Impact of interannual variations in sources of insoluble aerosol species on orographic precipitation over California’s central Sierra Nevada

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Abstract

Aerosols that serve as cloud condensation nuclei (CCN) and ice nuclei (IN) have the potential to profoundly influence precipitation processes. Furthermore, changes in orographic precipitation have broad implications for reservoir storage and flood risks. As part of the CalWater field campaign (2009-2011), the variability and associated impacts of different aerosol sources on precipitation were investigated in the California Sierra Nevada using an aerosol time-of-flight mass spectrometer for precipitation chemistry, S-band profiling radar for precipitation classification, remote sensing measurements of cloud properties, and surface meteorological measurements. The composition of insoluble residues in precipitation samples collected at a surface site contained mostly local biomass burning and long-range transported dust and biological particles (2009), local sources of biomass burning and pollution (2010), and long-range transport (2011). Although differences in the sources of insoluble residues were observed from year-to-year, the most consistent source of dust and biological residues were associated with storms consisting of deep convective cloud systems with significant quantities of precipitation initiated in the ice phase. Further, biological residues were dominant (up to 40%) during storms with relatively warm cloud temperatures (up to -15°C), supporting the important role bioparticles can play as ice nucleating particles. On the other hand, lower percentages of residues from local biomass burning and pollution were observed over the three winter seasons (on average 31% and 9%, respectively). When precipitation quantities were relatively low, these insoluble residues most likely served as CCN, forming smaller more numerous cloud droplets at the base of shallow cloud systems, and resulting in less efficient riming processes. Ultimately, the goal is to use such observations to improve the mechanistic linkages between aerosol sources and precipitation processes to produce more accurate predictive weather forecast models and improve water resource management.
1. Introduction

Aerosol particles serve as nuclei upon which cloud droplets and ice crystals form and thus can have profound impacts on climate. In particular, pollution aerosols in high number concentrations have been suggested to slow down cloud drop coalescence and accretion by creating large populations of small-sized cloud droplets that delay the conversion of cloud water into precipitation (Borys et al., 2000; Rosenfeld et al., 2008). In contrast, aerosols that form ice nuclei (IN), such as mineral dust and biological aerosols, have been shown to enhance precipitation via secondary ice formation and aggregation (Bergeron, 1935; Hosler et al., 1957; DeMott et al., 2003; Morris et al., 2004; Tobo et al., 2013). Once formed, crystals can develop rime after colliding with supercooled cloud droplets (≥10 μm) (Yuter and Houze, 2003), particularly in more turbulent clouds (Pinsky et al., 1998). In regions with orographically-enhanced cloud formation such as California’s Sierra Nevada (Pandey et al., 1999), IN are theorized to become incorporated into the top of high-altitude clouds to form ice crystals (Meyers et al., 1992), whereas cloud condensation nuclei (CCN) have been hypothesized to enhance cloud droplet formation at the base of orographic clouds (Rosenfeld et al., 2008). Under subfreezing conditions, a precipitating ice cloud overlaying a pristine marine liquid water cloud enables growth of precipitation particles through riming via the seeder-feeder process (Choularton and Perry, 1986; Saleeby et al., 2009). However, if the lower cloud contains high concentrations of CCN, such as those from pollution (Rosenfeld, 2000), ice crystal riming efficiency is reduced, and snow growth rates and deposition location are altered (Saleeby et al., 2009). Although the effects of CCN on precipitation suppression in the Sierra Nevada are well-documented (Colle and Zeng, 2004; Givati and Rosenfeld, 2004; Rosenfeld and Givati, 2006), the combined effects of CCN and IN simultaneously on precipitation in mixed-phase clouds are not well established (Muhlbauer et al., 2010). It is plausible that these effects can
offset one another to some degree, and thus past measurement campaigns that addressed one or
the other could not account for the combined effects.

The Sierra Nevada region is influenced by numerous sources of CCN, including regional
transport from biomass burning, urban, agricultural, and industrial emissions from the Central
Valley (Collett et al., 1990; Guan et al., 2010) in addition to in situ formation of particles that act
as CCN from transported gas phase species (Lunden et al., 2006; Creamean et al., 2011) (see Figure
1). In contrast, IN populations have been shown to be influenced by dust transported over long
distances from arid regions in Africa and Asia (McKendry et al., 2007; Ault et al., 2011; Uno et
al., 2011; Creamean et al., 2013; Creamean et al., 2014b). Furthermore, biological species (e.g.,
bacteria) have been shown to be more effective IN (Despres et al., 2012; Murray et al., 2012;
O'Sullivan et al., 2014) since they activate at temperatures as warm as -1°C (Morris et al., 2004)
compared to dust (~ -38 to -17°C) (Field et al., 2006; Marcolli et al., 2007). Conen et al. (2011)
demonstrated even biological fragments such as proteins can largely determine ice nucleation
properties of soil dust in a laboratory setting, while Pratt et al. (2009) confirmed the importance of
biological IN in orographic cloud ice formation using in situ aircraft measurements.

