all subsequent experiments.

In the MPHYS experiment, we only perturb the \( N_d \) distribution (see Figure 3d); this allows us to focus on the response of the domain to changes in \( N_d \) induced by the BBA and is analogous to the ACI effect. In the RAD experiment, only the 3D aerosol distribution in the radiation scheme is perturbed (Figures 3a and 3b) while \( N_d \) is kept at the background value; we use this to isolate the radiative effect of BBA which results in
perturbations to the temperature profile and is analogous to the ARI effect. Using this method, the semi-direct effect (Herbert et al., 2020; Johnson et al., 2004) is implicitly included in ARI effects. The RAD-scat experiment is similar to RAD but here the SSA has been increased to 1.0 and the AOD enhanced by a factor of 2.55 (see Section 2.2 for details); this allows us to isolate the impact of SW radiative effects on surface fluxes. For the TOTAL experiment, we perturb both the \( N \) and aerosol distribution to see the total effect of BBA on the domain.

3. Results

3.1. Characteristics of the Control Simulation and Comparison With Observations

We begin by describing the characteristics of the control simulation and compare with observations that show the model is reproducing a realistic diurnal cycle. Figure 4 shows the domain during a diurnal cycle, starting in the afternoon of day 3 and progressing throughout the evening and night. At 1400 LST the LWP exceeds 100 gm\(^{-2}\) in numerous small-scale features (Figure 4a) consistent with thermally triggered convective cells. The cells are advected westward and aggregate, producing larger mesoscale cells that continue to develop throughout the night (Figures 4b–4d). Extensive regions of stratiform cloud with low LWP are present over the coastal regions to the northeast of the domain and downwind of the convection in the northwest. Precipitation (Figures 4e–4h) coincides with regions of high LWP and convective activity, with extensive small-scale features present at 1400 LST that grow to larger, more isolated, cells during the evening and overnight. Maximum (99th-percentile) precipitation rates in the convective cells reach 22 mm h\(^{-1}\) at 1400 LST and drop to 16 mm h\(^{-1}\) at 0200 LST. The convection lofts condensate to the freezing level and produces extensive high-altitude ice clouds which can be seen predominantly in the northwest of the domain (Figures 4i–4p) with IWP exceeding 260 gm\(^{-2}\) in some regions. These clouds appear more extensive...
at 2000 LST (Figure 4n) and dissipate overnight, reflecting the time required for the convective cells to grow large enough to penetrate into the upper troposphere.

Figure 5 shows a comparison of precipitation rates across the domain from satellite-based estimates (IMERG data product 3B-HHR-E.MS.MRG.3IMERG V06 B; Huffman et al., 2019) and the simulation for the same time period (August, 31 to September 6, 2007). Domain-mean satellite-based estimates (Figure 5a) show a distinct diurnal cycle with a peak between 1400 and 1500 LST and a second, more variable, peak during the night at ~0500 LST. The afternoon peak is driven by deep convection with high precipitation rates (Figure 5b), and the overnight peak by mesoscale systems that develop within the domain or are advected into the domain. Our control simulation is able to reproduce the timing and the general characteristics of the diurnal cycle of precipitation. We see the largest peak between 1400 and 1500 LST (Figures 5a and 5c) and a second peak with more variability during the night. The simulation generally produces a lower precipitation rate than the IMERG data set, though this could be due to an over-estimation of moderate (0.4–10 mm h\(^{-1}\)) rainfall rates in IMERG that have been reported over the Amazon (Oliveira et al., 2016) and evident in Figure 5b. The simulated precipitation intensities (Figure 5c) highlight a period of model spin-up that produces pronounced low (<0.01 mm h\(^{-1}\)) precipitation rates for the first 30 h or so; this occurs as the model removes excess water from the domain in order to attain a quasi-steady state.

These results demonstrate that the ICON model and our configuration can reproduce the observed characteristics of the diurnal cycle of convection and precipitation over the Amazon basin, and provides a reliable control simulation for understanding BBA plume impacts on the domain.
Increasing $N_d$ in the MPHYS experiment (see Table 1) strongly suppresses precipitation in the initial stages (days 1 and 2) throughout the domain at all precipitation intensities (Figure 6b), accompanied by an increase in the domain mean LWP (Figure 7c) of $\sim 10$ gm$^{-2}$ ($\sim 20\%$). This is driven by a shift of the cloud droplet distribution to smaller sizes (Figure 9c) which suppress the warm rain process (Albrecht, 1989; Herbert et al., 2021).

**Figure 6.** Time-series of the precipitation intensity occurrence across the domain and response in each experiment. In the control simulation (a) the colors depict the percentage occurrence of the domain as a function of precipitation rate (note the log$_{10}$ scale used for the bins). For all other experiments (b–e) the colors depict the absolute change in percentage occurrence compared to the control (experiment – control), with blue depicting a suppression and red an enhancement. Panels correspond to responses from: (b) the MPHYS experiment; (c) the RAD-scat experiment; (d) the RAD experiment; and (e) the TOTAL experiment. In the control simulations (a) the black line shows the domain mean precipitation rate, and in the other experiments (b–e) the black solid line shows the domain mean cumulative precipitation response (experiment – control); corresponding values are shown on the right-hand side y-axis in black. Vertical dashed (solid) lines are at 1200 (0000) local solar time for each simulated day.

### 3.2. Response to Perturbations

#### 3.2.1. MPHYS Experiment

Increasing $N_d$ in the MPHYS experiment (see Table 1) strongly suppresses precipitation in the initial stages (days 1 and 2) throughout the domain at all precipitation intensities (Figure 6b), accompanied by an increase in the domain mean LWP (Figure 7c) of $\sim 10$ gm$^{-2}$ ($\sim 20\%$). This is driven by a shift of the cloud droplet distribution to smaller sizes (Figure 9c) which suppress the warm rain process (Albrecht, 1989; Herbert et al., 2021).
The enhanced liquid condensate and smaller droplets reduce surface downwelling radiation, and removes a source of water vapor and evaporative cooling below the cloud base. The consequence is enhanced convective inhibition and an overall weakening of moderate updrafts ($\omega_{\text{max}} \leq 5 \text{ m s}^{-1}$) in the initial response (Figure 8b). In the same time period, there is no significant change in the fraction of the domain covered by liquid cloud (Figure 9a), demonstrating the initial response of the domain to an $N_d$ perturbation primarily impacts the bulk cloud properties, rather than the thermodynamic properties that drive cloud formation. The LWP histogram (Figure 7c) shows a decrease at 100 gm$^{-2}$ and an increase at 1,000 gm$^{-2}$ on day 1, attributed to a deepening of clouds at the expense of shallow clouds, which supports this hypothesis. A second consequence of enhanced $N_d$ in the MPHYS experiment is an increase in the IWP in the evening (Figure 7d). This is possibly due to a combination of deeper clouds providing enhanced mass aloft and smaller droplets with lower fall speeds (Fan et al., 2013; Grabowski &
The domain response to enhanced $N_d$ settles into a consistent diurnal cycle after the first two days. Precipitation is suppressed in weakly precipitating clouds ($<0.01$ mm h$^{-1}$) throughout much of the simulation and up to rates of 1 mm h$^{-1}$ during the afternoon convection (Figure 6b). These precipitation rates are insufficient to strongly impact the domain-mean cumulative precipitation which remains at around −0.55 mm from day 3 onwards (Figure 6b). LWP sharply increases during the afternoon convection at 1400 LST (Figure 7c) though this is almost entirely due to increases from clouds with LWP exceeding 1,000 gm$^{-2}$.

