Putting the clouds back in Aerosol-Cloud Interactions

A. Gettelman
National Center for Atmospheric Research, 1850 Table Mesa Dr., Boulder, CO, 80305

Correspondence to: A. Gettelman
(andrew@ucar.edu)

Abstract. Aerosol Cloud Interactions (ACI) are the consequence of perturbed aerosols affecting cloud drop and crystal number, with corresponding microphysical and radiative effects. ACI are sensitive to both cloud microphysical processes (the ‘C’ in ACI) and aerosol emissions and processes (the ‘A’ in ACI). This work highlights the importance of cloud microphysical processes, using idealized and global tests of a cloud microphysics scheme used for global climate prediction. Uncertainties in key cloud microphysical processes examined with sensitivity tests cause uncertainties of nearly -30% to +60% in ACI, similar to or stronger than uncertainties identified due to natural aerosol emissions (-30 to +30%). The different dimensions and sensitivities of ACI to microphysical processes identified in previous work are analyzed in detail, showing that precipitation processes are critical for understanding ACI and that uncertain cloud lifetime effects are nearly 1/3 of simulated ACI. Buffering of different processes is important, as is the mixed phase and coupling of the microphysics to the condensation and turbulence schemes in the model.

1 Introduction

Aerosols represent the largest uncertainty in our estimates of current anthropogenic forcing of climate (Boucher et al., 2013), limiting our ability to constrain the sensitivity of the current climate to radiative forcing. Aerosols affect climate through direct effects of absorption or scattering, and indirect effects (Twomey [1977]) by changing the number of cloud drops and resulting complex microphysical interactions. Increased aerosol number concentrations are associated with more Cloud Condensation Nuclei (CCN) (Rosenfeld et al., 2008; Twomey and Squires, 1959), leading to higher cloud drop number concentrations (Nc). The relationship between aerosols and CCN is affected by a number of factors (Lohmann and Feichter, 2005), including the aerosol type and meteorological conditions. The result is a different population of cloud droplets, depending on aerosol distribution and meteorology.

But that is only the beginning of aerosol effects on clouds. Cloud microphysics (the interactions of a distribution of cloud drops at the micro-meter scale) determines how much water precipitates, the amount of water remaining in the cloud, and the resulting population of cloud drops. In global modeling experiments, ACI can be altered by the representation of cloud microphysical processes (the ‘C’ in ACI) while the aerosol processes (‘A’) remain largely unchanged. Menon et al. (2002), Rotstayn and Liu (2005), Penner et al. (2006) Wang et al. (2012) and Gettelman et al. (2013) all looked at changes to autoconversion, while Posselt and Lohmann (2008) looked at changes to precipitation.

ACI are typically quantified by the change in cloud radiative effect (Ghan et al., 2013). ACI occur most readily with liquid sulfate aerosol (H₂SO₄) derived from sulfur dioxide (SO₂) assisting the formation of cloud droplets, thus increasing cloud drop numbers. Higher drop numbers affect cloud albedo (Twomey, 1977), and potentially also affect cloud lifetime and dynamics (Albrecht, 1989; Pincus and Baker, 1994). Cloud lifetime and dynamics effects are highly uncertain (Stevens and Feingold, 2009).

Recent work (Carslaw et al., 2013; Ghan, 2013; Kiehl et al., 2000) found large sensitivities of ACI to uncertainty in natural emissions and thus pre-industrial aerosols: the ‘A’ in ACI. But these studies used fixed assumptions about how clouds interact with aerosols, assuming aerosols translated into cloud drop numbers based on fixed cloud dynamics and water content (Carslaw et al., 2013), largely ignoring the ‘C’ in ACI. The cloud microphysical state, defined as the combination of cloud liquid water path and drop number, determines cloud microphysical (precipitation rates) and radiative properties. As a result, perturbations to this state from aerosols (ACI) may depend on the base state, i.e. the re-
spatially of a cloud to a change in CCN may depend on the unperturbed CCN and resulting drop number.

In this work we quantify the sensitivity of ACI to cloud microphysics with detailed off-line tests and global sensitivity tests of ACI with a cloud microphysics scheme. First, detailed off-line tests will isolate the different components of ACI in a cloud microphysics scheme. Off-line tests will include exploration of lifetime effects and microphysical process rates. Then global simulations will analyze the sensitivity of ACI to many different aspects of cloud microphysics, including sensitivity to: (1) activation, (2) precipitation, (3) mixed phase processes (4) autoconversion treatment, (5) coupling to other parameterizations and (6) background aerosol emissions. These processes have been highlighted in previous studies.

The methodology is described in Section 2. Detailed off-line tests are in Section 3. Global results and sensitivity tests are in Section 4 and conclusions are in Section 5.

2 Methods

The double moment (mass and number predicting) bulk cloud microphysics scheme described by Morrison and Gettelman (2008) (hereafter MG1) and Gettelman and Morrison (2015) (hereafter MG2) is used for this study. The scheme handles a variable number of droplets specified from an external activation scheme (Abdul-Razzak and Ghan 2000). It can also run with a fixed droplet and crystal number. The scheme is implemented both in an off-line idealized Kinematic Driver (KiD) (Shipway and Hill 2012), as well as in a General Circulation Model (GCM), the Community Earth System Model (CESM) (Gettelman and Morrison 2015). The susceptibility of an earlier version of the scheme to aerosols has been shown by Gettelman et al. (2013) to be similar to detailed models with explicit bin microphysics that represent more accurately the precipitation process (Jiang et al. 2010).