Precipitation events in the Sierra Nevada are influenced largely by the combined effects of
transient synoptic-scale dynamics and terrain-locked orographic lift. Ralph et al. (2013a)
demonstrated that precipitation totals in land-falling atmospheric rivers (Ralph et al., 2004) depend
considerably on orographic lift associated with water vapor transport during storms that move
across the California Coastal Mountains. Their study showed that differences in storm-total water
vapor transport directed up the mountain slope contributed 74% of the variance in storm-total
rainfall across 91 storms from 2004-2010. One hypothesis is that the remaining 26% variance
results from influences by other processes, including aerosol impacts on precipitation, as well as
convection, synoptic and frontally forced precipitation and static stability. Aircraft and ground-based cloud seeding experiments in the Sierra Nevada suggest aerosols serving as IN are more frequently removed by forming ice crystals versus scavenging during snowfall, and increase precipitation rates by 0.1-1.0 mm/hr (Reynolds and Dennis, 1986; Deshler and Reynolds, 1990; Warburton et al., 1995). Frozen winter precipitation in the Sierra Nevada produces a deep snowpack which gradually feeds reservoirs in the spring (Dettinger et al., 2011). However, the presence of CCN may also influence the snowpack by creating smaller cloud droplets that are scavenged less efficiently by falling cloud ice crystals in the riming process, leading to reduced snowfall and thus significant implications for water resources (Borys et al., 2000; Saleeby et al., 2009). In short, the interplay between CCN and IN activity of aerosols and their impacts on precipitation in this region will influence the depth of the Sierra Nevada snowpack and, thus, the water resources available to California.

CalWater (http://www.esrl.noaa.gov/psd/calwater/overview/calwater1.html) was a field campaign designed to study aerosol-cloud-precipitation interactions in California during winter storms, as well as the dynamics of the inland penetration of atmospheric rivers from the coast. A unique combination of radar technology, ground-based aerosol measurements, and hydrometeorological sensors were stationed in the Sierra Nevada and nearby for up to 6 weeks during each of the three winter seasons from 2009-2011. This study focuses on identifying cloud seeds, interstitial aerosol, and scavenged aerosols in Sierra Nevada precipitation by examining individual particles as insoluble residues in precipitation samples collected at a ground-based site co-located with a precipitation radar and other meteorological sensors. Key elements of the unique hydrometeorological measurement network were obtained as part of the National Oceanic and Atmospheric Administration’s (NOAA) Hydrometeorology Testbed (Ralph et al.,
Precipitation composition studies regarding the insoluble components were employed for a number of CalWater events by Ault et al. (2011) and Creamean et al. (2013), providing valuable insight into the potential sources of aerosols acting as CCN and IN.

This study probes two unresolved questions from the previous 2009 and 2011 studies by Ault et al. (2011) and Creamean et al. (2013), respectively: 1) How do both local pollution (i.e., from Sierra Nevada and Central Valley) and long-range transported sources of the insoluble components of aerosols vary between winter seasons? 2) How do these sources impact precipitation processes? This study focuses on measurements from the 2010 winter season in addition to demonstrating the large interannual variability in sources of insoluble residues in the Sierra Nevada during all three winter field seasons, including both long-range transported and local emissions. Further, we evaluate how these sources impact precipitation formation through comparing the comprehensive set of cases and relating these to radar-observed precipitation characteristics. The links obtained here between sources of the insoluble components of aerosols and precipitation outcomes will ultimately be used as inputs into regional climate models to develop a longer-term mechanistic picture for how different aerosol sources influence clouds and precipitation processes in California.

2. Measurements

2.1. CalWater field campaign

The CalWater study centered at Sugar Pine Dam (SPD; 1064 m ASL; 39.13°N, 120.80°W; shown in Figure 1) involved a unique combination of meteorological (NOAA) and atmospheric measurements (University of California, San Diego; UCSD) to deconvolute how different factors affect precipitation quantity and type. Simultaneous atmospheric and meteorological measurements were made from 22 Feb – 11 Mar 2009, 27 Jan – 15 Mar 2010, and 28 Jan – 8 Mar
2011. Dates, times, and analysis statistics for each of the precipitation samples collected during
the storms from 2009-2011 at SPD are provided in Table 1. Multi-year measurements provide an
extensive dataset to determine the impact different aerosol sources have during winter storms in
California.

2.2. Surface meteorology and cloud properties above SPD

Hourly precipitation rates (mm h\(^{-1}\)) and 2-minute temperature (°C) at SPD were acquired
from NOAA’s Hydrometeorological Testbed Network (NOAA HMT-West). Storm-total
precipitation represents the total accumulated precipitation per storm throughout the CalWater
winter sampling season (provided in Table 1). NOAA’s S-band profiling radar (S-PROF, (White
et al., 2000)), a fixed dish antenna, was operated at 2875 MHz and directed vertically to study the
backscatter of energy from hydrometeors and cloud droplets and to monitor the radar brightband
melting layer (White et al., 2003). The S-PROF radar can distinguish between different
precipitation process types by detecting a “brightband”, where the phase of falling precipitation
changes from solid to liquid (White et al., 2002). The accumulation and percentages of
precipitation process type including non-brightband rain (NBB rain), brightband rain (BB rain),
and snow/graupel/hail (herein, simply referred to as “snow”) were estimated using the rainfall
process-partitioning algorithm developed by White et al. (2003; 2010), which was applied to the
S-PROF profiles. These measurements represent the types of precipitation aloft, not just at the
surface level. Both snow and BB rain were formed in the ice phase; however, BB rain reached the
surface by passing through a melting layer. NBB rain is precipitation that likely originated as liquid
droplets and is characterized by a larger number of small drops than BB rain (White et al., 2003;
Neiman et al., 2005; Martner et al., 2008). Echo top heights (km, MSL) were also estimated using
S-PROF radar data using methods employed by Neiman et al. (2005) and Martner et al. (2008)
and used to determine the depth of the clouds above SPD. Analysis was performed on all 30-minute periods when the precipitation rate exceeded \( \sim 1 \text{ mm hr}^{-1} \).