Figure 8. Time-series of the vertical wind intensity occurrence across the domain and response in each experiment. In the control simulation (a) the colors depict the percentage occurrence of the domain grid points as a function of maximum vertical windspeed below 6 km ($w_{\text{max}}$) at each grid point. For all other experiments (b–e) the colors depict the absolute change in percentage occurrence compared to the control (experiment − control). Panels correspond to responses from: (b) the MPHYS experiment; (c) the RAD-scat experiment; (d) the RAD experiment; and (e) the TOTAL experiment. Vertical dashed (solid) lines are at 1200 (0000) local solar time for each simulated day.

Morrison, 2016; Heikenfeld et al., 2019), and potentially CCN-induced convective invigoration (discussed in more detail in the following paragraph).
Similarly, the IWP response (Figure 7d) has a positive peak in the diurnal cycle centered at 1800 LST of ~ +20 gm$^{-2}$ which represents an increase of >60%. Due to the slower fall speed of ice particles, the peak is slower to decrease and results in a consistent increase in the IWP throughout the simulation. These results highlight the dominant role that small-scale convective features (representing <0.5% of the domain) have on the response of the mean properties of the large-scale environment. Following the initial suppression of $w_{\text{max}}$, there is little consistent response in the diurnal cycle (Figure 8b) except at 1400 LST when there is some suppression of weak ($w_{\text{max}} \leq 3$ ms$^{-1}$) updrafts. It is worth noting that updrafts greater than 6 ms$^{-1}$ are generally more prevalent throughout the simulation. This could be attributed to CCN-induced convective invigoration via latent heat released from the additional freezing (Figure 7d); this process would tend to produce a stronger signal in existing cloudy columns with high updraft speeds (and therefore more condensate to freeze), which would support our results, or be a consequence of the overall cloud deepening due to warm-phase invigoration processes. A definite causal attribution would require a more in-depth analysis of the updraft vertical momentum equation that is not attempted here. However, we note that a recent cloud resolving modeling intercomparison study (Marinescu et al., 2021) under the umbrella of the aerosol, clouds, precipitation, and climate (ACPC) initiative revealed a dominant contribution of enhanced condensational latent heating of the warm phase of convective clouds, even when saturation adjustment schemes are used, attributable to increased environmental instability in high-CCN conditions.

Why do the first two days behave differently? As highlighted, increasing $N_d$ primarily perturbs the cloud properties and has little impact on the large-scale thermodynamic environment (Figure S1). Comparing the precipitation in the control and the MPHYS experiment (Figures 6a and 6b) shows that the marked suppression on days 1 and 2 occurs simultaneously with the high precipitation rates in the control simulation (largely due to spin-up of the model). The same can be seen with the positive LWP response (Figures 7a and 7c). It is only when the perturbation is this strong that we see a response in updraft speeds (Figure 8b). This suggests that the response of the domain to increasing $N_d$ is consistent throughout the simulation, but that the impact is nonlinear.

Figure 9. Time-series showing the change in percentage coverage of the domain with (a) liquid clouds and (b) ice clouds for the experiments compared to the control simulation (experiment – control). Panel (c) shows the domain-mean cloud droplet effective radius for the control simulation and MPHYS experiment. Vertical dashed (solid) lines are at 1200 (0000) local solar time for each simulated day.
The presence of a layer of purely scattering aerosol in the RAD-scat experiment (see Table 1) strongly reduces downwelling shortwave radiation reaching the surface (∼ −50 Wm$^{-2}$ domain mean at 1200 LST; not shown) and produces anomalous cooling in the BL of several degrees across the domain (see Figure S1b). Daytime surface fluxes of sensible and latent heat are reduced by ∼10% (latter shown in Figure S1d). The cooling strongly suppresses convection during the afternoon and considerably weakens updraft intensities across the domain (Figure 8c) and throughout the simulation. As a result, precipitation decreases during the afternoon (Figure 6c) resulting in a reduction in domain-mean cumulative precipitation by −1.2 mm at the end of the simulation at a rate of almost −0.2 mm day$^{-1}$. LWP (Figure 7e) similarly decreases in the afternoon and is attributed to clouds with LWP between 100 and 1,000 gm$^{-2}$; along with the pronounced suppression of $w_{\text{max}} < 3$ ms$^{-1}$ (Figure 8c) this demonstrates that the shallow convective cells with moderate...
loadings of condensate are predominantly affected by the anomalously cool surface. With weaker updrafts there is a reduction in the amount of condensate reaching the freezing level and an overall decrease in IWP throughout the simulation of $\sim 8 \text{ g m}^{-2}$; the IWP response does not have a pronounced diurnal cycle.

The domain-mean low-level ($P > 800 \text{ hPa}$) cloud cover (Figure 10e) is generally enhanced by 1 or 2%, although suppression of afternoon convection often counteracts this and results in a slight diurnal cycle in the response. This helps drive a slight increase in precipitation overnight (Figure 6c) which extends over all precipitation intensities. The cloudiness increases due to an anomalously cooler BL and enhanced moisture in the BL (see Figure 11g), the latter of which occurs due to reduced mixing with the dry free troposphere (FT). An increase in weak updrafts ($w_{\text{max}} < 3 \text{ m s}^{-1}$; Figure 8c) overnight is likely attributed to this enhanced low-level cloudiness and precipitation. Mid-level (800–400 hPa) cloud cover is considerably reduced throughout the simulation, especially at lower cloud fractions (Figure 10f). The daily suppression of convection accounts for the minimum at 1400 LST and is dominated by the decrease of cloud fraction $>0.95$; a simultaneous decrease in liquid cloud coverage by $\sim 5\%$ across the domain (Figure 9a) suggests these are entire convective cells that have failed to initiate.

### 3.2.3. RAD Experiment

In the RAD experiment, the aerosol layer not only reduces the downwelling shortwave at the surface (as with RAD-scat) but also produces a localized source of heating during the day throughout the BL (see Figure 3a for the aerosol profile). The primary role that this heating aloft has on the domain is to enhance cloudiness, precipitation, and convective activity overnight and increase the magnitude of the response during the day.

