The observed influence of local anthropogenic pollution on northern Alaskan cloud properties

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Abstract.

Due to their importance for the radiation budget, liquid-containing clouds are a key component of the Arctic climate system. Depending on season, they can cool or warm the near-surface air. The radiative properties of these clouds depend strongly on cloud drop sizes, which are governed by the availability of cloud condensation nuclei. Here, we investigate how cloud drop sizes are modified in the presence of local emissions from industrial facilities at the North Slope of Alaska. For this, we use aircraft in-situ observations of clouds and aerosols from the 5th Department of Energy Atmospheric Radiation Measurement (DOE ARM) Program’s Airborne Carbon Measurements (ACME-V) campaign obtained in Summer 2015. Comparison of observations from an area with petroleum extraction facilities (Oliktok Point) with data from a reference area relatively free of anthropogenic sources (Utqiagvik/Barrow) represents an opportunity to quantify the impact of local industrial emissions on cloud properties. In the presence of local industrial emissions, the mean effective radii of cloud droplets are reduced from 12.2 to 9.8 µm, which leads to a suppression of drizzle production and precipitation. At the same time, concentrations of refractory black carbon and condensation nuclei are enhanced below the clouds. These results demonstrate that the effects of anthropogenic pollution on local climate need to be considered when planning Arctic industrial infrastructure in a warming environment.

1 Introduction

Liquid-containing clouds are a significant modulator of the Arctic climate system’s radiation budget. Their properties impact both shortwave and longwave radiative transfer, resulting in seasonally-dependent influences that include both net cooling and warming of the Arctic surface (Intrieri et al., 2002; Shupe and Intrieri, 2004), and various forms of cloud feedbacks (Colman, 2003). At the same time, liquid cloud droplet number concentration and size are influenced by the number of available cloud condensation nuclei. It has been proposed that this has an effect on cloud albedo, life cycle and longwave emissivity (Twomey, 1976; Albrecht, 1989; Garrett and Zhao, 2006). Long range transport of aerosol particles from lower latitudes in winter and
early spring (Arctic haze) and episodic forest fires in summer can lead to higher aerosol concentrations (Shaw, 1995; Law and Stohl, 2007), which have been found to modify liquid and mixed phase cloud properties (Garrett et al., 2004; McFarquhar et al., 2011; Jackson et al., 2012; Zamora et al., 2016). Besides these transported emissions, the Arctic is an environment that is generally relatively clean (Quinn et al., 2002, 2009) which makes clouds particularly susceptible to an increase in aerosol concentration (Platnick and Twomey, 1994). In comparison to other regions, there are only few sources of local anthropogenic emissions north of the Arctic Circle, which are mainly related to ship traffic and petroleum as well as natural gas extraction facilities (Law and Stohl, 2007). While emissions from ships are expected to rise due to the retreating sea ice, emissions from resource extraction are expected to remain at present day levels (Peters et al., 2011) with an estimated 13% of the world’s untapped oil resources located in the Arctic (Gautier et al., 2009). Local emissions by Arctic petroleum and natural gas extraction facilities have been observed and quantified by aircraft campaigns (Brock et al., 2011; Roiger et al., 2015). These emissions are mostly associated with flaring, but also by regular internal combustion engines. Ødemark et al. (2012) found that black carbon (BC), which is particularly created by flaring (Stohl et al., 2013), results in a modeled positive net radiative forcing of petroleum and natural gas extraction, mainly due to deposition of BC on the snow. Kolesar et al. (2017) showed that emissions from the Prudhoe Bay area result in in-situ particle growth events in Barrow, located around 300 km west of the Prudhoe Bay region. Although these previous studies have demonstrated the potential impact from industrial activities in the Arctic, in-situ aerosol and cloud observations have not been combined in order to study local sources of emissions.

In this work, we show how cloud properties are altered by aerosol particles originating from local anthropogenic pollution from industrial activities in the Prudhoe Bay area in northern Alaska (Fig. 1), and investigate the influence on processes impacting the cloud life cycle. Even though the work is limited to observations from the North Slope of Alaska, the results are broadly applicable to other Arctic regions with significant industrial activities (e.g. Siberia), although exact details of the types of aerosol effects will be influenced by aerosol concentration, size, and composition. Because of their importance in regulating the surface and top-of-atmosphere energy budgets, we focus here on liquid clouds. Increased cloud droplet concentrations in the Prudhoe Bay, Alaska area were previously reported by Hobbs and Rangno (1998) although that study could not directly connect these increased concentrations to locally produced aerosol particles due to a lack of aerosol measurements. In this study, we fill this gap by using airborne cloud property and aerosol observations obtained during the US Department of Energy Atmospheric Radiation Measurement (DOE ARM) program’s 5th ARM Airborne Carbon Measurements (ACME-V) campaign to study the influence of local pollution on Arctic liquid clouds. An enhanced understanding of the influence is crucial to evaluate the role of clouds and aerosols in changing Arctic which is warming faster than other regions (Jeffries et al., 2013).