Data from the 11\textsuperscript{th} Geostationary Operational Environmental Satellite (GOES-11) were used to define effective cloud temperature, which is close to the cloud-top temperature, and the cloud top phase over SPD. GOES-11 was centered at 135°W over the eastern Pacific Ocean. Cloud properties from 22 Feb – 4 Mar 2009, 27 Jan – 13 Mar 2010, and 28 Jan – 8 Mar 2011 were retrieved for CalWater. The five channels on the GOES-11 imager include a visible channel (0.65 \( \mu \)m), which was calibrated to the Aqua MODIS 0.64-\( \mu \)m channel, as well as four infrared channels. The 4-km pixel GOES-11 data were analyzed each hour for a domain bounded by 30°N – 42.5°N latitude and 112.5°W – 130°W longitude using the methods described by Minnis et al. (2008; 2011). Data from all parallax-corrected pixels within a 10-km radius of the SPD were used to compute mean effective cloud temperature and percentage of cloud ice.

2.3. Analysis of insoluble precipitation residue particles and ambient aerosols

Methods for collection and analysis of insoluble precipitation residues are described elsewhere (Holecek et al., 2007; Ault et al., 2011; Creamean et al., 2013; Creamean et al., 2014a). Briefly, precipitation samples were manually collected using beakers cleaned with ultrapure Milli-Q water (18 MΩ/cm) and methanol. Most samples were analyzed immediately after collection, while others were transferred to 500-mL glass bottles, frozen, and stored for 6-10 days before chemical analysis. Insoluble residues in the precipitation samples were resuspended using a Collison atomizer, dried using two silica gel diffusion driers, and sampled by an aerosol time-of-flight mass spectrometer (ATOFMS) (Gard et al., 1997). This aerosolization method can produce single soluble and insoluble particles, agglomerates of different particle types, and coatings of soluble species on insoluble residues. Thus, the composition is likely somewhat altered from how
the particles would have existed in the atmosphere (Holecek et al., 2007). Even with the caveats associated with the aerosolization process as discussed in Creamean et al. (2013; 2014a), this method provides useful information on chemical differences in the aerosols seeding clouds.

Insoluble precipitation residues between 0.2-3.0 μm in diameter were individually sized and chemically analyzed by the ATOFMS. In this instrument, single particles traverse between and scatter the light from two continuous wave lasers (532 nm) at a set distance apart from which particle size is calculated based on particle velocity upon calibration using known size polystyrene latex spheres. A third pulsed Nd:YAG laser (266 nm) is then triggered and simultaneously desorbs and ionizes each sized particle, generating positive and negative ions which are analyzed using a dual-polarity time-of-flight mass spectrometer. The mass spectra from individual particles were classified into different types based on combinations of characteristic ion peaks as discussed in detail by Creamean et al. (2014a). Peak identifications correspond to the most probable ions for a given mass-to-charge (m/z) ratio based on previous ATOFMS precipitation studies (Holecek et al., 2007; Ault et al., 2011; Creamean et al., 2013; Creamean et al., 2014a).

Ambient aerosols were analyzed using ATOFMS simultaneous to precipitation sample collection time periods. The instrument operates in the same manner as with the insoluble residues, however, ambient air was drawn in the inlet instead of resuspended particles from atomized precipitation samples. Due to the sheer number of ambient aerosols analyzed by ATOFMS, particles were classified via a clustering algorithm as opposed to hand classification. Single-particle mass spectra were imported into YAADA (Allen, 2004), a software toolkit in Matlab (The Mathworks, Inc.), for detailed analysis of particle size and chemistry. ART-2a, an adaptive resonance theory-based clustering algorithm (Song et al., 1999), was then used to classify particles into separate groups depending on the presence and intensity of ion peaks within an individual
particle’s mass spectra. The most populated 50-70 clusters accounted for >90% of the total ART-2a classified particles and are considered representative of the overall aerosol composition. Peak identifications within this paper correspond to the most probable ions for a given mass-to-charge ratio.