2.1 Off-Line Tests

To isolate and test the microphysics we use a simple one dimensional off-line driver, the Kinematic Driver (KiD) (Shipway and Hill 2012) with the same microphysical parameterization as used in the global model. We use a 1 second time step, 25m vertical resolution and a 3km vertical domain in KiD. In the off-line implementation, specified drop numbers are assumed. Here we focus only on warm rain cases. We use several different cases for analysis. The basic case (Warm 2 or W2) features multiple 2 ms\(^{-1}\) updrafts over 2 hours (Gettelman and Morrison 2015, Shipway and Hill 2012). We have examined 3 other cases as well, with notation following Gettelman and Morrison (2015). These cases represent some basic idealized clouds commonly used to evaluate cloud microphysical processes such as condensation, precipitation and evaporation. Case 1 (W1) is a single 2 ms\(^{-1}\) updraft that decays in time (1 hr). Case 3 (W3) features multiple updrafts that weaken over time. Case 7 (W7) has shallower updrafts of maximum 0.5 ms\(^{-1}\) over 8 hours. To assess the impact of aerosols, experiments are conducted with variable drop number from 10-2000 cm\(^{-3}\). This spans the range from pristine to very polluted conditions.

In off-line tests, we estimate first the cloud albedo, and then divide albedo (\(A\)) changes into contributions from: (1) Liquid Water Path (\(LWP\)), (2) Cloud drop Number Concentration (\(N_c\)) and (3) Cloud Coverage (\(C\)). To estimate albedo (\(A\)) we make the assumption that

\[
A = C \ast \tau / (\beta + \tau)
\]

Where \(\beta = 6.8\), \(\tau = \alpha * LWP^{5/6} * N_c^{1/3}\) and \(\alpha = 0.19\) (Zhang et al. 2005, eqn. 19-20). Strictly speaking the albedo should include a surface reflectance term, which over ocean would be (1-C) * \(A_{sfc}\), where for ocean \(A_{sfc}=0.05\). For these idealized cases we assume \(A_{sfc}=0\). The change in albedo (\(dA\)) can then be represented as:

\[
\Delta A = \frac{dA}{dN_c} \Delta N_c + \frac{dA}{dLWP} \Delta LWP + \frac{dA}{dc} \Delta C + \tau
\]

\(C\) (cloud cover or cloud fraction) has one value for each simulation. \(N_c\) has one specified value for each simulation and \(LWP\) is an average over the simulation period. \(r\) is a residual. The changes are discrete differences between simulations with different specified \(N_c\) for each case.

The idealized one-dimensional kinematic driver is designed to test different microphysical schemes in the same framework. Results of such idealized off-line tests are qualitatively useful for examining the relative importance of individual processes for ACI. We use them for illustration, and will use global sensitivity tests of the full GCM for quantification.

2.2 Global Sensitivity Tests

The MG2 scheme is implemented in version 5 of the Community Atmosphere Model (CAM5.3, Neale et al. 2010) as described by Gettelman et al. (2015). The MG2 scheme in CAM is coupled to aerosol activation on liquid (Abdul-Razzak and Ghan 2002) and ice (Liu et al. 2007) hydrometeors, and can also take specified number concentrations for liquid and ice. CAM5 features a modal aerosol model (Liu et al. 2012). The MG scheme has prognostic drop number with no minimum drop number.

For the global model, we run simulations with specified climatological sea surface temperatures (SSTs) and greenhouse gases representing year 2000 conditions. We then vary aerosol emissions in two simulations for the year 2000 and 1850; differences represent only the effects of aerosol emis-
sions. ‘1850’ refers only to the aerosol emissions, greenhouse gases and SSTs remain at year 2000 conditions. Simulations are 1.9° latitude by 2.5° longitude horizontal resolution. Simulations are 6 years long, and the last 5 years are analyzed. Simulations are similar to previous work (Gettelman et al., 2012, 2015). Sensitivity tests are described below.

To understand the uncertainty in using 5 years of simulation, we performed an uncertainty analysis. This consisted of running the MG2 experiment out for 20 years (for 2000 and 1850 conditions). Analysis of separate 5 year periods indicates uncertainty of 0.08 Wm\(^{-2}\) for ACI and LW/SW components (about 10%) and within 0.04 Wm\(^{-2}\) for direct effects relative to 20 year means. We also performed nudged experiments where winds or winds and temperatures were fixed to a previous CAM simulation, but these produced lightly different cloud radiative effects, and thus slightly different quantitative values for ACI (different by 20–40%). Qualitative patterns and zonal mean structure of ACI are similar to the free running experiments.

In global simulations, ACI can be defined as the change in cloud radiative effects (CRE) in the long wave (LW) and shortwave (SW), where CRE is equal to the all sky top of atmosphere (TOA) radiative flux minus an estimate of what the clear sky flux would be without clouds, but with the same state (temperature, humidity and surface structure). CRE is adjusted following Ghan et al. (2013) to use the ‘clean-sky’ effects based on TOA fluxes estimated with a diagnostic call to the radiation code without aerosols. Results are similar, but with a slightly higher magnitude, to a direct estimate of ACI using CRE. Direct absorption and scattering by aerosols is also estimated by differencing the TOA radiative fluxes to TOA fluxes estimated with a diagnostic call to the radiation code without aerosols.

Table 1 describes the different sensitivity tests. As noted below, tests are motivated by previous studies identifying microphysical sensitivities. All tests are pairs of simulations with emissions of aerosols set to 2000 and 1850, except for the MG2-2000-1750 and MG2-1850-1750, which use different emissions years to explore different magnitudes of emissions changes. To explore how linear the changes in emissions are we look at emissions without any human influence (no biomass burning, domestic or industrial emissions) and term this ‘1750’. We also explore modifying background natural emissions in both 1850 and 2000 by a factor of 0.5 or 2. These experiments test the impact of emissions (Carslaw et al., 2013), not cloud microphysics.

Tests also track the evolution of the cloud microphysics in CAM from MG1 (Morrison and Gettelman, 2008) to MG2 (Gettelman and Morrison, 2015). MG1.5 is an interim version that has (a) changes to the location where activated numbers are applied to before estimation of microphysical processes (which thickens the stratiform clouds) and (b) compensating increases in the threshold relative humidity for cloud formation to thick clouds back to radiative balance. The difference between MG1 and MG1.5 tests the changes to the activation scheme. The impact of prognostic precipitation is tested by the differences between MG2 and MG1.5.