In Section 2 we provide background information on the ACME-V campaigns along with details on the various data sets used to conduct our analysis. Following this, we analyze observed aerosol particle (Section 3) and cloud (Section 4) properties, before combining these to evaluate the interactions between locally-produced aerosols and clouds in Section 5. This evaluation is carried further in Section 6 where we attempt to quantify observed aerosol-cloud interactions. Finally, we provide a summary and concluding remarks in Section 7.
Figure 1. Overview of all flights of the ACME-V campaign. Color shows altitude m MSL. The dashed circles indicate 90 km radii around the sites (black crosses), the red dots indicate active oil wells (Data obtained from http://doa.alaska.gov/ogc/publicdb.html in March 2017)

2 Data set

The ACME-V aircraft campaign took place from June to September 2015 (Biraud et al., 2016) and consisted of 38 research flights of the ARM Gulfstream G-159 (G-1) aircraft of the ARM aerial facility (Schmid et al., 2014, 2016). Since the campaign targeted trace gas measurements from local and regional sources, a majority of the flight time was spent below 200 m above mean sea level (MSL). However, spirals up to an altitude of 6,000 m were flown in the vicinity of two ARM surface observatories in northern Alaska, Utqiagvik (formerly known as Barrow or North Slope of Alaska, NSA, 71.323°N, 156.616°W) and Oliktok Point (OLI, 70.495°N, 149.886°W). Additional spirals were flown at Toolik (68.628°N, 149.598°W), Ivotuk (68.483°N, 155.754°W), and Atqasuk (70.467°N, 157.436°W) in order to characterize cloud and aerosol properties (Fig. 1). In this work, we compare data within 90 km of OLI and NSA. These two sites form an ideal opportunity to study the effects of local emissions on cloud properties: While OLI is surrounded by industrial activities related to oil and natural gas extraction (with the majority closer than 90 km), no substantial local sources exist in the vicinity of NSA and previous studies have shown only limited advection ($8 \pm 2\%$) of air masses passing through the Prudhoe Bay area to NSA (Kolesar et al., 2017). Despite substantial differences in aerosol properties, the two coastal sites lie only 250 km apart, resulting in very similar synoptic scale forcing, as can be seen from the high correlation between both sites for pressure, temperature, humidity, and wind (Fig. 2).

For both sites, north-easterly to easterly winds prevailed during ACME-V (Holdridge and Kyrouac, 1993). Additionally, we grouped observations closer than 90 km to the two more continental sites Toolik and Atqasuk into a third data set (labeled TOI).

Cloud properties were observed using a combination of forward scattering, optical array, and bulk probes. The particle size distributions were measured using the forward scattering Cloud Droplet Probe (CDP) manufactured by Droplet Measurement Technologies (DMT), Inc.; the Fast Cloud Droplet Probe (FCDP) from Stratton Park Engineering Company (SPEC), Inc.; Two

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Figure 2. Comparison of daily mean values for a) surface pressure, b) 2 m temperature, c) 2 m humidity, and d) 10 m wind speed.