3. Results and discussion

3.1. Interannual variability of precipitation residue composition measured by ATOFMS

The insoluble residue chemical composition during the three winter sampling seasons was mainly composed of dust, biological material, and organic carbon (OC). The OC residues were predominantly from biomass burning (Ault et al., 2011; Creamean et al., 2014a) with minor contributions from agricultural and pollution aerosols from the Central Valley (hereafter referred to simply as “pollution”) (McGregor and Anastasio, 2001; Gaston et al., 2013). Mass spectra for each of these types are shown by Creamean et al. (2013; 2014a) and Ault et al. (2011). Other types contributed to ≤8% of the total residues each year. Control experiments of specific mixtures and solutions—including dust, leaf litter, smoke, and sea salt—were conducted using ATOFMS to accurately identify residue types observed in precipitation samples. These are discussed in detail by Creamean et al. (2014a), in addition to the chemical speciation of the major residue types from precipitation samples. The ATOFMS is less sensitive to soluble species, such as sea salt, as they form residues that are too small to detect and chemically analyze when concentrations are low due to dilution that occurs in precipitation samples (Creamean et al., 2014a). Briefly, in ATOFMS analysis, dust particles typically contain a combination of different metal and metal oxides, including but not limited to aluminosilicates, iron, and titanium. Biological residues typically contain a combination of sodium, magnesium, potassium, calcium, organic nitrogen markers, and/or phosphate. In many cases, dust residues were mixed with biological material as indicated
by the combination of ion markers. The mixed nature of the dust with biological material is likely a result of soil dust (Conen et al., 2011) or other sources such as dust interacting with marine biomaterial during transport (Prather et al., 2013), and to a lesser extent agglomerates produced during the analysis resuspension process (Creamean et al., 2014a). Thus these mixed particles were grouped in the “dust” category. Biomass burning residues varied in composition, but typically contain sodium, potassium, aged organic carbon fragments, high mass organic carbon markers, and/or polycyclic aromatic hydrocarbon markers. Pollution residues contained aged organic carbon and/or amine markers, with a dearth of common biomass burning markers. Ault et al. (2011) illustrated the ubiquitous presence of local biomass burning in precipitation at SPD during the 2009 winter sampling and highlighted the potential importance of these aerosols as CCN (Holecék et al., 2007). In particular, biomass burning aerosols containing potassium and sodium have been shown to be hygroscopic in CCN measurements (Carrico et al., 2010; Engelhart et al., 2012). Ault et al. (2011) also suggested the source of the dust in 2009 was from high-altitude, long-range transport as opposed to local or regional sources. Further, Creamean et al. (2013) demonstrated that dust and biological aerosols during the 2011 measurements were long-range transported particles which became incorporated into the tops of high-altitude clouds. Dust from Asia has been shown to reach the U.S. west coast consistently throughout the late winter/early spring (Husar et al., 2001; VanCuren and Cahill, 2002; Liu et al., 2003; Jaffe et al., 2005; Creamean et al., 2014b).

Large variations existed between the major precipitation residue types during the three winter seasons (Table 1). The results from 2009 were presented in detail by Ault et al. (2011), and therefore will only be briefly discussed here. It is important to note that only two of the three 2009 storms (storms 1 and 3 here) were presented in Ault et al. (2011) due to their meteorological
similarities. As shown in Table 1 during storms 1 and 2, the residues were mainly composed of biomass burning (70% and 76% for samples 1 and 3, denoted as “S1” and “S3”, respectively), with some dust present (up to 38% in S2). However, during storm 3, the residue composition shifted to predominantly dust (46-80%, S6-S10). Even though meteorological conditions were relatively similar during the most intense storms (storms 1 and 3), the precipitation shifted to snow during storm 3 due to colder conditions later in that event. This storm produced 40% more precipitation than the first storm (Ault et al., 2011). During the 2010 winter sampling season, high percentages of biomass burning particles were present throughout the entire study (up to 61%, 38% on average) and constituted the dominate residue type during almost all of the storms. In contrast, in 2011 dust residues were dominant during the first storms (44-94%, storms 12-14), while biological percentages were highest during most of the latter storms (37-83%, storms 15-17). The results from 2011, presented in detail in Creamean et al. (2013), are only briefly discussed. Overall, each winter sampling season was impacted by very different aerosol sources, which we hypothesize impacted the type and quantity of precipitation as discussed in the following section.

Although we cannot determine with great certainty, we hypothesize that the residue types from each winter sampling season were most likely present due to nucleation in cloud with a smaller contribution from scavenging of ambient aerosols during rainfall/snowfall. Figure 2 shows the composition of the precipitation residues compared to the relative abundance of the ambient aerosols present during each sampling time period for 2010 (2009 and 2011 are shown and/or discussed in Ault et al. (2011) and Creamean et al. (2013), respectively). Dust, biomass burning, and pollution were present in both in the ambient aerosol as well as the residues. Sea salt was not observed in the precipitation due to its soluble nature, while biological particles were not observed as ambient aerosols likely due to the fact that the majority of these particles originated from soil
dust and were separated during the resuspension process (Creamean et al., 2013; Creamean et al., 2014a). For all three sampling seasons, the time periods with the highest relative amount of dust, biomass burning, or pollution residues in the precipitation samples did not correspond to highest relative amount of the same type of ambient aerosol (i.e., almost all of the Spearman’s correlation coefficients ($\rho$) were low or negative and did not demonstrate statistical significance as shown in Table 2). Herein, we employ the use of $\rho$ to show the monotonic relationships between the residue composition and ambient aerosol or cloud and precipitation properties, since the relationship between aerosols and precipitation is not a linear function of two variables and other factors play a role. The absence of correlation between similar types of ambient aerosol versus precipitation residue particles suggests the majority of the residues were from nucleation of cloud particles, with a possible smaller contribution from scavenging during precipitation particle descent.