Two experiments test sensitivity to mixed phase cloud processes. MG1-Hoose contains a representation of mixed phase ice nucleation that is tied to aerosols (Hoose and Kristjanson, 2010), instead of the temperature-dependent scheme in MG1 (Meyers et al., 1992). This change tests the mixed phase ice scheme. The MG2-Berg0.1 simulations reduce the efficiency of the vapor deposition process by a factor of 10. This sensitivity test is motivated by the work of Korolev (2007) and Korolev (2008) who suggested that due to updraft rates in clouds at least half the time the vapor deposition rate may not apply. It is also motivated by tests in Lawson and Gettelman (2014) extending this to a large scale model that would also assume inhomogeneity in a grid box, and found improvements in Antarctic radiative fluxes.

Perturbations to the MG2 microphysics itself are also explored. First, removing evaporation of rain number (MG2-NoER) present in MG2 but not MG1. Then removing lifetime effects by fixing cloud drop numbers in autoconversion, sedimentation and freezing (MG2-NoLif). A fixed number of 100 cm\(^{-3}\) for liquid drops and 0.1 cm\(^{-3}\) for ice crystals is used. An additional simulation with 300 cm\(^{-3}\) for liquid drops yields quantitatively and qualitatively similar results. A simulation is performed changing the moist turbulence scheme and coupling to cloud microphysics using a higher order closure scheme called Cloud Layers Unified By binormals (CLUBB, (Bogenschutz et al., 2013)) in MG2-CLUBB (Gettelman et al., 2015). As noted in the introduction, several previous studies have focused on sensitivity of ACI to the autoconversion process. Accordingly, we alter the autoconversion scheme in the simulations MG2-K2013 (Kogan 2013) and MG2-SB2001 (Seifert and Beheng, 2001).

These tests and the parameter values are motivated by previous work. Zhao et al. (2013) conducted a perturbed parameter ensemble with a similar version of CESM and focused on radiative effects. However, Zhao et al. (2013) and other perturbed parameter ensembles have not focused on the radiative perturbations due to aerosols, and here the experiments are all pairs of simulations with pre-industrial and present day aerosols.

### 3 Results: Off-Line Tests

Figure 1 illustrates basic results from the off-line experiments with different specified drop numbers. As drop number increases, average cloud condensate mass increases (Figure 1A) and the surface rain rate (Figure 1B) and rain mass (Figure 1C) drop rapidly to zero for \(N_c > 500\) cm\(^{-3}\). The cloud albedo (estimated using Equation 1) increases substantially (Figure 1D) for increasing drop number. The mechanism for the microphysical changes as described by Gettelman and Morrison (2015) is the decrease in the autoconversion rate with increasing drop number (Figure 1E), which...
also causes decreases in accretion rate as the rain mass decreases (Figure 1F).

The W2 case initiates two separate layers of cloud in subsequent updrafts after the first. There is larger autoconversion and accretion in the lower layer, creating the peaks in cloud mass (Figure 1A), rain mass (Figure 1C), autoconversion (Figure 1E) and accretion (Figure 1F). Autoconversion and accretion are NOT increasing at the bottom of the cloud. Instead, this is a different layer of cloud not seen as separate in the time average.

The impact of these changes on albedo is highlighted in Figure 2. The albedo increases with higher drop numbers (Figure 2A). At low liquid water paths, the albedo changes are more sensitive to LWP. In Figure 2A, the slope (dA/dLWP) is constant at low LWP, but shifts to reduced sensitivity at high LWP. Using the decomposition of the albedo change in Equation 2, we can break down the change between pairs of simulations (Nc=20 to Nc=10, etc) by the different components: the total change in albedo (Tot), the change due to LWP (dA/dLWP), the change due to changes in Nc (dA/dNc × ΔLWP) and the change due to cloud cover changes (dA/dC × ΔC). Differences are calculated based on the difference in time averaged albedo between two simulations. The residual is the difference between the total and the sum of the 3 terms, which is small. In the W2 case with an oscillating updraft, the change in cloud coverage dominates the albedo change (Figure 2B). Note that cloud mass (Figure 1A) is changing along with cloud coverage (Figure 1D). Most of the difference in Figure 2A (Cloud Mass) is change to the extent of clouds with the same in-cloud water content, hence for this case, the coverage is identified as being critical.

Figure 3 illustrates the same set of albedo sensitivity terms for 4 different cases. The mean and one standard deviation of pairs of adjacent drop numbers (7 pairs from 8 values of drop number) is indicated by the error bar range and midpoint, and the median is shown as a diamond. The W2 case from Figure 2B is illustrated in Figure 3B (Black line), where cloud coverage dominates the change in albedo. Some cases have mostly small differentials for the terms, and only some values of Nc have large differentials, so the median is often near zero but the average (dominated by 1–2 cases) is non-zero. The ‘Base’ case (black) is the basic case using the autoconversion scheme of [Khairoutdinov and Kogan (2000)], hereafter KK2000. KK2000 represent autoconversion from a fit to cloud resolving model experiments as a function of the cloud mass and an inverse function of drop number, the autoconversion rate (Au) is Au = 1350q−1.47 N−1.79. This is also true for W1 (Figure 3A), with lower sensitivity. However, the LWP and Nc changes are important in the W3 and W7 cases (Figure 3C and D). These are weaker multiple updraft cases.