Dimensional Stereo optical array probes (OAPs) (2DS, Lawson et al., 2006); and the High Volume Precipitation Spectrometer (HVPS, Lawson et al., 1993) from SPEC, Inc. The raw OAP datasets were processed by the University of Illinois Optical Array Probe Processing Software (Wu and McFarquhar, 2016). In order to merge the cloud droplet size distributions, FCDP data were used for particles less than 50 $\mu$m size, the 2DS was used for particles between 50 and 605 $\mu$m, and the HVPS was used for all particles exceeding 605 $\mu$m. In this study, particle diameter is used in general to describe the size of cloud and aerosol particles. Liquid clouds were required to have at least $10^7$ m$^{-3}$ droplets (Lance et al., 2011). In order to remove ice clouds from the data set, the Holroyd habit classification was applied to 2DS and HVPS observations with 1 s temporal resolution, which classifies particles mainly based on a fine detail ratio $F = pd/a$, where $p$ is perimeter, $d$ is diameter and $a$ is area (Holroyd, 1987). The habit classification scheme differentiates between spherical particles, tiny particles which are too small to be classified and various forms of ice crystals. Spherical particles were assumed to be liquid. Tiny particles appear only at the lower end of the 2DS ($< 105$ $\mu$m) and HVPS ($< 1125$ $\mu$m) size range. They were classified as ice only if other size ranges were not dominated by spherical particles. Otherwise, tiny particles were assumed to be liquid. Data points with more than $100$ m$^{-3}$ particles larger than 400 $\mu$m (Lance et al., 2011) classified as ice were removed from the data set. This ensures that observations of spherical ice particles falsely classified as liquid, which likely occur together with larger, more
complex shaped ice particles, were removed from the data set as well. Liquid water content (LWC) was obtained by integrating the merged droplet size distribution (DSD), because direct observations of LWC from the King probe (King et al., 1978) are affected by a decreasing sampling efficiency for (drizzle) drops greater than 30 \(\mu\)m diameter. Clouds that were observed for less than ten continuous seconds were discarded, while gaps of up to 5 s were permitted once in cloud. Considering the typical true airspeed of the G-1 of 95 m/s, this corresponds to 950 m and 475 m when flying in a straight line, respectively. Additionally, only vertically sampled clouds (i.e. the aircraft was constantly ascending or descending) with a sampled vertical extent of at least 20 m were included in this evaluation to allow for comparison of in-cloud microphysical observations with below-cloud aerosol properties in sections 5 and 6. Therefore, very thin and/or small clouds might be discarded inadvertently. To make the detection of cloud boundaries more robust, the cloud probe data were smoothed using a 10s running average. Except for the detection of the cloud boundaries, effects of the smoothing are negligible for the presented analysis. For liquid clouds, aerosols of the size range 100 nm to 3 \(\mu\)m were observed with the Passive Cavity Aerosol Spectrometer (PCASP model 100X, DMT Inc.). We expect particles measured by the PCASP to be mostly dry, because it was operated with an anti-ice heater. Kassianov et al. (2015) showed for the very same aircraft that this assumption leads to good agreement between calculated (using, among others, PCASP observations) and measured scattering properties. Unfortunately, another aerosol sampler (Ultra-High Sensitivity Aerosol Sizer, UHSAS) which is able to detect aerosols below the PCASP detection threshold of 100 nm did not operate during the majority of the ACME-V flights. Two Condensation Particle Counters (CPC, TSI, Inc. models 3025 and 3010, respectively) were used to observe total number concentrations of condensation nuclei (CN) for the size ranges 3 nm - 3 \(\mu\)m and 10 nm - 3 \(\mu\)m, respectively. Unless otherwise stated, only the CPC 3025 featuring a size range of 3 nm - 3 \(\mu\)m was used in this evaluation. Black carbon (BC), which results from incomplete combustion of biomass and fossil fuels (Schwarz et al., 2008; Bond et al., 2013; Lack et al., 2014), was measured with the Single Particle Soot Photometer (SP2, from DMT Inc.), via incandescence. Thus, only refractory black carbon (rBC) is observed by the instrument. A counter for cloud condensation nuclei (CCN) was not deployed during ACME-V.

While Arctic Haze was not observed during ACME-V, transported emissions from forest fires can contribute significantly to summertime aerosol loading in the Arctic (Law and Stohl, 2007). Therefore, we manually inspected the vertical profiles of rBC and carbon monoxide (CO), which together constitute a good tracer for biomass burning (Warneke et al., 2009, 2010). Typically, these layers are found aloft (Roiger et al., 2015), allowing us to use vertical profiles to aid in their identification. The manual inspection was supported by aerosol dispersion simulations executed using version 4 of the Hybrid Single Particle Lagrangian Integrated Trajectory (HYSPLIT) model (Stein et al., 2015). These simulations were forced using 1° data from the NOAA/NCEP Global Data Assimilation System (GDAS) (Kalnay et al., 1996). Five locations were included as sources ((1) 62.096°N, 163.632°W, (2) 63.843°N, 159.046°W, (3) 65.294°N, 154.386°W, (4) 66.631°N, 149.023°W, and (5) 67.631°N, 144.087°W) and toggled on or off on a daily basis in correspondence to thermal anomaly observations from the Moderate Resolution Imaging Spectroradiometer (MODIS) on the Aqua and Terra satellites obtained using brightness temperature measurements in the 4 and 11 \(\mu\)m channels. (Giglio et al., 2003; Giglio, 2013). From each fire location, particle mass concentrations were simulated for 72 h at 100-m intervals from 0 to 5,000 m above ground level (m AGL). Both dry and wet deposition were considered for particles using the default HYSPLIT parametrizations (particle density 6 g cm\(^{-3}\), shape factor 1.0). The particle
diameter of 0.2 µm used for the simulations is based on previous observations from fossil fuel and biomass burning sources (Brock et al., 2011; Eck et al., 1999; Rissler et al., 2006; Sakamoto et al., 2015). A dry deposition velocity of 1 x 10^{-4} m s^{-1} was assumed according to Warneck (1999) while 4 x 10^4 L L^{-1} and 5 x 10^{-6} s^{-1} were used to account for in-cloud and below-cloud scavenging, respectively. Radioactive decay and pollutant resuspension were not considered. Note that the data impacted by forest fires were only removed for spirals above OLI, NSA, and TOI. For clear-air observations during level flight legs between sites, it is generally impossible to distinguish aerosols originating from forest fires as compared to other sources. Therefore, data potentially impacted by forest fires have only been removed from the cloud observations which have been associated with vertical profiles (Sec. 4f), but not from the aerosol observations presented in Sec. 3.