The suppression of afternoon convection from the anomalously cool BL is further enhanced by the heating aloft, which acts to stabilize the BL (Figure 8d); precipitation in the afternoon is also further decreased at all intensities (Figure 6d) especially at rates $\geq 1 \text{ mm h}^{-1}$, as is the LWP which reaches a domain mean response of $\sim 10 \text{ g m}^{-2}$ at 1400 LST on most days of the simulation. During the afternoon, the anomalous heat aloft is mixed throughout the BL and increases CAPE (see Figure S1f); convective activity from the organized mesoscale cells is enhanced overnight (Figure 8d), increasing LWP (Figure 7g) and precipitation (Figure 6d) until midday. The enhanced precipitation overnight counteracts the daytime suppression, preventing the gradual anomalous decrease in cumulative precipitation as observed in the RAD-scat experiment. The enhanced convective activity overnight increases IWP at high loadings (Figure 7h), which gradually spreads to lower values throughout the following 24 h (note the diagonal features), and produces an overall increase in IWP and high-altitude cloud coverage (Figure 9b) throughout the simulation.

In the RAD-scat experiment, anomalous BL cooling enhanced low-level clouds throughout the simulation. The addition of heating aloft prevents this via a partial offset in the BL cooling (Figure S1b) and evaporation of cloud, resulting in suppression of low-level cloud in the afternoon (Figure 10g). As the simulation progresses the low-level cloud cover displays a positive trend on top of the afternoon suppression. This is driven by a steady increase in the water vapor content ($Q_{\text{vapor}}$) in the BL (see Figure S1h) that reaches a domain mean value of $+0.75 \text{ g kg}^{-1}$ ($\sim 7\%$); this is partially driven by suppressed mixing with the FT but also anomalous convergence centered beneath the BBA plume (Figure 11d). The gradual increase in $Q_{\text{vapor}}$ also enhances CAPE (see Figure S1f) but primarily in the region of anomalous convergence beneath the plume. Mid-cloud cover response (Figure 10h) is primarily driven by the afternoon suppression of convection but occasional small positive anomalies overnight demonstrate an impact from the enhanced activity overnight, similar to the LWP response.

Overnight on day 6/7 there is a considerable anomalous increase in precipitation (Figure 6d) and IWP (Figure 7h), with an associated increase in updraft speeds (Figure 8d). This corresponds to the intensification of a mesoscale convective cell in the west of domain, downwind of the plume center. The cell, present in the control simulation, becomes larger and deeper and produces widespread intense precipitation and produces extensive convective outflow that persists for $\sim 36$ h (Figure 7h). Figure S2 in the supporting information provides a snapshot of the feature at 0500 LST on day 7.

### 3.2.4. TOTAL Experiment

In this experiment, the full effect of the BBA is examined. We find that the two effects (microphysical and radiative) largely act independently, with the resulting total response approximating a combined effect.
Having said that, the radiative effect is dominant and controls much of the mean response over the domain and in some instances the microphysical effects buffer the radiative effects. The response of updraft intensities (Figure 8e) and large-scale circulation (Figure 11e) is almost entirely driven by the radiative effects except for the initial hours before sunrise on day 1. With minimal impact on cloud cover seen in the MPHYS experiment (Figures 10c and 10d) it follows that the total response of low and mid-cloud cover is controlled by the radiative effect (Figures 10i and 10j), which is illustrated by overlapping time-series for the response in cloud-free fraction in the RAD and TOTAL experiments (Figure 9a).

Cloud formation and development is impacted by radiative effects, and cloud properties are directly impacted by changes to N_d. This is reflected in the LWP, IWP, and precipitation responses (Figures 6e, 7i and 7j).
that demonstrate a combined effect from both microphysics and radiation. The diurnal cycle from the RAD precipitation response is still apparent (Figure 6e), but has become offset by an overall suppression by the enhanced $N_a$, resulting in a more pronounced decrease in domain-mean cumulative precipitation which decreases by $-0.6$ mm on the first day and then remains roughly at this value for much of the simulation. The precipitation histogram highlights that radiative effects primarily impact the higher rates ($>1$ mm h$^{-1}$) from convective cells, while microphysical effects are more apparent at lower rates ($<1$ mm h$^{-1}$) associated with shallower convection and stratiform cloud. However, the two effects are not entirely independent: a suppression of the highest precipitation rates overnight (comparing RAD with TOTAL experiment) are caused by the enhanced $N_a$ which acts to dampen the response. This is most evident overnight on day 6/7 as the intensification of the mesoscale convective cell is weakened.

The response of the LWP is a combination of both radiative and microphysics effects. During afternoon convection at 1400 LST, the two effects have opposite responses of equal magnitudes, yet the total effect is dominated by the radiative effect. This occurs because the radiative effect causes the primary response, which is a suppression of convection resulting in fewer and less intense convective cells. The enhanced $N_a$ acts on the bulk cloud properties and can be seen as a secondary response that weakly buffers the primary response, hence why the response in the TOTAL experiment is more aligned to the RAD experiment. The IWP response (Figures 7 and 9b) is closely aligned with an additive effect from both radiative and microphysical effects, resulting in a large increase in domain-mean IWP and ice-cloud coverage throughout much of the simulation that often doubles the IWP in the control simulation. The linearity of the LWP and IWP responses from RAD and MPHYS are shown in Figure S3.

### 3.3. TOA Radiative Effects

Figure 12 shows the domain-mean TOA effect on radiation due to the presence of the BBA. Values from the MPHYS, RAD, and TOTAL experiments are shown for shortwave (SW), longwave (LW), and net (NET; SW + LW) components, with negative values representing a cooling effect on the climate.

In the MPHYS experiment, the SW$_{TOA}$ response is consistently negative, reaching a maximum magnitude of $\sim -15$ Wm$^{-2}$ at 1200 LST on most days. There is little change in the low and mid-cloud cover (Figures 10c and 10d) and the sharp increase in domain-mean LWP (Figure 7c) is attributed to clouds with high loadings of water that already have a high albedo, demonstrating the response is primarily driven by a large-scale reduction in cloud droplet effective radius (Figure 9c) that increases the albedo of clouds with moderate loadings of water. In the RAD experiment, the SW$_{TOA}$ response shows considerable temporal variation. Initially, the response is positive with a magnitude of $\sim +15$ Wm$^{-2}$ but a diurnal response develops and evolves into a consistent negative response. These responses are a combination of the positive direct effect from the BBA absorption and the temporally varying low-cloud response (Figure 10g). In both MPHYS and RAD experiments, the LW$_{TOA}$ response is positive throughout the simulation due to enhanced IWP. The sharp diurnal cycle in IWP response for the MPHYS experiment (Figure 7d) is not duplicated in the LW$_{TOA}$ response (Figure 12a); as with the SW$_{TOA}$ this suggests that the LW$_{TOA}$ response is most sensitive to the clouds that have relatively moderate loadings of ice. This is in contrast to the RAD experiment which shows the LW$_{TOA}$ response is closely aligned to the IWP response (Figure 7h); this occurs because the cloud field itself is being perturbed, with more convective activity overnight that results in more high-altitude cloud coverage (Figure 9b).