3.2. Linking residue composition to precipitation type and quantity using ATOFMS and S-PROF

As observed by Ault et al. (2011), aerosols can produce up to 40% more precipitation during storms in the Sierra Nevada. Fan et al. (2014) showed the large impact that dust and biological aerosols can have on Sierra Nevada snowpack, where they simulated these aerosols increasing snowpack by 40%. Further, Martin et al. (2014) simulated storms during CalWater in 2011 and demonstrated how the storms with more dust and biological particles incorporated into upper cloud levels produced 23% (but as much as 67%) more precipitation than storms with a greater influence from regional pollution aerosols. Variations in meteorological forcing also play a role in the precipitation type and quantity (Martin et al., 2014), but the rather systematic correlations between different aerosol sources and precipitation processes previously shown and described herein suggest the aerosol sources can still play a vital role.
3.2.1. Dust and biological residues were dominant when precipitation formed as ice

Here, we demonstrate how the variability in the different sources of insoluble residues from aerosols influence both the type and quantity of precipitation during the CalWater storms in the Sierra Nevada. In most cases, the source of the ATOFMS residues were correlated with the precipitation process type as delineated by the meteorological (S-PROF radar) measurements. This is demonstrated by the 2010 samples in Figure 3 (2009 and 2011 are shown in Ault et al. (2011) and Creamean et al. (2013), respectively, but follow similar trends to the 2010 samples). Overall, BB rain or snow events (when surface temperatures dropped to ~0°C) were typically detected during time periods when precipitation samples contained higher percentages of dust plus biological residues (hereafter referred to as %Dust+Bio), particularly when Dust+Bio was >40% of the total residues. Throughout this discussion, the dust and biological residues are combined to simulate the percentage of residue types that likely served as IN and because they are likely from a similar source (Creamean et al., 2013). However they are shown separately in the figures to demonstrate the relative contribution of each, which is particularly important for the biological residues as discussed in more detail below. Sample time periods with the most biomass burning and pollution residues typically corresponded to the most NBB rain during 2010, suggesting precipitation was formed as liquid due to the lesser influence from Dust+Bio. For instance, storm 5 in 2010 corresponded to samples with some of the lowest percentages of Dust+Bio (down to 20%), and frequent detection of NBB rain (5 out of 13.5 h), particularly towards the end of the storm. BB rain was detected during the precipitation sampling at the end of this storm as well, possibly because Dust+Bio residues were still present and thus ice was still nucleated in the clouds above SPD. The sample from storm 7 (S18) also contained low %Dust+Bio (30%), and frequent detection of NBB rain (6.5 out of 15 h). Overall, these results show that dust and biological residues
were dominant during time periods when precipitation formed in the ice phase based on ATOFMS
and S-PROF measurements.

Although 2010 samples contained very different relative contributions of residue types
when compared to 2009 and 2011, the different residue types followed very similar relationships
with cloud ice amounts, precipitation type and quantity, and cloud depth. Figures 4 and 5 provide
a summary of observed meteorological conditions during each of the three winter sampling seasons
in addition to precipitation residue composition averaged per storm and properties of clouds above
SPD. Snow and BB rain are combined and denoted as “ice-induced precipitation,” i.e.,
precipitation that was initially formed as ice (Creamean et al., 2013). The echo top heights and
storm-total precipitation are shown as deviations from their averages during all of CalWater storms
to demonstrate the range of their variations: the echo top height average and storm-total
precipitation averages were 3.51 km and 55.46 mm, respectively, based on data from 43 days
during sample collection time periods provided in Table 1. Data from GOES-11 were removed if
the cloud effective temperature was within the homogeneous nucleation regime (≤ -36°C; during
storms 7 and 8) to enable the investigation of heterogeneous ice nucleation processes only. It is
important to note that correlations are not statistically significant due to the low number (17) of
events, however, they still provide a useful context to the trends between the residue composition
and cloud and precipitation properties. As shown in Figure 4, events with more ice-induced
precipitation and cloud ice typically correspond to samples with more dust and/or biological
residues (ρ = 0.58 and 0.67, respectively, for Dust+Bio). This relationship supports our hypothesis
that the majority of the residues were nucleated versus scavenged. If, for example, most of the
residues were scavenged, we would not expect such strong relationships of dust and biological
residues with the amount of cloud ice and ice-induced precipitation.
In particular, the storms with the highest Dust+Bio (storms 14 and 15; 93% and 95% respectively) correspond to some of the highest values of ice-induced precipitation (82% and 96%, respectively). Interestingly, these two storms had very different residue composition: storm 14 had more dust (81%) whereas storm 15 had more biological residues (83%). The effective cloud temperatures were -32°C and -25°C, respectively, suggesting that the dust IN were more effective at colder temperatures, while the biological IN were active at warmer temperatures. Other interesting cases are storms 4 and 10 from 2010, where biological residues composed 80% and 77% of the potential IN and ice-induced precipitation was 87% and 92%, respectively. Cloud temperatures were also relatively warm during these storms (-16°C and -15°C, respectively), further demonstrating that biological IN are active at warmer temperatures. In the cases where biological residues were dominant during storms 3, 10, and 15 and likely served as IN at warmer cloud temperatures, the cloud ice content was ≥50% based on GOES-11 measurements. It is important to note that the purely biological residues could be a result of the aerosolization process, thus might have originally been components of the dust particles. Although biological particles were not observed as ambient aerosol at the ground, they were observed as interstitial aerosol and in individual cloud particles during the 2011 in-cloud aircraft measurements (Creamean et al., 2013). However, when examining the 2011 measurements, the fact that: 1) a higher abundance of purely biological residues was observed in the precipitation samples compared to the interstitial aerosol or cloud particles and 2) a higher abundance of dust mixed with biological material was observed in the aircraft measurements compared to the precipitation collected on the ground, supports the fact that the majority of the biological residues are likely separated from the dust during the aerosolization process. Even considering this issue, the dust particles that were present in cloud still contained more biological material during time periods with warmer cloud
temperatures, thus would have enabled the dust to serve as more efficient IN as delineated by Conen et al. (2011) and O’Sullivan et al. (2014).