Also shown in Figure 3 are three additional sets of experiments where the microphysics has been modified to limit the ‘lifetime’ effects. This has been done by specifying a constant fixed drop number of 100 cm−3 to (a) the autoconversion scheme (Au), and (b) the sedimentation (Sed) or both (Nolif). Different drop numbers ranging from 10–2000 cm−3 are used for all other processes in the microphysics. The ‘Nolif’ cases (dark blue in Figure 3), are similar to the ‘Au’ cases (green: autoconversion effects only) indicating that autoconversion is the dominant process for lifetime effects. In particular, removing the lifetime effects by specifying the number concentration going into autoconversion removes the cloud coverage effects in the W2 case (Figure 3B), and perhaps more significantly removes the LWP effects on albedo in all cases. This leaves only the drop number effects on albedo. Thus for some cases with partial cloud cover (e.g. like the W2
Fig. 1. Warm 2 (W2) off-line tests of (A) time averaged cloud mass (g kg\(^{-1}\)), (B) Timeseries of surface rain rate (mm hr\(^{-1}\)), (C) time averaged rain mass (g kg\(^{-1}\)), (D) time series of Albedo, time averaged (E) autoconversion and (F) accretion rates (kg kg\(^{-1}\) s\(^{-1}\)). Different colors correspond to different fixed cloud drop number concentrations.

Recognizing that the representation of autoconversion is important, we explore two alternatives. Kogan (2013), hereafter K2013, use a similar representation as KK2000 and derive \(A_u = 7.90 \times 10^{10} q_r^{4.22} N_c^{-3.01}\). Seifert and Beheng (2001), hereafter SB2001, derive expressions for autoconversion and accretion that include the rain water mixing ratio as a proxy for large cloud droplets to describe the broadening of the drop size distribution and reduce the efficiency of accretion in the early stage of the rain formation. We have implemented both of these parameterizations into the microphysics scheme.

Figure 4 shows the impact of the SB2001 scheme in the single updraft W1 case with fixed drop number of 200 cm\(^{-3}\). Relative to KK2000 (black), the use of SB2001 (red) for autoconversion results in higher cloud mass (Figure 4A), significantly less precipitation (Figure 4B) and delayed and smaller rain formation (Figure 4C) and rain number concentration (Figure 4D). Autoconversion (Figure 4E) is delayed, but has a higher magnitude, and accretion is also delayed (Figure 4F), but has a lower magnitude. The changes are significant. At lower number concentrations the differences are smaller, and they are larger at higher number concentrations (not shown).

The impact of these changes on the albedo changes in the off-line driver cases is illustrated in Figure 5. KK2000 is the same as the ‘Base’ case in Figure 3. Results are similar with different autoconversion schemes in case W1 (single updraft: Figure 5A) and case W7 (shallow updraft: Figure 5D). In case W2, there is a significant reduction in the cloud coverage and LWP effects on albedo with SB2001 (Figure 5B). And there is a significant reduction in the LWP effect in case W3 for SB2001 and K2013, (Figure 5C), which is compensated for in cloud cover changes. Auto-conversion matters in the cases with multiple updrafts where cloud coverage is most sensitive (W2 and W3), and it matters more for the oscillating (W2) than decaying (W3) or weak (W7) updraft case. This is likely because with a limited temporal updraft, the timing of precipitation matters.
4 Results: Global Sensitivity Tests

Global sensitivity tests with CESM explore how different perturbations to cloud microphysics impact ACI. All tests are pairs of simulations with emissions of aerosols set to 2000 and 1850, except for the 2000-1750 and 1850-1750 cases, which use different emissions years to explore different magnitudes of emissions changes. The experiments described in Section [2.2] and Table [1] fall into several categories chosen to span key sensitivities in different microphysical processes. These are based on a number of previous studies that have identified these different processes as critical for the interaction of aerosols with clouds. These studies are highlighted below. The different processes include: (1) Aerosol activation (MG1) [Ghan et al., 2013] [Carslaw et al., 2013], (2) Precipitation (MG1.5, NoER: evaporation of rain) [Wood et al., 2009] [Jiang et al., 2010], (3) Mixed Phase (Berg0.1: vapor deposition and Hoose: ice nucleation) [Hoose and Kristjansdóttir, 2010] [Lawson and Gettelman, 2014], (4) Autoconversion (Lifetime effects and 2 other autoconversion schemes: K2013, SB2001) [Wood et al., 2009] [Gettelman et al., 2013], (5) Coupling to other schemes (CLUBB) [Quo et al., 2011], (6) Natural emissions (Nat 0.5 and Nat2) [Carslaw et al., 2013]. In particular, the range of ‘natural’ aerosol emissions is identical to the range in Carslaw et al. [2013].

The radiative changes between the pairs of simulations in each sensitivity experiment are indicated in Figure [3A]. ACI are defined as the change in clean sky Cloud Radiative Effect (ΔCRE) between pairs of simulations with different aerosol emissions [Ghan et al., 2013]. Directly using CRE yields similar quantitative (%) differences between simulations. ACI for 2000-1850 emissions are -1.57 W m⁻² with MG1, -1.13 W m⁻² with MG1.5 and -0.98 W m⁻² with MG2 (Table [2]). Maximum ACI is found in N. Hemisphere midlatitudes (Figure [3A]), where most anthropogenic emissions occur. N. Hemisphere midlatitudes is also where the largest changes to LWP (Figure [3B]) and cloud top drop number concentration (ΔNc, Figure [3C]) occur. Interestingly the changes to cloud top drop effective radius (ΔRe, Figure [3D]) spread farther into high latitudes.

Most of the ACI are due to the shortwave (SW; solar) wavelengths: ‘brighter’ clouds (Table [2]). However, there is a significant component of positive ACI in the longwave (LW; terrestrial) wavelengths. This is a result of two factors. First is the effect of aerosols on cirrus clouds, where more ice nuclei are formed, and clouds become more opaque in the longwave wavelengths are estimated by taking the highest level (first from the top of the model going down) where cloud condensate is found. This is done at each point in the model and averaged over those points which are non-zero. The values and figures in the text come from these simulations. The net CRE for all the simulations (Table [3]) is broadly similar, within about ±1 W m⁻², except for the MG2-CLUBB simulation, which has a different balance of CRE, drop number and effective radius (Table [3]).