The NET$_{TOA}$ response for each experiment, also referred to as effective radiative forcing (ERF), is shown in Figure 12d and daily mean values presented in Table 2. The ERF (determined from the TOTAL experiment) has contributions from ERF$_{ACT}$ (determined from the MPHYS experiment) and ERF$_{ARI}$ (determined from the RAD experiment). The initial response is characterized by strong opposing SW effects from ERF$_{ARI}$ and ERF$_{ACT}$ (Figure 12c), resulting in an overall weak ($\leq 1$ Wm$^{-2}$) and positive daily mean ERF. As the simulation progresses the daily mean ERF becomes increasingly positive due to an increasing LW component, driven by enhanced ice, which results in a positive ERF$_{ARI}$ and ERF$_{ACT}$. A positive ERF$_{ACT}$ due to the ice phase response was also reported by Dagan et al. (2020). By day 5 the ERF is close to 10 Wm$^{-2}$, driven predominantly by a strong LW component of 13 Wm$^{-2}$. 

These results demonstrate two important findings. First, the response of IWP and high-altitude cloud cover to BBA plays a significant role in determining its overall ERF: if the IWP response is weaker, as reported by Liu et al. (2020) and discussed later, then the ERF is instead controlled by perturbations to the liquid cloud field and has an opposite sign, i.e., a cooling rather than warming effect on climate. Second, the ERF from BBA is highly time-dependent: all TOA radiative effects (LW, SW, and NET) are initially weak but then tend to grow in magnitude throughout the simulation, and are therefore highly sensitive to the longevity and horizontal coverage of the BBA plume.

Figure 12. Time-series showing the domain-mean TOA effective radiative forcing (ERF) in the MPHYS (a; ERF\textsubscript{ACI}), RAD (b; ERF\textsubscript{ARI}), and TOTAL (c; ERF) experiments. Blue dotted lines show the SW component, red dashed lines show the LW component, and solid back lines show the NET (SW + LW) response. The NET component from each experiment in (a–c) are individually re-plotted in (d) and show the ERF\textsubscript{ACI} (blue line), ERF\textsubscript{ARI} (red line), ERF (black line), and ERF\textsubscript{ACI} + ERF\textsubscript{ARI} (cyan dashed line). Note the change in scale on the y-axis for plot (d). Negative values represent a cooling effect and vice versa. Vertical dashed (solid) lines are at 1200 (0000) local solar time for each simulated day.

These results demonstrate two important findings. First, the response of IWP and high-altitude cloud cover to BBA plays a significant role in determining its overall ERF: if the IWP response is weaker, as reported by Liu et al. (2020) and discussed later, then the ERF is instead controlled by perturbations to the liquid cloud field and has an opposite sign, i.e., a cooling rather than warming effect on climate. Second, the ERF from BBA is highly time-dependent: all TOA radiative effects (LW, SW, and NET) are initially weak but then tend to grow in magnitude throughout the simulation, and are therefore highly sensitive to the longevity and horizontal coverage of the BBA plume.
Table 2

| ERF   | Day 1 | Day 2 | Day 3 | Day 4 | Day 5 | Day 6 | Day 7 | Day 8 |
|-------|-------|-------|-------|-------|-------|-------|-------|-------|
| Longwave | −1.0  | 1.8   | 7.8   | 11.8  | 13.2  | 16.7  | 26.6  | 14.7  |
| Shortwave | 0.1   | 1.0   | −4.5  | −4.5  | −3.4  | −6.4  | −15.9 | −6.5  |
| Net    | −1.0  | 2.8   | 3.3   | 7.3   | 9.8   | 10.3  | 10.6  | 8.2   |

Note. All values are for the domain-mean ERF (experiment – control) in Wm$^{-2}$.

3.4. AOD-Dependence

The results so far have demonstrated that the response of the domain to the BBA is highly complex and dependent on the longevity of the plume, which makes the representation of its effects in coarser models difficult. Here we identify a relationship between the BBA perturbation (ΔAOD) and the suppression of afternoon convection, a key driver of BBA impacts in our simulations.

In Figure 13, we use the prescribed horizontal distribution of AOD provided by the plume model (Figure 3b) to subset the column-mean updraft speed between the surface and 700 hPa ($w_{up}$) and its response in the RAD and RAD-scat experiments. Data is downscaled to a resolution of 0.5° and mean values for each subset of pixels, stratified by ΔAOD, determined at each output timestep. Figure 13a shows the distribution of $w_{up}$ as a function of ΔAOD in the control simulation. The daily afternoon convection is visible at 1400 LST at all ΔAOD values, demonstrating that the highest updraft speeds are dispersed around the domain, rather than co-located with the simulated BBA plume. The MPHYS experiment response (Figure 13b) shows no evidence of a consistent impact to $w_{up}$ nor a clear dependence on ΔAOD, which supports our earlier hypothesis that there is no widespread CCN-induced cloud invigoration occurring in these simulations. The response of $w_{up}$ in the RAD-scat experiment (Figure 13c) clearly demonstrates an overall suppression throughout the simulation up to a value of −60% during the periods of afternoon convection. There is also evidence of a relationship between ΔAOD and $w_{up}$, with greater AOD perturbations occurring alongside a greater suppression of $w_{up}$. This is most evident at 1400 LST when the strongest updrafts occur (Figure 13a). Including the in situ heating in the RAD experiment (Figure 13d) enhances convective activity overnight but does not display a consistent dependence on ΔAOD. Importantly, the addition of heating does not impact the suppression of convection at 1400 LST.

The convection that occurs at 1400 LST produces the greatest values of LWP, IWP, and precipitation during the diurnal cycle, therefore we focus on this time period in Figure 13c. This data is taken directly from Figure 13d (for the RAD experiment) and plotted individually for each day of the simulation. The data provide more evidence of a relationship between the plume ΔAOD and the suppression of convection. At values of ΔAOD = 0.2 there is a suppression of mean updraft speeds by roughly −10%, and at ΔAOD = 0.8 the suppression has increased to roughly −40%. There is some day-to-day variability but all data tends toward a similar relationship. The resulting relationship is likely driven by the formation of fewer convective cells and a weakening of existing cells, a hypothesis that requires further work to test.

This result provides a quantitative means to relate the large-scale suppression of convection to AOD from BBA smoke, and additionally shows that this relationship is caused by the anomalous surface cooling that occurs due to the extinction of shortwave radiation by the aerosol and is therefore applicable to other aerosol species such as dust or smoke with higher SSA values. Figure 13d also demonstrates that the intensification of convection due to the heating aloft is not clearly dependent on ΔAOD, though there are signs of intensification during the night at higher values of ΔAOD.