The percentages of dust and biological residues were also generally in phase with the echo top height deviation as shown in Figure 4 ($\rho = 0.39$): when the clouds were deeper, i.e., larger positive echo top height deviation (shallower, i.e., larger negative echo top height deviation), the %Dust+Bio was higher (lower) as was the relative amount of ice-induced precipitation. However, storm 10 was atypical; the %Ice-induced precipitation was high (92%), while %Dust+Bio was not as high (52%), which could be a result of the clouds being shallower. Based on these results, we suggest that when the clouds were sufficiently deep, they were more likely to have incorporated long-range transported dust and biological aerosols that were present only at higher altitudes (above ~3 km), such as in the cases documented by Ault et al. (2011) and Creamean et al. (2013), and the simulations of storms 13 and 14 by Martin et al. (2014). These dust and biological aerosols likely initiated ice formation and thus influenced the relative amount of ice-induced precipitation.

3.2.2. Shallow clouds associated with aerosols from local biomass burning and pollution produced less precipitation

In contrast, when clouds were more shallow: 1) dust and biological aerosols likely traveled over the cloud tops, and thus did not become incorporated, and/or 2) less dust and biological aerosols were transported into the region. Thus a larger influence from local aerosols in the form of biomass burning and pollution residues was observed, as shown in Figure 5. Local biomass burning residues composed most of the OC residues (78%) compared to pollution (22%), particularly in 2009 and 2010. On average, biomass burning (31%) and pollution residues (9%) did not constitute as many of the residues as Dust+Bio (55%). Figure 5 also shows the relationship between OC residues (biomass burning and pollution) and storm-total precipitation deviation.
Generally, events with a negative storm-total precipitation deviation corresponded to precipitation samples containing more OC residues ($\rho = -0.38$), i.e., the combined percentage of biomass burning and pollution residues, was out-of-phase with the storm-total precipitation deviation. For instance, the highest percentage of OC residue types (storm 2) had the largest negative storm-total precipitation deviation. Further, storms 13-15 in 2011 had some of the lowest percentages of OC residues and some of the largest positive storm-total precipitation deviations compared to the remaining 2011 storms. The OC residues from local biomass burning and pollution likely served as CCN and seeded the lower levels of orographic clouds, resulting in smaller cloud droplets that are less efficiently scavenged during the riming process (Borys et al., 2000; Rosenfeld and Givati, 2006; Saleeby et al., 2009).

Although CCN are typically thought to be soluble in nature, partially soluble or insoluble organic-containing aerosols have been shown to serve as CCN as well. For instance, CCN closure studies have found better agreement between predicted and observed CCN concentrations when insoluble organic particles were used in their simulations (Broekhuizen et al., 2006; Wang et al., 2008; Ervens et al., 2010). Further, previous studies have shown that relatively insoluble organic particles with small amounts of soluble inorganic material, such as sodium chloride, can drive the CCN activity of the organic particles (Broekhuizen et al., 2004; Ervens et al., 2010). Even partially or slightly soluble organics have been shown to serve as CCN (Bilde and Svenningsson, 2004), particularly if the particles were wet versus dry (Henning et al., 2005). For the 2009 samples, measurements of select soluble species were acquired and presented by Creamean et al. (2014a). Results presented there showed correlations between sodium, potassium, sulfate, chloride, nitrate, and phosphate and insoluble OC residues, thus signifying these insoluble OC residues were likely cores of the original particles from biomass burning and/or pollution. Therefore, the OC residues
observed in all the CalWater samples, although insoluble, could have potentially originated as cores with soluble species on the surfaces or partially soluble organic particles that remained intact while in solution, enabling them to serve as CCN and lead to the relationships with shallow clouds and negative precipitation deviation.