### 3 Aerosol properties

The spatial distributions of aerosol observations below 500 m MSL are presented for the CPC, the SP2, and the PCASP in Fig.3. The data presented are limited to observations obtained below 500 m in order to demonstrate the impact of local emissions and reduce the impact of forest fires. Furthermore, data flagged as sampled in cloud using the thresholds described in the previous section have been discarded in the analysis of aerosol properties due to concerns of contamination of the statistics by cloud droplets. For the SP2, a clear local maximum of rBC concentration is visible east of OLI within the 90 km radius where most petroleum and gas extraction facilities are located (Fig. 1). A comparison of the distributions measured within a 90 km radius around the facilities reveals that the median of rBC concentration is the same for both regions (4 ng kg^{-1}). The tail of the distributions towards larger concentrations, however, is much greater at OLI (90th and 99.9th percentile 15 ng kg^{-1} and 42 ng kg^{-1}, respectively) than at NSA (17 ng kg^{-1} and 198 ng kg^{-1}, respectively). This enhancement is most likely connected to local emissions. CN measurements from the CPC show a similar pattern even though the increased values are distributed over a larger area. For both instruments, the distributions within the 90 km circle belonging to each site are skewed towards higher concentrations. Further, the difference between both CPC instruments, which depends on the concentration of CN between 3 and 10 nm diameter, is enhanced east of OLI (not shown). Freshly emitted soot has been shown to be larger than this (> 20 nm), so this range is likely due to in situ nucleation of aerosol particles from gas phase precursors (i.e., formation of new particles as compared to secondary aerosol formation, where gases condense onto preexisting aerosol, Kulmala et al., 2012). Nucleated aerosols typically have sizes below 3 nm, but quickly grow via condensation and coagulation to sizes > 3 nm (Colbeck and Lazaridis, 2014). This source of nucleated aerosol particles from petroleum and gas extraction activities has been reported by Kolesar et al. (2017) for emissions transported from OLI to NSA. Unfortunately, we cannot analyze this aerosol nucleation process in more depth given limitations with the instrumentation operated during ACME-V. rBC concentrations appear to be similar to values found by Zamora et al. (2016) (1-16 ng/m3) and Roiger et al. (2015) (median 20-30 ng/kg) for summertime transported forest fire plumes in the Arctic. Other studies (Warneke et al., 2009; Schwarz et al., 2010) found up to one order of magnitude higher rBC concentrations in the Arctic which is more similar to the maximum values we observed around OLI.
contrast to traces from forest fires, carbon monoxide (CO) concentrations were not found to be significantly enhanced in the OLI region (not shown).

The PCASP, which detects only particles larger than 100 nm, shows no spatial trends in the vicinity of the two sites. The comparison of the distributions around the facilities shows that the number of aerosols observed by the PCASP is on average actually slightly larger for NSA than for OLI. This is related to the fact that the median of the distribution is at 97 cm$^{-3}$ at NSA and at 76 cm$^{-3}$ at OLI. Similar to rBC, the tail of the distributions towards larger concentrations is greater at OLI (90th percentile 200 cm$^{-3}$) than at NSA (184 cm$^{-3}$). While it is challenging to clarify the precise cause of the increased mean concentration in detail, we speculate it might be related to transported emissions, including those from forest fires, that have not been properly removed from the data set. An alternative explanation could be the fact that collision-coalescence and precipitation rates are larger at NSA than at OLI (see next Sec. 4.) resulting in more aerosol processing by precipitation (e.g. Hoppel et al., 1990). Cloud-based processing leads to a reduction in aerosol concentration and an increase in aerosol size through conglomeration of cloud droplets (and corresponding aerosol particles) in the drizzle formation stage and subsequent evaporation. Such effects could increase the number of aerosols within the PCASP measurement size range.

To investigate why there is a clear enhancement in the CPC concentrations but not in the PCASP concentrations, both instruments sets are compared for all non-cloud observations below 500 m MSL during ACME-V. Fig. 4 shows CN concentration versus PCASP mean particle size and PCASP number concentrations for the two 90 km radii around OLI and NSA. It clearly illustrates that elevated CPC concentrations (> 600 cm$^{-3}$) found at NSA are typically connected with increased PCASP concentrations (> 100 cm$^{-3}$). While mean particle size generally increases with decreasing CN concentration, the variability of PCASP mean size is rather low for CN concentrations > 600 cm$^{-3}$, which is consistent with the idea that particles have already experienced growth. For OLI, on the other hand, variability in PCASP mean size is much larger. In addition, increased CPC CN concentrations do not necessarily correlate to increased PCASP particle concentrations at OLI. This means that for enhanced CN concentrations, which likely indicate a local source, PCASP particle concentrations can be small, but mean particle size is still significantly larger than 100 nm. The lower concentration and the higher size variability observed at OLI is consistent with the idea that particles in this region are younger. The value of 100 nm is important because it is assumed to be the size threshold between particles in Aitken mode and accumulation mode and typically only the latter can act as a CCN. (We do, however, note that smaller particles can act as CCN in very clean conditions (Leaitch et al., 2016).) In summary, aerosol particle sizes are more diverse at OLI and elevated CN concentrations are accompanied by enhanced PCASP mean particle sizes even though total PCASP particle concentrations are lower at OLI. Additional evidence for significant differences in aerosol concentrations between the sites is found in the fact that particle concentration of 600 cm$^{-3}$ is exceeded in 62% of all cases at OLI, but only 35% at NSA.