4. Discussion and Conclusions

In this study, we used the ICON model in a regional configuration to study the impact of BBA smoke plumes on clouds and precipitation processes over the Amazon rainforest, with a domain covering 3,000 × 2,000 km at a resolution of 1,500 m that allows us to resolve the convective processes that drive the diurnal cycle of clouds and precipitation over the region. The MACv2-SP model (Stevens et al., 2017) was used to simulate a persistent BBA plume with a horizontal and vertical distribution consistent with observations, and a realistic N$_o$ distribution was produced using an observationally informed relationship between AOD and N$_o$. A series of experiments were designed to isolate the different pathways through which the BBA plume influences the atmosphere in the domain: 1) enhanced CCN and N$_o$ (isolating ACI); 2) reduction in surface downwelling radiation (isolating surface cooling); 3) surface cooling and in situ heating of BBA (isolating...
Our main findings are as follows:

- **BBA** induces pronounced changes to the diurnal cycle of cloud and precipitation over the Amazon rainforest, with contrasting responses before and after midday.
- The two routes by which BBA influences clouds have distinct mechanisms: ACI primarily perturbs the bulk cloud properties with evidence of cloud deepening, whereas ARI influences the formation and development of convective cells and the convergence of moisture. ARI predominantly impacts high
precipitation rates (>1 mm h\(^{-1}\)), while ACI predominantly impacts low-to-moderate precipitation rates (<1 mm h\(^{-1}\)).

- The diurnal cycle of convection is suppressed in the afternoon due to a cooler surface and in situ heating of the BBA layer, and enhanced overnight due to heating aloft; these responses are entirely driven by ARI.
- Low-level moisture increases with plume longevity and gradually enhances low-level cloud cover. Suppressed convection during the daytime weakens mixing between the BL and FT, while heating aloft induces anomalous convergence beneath the plume.
- Both ACI and ARI result in enhanced IWP. ARI-induced intensification of cells overnight transports more condensate to freezing level, while ACI-induced reduction in cloud droplet size results in more IWP from existing deep convective clouds. There was no evidence of significant CCN-induced cloud-in-vigoration (enhanced updraft speed) from ACI.
- TOA radiative effects are complex and show considerable sensitivity to the BBA plume longevity. In the shortwave, much of the variability is driven by ARI-induced changes to the low-level cloud, while enhanced IWP throughout simulation from both ARI and ACI drives a positive longwave effect that grows in magnitude. The daily mean net TOA radiative effect is dominated by the longwave effect and although starting slightly negative (−1 Wm\(^{-2}\) on day 1) becomes increasingly positive (+3.5 Wm\(^{-2}\) on day 3, +10 Wm\(^{-2}\) by day 6).

Our results largely support the findings in previous modeling studies though there are some differences that highlight the importance of resolving convective processes and the response of the IWP. The primary impact of BBA on the domain is a suppression of convection beneath plume, which is consistent with previous studies (Liu et al., 2020; Martins et al., 2009; Wu et al., 2011). Other modeling studies have reported an overall decrease in cumulative precipitation due to smoke presence (Liu et al., 2020; Martins et al., 2009; Thornhill et al., 2018; Wu et al., 2011) but the possibility of increased peak intensities (Martins et al., 2009). These results agree with our study: the suppression of convection (ARI) and the warm-rain process (ACI) both contribute to suppressed precipitation during the afternoon; however, we observed an increase in precipitation overnight that partially counteracted the daytime suppression; we also observed the intensification of a mesoscale convective cell toward the end of the simulation. Birch et al. (2014) have shown that the diurnal cycle of convection and associated precipitation over West Africa is not well represented by coarse-resolution models that parameterize convection; therefore, the diurnal response of convection that we observe would not be reproduced by climate models. An analysis of precipitation over the Amazon region in CMIP5 models shows a dry model bias throughout the dry season (Yin et al., 2013); it is possible that this is driven by a poorly represented diurnal cycle of convection. A replication of our simulations using a coarser resolution and parameterized convection would allow us to understand whether key responses to smoke are reproduced in this region, and bridge the gap between regional- and global-scale models.

Our simulations provide evidence that large-scale impacts on circulation and moisture transport are driven by perturbations to the radiation, rather than \( N_p \). Surface cooling promotes anomalous divergence, whereas heating aloft promotes anomalous convergence, with the latter dominating. Lee et al. (2014) also reported the two processes often result in opposing impacts on the circulation. Hodzic & Duvel (2018) studied the impact of BBA over the forested island of Borneo and found that moderately absorbing BBA induced anomalous divergence, whereas strongly absorbing BBA induced anomalous convergence; as in our study, this indicates a competition between the surface cooling effect and heating aloft driven by radiation. In contrast to our results, Zhang et al. (2008) found anomalous moisture divergence below the smoke region due to surface cooling, despite the heating aloft. This may be attributed to a coarser resolution (60 km) and differences in the representation of convective processes: along with enhanced stability during the daytime we also observed enhanced CAPE aloft and intensification of turbulence overnight which was not reported by Zhang et al. (2008). Enhanced CAPE aloft and invigorated updrafts were reported by Wu et al. (2011); our results support this hypothesis and show that this is an important response in the diurnal response of precipitation.

A recent study by Liu et al. (2020) provides a good comparison to our study. Using 7-day WRF-Chem simulations at convection-permitting resolution (3 km) the authors isolated the impacts of smoke on the Ama-
zon to ARI and ACI. Focusing on the final day of the simulation, they found that ARI suppressed low-level clouds due to cloud burn-off and enhanced updrafts aloft, thus increasing IWP; we observed a similar response but additionally saw an overall moistening of the BL which offset the low-level cloud suppression. The authors found that ACI enhanced low-level clouds and suppressed updrafts and IWP; in our study, the IWP was enhanced despite a weak suppression of updrafts. We observe pronounced increases (often exceeding 20 gm$^{-2}$) in IWP due to both ARI and ACI (Figure 7), whereas Liu et al. (2020) observed perturbations below 0.5 gm$^{-2}$. This discrepancy in IWP response between models could be attributed to previously reported underestimations in WRF (Baró et al., 2018) which are illustrated by the poor agreement between model and observations in Liu et al. (2020). Given that the IWP response in our simulations drives a strong positive LW$_{TOA}$ radiative effect this highlights an important sensitivity of the simulated response of cloud and precipitation over Amazon rainforest to smoke. The discrepancy between studies may also be due to inherent differences between models and microphysics schemes; a recent model intercomparison study under the umbrella of the ACPC initiative (Marinescu et al., 2021) demonstrates that there is uncertainty in ACI effects on deep convective clouds due to the different models and/or microphysics scheme employed, especially at higher altitudes (>8 km). It would be beneficial to repeat this study with other models to test our conclusions.

Our simulations are supported by observations from in situ and space-borne observations, and additionally, highlight the importance of observing the full diurnal cycle for quantifying the overall impact of smoke on the Amazon. In situ observations point toward smoke generally suppressing precipitation during the dry season (Andreae et al., 2004; Bevan et al., 2008; Camponogara et al., 2014; Gonçalves et al., 2015) alongside convective invigoration within unstable atmospheres (Gonçalves et al., 2015) and existing updrafts (Andreae et al., 2004). We observed an overall suppression of precipitation, though we occasionally observed an intensification of convection in the evening and overnight. Space-borne observations show a relationship between high AOD and the suppression of convection (Jiang et al., 2018; Kaufman & Nakajima, 1993) alongside a suppression of low-cloud cover (Koren et al., 2004) and cloud optical depth (Ten Hoeve et al., 2012). These observations were made with instruments within the A-train satellite constellation which has a local overpass time of ~1330 LST. During this time frame, the results from our simulations are consistent with these observations. However, our simulations suggest there are strong diurnal responses in LWP, IWP, cloud cover, convection, and precipitation, that cannot be observed by polar orbiting satellites such as those used in the aforementioned studies. The use of geostationary satellites would be ideal for assessing the conclusions of our study.