4. Broader implications

Overall, the results from this study demonstrate the interannual variability in the sources of aerosols seeding clouds over the Sierra Nevada as indicated by the insoluble residue composition. The combination of dust and biological residues, aerosols that likely served as IN, increased over time from 2009 to 2011, whereas the organic carbon residues (including local biomass burning and pollution residues) decreased over time. Further, the level at which the cloud nuclei impact cloud formation is important for resulting effects on precipitation processes: dust and biological residues likely serve as IN at higher altitudes in-cloud while organic carbon residues serve as CCN at cloud base. However, this study presents a limited number of data points and thus needs to be extended by future, additional measurements. It has been shown that dust and biological aerosols originate from long-range transport to the Sierra Nevada, whereas biomass burning and pollution residues are more likely from local sources (Rosenfeld and Givati, 2006; Ault et al., 2011; Creamean et al., 2013). Dust and biological residues were ubiquitous in the most of the samples, which induced the formation of ice precipitation, particularly corresponding to time periods where the samples contained a relatively high amount of biological residues. This suggests the residues containing biological material served as more efficient IN than dust. The two storms with the highest percentages of either dust (storm 14) or biological (storm 15) residues demonstrate this effect, where storm 15 produced more ice-induced precipitation and had higher cloud temperatures, whereas much lower cloud temperatures were observed during storm 14.
Sample 35 (S35 from storm 14) contained mainly mineral dust with little-to-no biological material as shown from IN measurements and heat treatment of the sample by Creamean et al. (2014a). Creamean et al. (2014a) also conducted the same measurements on the sample from storm 15 (S38), which contained IN active at high temperatures. Thus, the comparison of the samples from storms 14 and 15 enables us to determine the IN efficiency of dust versus biological material, both from previous laboratory measurements and in situ observations. Storms 4 and 10 contained more biological residues and produced substantial amounts of precipitation formed as ice under high cloud temperatures, further corroborating the fact that biological aerosols are more effective IN.

The source of the insoluble residues not only influenced whether precipitation formed in the ice or liquid phase, but also likely affected the quantity of precipitation that fell at SPD. Larger quantities of precipitation in comparison to the average from all three sampling seasons were observed during time periods where dust and biological residues were predominant in the samples. The most plausible explanation for this, as described previously, is that these residues likely served as IN which led to efficient riming processes and enhanced precipitation formation (Ault et al., 2011; Creamean et al., 2013; Creamean et al., 2014a). In contrast, OC residues from both biomass burning and to some extent pollution were observed during time periods with less precipitation. One possibility is that the local biomass burning and pollution residues served as CCN, which enhanced cloud droplet formation after being incorporated into lower levels of the orographic clouds and led to less precipitation (Weaver et al., 2002; Rosenfeld and Givati, 2006; Rosenfeld et al., 2008; Saleeby et al., 2009). A modeling study of aircraft measurements from 2011 presented by Martin et al. (2014) shows the presence of organic carbon residues at lower cloud levels during prefrontal storm conditions in the Sierra Nevada, demonstrating the significance of our observations and how they validate model results. The cloud droplets formed from biomass
burning and pollution likely decreased the riming efficiency of the ice crystals formed at higher altitudes in the presence of dust and biological aerosols, subsequently contributing to time periods with less ice-induced precipitation. With fewer aerosol seeds, cloud droplets and ice crystals form much less frequently under typical atmospheric conditions in the lower troposphere over the Sierra Nevada, altering the quantity of precipitation. Previous studies have shown that aerosols can have a significant impact on precipitation quantity and type in the Sierra Nevada during strong winter storms (Ault et al., 2011; Creamean et al., 2013; Fan et al., 2014; Martin et al., 2014). Based on this, the results presented here are in alignment with previous research.

Fan et al. (2014) and Martin et al. (2014) demonstrate the reproducibility of the observations in the Weather Research and Forecasting (WRF) model by focusing on particular case studies from CalWater 2011. Observations presented herein for all CalWater storms will be incorporated into future modeling work to improve simulations. However, future work is needed to better isolate the impacts of storm dynamics, aerosol microphysics, and precipitation, particularly when incorporating observations into regional climate models. Ultimately, the goal is to develop a mechanistic understanding of how, when, and where different aerosol sources influence cloud microphysics and the resulting precipitation in the Sierra Nevada. Improvement of these models can be used as predictive tools for future weather forecasts.

5. Conclusions

Observed variations of sources of the insoluble residues from aerosols serving as CCN and IN in Sierra Nevada precipitation were documented during three winter sampling seasons as part of the CalWater field program. These variations were then compared with meteorological observations of precipitation characteristics aloft during the same events. Insoluble residues in precipitation samples were used to link aerosol sources with trends in precipitation characteristics.
The unique multi-year, multi-event, and co-located aerosol and meteorological observations enabled the development of the following main conclusions:

- Differences in aerosol sources seeding the clouds based on the composition of insoluble residues were observed from year to year and between storms. We present cases with predominantly long-range transported dust and biological residues (2011), local biomass burning and pollution residues (2010), or a combination of these sources (2009).

- Although the residues in the 2010 samples were vastly different (i.e., influence more by biomass burning), the relationships between the dust and biological residues and cloud ice, precipitation type and quantity, and cloud depth were consistent with 2009 and 2011 samples.

- Dust and biological residues serve as IN, becoming incorporated into deeper cloud systems at cloud top and subsequently influencing the formation of ice-induced precipitation at SPD. This effect was documented in the CalWater 2011 modeling study by Fan et al. (2014).