The radiative changes between the pairs of simulations in each sensitivity experiment are indicated in Figure [3A]. ACI are defined as the change in clean sky Cloud Radiative Effect (ΔCRE) between pairs of simulations with different aerosol emissions [Ghan et al., 2013]. Directly using CRE yields similar quantitative (%) differences between simulations. ACI for 2000-1850 emissions are -1.57 W m⁻² with MG1, -1.13 W m⁻² with MG1.5 and -0.98 W m⁻² with MG2 (Table [2]). Maximum ACI is found in N. Hemisphere midlatitudes (Figure [3A]), where most anthropogenic emissions occur. N. Hemisphere midlatitudes is also where the largest changes to LWP (Figure [3B]) and cloud top drop number concentration (ΔNc, Figure [3C]) occur. Interestingly the changes to cloud top drop effective radius (ΔRe, Figure [3D]) spread farther into high latitudes.
Fig. 3. Albedo change by different sensitivity (dA) terms from different warm rain cases. (A) Warm 1, (B) Warm 2, (C) Warm 3 and (D) Warm 7. Albedo terms in each panel correspond to the total (Tot) change and the portion due to Liquid Water Path (LWP), number concentration (Nc), Cloud Coverage (CC) and a residual (Res). Standard case (Base) in black. Also shown are the no lifetime effects case (dark blue) and the two components of the lifetime effect: Sedimentation (cyan) and Autoconversion (green).

than they become brighter in the shortwave (Gettelman et al., 2012). Second is a compensation effect between LW and SW for cirrus clouds due to movement of cirrus cloud fraction in the tropics. The second effect accounts for a good amount of the variance in the magnitude of the LW and SW between sensitivity tests: increases in LWCRE are compensated for by decreases in (increased magnitude of negative) SWCRE.

To understand how ACI change with cloud microphysics, we explore how the radiative effects of ACI are related to microphysical properties strongly related to radiative effects. Figure 4 illustrates some of the broad scale patterns across the simulations, by relating the changes in cloud radiative effect (ACI = ΔCRE) to other properties of the simulations, namely changes to LWP (in percent, Figure 4A), changes to the cloud top drop number concentration (Figure 4B), changes to cloud top effective radius (Figure 4C), or changes in total (vertically integrated) cloud coverage or fraction (Figure 4D). There is a strong correlation between ΔLWP and ACI (Figure 4A). The only simulations that differ from the correlation are those with CLUBB and the simulation without lifetime effects (NoLif). The CLUBB simulation has a very different coupling of large scale condensation and cloud microphysics, as described by Gettelman et al. (2015), where the microphysics is sub-stepped with the CLUBB condensation scheme 6 times in each time step. The NoLif simulation has basically no change in LWP, which is consistent with the off-line KiD tests with a similar formation. The ACI go from -0.98 (MG2) to -0.72 Wm$^{-2}$ (NoLif) in Table 2. There is no correlation between the change in cloud top drop number (Figure 4B) or effective radius (Figure 4C) and ACI. Changes in effective radius are negative, indicating smaller drops in the present with more aerosols, than
in the past (pre-industrial). There are small changes in total cloud cover that correlate slightly with ACI (Figure 7D), but mostly because there are large changes (increases in cloud coverage) in 3 simulations with large ACI (CLUBB, MG1, MG1-Hoose).

The simulation without lifetime effects (NoLif) actually has the largest change (reduction) in averaged drop radius (Figure 7C), despite no change in LWP (Figure 7A) and small changes in ACI. Most simulations have an increase in cloud drop number of $\sim 30 \text{cm}^{-3}$. This is an interesting result because many models still prescribe the radiative effects of aerosols by linking aerosol mass to a change in cloud drop number or size. Whereas in CAM, the clearest effects seem to be due to LWP, though ACI are non-zero even if $\Delta LWP = 0$.

The following sub-sections detail each of the ‘dimensions’ of changes to understand the magnitude of the effects.

### 4.1 Activation

The change to drop activation (moving it before the cloud microphysical process rates) is seen in the difference between MG1 v. MG1.5. This is a substantial reduction in ACI from $-1.57$ to $-1.13 \text{ Wm}^{-2}$ or 28% (Table 2). The likely cause is that by activating number first, other processes in the microphysics act on the revised number, and this likely “buffers” the changes in the indirect effects (Stevens and Feingold, 2009). Note that there is basically no difference in LWP
change between MG1 and MG1.5 (Table 3). Effects are not simply linear however since MG1.5 with lower ACI has a larger $\Delta$REL and $\Delta N_c$.

4.2 Prognostic Precipitation and Rain Evaporation

The major scientific changes between MG1.5 (MG1 with the activation change) and MG2 as described by Gettelman and Morrison (2015) are the addition of prognostic precipitation, and the addition of evaporation of number when rain evaporates. The latter change to evaporation of rain number actually does seem to make a difference: a reduction in ACI (Table 2) due to a reduction in $\Delta LWP$ (Table 3). The total reduction between MG1.5 and MG2 is -1.13 to -0.98 Wm$^{-2}$, or about 14%. This occurs through reductions in the $\Delta LWP$ (Table 3), especially between 10–60°N latitude (Figure 6B).

4.3 Mixed Phase Clouds

Two different sets of experiments were conducted to look at the impact of altering mixed phase clouds. The changes in MG1-Hoose make the simulations sensitive to anthropogenic aerosols in the mixed phase regime where they were not before. This causes increases in the magnitude of the LW and SW components of ACI (Table 2), but a small change in the net ACI (~4%). The sensitivity of LWP goes up ($\Delta LWP$: Table 3). This experiment has the largest LW ACI, which is expected since it adds ACI in the mixed phase cloud regime between 0 and -20°C, which will have a significant effect on the LW radiation.