Yu et al. (2007) used satellite observations (local overpass of ~1330 LST) during two years with contrasting meteorological conditions. The authors found that large-scale cloud fraction decreased with AOD in the drier and more-stable year, but increased for the moister, more unstable year; our simulations may help understand how such a dichotomy could be achieved. In a “stable” year, the suppression of convection is less effective, which would result in cloud-burn off being a prominent impact from ARI, and resulting in decreased cloud cover. In a more “unstable” year the suppression of convection is more active and sets up anomalous convergence below the smoke plume and reduces mixing of drier FT air into the BL. In our simulations, this gradually enhanced moisture content in the BL and enhanced low-cloud cover. Although difficult to evaluate this does offer a possible explanation for the contrasting response.

A novel aspect of our study was isolating the surface cooling effect and the heating aloft. Although primarily used to understand the different processes, it also allows us to estimate the impact of the increasing single-scattering albedo that accompanies the recently observed shift of burning from forests to grassland (Ten Hoeve et al., 2012). Higher values of SSA result in the same extinction but reduce the absorption of solar radiation, and relatively enhance the role of surface cooling. Our simulations suggest that this would weaken the intensification of convection overnight, drive a stronger suppression in precipitation, and reduce cloud cover.

Observations demonstrate that there is considerable variation in the duration of time that smoke plumes are present over a single location (see Figure 2). By presenting the full 8-day time-series of the domain response to a persistent layer of smoke we can consider the role of smoke plume lifetime, i.e., do we expect the cloud and precipitation to respond differently to a short-lived smoke plume compared to a long-lived plume. In
our simulations, the response of low-cloud cover was driven by a gradual moistening of the BL, resulting in contrasting SW$_{\text{TOA}}$ radiative effects at the beginning and end of the simulation. The gradual increase in IWP throughout the simulation similarly drove contrasting LW$_{\text{TOA}}$ radiative effects. Table 2 highlights the consequences of this: the daily mean NET$_{\text{TOA}}$ radiative effect is $-1\, \text{Wm}^{-2}$ on day 1 and $+10\, \text{Wm}^{-2}$ by day 6, which suggests that short-lived smoke plumes have a weaker impact on the climate than long-lived smoke plumes. This conclusion also holds true if we neglect the IWP response and instead focus on the SW$_{\text{TOA}}$ radiative response that is driven by the low-cloud response.

The idealized treatment of aerosols and $N_a$ in our simulations allows us to reduce uncertainties from aerosol microphysical processes and aerosol activation. However, in reality sources and sinks of aerosols will evolve spatially and temporally, and the activation of aerosols to cloud droplets will be dependent on not only the aerosol population but also on the thermodynamic environment and updraft speed (e.g., Ovchinnikov et al., 2013; Reutter et al., 2009). Precipitation washes out the airborne aerosol, therefore we may expect considerable heterogeneity in the spatial distribution of aerosol. Variations in meteorological conditions, such as air mass source, will impact the aerosol population and potentially ACI (e.g., Grysspeerdt et al., 2017). The activation of aerosol is strongly dependent on the degree of supersaturation which increases with updraft speed; our simulations suggest a suppression of convection and therefore a suppression in cloud droplet number (assuming updraft-limited regime; Reutter et al., 2009). This may then impact the role of ACI and ARI on the evolution of the cloud. The vertical structure of the aerosol, fixed in our simulations, is also reported to play an important role in ARI (Herbert et al., 2020) and ACI (Lebo, 2014). Using a prognostic aerosol scheme coupled to the microphysics scheme would provide additional realism to test our conclusions; however, include the uncertainty of the aerosol simulation itself; we are currently working on implementing the HAM aerosol module (Tegen et al., 2019) into this cloud-resolving model to conduct further analyses along those lines.

Data Availability Statement

IMERG data was obtained via the NASA Goddard Earth Sciences Data and Information Services Center (https://gpm1.gesdisc.eosdis.nasa.gov/data/GPM_L3/GPM_3IMERGHH.06). ERA-Interim and ERA-5 data sets provided by ECMWF were obtained via the MARS client and are available online (https://apps.ecmwf.int/datasets/). AERONET data is available online (https://aeronet.gsfc.nasa.gov). Data and scripts for generating figures from this manuscript can be found at http://doi.org/10.5281/zenodo.4694550. Please contact the authors for access to the full data set.

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References

Albrecht, B. A. (1989). Aerosols, cloud microphysics, and fractional cloudiness. Science, 245(4923), 1227–1230. https://doi.org/10.1126/science.245.4923.1227

Andreae, M. O., Rosenfeld, D., Artaxo, P., Costa, A. A., Frank, G. P., Longo, K. M., & Silva-Dias, M. A. F. (2004). Smoking rain clouds over the Amazon. Science, 308(5709), 1337–1342. https://doi.org/10.1126/science.1092779

Aragão, L. E. O. C., Malhi, Y., Roman-Cuesta, R. M., Saatchi, S., Anderson, L. O., & Shimabukuro, Y. E. (2007). Spatial patterns and fire response of recent Amazonian droughts. Geophysical Research Letters, 34(7). https://doi.org/10.1029/2006GL028946

Baró, R., Jiménez-Guerrero, P., Stengel, M., Brunner, D., Curci, G., Forkel, R., et al. (2018). Evaluating cloud properties in an ensemble of regional online coupled models against satellite observations. Atmospheric Chemistry and Physics, 18(20), 15183–15199. https://doi.org/10.5194/acp-18-15183-2018

Bevan, S. L., North, P. R. J., Grey, W. M. F., Los, S. O., & Plummer, S. E. (2008). The impact of atmospheric aerosol from biomass burning on Amazon dry-season drought. In: European Space Agency, (Special Publication) ESA SP (Vol. 114). John Wiley & Sons, Ltd. https://doi.org/10.1029/2008jd011112

Birch, C. E., Parker, D. J., Marsham, J. H., Copsey, D., & García-Carreras, L. (2014). A seamless assessment of the role of convection with updraft speed; our simulations suggest a suppression of convection and therefore a suppression in cloud droplet number (assuming updraft-limited regime; Reutter et al., 2009). This may then impact the role of ACI and ARI on the evolution of the cloud. The vertical structure of the aerosol, fixed in our simulations, is also reported to play an important role in ARI (Herbert et al., 2020) and ACI (Lebo, 2014). Using a prognostic aerosol scheme coupled to the microphysics scheme would provide additional realism to test our conclusions; however, include the uncertainty of the aerosol simulation itself; we are currently working on implementing the HAM aerosol module (Tegen et al., 2019) into this cloud-resolving model to conduct further analyses along those lines.