- Our observations support the hypothesis that biomass burning and pollution residues likely served as CCN in shallower orographic clouds, which coincided with periods of less precipitation as simulated by Martin et al. (2014) during two CalWater 2011 storms in the Sierra Nevada.

- When dust/biological residues and pollution/biomass burning residues were both present, orographic clouds also were typically shallow and coincided with periods of less precipitation. This aligns with the hypothesis that IN and high concentrations of CCN at different altitudes in the same cloud system inhibit precipitation formation (Saleeby et al., 2009).
Results presented herein represent a noteworthy advancement in understanding the effects of sources of insoluble aerosol species on the type and quantity of precipitation in the California Sierra Nevada, by building on previous case studies presented by Ault et al. (2011) and Creamean et al. (2013). Aerosol impacts on clouds and precipitation derived from insoluble residue links with cloud and precipitation properties have important implications for the Sierra Nevada, by serving as one of the key factors that influence water supply in the region. The relationships between insoluble precipitation residues and their potential climate impacts could translate to a global scale, i.e., apply to other orographic regions where such insoluble particles are found in and impact the formation of clouds and precipitation. Thus, understanding insoluble residue sources has implications on a global level, particularly when modeling their impacts on clouds. However, additional studies are needed to better quantify these relationships, which served as a major motivation for the more recent CalWater 2 field campaign which started in early 2015. The findings presented here from CalWater served as the foundation for the flight planning and execution of field measurements during CalWater 2, demonstrating the importance of our results for not only constraining future modeling work but also serving as a driver to continue similar measurements to develop a longer-term record. Results from both studies will enable improvements in models to better assess how weather patterns and/or regional climate may change due to the effects from different aerosol sources, particularly those from long-range transport which have a major impact on the seeder-feeder mechanism long observed over the Sierra range. Improving our ability to model the interactions between aerosols, clouds, and precipitation can contribute to better winter storm preparedness, water resource management, and flood mitigation.

**Author Contribution**
J. C. collected and analyzed ATOFMS data from precipitation samples in 2009, 2010, and 2011, interpreted all data, and prepared the manuscript with contributions from all co-authors. A. A. collected and analyzed ATOFMS data from precipitation samples in 2009. A. W., P. N., and F. R. collected and analyzed S-PROF data and surface meteorology measurements at SPD. P. M. analyzed GOES-11 data. F. R. was additionally involved with experimental design. K. P. was the principal investigator of this study, involved with experimental design, and preparation and editing of this manuscript. All authors reviewed and commented on the paper.

Acknowledgements

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Figure 1. Map showing potential aerosol sources and the topography in the region surrounding Sugar Pine Dam (SPD), where precipitation sample collection and meteorological measurements occurred during CalWater.
Figure 2. Comparison of average ambient aerosol versus precipitation residue composition per sampling time period during CalWater 2010. Percentages represent either the number of each type of aerosol or residue per total number of aerosols or residues analyzed per sample. Sea salt was not observed in precipitation and biological particles were not observed in the ambient data.
Figure 3. Precipitation process type (30-min), residue type (per sample), precipitation accumulation (1-hr), and surface temperature (2-min) during all storms from 2010. Time periods without precipitation process measurements correspond to no falling precipitation or missing S-PROF data. Each precipitation sample bar of the residue types represents one sample and the width of the bar reflects the sample collection time period. Sample IDs are provided above each sample bar and correspond to those in Table 1. Note that the sample length is only shown during rain or snow, thus may not directly correspond to times provided in Table 1. The horizontal black dashed line represents the 40% mark for ATOFMS.
Figure 4. Summary of IN precipitation residue composition, observed surface meteorology at SPD, and cloud properties above SPD. a) The percentages of dust and biological residues and the % ice-induced precipitation (snow plus BB rain). b) Echo top height deviation (km) calculated from all storms during CalWater (average = 3.51 km based on data from 43 days during ATOFMS sample collection time periods provided in Table 1). Positive (negative) deviations correspond to higher (lower) than average echo top heights. Effective cloud temperature and percentage of cloud ice are also shown. Data were removed if in the homogeneous nucleation regime (≤ -36°C). The respective instruments in which each measurement was acquired is provided in the axis labels.
Figure 5. Summary of organic carbon precipitation residue composition and storm total precipitation deviation. Organic carbon residues are separated into those from biomass burning and those from local pollution. Storm-total precipitation deviation (mm) is calculated from all storms during CalWater (average = 55.46 mm based on data from 43 days during ATOFMS sample collection time periods provided in Table 1). Positive (negative) deviations correspond to higher (lower) than average echo top heights.
Table 1. Statistics for precipitation sample collection during storms from 2009-2011 at SPD. The start and end dates reflect when the beakers were placed outside; they do not always correspond to the exact start and end of falling precipitation. The percentages of each insoluble residue type per sample are provided (bolded and colored percentages show dominant type).

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<th>Start (UTC)</th>
<th>End (UTC)</th>
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<td>13%</td>
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Table 2. Spearman’s correlation coefficients P-values, and statistical significance of relationships between similar particle types that were found in the precipitation samples and ambient aerosols during the same time period of precipitation collection.

<table>
<thead>
<tr>
<th>Year</th>
<th>Spearman</th>
<th>P</th>
<th>Significant</th>
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<td>Dust</td>
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<tr>
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