The second experiment used the MG2 configuration to reduce the efficiency of the vapor deposition onto ice (Bergeron-Findeisen process) by a factor of 10. This simulates inhomogeneity in cloud liquid and ice (or effectively inhomogeneity for in-cloud supersaturation or vertical ve-
Table 2. Radiative impacts of ACI for the different sensitivity tests. Change in Top of Atmosphere (TOA) flux (ΔR), ACI as change in clean sky Cloud Radiative Effect (ΔCRE), and its long wave (LW) and shortwave (SW) components following Chan et al. (2013). Direct Effects (DE) of aerosols as described in the text. Finally, a residual (Res=ACI + DE - ΔR).

<table>
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<th>Case</th>
<th>ΔR Wm$^{-2}$</th>
<th>ACI Wm$^{-2}$</th>
<th>ΔLWCRE Wm$^{-2}$</th>
<th>ΔSWCRE Wm$^{-2}$</th>
<th>DE Wm$^{-2}$</th>
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<td>-0.01</td>
<td>0.04</td>
</tr>
<tr>
<td>MG2-Nat0.5</td>
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<td>0.21</td>
<td>-1.44</td>
<td>-0.11</td>
<td>-0.12</td>
</tr>
<tr>
<td>MG2-Nat2</td>
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<td>-0.86</td>
<td>0.09</td>
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</tr>
<tr>
<td>MG2-CLUBB</td>
<td>-1.43</td>
<td>-1.36</td>
<td>-0.05</td>
<td>-1.50</td>
<td>-0.02</td>
<td>0.14</td>
</tr>
<tr>
<td>MG2-NoLif</td>
<td>-0.78</td>
<td>-0.72</td>
<td>0.36</td>
<td>-1.08</td>
<td>-0.08</td>
<td>0.02</td>
</tr>
<tr>
<td>MG2-K2013</td>
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<td>-0.08</td>
<td>-0.01</td>
</tr>
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</tr>
<tr>
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<td>-0.08</td>
<td>0.00</td>
</tr>
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<td>-1.41</td>
<td>0.26</td>
<td>-1.67</td>
<td>-0.06</td>
<td>-0.06</td>
</tr>
</tbody>
</table>

4.4 Autoconversion and Lifetime Effects

As in Section 3, we can also explore the sensitivity of the microphysics to autoconversion scheme. Gettelman et al. (2013) noted that the description of autoconversion and accretion matters for ACI, consistent with a series of previous studies (Menon et al., 2002; Rotstayn and Liu, 2005; Penner et al., 2006; Wang et al., 2012). One of the reasons for lower ACI in MG2 is due to the reduction of the ratio of autoconversion to accretion (more accretion and less autoconversion) with prognostic rain in MG2 (Gettelman et al., 2015).

Here we explore the impact of different autoconversion schemes on ACI. The K2013 scheme actually increases slightly the ACI over MG2 with KK2000 (Table 2), again consistent with an increase in ΔLWP (Table 3). Conversely, the SB2001 scheme, with a smaller ΔLWP, reduces ACI from -0.98 to -0.77 Wm$^{-2}$, or 22%, and the ‘NoLif’ simulation reduces ACI to -0.72 Wm$^{-2}$, (nearly -30%) largely through more compensation between LW and SW effects. (larger LW effects, indicating clouds with cold cloud tops may have higher LW emissivity). This indicates that the life-time effects themselves may approach 1/3 of ACI (the total change in radiative flux changes from -1.04 in MG2 to -0.78 in the NoLif simulation, a reduction of 33%). The lifetime effects are not that sensitive to the drop number threshold chosen. Results of a ‘NoLif’ simulation with 300 cm$^{-3}$ rather than 100 cm$^{-3}$ for liquid drops yield similar results for ΔR or ACI.

The regional pattern of ACI, based on the total change in top of atmosphere fluxes is illustrated in Figure 8 for (A) the base MG2 case and (B) the ‘NoLif’ case. The average local standard deviation of annual TOA flux is about 3 Wm$^{-2}$, so Figure 8 shows regions with differences larger than one standard deviation. ACI effects are mostly in the N. Hemisphere, and mostly over the oceans. There are some tropical effects in S. E. Asia and off the equatorial E. Pacific, the latter due to anthropogenic emissions over the Amazon. The removal of lifetime effects in Figure 8B indicates they are strong over the N. Hemisphere mid-latitude storm tracks, especially in the N. Pacific. Lifetime effects also are strong in the equatorial E. Pacific. Lifetime effects do not seem to impact the stratocumulus region off the coast of California, which has strong ACI without lifetime effects.

The effect of autoconversion and accretion is illustrated in Figure 9. Figure 9 shows autoconversion and accretion rates as change in clean sky CRE, or ACI. The figure represents 60°S-60°N averages for all liquid.
clouds treated by the stratiform cloud scheme, so it does not include convective clouds. A similar figure for just the S.E. Pacific region yields similar results, but not as good statistics.

Accretion rates (Ac) are well represented in MG2 with the KK2000 autoconversion (Figure 9C), but autoconversion rates (Au) at low LWP are very large (Figure 9B), leading to a low Ac/Au ratio (Figure 9). With the SB2001 scheme, accretion is high at low LWP, and autoconversion is 2 orders of magnitude lower. Autoconversion in particular is much closer to estimates from VOCALS (Terai et al., 2012). The result is a higher Ac/Au ratio, which may be too high at low LWP. The K2013 scheme (cyan in Figure 9) yields similar results to KK2000: autoconversion is almost the same, and accretion is a little bit higher. The no lifetime simulation (green in Figure 9] has accretion rates similar to KK2000, but lower autoconversion rates due to fixing the drop number in the autoconversion scheme. The no lifetime simulation has perhaps the closest representation to the Ac/Au ratio (Figure 9A).

4.5 Coupling to Other Schemes

We can also examine the effect of coupling of the microphysics to other cloud schemes in the model. The CLUDB simulation uses a different unified higher-order closure scheme to replace the CAM large scale condensation, shallow convection and boundary layer scheme, as described by Bogenschutz et al. (2013). It uses MG2 with a different sub-stepping strategy of 5 minute time steps, called 6 times per model time step.