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References

Albrecht, B. A. (1989). Aerosols, cloud microphysics, and fractional cloudiness. Science, 245(4923), 1227–1230. https://doi.org/10.1126/science.245.4923.1227

Andreae, M. O., Rosenfeld, D., Artaxo, P., Costa, A. A., Frank, G. P., Longo, K. M., & Silva-Dias, M. A. F. (2004). Smoking rain clouds over the Amazon. Science, 308(5662), 1337–1342. https://doi.org/10.1126/science.1092779

Aragão, L. E. O. C., Malhi, Y., Roman-Cuesta, R. M., Saatchi, S., Anderson, L. O., & Shimabukuro, Y. E. (2007). Spatial patterns and fire response of recent Amazonian droughts. Geophysical Research Letters, 34(7). https://doi.org/10.1029/2006GL028946

Baró, R., Jiménez-Guerrero, P., Stengel, M., Brunner, D., Curci, G., Forkel, R., et al. (2018). Evaluating cloud properties in an ensemble of regional online coupled models against satellite observations. Atmospheric Chemistry and Physics, 18(20), 15183–15199. https://doi.org/10.5194/acp-18-15183-2018

Bevan, S. L., North, P. R. J., Grey, W. M. F., Los, S. O., & Plummer, S. E. (2008). The impact of atmospheric aerosol from biomass burning on Amazon dry-season drought. In: European Space Agency, (Special Publication) ESA SP (Vol. 114). John Wiley & Sons, Ltd. https://doi.org/10.1029/2008jd011112

Birch, C. E., Parker, D. J., Marsham, J. H., Copsey, D., & García-Carreras, L. (2014). A seamless assessment of the role of convection with updraft speed; our simulations suggest a suppression of convection and therefore a suppression in cloud droplet number (assuming updraft-limited regime; Reutter et al., 2009). This may then impact the role of ACI and ARI on the evolution of the cloud. The vertical structure of the aerosol, fixed in our simulations, is also reported to play an important role in ARI (Herbert et al., 2020) and ACI (Lebo, 2014). Using a prognostic aerosol scheme coupled to the microphysics scheme would provide additional realism to test our conclusions; however, include the uncertainty of the aerosol simulation itself; we are currently working on implementing the HAM aerosol module (Tegen et al., 2019) into this cloud-resolving model to conduct further analyses along those lines.

Data Availability Statement

IMERG data was obtained via the NASA Goddard Earth Sciences Data and Information Services Center (https://gpm1.gesdisc.eosdis.nasa.gov/data/GPM_L3/GPM_3IMERGHH.06). ERA-Interim and ERA-5 data sets provided by ECMWF were obtained via the MARS client and are available online (https://apps.ecmwf.int/datasets/). AERONET data is available online (https://aeronet.gsfc.nasa.gov). Data and scripts for generating figures from this manuscript can be found at http://doi.org/10.5281/zenodo.4694550. Please contact the authors for access to the full data set.
Lee, S. S., Feingold, G., Comissey, A., Yamaguchi, T., Koren, I., Vanderlei Martins, J., & Yu, H. (2014). Effect of gradients in biomass burning aerosol on shallow cumulus convective circulations. Journal of Geophysical Research: Atmospheres, 119(16), 9948–9964. https://doi.org/10.1002/2014JD021819

Leuenberger, D., Koller, M., Fuhrer, O., & Schär, C. (2010). A generalization of the SLEVE vertical coordinate. Monthly Weather Review, 138(9), 3683–3698. https://doi.org/10.1175/2010MWR3307.1

Liu, L., Cheng, Y., Wang, S., Wei, C., Pöhlker, M., Pöhlker, C., et al. (2020). Impact of biomass burning aerosols on radiation, clouds, and precipitation over the Amazon during the dry season: Dependence of aerosol-cloud and aerosol-radiation interactions on aerosol loading. Atmospheric Chemistry and Physics, 20, 1–50. https://doi.org/10.5194/acp-2020-191

Lovejoy, T. E., & Nobre, C. (2019). Amazon tipping point: Last chance for action. Science Advances, 5(12), eaab2949. https://doi.org/10.1126/sciadv.aab2949

Malavalle, F. F., Haywood, J. M., Mercado, L. M., Folberth, G. A., Bellouin, N., Sitch, S., & Artaxo, P. (2019). Studying the impact of biomass burning aerosol radiative and climate effects on the Amazon rainfall productivity with an Earth system model. Atmospheric Chemistry and Physics, 19(2), 1301–1326. https://doi.org/10.5194/acp-19-1301-2019

Marinescu, P. I., van den Heever, S. C., Heikenfeld, M., Barrett, A. I., Barthlott, C., Hoose, C., et al. (2021). Impacts of varying concentrations of cloud condensation nuclei on deep convective cloud updrafts—a multimodel assessment. Journal of the Atmospheric Sciences, 78(4), 1147–1172. https://doi.org/10.1175/JAS-D-20-0201.1

Martin, S. T., Artaxo, P., Machado, L. A. T., Manzi, A. O., Souza, R. A. F., Schumacher, C., et al. (2016). Introduction: Observations and modeling of the green ocean amazon (GoAmazon2014/5). Atmospheric Chemistry and Physics, 16(8), 4785–4797. https://doi.org/10.5194/acp-16-4785-2016

Martins, J. A., Silva Dias, M. A. F., & Gonçalves, F. L. T. (2009). Impact of biomass burning aerosols on precipitation in the Amazon: A modeling case study. Journal of Geophysical Research, 114(D2). https://doi.org/10.1029/2009JD009587

Mlawer, E. J., Taubman, S. J., Brown, P. D., Iacono, M. J., & Clough, S. A. (1997). Radiative transfer for inhomogeneous atmospheres: RRTM, a validated correlated-k model for the longwave. Journal of Geophysical Research, 102(14), 16663–16682. https://doi.org/10.1029/97JD00237

Morton, D. C., Defries, R. S., Randerson, J. T., Giglio, L., Schroeder, W., & van der Werf, G. R. (2008). Agricultural intensification increases deforestation fire activity in Amazonia. Global Change Biology, 14(10), 2262–2275. https://doi.org/10.1111/j.1365-2486.2008.01652.x

Nobre, C. A., Sampaio, G., Borma, L. S., Castilla-Rubio, J. C., Silva, J. S., & Cardoso, M. (2016). Land-use and climate change risks in the Amazon region. Remote Sensing, 8(7), 544. https://doi.org/10.3390/rs8070544

Ovchinnikov, M., Easter, R. C., & Gustafson, W. I. (2013). Untangling dynamical and microphysical controls for the structure of stratocumulus. Geophysical Research Letters, 40(16), 4432–4436. https://doi.org/10.1002/2013GL058010