4 Cloud properties

Here, cloud properties are compared for flights occuring near NSA and near OLI. In order to evaluate a sufficiently large sample, all observations obtained closer than 90 km to NSA, OLI and the two sites comprising TOI are assigned to the corresponding
Figure 3. Left column: Spatial distribution of SP2 refractory black carbon concentration (a), CPC3025 CN concentrations (b), and PCASP aerosol concentration (c). Only non-cloudy observations below 400 m MSL have been considered. The size of the dots is proportional to the number of observations. The dashed circles correspond to a distance of 90 km. Right column: Here, the distribution of measurements within the 90 km circles are shown. The horizontal bar denotes the median value.

site (Fig. 1). As mentioned above, clouds impacted by forest fires have been removed. This limits the number of observations to 1608 1 s data points for OLI, 942 for NSA, and 579 for TOI.

When comparing 2D histograms of liquid effective radius and liquid water content for OLI and NSA (Fig. 5, a, b), a shift towards smaller $r_{\text{eff}}$ can be clearly seen in the measurements obtained in close proximity to OLI. This supports our hypothesis that CCN concentrations are elevated in the OLI region, since the first aerosol-cloud indirect effect proposes that droplet size is reduced when more CCN are available (all else equal). While droplet $r_{\text{eff}}$ observed at NSA cover the full range from droplet nucleation to drizzle (3 to 25 µm, mean 12.2±6.9 µm), $r_{\text{eff}}$ values are typically smaller than 16 µm at OLI (mean 9.8±4.0 µm).
and observations of drizzle-sized droplets are rare. The value of 16 µm is of special interest because it was proposed by Gerber (1996) as a minimal effective radius required to initiate collision-coalescence. For comparison, data obtained in a 90 km radius around Toolik and Ivotuk (TOI) (Fig. 5, c), reveal that the distribution of observed $r_{\text{eff}}$ at the coastal site in OLI is more similar to the inland sites comprising TOI (mean 7.2±3.1 µm) than to the second coastal site NSA.

The Albrecht effect proposes that more polluted clouds have longer cloud lifetime due to less efficient collision-coalescence (Albrecht, 1989). It is not possible to study the cloud life cycle using aircraft in-situ observations, but the potential for impact on cloud life cycle can be estimated by calculating the collection growth rate $C$ (Long, 1974). $C$ describes the mass of drops collected by a unit mass in a unit volume $M$ per time interval $t$. It is the key process for converting cloud drops into precipitation and is estimated by integrating the mass collected by particles with diameter $D_1$ and mass $m_1$ over all size bins:

$$C = \frac{dM}{dt} = \int_{D_{\text{min}}}^{D_{\text{max}}} \frac{dm_1}{dt} N(D_1) dD_1$$

where $N(D_1)$ is the particle size distribution and $D_{\text{min}}$ and $D_{\text{max}}$ are the bounding drop diameters as determined by the cloud probes (0.75 µm and 8.7 mm). $\frac{dm_1}{dt}$ is obtained by integrating the collection kernel $K$ for all smaller size bins (i.e. $D_1 > D_2$) described by the diameter of the collected drops $D_2$

$$\frac{dm_1}{dt} = \frac{\pi \rho_w}{6} \lim_{D_{1} \rightarrow D_{2}'} \int_{D_{\text{min}}}^{D_{2}'} K(D_1, D_2) N(D_2) D_2^3 dD_2$$

where $\rho_w$ is the density of liquid water. For simplicity, here we use a simple polynomial approximation of $K$