Notably, CLUDB provides a unified condensation scheme for the boundary layer, stratiform and shallow convection regimes, so that ACI are included in shallow cumulus regimes in this formulation. This results in a substantial in-
Fig. 7. ACI (Change in CRE, Wm$^{-2}$) v. (A) percent change in LWP, (B) change in cloud top drop number concentration ($\Delta N_c$, cm$^{-3}$), (C) change in cloud top effective radius ($\Delta REL$, m$^{-6}$ or microns) and (D) change in total cloud coverage (CTOT, %) for different sensitivity tests. Red colors are tests of different emissions, blue is the ‘base’ MG2 case and black are the other sensitivity tests.

The change in LWP ($\Delta LWP$) is moderate (Figure 7A), and less than would be expected based on the ACI (Figure 7A). CLUBB has a lower change in cloud top drop number (Figure 7C), which likely is contributing to ACI. The increase appears to be occurring in the sub-tropics of the N. and S. Hemispheres (Figure 7A) mostly from 20-40°N over the Pacific and Atlantic (not shown). The increase in ACI over the sub-tropical N. Pacific and Atlantic is consistent with ACI being added in shallow cumulus regimes. Further future exploration of the impacts of this change is warranted but is beyond the scope of this work. Also notable is that CLUBB simulations have a decrease in the positive LW ACI. This occurs in the tropics, and may be related to changes in transport of water vapor into the upper troposphere, reducing high cloudiness and any positive ACI associated with high (cirrus) clouds. These changes may also be due to differences in how CLUBB treats aerosols and aerosol scavenging in these simulations: it appears that the change in AOD is larger in CLUBB than in other simulations, perhaps due to different treatment of aerosol scavenging and transport in CLUBB. Thus a very different physical parameterization suite with the same microphysical process rates can lead to very different ACI.

4.6 Background Emissions

Finally we explore the impact of background emissions on ACI. For these experiments no changes to the model are made. The experiments here all use the MG2 code. The only changes are to the emissions files. First, we just explore what happens with different baselines: a larger period (2000-1750) or a smaller period (1850-1750) than the basic 2000-1850
Fig. 8. ACI as the total change in the top of atmosphere clean sky Cloud Radiative Effect (CRE) between simulations with 2000 and 1850 aerosol emissions for (A) base (MG2) and (B) no lifetime effect (NoLif) cases.

Fig. 9. 60°S to 60°N latitude (A) ratio of Accretion to Autoconversion, (B) Autoconversion rate and (C) Accretion Rate using KK2000 (red), SB2001 (dark blue), K2013 (cyan) autoconversion and the no lifetime (NoLif) simulation (green). Estimates derived from observations from the VOCALS experiment shown as black crosses (see text for details).
(MG2). As noted, for ‘1750’ emissions, we remove all human sources from the 1850 emissions. So this is really a ‘no Anthropogenic’ emissions case. Figure 7A illustrates that the 2000-1750, MG2 and 1850-1750 changes are fairly ‘linear’, with LWP changing about 1% per -0.1 Wm\(^{-2}\) change in ACI. The changes are also somewhat linear for changes to cloud top drop number (Figure 7B) and effective radius (Figure 7C). Larger changes occur with higher emissions differences. This is not true for cloud coverage changes however (Figure 7D), where MG2 (2000-1850) and 2000-1750 have about the same decreases in cloud coverage, while there is no change for 1850-1750.

Carslaw et al. (2013) found ±20% effects on ACI from the assumed level of background emissions. Similar to Carslaw et al. (2013) we conducted experiments by halving (Nat0.5) or doubling (Nat2) the ‘natural’ emissions of aerosols from dust, volcanoes, ocean dimethylsulfide (DMS) and natural organic carbon (terpenes and other biological aerosols). This was done for both pre-industrial and present emissions. Halving natural emissions makes the model more sensitive to anthropogenic aerosols (-1.13 to -1.24 Wm\(^{-2}\) ACI in Table 2 a 27% increase) Doubling emissions decreases the sensitivity significantly (-0.98 to -0.68 Wm\(^{-2}\) ACI in Table 2 a 30% decrease). The total change in TOA flux (dR) ranges from -1.46 (+34%) to -0.87 Wm\(^{-2}\) (-17%) in Table 2. There is little change in LW ACI. Thus we can conclude that the background natural aerosols are important for determining the total ACI.

The variation in ‘natural’ emissions alters present day Aerosol Optical Depth (AOD). Global mean AOD for Nat2, baseline (Nat1) and Nat0.5 is 0.175, 0.137 and 0.117 respectively, with most of the difference caused by the imposed change to the efficiency of dust production and the dust AOD of 0.042, 0.024 and 0.013 respectively for natural emissions scaling of 2, 1 and 0.5. This highlights and confirms the need to better constrain background aerosols identified by Carslaw et al. (2013).

4.7 Summary of Sensitivity Tests

The sensitivity of ACI in the global model in terms of the percent change in ACI (ΔCRE) is illustrated in Figure 10. Different categories correspond to groups of sensitivity tests noted above. The autoconversion scheme is particularly important, also manifested through ‘lifetime’ effects (Figure 3) that change the overall mean LWP in simulations.

The SB2001 parameterization that reduces autoconversion at low LWP reduces ACI, and also reproduces estimates of autoconversion rates better (Figure 9). Different autoconversion parameterizations can change ACI by 35%, and lifetime effects in CESM account for about 1/3 of total ACI. The use of prognostic precipitation, and the evaporation of rain also affect ACI, largely through a similar mechanism of changing the balance between accretion and autoconversion: with more accretion using prognostic rain.