Polonik, P., Knot, B., Ewald, F., Kölling, T., Mayer, B., et al. (2020). The challenge of simulating the sensitivity of the Amazonian microstructure to cloud condensation nuclei number concentrations. Atmospheric Chemistry and Physics, 20(3), 1591–1605. https://doi.org/10.5194/acp-20-1591-2020

Reddington, C. L., Morgan, W. T., Darbyshire, E., Brito, J., Cooke, P. J., Artaxo, P., et al. (2019). Biomes burning aerosol over the Amazon: Analysis of aircraft, surface and satellite observations using a global aerosol model. Atmospheric Chemistry and Physics, 19(14), 9125–9152. https://doi.org/10.5194/acp-19-9125-2019

Reutter, P., Su, H., Trentmann, J., Simmel, M., Rose, D., Gunthe, S. S., et al. (2009). Aerosol- and updraft-limited regimes of cloud droplet formation: influence of particle number, size and hygroscopicity on the activation of cloud condensation nuclei (CCN). Atmospheric Chemistry and Physics, 9(18), 7067–7080. https://doi.org/10.5194/acp-9-7067-2009

Rosenfeld, D., Lohmann, U., Raga, G. B., O’Dowd, C. D., Kulmala, M., Fuzzi, S., et al. (2006). Flood or drought: How do aerosols affect precipitation? Science, 312, 1309–1313. https://doi.org/10.1126/science.1126006

Rothman, D., Avramov, A., & Wang, C. (2018). On the representation of aerosol activation and its influence on model-derived estimates of the aerosol indirect effect. Atmospheric Chemistry and Physics, 18(11), 7961–7983. https://doi.org/10.5194/acp-18-7961-2018

Schär, J. S., Eck, T. F., Holben, B. N., Artaxo, P., & Duarte, A. F. (2008). Characterization of the optical properties of atmospheric aerosols in Amazonia from long-term AERONET monitoring (1993–1995 and 1999–2006). Journal of Geophysical Research, 113(D4). https://doi.org/10.1029/2007JD009319

Schrodin, E., & Heise, E. (2002). A New Multi-Layer Soil Model. COSMO Newsletter, pp. 149–151.

Seifert, A., & Beheng, K. D. (2006). A two-moment cloud microphysics parameterization for mixed-phase clouds. Part 2: Maritime vs. continental deep convective storms. Meteorology and Atmospheric Physics, 92(1–2), 67–82. https://doi.org/10.1007/s00703-005-0011-3

Simpson, E., Conolly, P., & McFiggans, G. (2014). An investigation into the performance of four cloud droplet activation parameterisations. Geoscientific Model Development, 7(4), 1535–1542. https://doi.org/10.5194/gmd-7-1535-2014

Stevens, B., Fiedler, S., Kinne, S., Peters, K., Rast, S., Müsse, J., et al. (2017). MACv2-SP: A parameterization of anthropogenic aerosol optical properties and an associated twomey effect for use in CMIP6. Geoscientific Model Development, 10(1), 433–452. https://doi.org/10.5194/gmd-10-433-2017

Stocker, T. F., Qin, D., Plattner, G.-K., Tignor, M., Allen, S. K., Boschung, J., et al. (2013). AR5 climate change 2013: The physical science basis — IPCC. Retrieved from https://www.ipcc.ch/report/ar5/wg1/

Tegen, I., Neubauer, D., Ferrachat, S., Siegenthaler-Le Drian, C., Bey, I., Schutgens, N., et al. (2019). The global aerosol–climate model ECHAM6.3–HAM2.3 – Part 1: Aerosol evaluation. Geoscientific Model Development, 12(4), 1643–1677. https://doi.org/10.5194/gmd-12-1643-2019

Ten Hoeve, J. E., Remer, L. A., Correia, A. L., & Jacobson, M. Z. (2012). Recent shift from forest to savanna burning in the Amazon Basin observed by satellite. Environmental Research Letters, 7(2), 024020. https://doi.org/10.1088/1748-9326/7/2/024020

Ten Hoeve, J. E., Remer, L. A., & Jacobson, M. Z. (2011). Microphysical and radiative effects of aerosols on warm clouds during the Amazon biomass burning season as observed by MODIS: Impacts of water vapor and land cover. Atmospheric Chemistry and Physics, 11(7), 3021–3036. https://doi.org/10.5194/acp-11-3021-2011

Thorndell, H. K., Wright, P. A., Moore, J. R., Lepley, K. A., Michelsen, M. A., & Knutson, J. R. (2021). Acidic con- chuva campaign: Studying tropical deep convective clouds and precipitation over Amazonia using the New German research aircraft HALO. Bulletin of the American Meteorological Society, 97(10), 1885–1908. https://doi.org/10.1175/BAMS-D-14-00255.1

https://doi.org/10.1029/2014JD024874
Wu, L., Su, H., & Jiang, J. H. (2011). Regional simulations of deep convection and biomass burning over South America: 2. Biomass burning aerosol effects on clouds and precipitation. *Journal of Geophysical Research, 116*(17). https://doi.org/10.1029/2011JD016106

Xue, H., Feingold, G., & Stevens, B. (2008). Aerosol effects on clouds, precipitation, and the organization of shallow cumulus convection. *Journal of the Atmospheric Sciences, 65*(2), 392–406. https://doi.org/10.1175/2007JAS2428.1

Yin, L., Fu, R., Shenviakova, E., & Dickinson, R. E. (2013). How well can CMIP5 simulate precipitation and its controlling processes over tropical South America? *Climate Dynamics, 41*(11–12), 3127–3143. https://doi.org/10.1007/s00382-012-1582-y

Yu, H., Fu, R., Dickinson, R. E., Zhang, Y., Chen, M., & Wang, H. (2007). Interannual variability of smoke and warm cloud relationships in the Amazon as inferred from MODIS retrievals. *Remote Sensing of Environment, 111*(4), 435–449. https://doi.org/10.1016/j.rse.2007.04.003

Zängl, G., Reinert, D., Rípodas, P., & Baldauf, M. (2015). The ICON (ICOshedral Non-hydrostatic) modeling framework of DWD and MPI-M: Description of the non-hydrostatic dynamical core. *Quarterly Journal of the Royal Meteorological Society, 141*(687), 563–579. https://doi.org/10.1002/qj.2378

Zhang, Y., Fu, R., Yu, H., Dickinson, R. E., Juarez, R. N., Chin, M., & Wang, H. (2008). A regional climate model study of how biomass burning aerosol impacts land-atmosphere interactions over the Amazon. *Journal of Geophysical Research, 113*(D14). https://doi.org/10.1029/2007jd009449

Zhang, Y., Fu, R., Yu, H., Qian, Y., Dickinson, R., Silva Dias, M. A. F., et al. (2009). Impact of biomass burning aerosol on the monsoon circulation transition over Amazonia. *Geophysical Research Letters, 36*(10), L10814. https://doi.org/10.1029/2009GL0377180