$$K(D_1, D_2) \approx \begin{cases} 
5.78 \times 10^3 (v_1 + v_2) & 20 \leq D_1 \leq 100 \mu m \\
9.44 \times 10^9 (v_1^2 + v_2^2) & D_1 > 100 \mu m 
\end{cases}$$
where $v_i$ is the drop volume corresponding to $D_i$ (Long, 1974; Pruppacher and Klett, 2010). Typical values range from $1 \times 10^{-16} \text{ kg m}^{-3} \text{s}$ for $\text{LWC} = 0.001 \text{ g m}^{-3}$ to $1 \times 10^{-5} \text{ kg m}^{-3} \text{s}$ for $\text{LWC} = 1 \text{ g m}^{-3}$. Because we are interested how $C$ is modified in the OLI region, we show the difference of $C$ between both sites in Fig. 6 as a function of $r_{\text{eff}}$ and $\text{LWC}$. It can be seen that $C$ is decreased at OLI in comparison to NSA by up to one order of magnitude for constant $\text{LWC}$ and $r_{\text{eff}}$. This is caused by reduced broadening of the drop size distribution towards large drops at OLI (not shown), consistent with cloud chamber experiments (Gunn and Phillips, 1957). Interestingly, differences in $C$ are largest for $r_{\text{eff}}$ smaller than 16 $\mu\text{m}$ where absolute values of $C$ are small according to Gerber (1996). However, small absolute increases in $C$ for small $r_{\text{eff}}$ are also crucial for triggering the positive feedback of drop growth due to collision-coalescence. When evaluating the potential impact of reduced $C$ on cloud life cycle, one also has to consider that typical $r_{\text{eff}}$ values are reduced at OLI in comparison to NSA for the same $\text{LWC}$ (Fig. 5, a, b). Therefore, we estimate the mean growth rate $\bar{C}$ averaged over $r_{\text{eff}}$ as a function of $\text{LWC}$ (Fig. 6, a red lines). Doing so reveals that, for constant $\text{LWC}$, $\bar{C}$ is reduced by 1 to 1.5 orders of magnitude at OLI. The offset is surprisingly constant for $\text{LWC}$ larger than 0.01 $\text{g m}^{-3}$. Differences in $C$ also translate to different rain rates $R$, which can be estimated by integrating the measured DSD and applying the fall velocity parametrization of Khvorostyanov and Curry (2002). Like $C$, $R$ is reduced by up to one order of magnitude for constant $\text{LWC}$ and $r_{\text{eff}}$ (Fig. 6, b). Averaging over all $r_{\text{eff}}$ enhances the effect and leads to differences of up to two orders of magnitude for $R$ as a function of $\text{LWC}$.

Parameterizations of $C$ and $R$ are crucial in numerical models to transform cloud liquid water into rain droplets and to remove condensate from the atmosphere. Typically, numerical weather and climate models include either one ($\text{LWC}$, one-moment schemes) or two ($\text{LWC}$ and drop concentration or $r_{\text{eff}}$, two-moment schemes) prognostic variables per hydrometeor species. Our comparison of $C$ and $R$ for both sites reveals, however, that these quantities vary by up to one order of magnitude for constant $\text{LWC}$ and $r_{\text{eff}}$ (which would be equivalent to a two-moment scheme). Considering only $\text{LWC}$ (i.e. one-moment scheme) increases the differences to 1.5 to 2 orders of magnitude. As a consequence, additional moments or the full particle size distribution need to be considered in order to accurately estimate $C$ and $R$ in these models. Otherwise, a parametrization of $C$ or $R$ relying only on $\text{LWC}$ ($\text{LWC}$ and $r_{\text{eff}}$) might be biased up to 1.5 to 2 (1) orders of magnitude for one-moment (two-moment) schemes.

Even though the rate of mass removal from a cloud is an important process impacting cloud life cycle, it is important to note that modifications to $C$ and $R$ cannot be directly translated into modifications in cloud lifetime. This is because a reduction in $R$ could result in a number of feedbacks such as cloud deepening (Stevens and Feingold, 2009) or reduced evaporation just below cloud base (Jiang and Feingold, 2006; Feingold and Siebert, 2009) that would act to counter the the extending effect of reduced precipitation rate on cloud lifetime.

5 Cloud-Aerosol interaction

So far, we have demonstrated that there are differences in aerosol properties and cloud properties between NSA and OLI. This is in general agreement with the findings of Hobbs and Rangno (1998) who found an increase in droplet number concentration.
when flying over Prudhoe Bay. In this section, we present evidence that these changes are indeed connected to local industrial activities centered around the Prudhoe Bay oil fields.

In order to evaluate the likelihood that Prudhoe Bay emissions impacted different portions of the ACME-V flights, we use the HYSPLIT dispersion model. Simulations were completed using one continuously emitting source located over the Prudhoe Bay oilfields (70.2556°N, 148.3384°W), using a configuration similar to that discussed above for evaluation of wildfire emissions. For OLI (NSA), 62% (16%) of all ACME-V cloud observations can be traced back to surface emissions originating from the Prudhoe Bay oilfields. The 16% determined for NSA is roughly twice that presented in Kolesar et al. (2017). However, they studied aerosol concentration at the surface instead of aloft and used a multi-year data set, which could introduce substantial variability from the 3-month period evaluated here. The HYSPLIT simulations (Fig. 7) show that the mass concentration

Figure 5. 2D-Histogram of number of observations as a function of effective radius $r_{\text{eff}}$ and liquid water content LWC for Utqiagvik/Barrow (a, 942 observations), Oliktok Point (b, 1608 observations) and Ivotuk/Toolik (c, 579 observations).