Changes to the mixed phase of clouds, in particular a reduction of the rate of vapor deposition (Berg0.1) to account for sub-grid inhomogeneity, results in an increase in the sensitivity of ACI to LWP. Reducing vapor deposition in the mixed phase increases the occurrence of liquid over ice. Liquid has a longer lifetime (and hence larger average shortwave radiative effect), and liquid clouds are more readily effected by sulfate aerosols than ice clouds are (only homogeneous freezings is effected by sulfate). The change to mixed phase ice nucleation (Hoose) has little impact on the net ACI, but a big impact on the LW. LW and SW effects for colder clouds tend to nearly cancel, with a slightly positive residual (similar to the net cloud forcing for cold clouds), so the LW does not have a strong effect typically on the net ACI in the sensitivity tests, but it does show that changes to colder cloud that effect the LW may increase the gross magnitude of ACI.

Coupung of the microphysics to different turbulence closures and adding the treatment of ACI in shallow convection (CLUBB) alters ACI by over 50% (Figure 10). ACI in deep convection is still not treated, and this may also be important for ACI (Lohmann et al. 2008).

Changing activation to allow all processes to see revised number concentrations lowers ACI by 25% (MG1 v. MG1.5), likely due to buffering of the change to activation by other processes in the microphysics. These microphysical effects are larger than aerosol processes or emissions uncertainties (the ‘A’ in ACI). Natural (or background) emissions can alter the ACI significantly with the same cloud microphysics code, seen in the ‘Emissions’ bar in Figure 10 with variability from -30% to +30%, consistent with previous work (Carslaw et al. 2013), indicating ±20% sensitivity of ACI to similar perturbations of natural emissions. Carslaw et al. (2013) also noted ACI sensitivity of ±10% to aerosol processes, much smaller than the sensitivity to microphysical processes noted here.

5 Discussion/Conclusions

Results of idealized and global model tests of a cloud microphysics scheme indicate strong sensitivity of ACI, the radiative response of clouds to aerosol perturbations, to cloud microphysics. Idealized experiments illustrate the different dimensions of aerosol cloud interactions, and how different cloud regimes may be affected in different ways by idealized aerosol perturbations. The idealized tests show that the representation of the autoconversion process is critical for cloud microphysical response to different drop numbers. These tests are consistent with, and help motivate global sensitivity tests.

The sensitivity of ACI to the cloud microphysics with MG2 is -30% to +55%, larger than the effects of background emissions (-30% to +30%). Better representations of cloud microphysical processes (the ‘C’ in ACI) are critical for representing the total forcing from changes in aerosols. These
Table 3. Microphysical impact of different sensitivity tests. Mean CRE, Change in LWP (%), mean LWP, change in cloud top (CT) effective Radius (REL), mean CT REL, change in CT drop number concentration (Nc) and mean CT Nc.

<table>
<thead>
<tr>
<th>Case</th>
<th>CRE (W m(^{-2}))</th>
<th>ΔLWP (%)</th>
<th>LWP (g m(^{-2}))</th>
<th>ΔREL (CT) (m(^{-6}))</th>
<th>REL (CT) (m(^{-6}))</th>
<th>ΔNc (CT) (cm(^{-3}))</th>
<th>Nc (CT) (cm(^{-3}))</th>
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<tbody>
<tr>
<td>MG1</td>
<td>-28.0</td>
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<td>44.6</td>
<td>-0.4</td>
<td>9.5</td>
<td>22.5</td>
<td>89.0</td>
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<td>-28.4</td>
<td>8.6</td>
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<td>-0.5</td>
<td>9.5</td>
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<td>45.0</td>
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<td>107.2</td>
</tr>
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<td>35.0</td>
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<td>37.6</td>
<td>-0.6</td>
<td>8.9</td>
<td>28.6</td>
<td>107.4</td>
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<td>MG2-SB2001</td>
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<td>28.6</td>
<td>106.9</td>
</tr>
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<td>-0.6</td>
<td>8.8</td>
<td>27.1</td>
<td>101.7</td>
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</table>

Fig. 10. Percent change in ACI for different dimensions of sensitivity tests as described in the text.
effects are stronger than uncertainties in aerosol emissions or processes. These sensitivity tests are not exhaustive in any statistical sense, but form a baseline based on expert judgement, including processes identified by previous work that have been found to be important. We also note that the relative importance between these dimensions of microphysics and aerosols is important. A more significant perturbed parameter ensemble, similar in spirit to Carstew et al. (2013), but including cloud microphysical uncertainties is currently being developed.

Uncertain ‘lifetime effects’ are manifest in CESM through changes to LWP with changes in aerosols. Lifetime effects in CESM represent about 1/3 of the total ACI. The mixed phase and the shallow convective regimes are also important, indicating that aerosol effects in convective clouds should be considered. Autoconversion parameterizations in particular seem to specify ‘lifetime’ effects that are highly uncertain. Many global models still prescribe cloud drop number or size based on aerosol mass. This may be problematic as interactions with different microphysical processes are important for the magnitude of ACI.

How general are these results across models? The model framework with MG2 is a ‘typical’ two-moment bulk microphysics scheme with a framework similar to other schemes. Many of the process rate formulations for autoconversion examined here (e.g. KK2000) are used by other schemes as well. The sensitivity to background aerosol emissions is very similar to that diagnosed by Carstew et al. (2013). In addition, the sensitivity of the microphysical process rates to autoconversion and accretion that occurs with prognostic precipitation is qualitatively similar to Posselt and Lohmann (2008). However, adding aerosol effects in all convective clouds (deep and shallow) in a different GCM reduced the ACI (Lohmann et al., 2008).

Similar tests with different microphysics schemes, and using different GCM’s would be valuable to confirm the conclusion that ACI sensitivity to cloud processes is large. We are in the process of developing such a cross model comparison. The overall conclusion is that getting a better representation of ACI is critical for reducing uncertainty in anthropogenic climate forcing: cloud microphysical development needs to go hand in hand with better constraints on aerosol emissions to properly constrain ACI and total forcing.

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