Figure 6. As Fig. 5, but with the difference OLI - NSA of the collection kernel growth rate $C$ (a) and the rainrate $R$ (b) as color. The rates averaged over $r_{\text{eff}}$ are shown in red for OLI (dotted) and NSA (dashed). The green dots highlight data points with less than five observations.
originating from local pollution sources is on average more than two order of magnitudes higher at OLI than at NSA. These simulations indicate that relative to NSA, the number of clouds impacted by local emission is increased at OLI and these clouds are impacted by a larger amount of aerosol particles by mass.

**Figure 7.** As Fig. 5, but with mass concentration of local emission according to the HYSPLIT model for OLI (a) and NSA (b). Note that the color scales for OLI and NSA differ by a factor of 100. The red dots indicate data points with less than five observations.

Fig. 8 relates in-cloud observations of LWC and $r_{\text{eff}}$ to below-cloud observations of rBC similar to the approach of Jackson et al. (2012). This means that a single below-cloud aerosol value is assigned to every data point within the same cloud based on the assumption that aerosol properties are not changing on the scale of individual cloud profiles. The below-cloud values are averaged, whenever possible, over 30 seconds with a 3 second gap to the cloud base to avoid issues with time synchronisation across instruments or cloud particle contamination of aerosol probe measurements. Similar to Fig. 8, the below-cloud CPC CN concentration is shown in Fig. 9. These figures demonstrate clear relationships between cloud microphysics and rBC and CN (compare also Fig. 3). Around OLI, below-cloud rBC and CN values are increased, more than 10 ng kg$^{-1}$ and 2000 m$^{-3}$, respectively. The coincidence of increased rBC concentrations with reduced $r_{\text{eff}}$ might indicate that the observed rBC acted as a CCN. However this would require the rBC to be coated with more hygroscopic material (e.g. sulfate), because pure rBC does not serve as efficient CCN (Weingartner et al., 1997). While many of the CN detected by the CPC are likely too small to act as a CCN, these small particles can grow to accumulation mode quickly, potentially creating a particle population capable of acting as CCN (Jaenicke, 1980). It is interesting to note that despite the notion that elevated concentrations should result in smaller $r_{\text{eff}}$, these measurement indicate that local pollution are not connected to $r_{\text{eff}}$ smaller than 9 µm. This is additionally supported by the reduced HSYPLIT mass concentrations (Fig. 7) for $r_{\text{eff}}$ below 9 µm.

For the PCASP (Fig. 10), the aerosol concentration decreases more monotonically from small to large $r_{\text{eff}}$ values. The fact that the response of $r_{\text{eff}}$ to PCASP aerosol concentrations is very similar for both sites is likely because the PCASP covers the aerosol size range most relevant to droplet nucleation. It should be noted that the monotonic decrease in PCASP concentration with increasing droplet size is consistent with the first indirect effect. However, since no enhanced PCASP particle concentrations are found to be correlated to droplet sizes in the emissions-impacted 9 to 12 µm range (unlike for rBC
and CN), there is no indication that local emissions are directly altering liquid clouds to have smaller $r_{\text{eff}}$ as a result of PCASP-sized particles. Even though similar PCASP concentrations lead to similar $r_{\text{eff}}$ for both sites, differences still exist relating to the breadth and tail of the distributions, as can be seen from differences in $C$ and $R$ (Fig. 6).

Fig. 4 also includes the below-cloud concentrations for CN and PCASP assigned to the clouds presented in this study (Note that—in contrast to the aerosols—clouds from all altitudes are shown). While for NSA, most clouds correspond to small CN concentrations, for OLI there are also clouds present corresponding to high CN values (i.e. potentially fresh emissions) most of which also feature high PCASP concentrations ($> 200 \text{ cm}^{-3}$) with mean sizes $> 200 \text{ nm}$. This indicates that even though CN measurements are dominated by Aitken mode particles, there are frequently sufficient CCN concentrations present to allow cloud formation. Because small particles can grow quickly into accumulation mode particles, these CCN might also originate from local sources.

![Figure 8](image-url)

**Figure 8.** As Fig. 5, but with absolute values for SP2 refractory black carbon rBC concentration for (a) OLI and (b) NSA.

![Figure 9](image-url)

**Figure 9.** As Fig. 8, but with absolute values for CPC3025 condensation nuclei CN concentration.
Finally, comparison of rBC size (Fig. 11) shows that black carbon particles are generally 50 to 300 nm smaller at OLI than at NSA. This is consistent with aging of rBC during atmospheric transport, and supports the idea that rBC measurements around OLI are associated with local emissions from Prudhoe Bay and not transported fire emissions. At NSA, this pattern of larger SP2 sizes is only interrupted for a small range of droplet $r_{\text{eff}}$ between 9 and 12 $\mu$m, where both sites have similar mean sizes. This might be related to sampling issues, because this range coincides with very low rBC concentrations (Fig. 8) and low number of observations at NSA.

Figure 11. As Fig. 8, but with the mean size of refractory black carbon rBC measured below cloud.
6 Quantification of cloud aerosol interaction

Various attempts have been carried out to quantify aerosol cloud interaction (ACI) in Arctic regions (Coopman et al., 2016; Zamora et al., 2016) and its impact on radiation (Earle et al., 2011; Tietze et al., 2011). One common definition used for quantification purposes is:

\[ ACI = \frac{1}{3} \frac{d \ln N_{\text{tot}}}{d \ln N_a} \]  

(4)

with \( N_{\text{tot}} \) the number concentration of cloud droplets and \( N_a \) the number concentration of aerosols (Feingold et al., 2001; McComiskey et al., 2009). For observations, ACI is obtained using a linear regression of \( \ln N_{\text{tot}} \) and \( \ln N_a \). We prefer defining ACI using \( N_{\text{tot}} \) instead of \( r_{\text{eff}} \), because the latter would require to classify the clouds by LWP, significantly reducing the size of the data set. Fig.12 shows \( N_{\text{tot}} \) and \( N_a \) for both sites. \( N_a \) is obtained form the PCASP because it covers the size range of active aerosols best. The ACI value for all clouds is 0.13 with \( R^2 = 0.23 \). Even though \( R^2 \) is small, the ACI value found here is similar to Zamora et al. (2016) who found ACI values of 0.15 for PCASP using a multi-campaign data set focused on biomass burning. McComiskey and Feingold (2012) found that the choice of platform and observational scales can have a significant impact on the estimation of ACI making comparisons between data sets challenging. Zamora et al. (2016), however, also used cloud-averaged in-situ aircraft observations and as a consequence we expect them to be comparable. When applying the linear regression to the data sets corresponding to the two sites separately, the obtained ACI values differ (Table 1), with OLI having a lower ACI value (0.1) than NSA (0.2). Given the small sample size (35 and 16 cases for OLI and NSA, respectively), it is not possible to answer the question of whether this is related to a difference in nucleation efficiency between aerosols at the two sites. The lower \( R^2 \) value for OLI (0.18) in comparison to NSA (0.40) could indicate that the assumption that PCASP particle concentrations represent a good approximation for CCN concentrations is partly violated at OLI. This could result from those particles being less aged and consequently less coated by sulfates and organics in comparison to those observed around NSA.

For comparison, we also evaluate ACI calculated using data points associated with forest fires. Based on the flight patterns executed, all of the cloud measurements associated with forest fire emissions were sampled in the vicinity of OLI. While fire emissions were also advected to the area surrounding NSA, cloud measurements from these time periods did not pass the quality control measures implemented (continuously ascending or descending profiles). Based on the measurements collected, aerosols associated with forest fires generally feature higher accumulation mode concentrations (and in consequence smaller \( r_{\text{eff}} \)), which is consistent with aging of these particles during transport, and in contrast to the freshly emitted particles generally found around OLI. When including cases associated with forest fire emissions, ACI is found to be 0.14 for both OLI and the complete data set, and is similar to results obtained when omitting forest fire influenced cases. Therefore, we conclude that a significant difference of ACI between local emissions and forest fires cannot be found given the limited data set.

7 Conclusions

The impact of local emissions from industrial activities in northern Alaska on liquid clouds has been investigated based on vertical profiles of aerosol and cloud properties during the ACME-V aircraft campaign, together with measurements from the
Table 1. ACI values for the subsets presented in Fig. 12

<table>
<thead>
<tr>
<th>Data set</th>
<th>ACI</th>
<th>R²</th>
<th># clouds</th>
</tr>
</thead>
<tbody>
<tr>
<td>all data</td>
<td>0.13</td>
<td>0.23</td>
<td>51</td>
</tr>
<tr>
<td>Oliktok Point</td>
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<td>0.18</td>
<td>35</td>
</tr>
<tr>
<td>Utqiağvik/Barrow</td>
<td>0.20</td>
<td>0.40</td>
<td>16</td>
</tr>
<tr>
<td>all data (with fires)</td>
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<td>0.43</td>
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<tr>
<td>Oliktok Point (with fires)</td>
<td>0.14</td>
<td>0.44</td>
<td>59</td>
</tr>
</tbody>
</table>

Figure 12. Aerosol indirect effect defined using cloud averaged cloud drop concentration \(N_{tot}\) and PCASP aerosol concentration \(N_a\). The color is to differentiate between OLI (green) and NSA (purple). Clouds related to forest fires are marked with a ‘+’. The trend lines indicate the linear regressions to obtain ACI (excluding forest fires) for the complete data set (gray), NSA (purple) and OLI (green).

ARM sites in Northern Alaska: Oliktok Point (OLI) and Utqiağvik (formerly known as Barrow or North Slope of Alaska, NSA). Our main findings can be summarized as follows:

1. Concentrations of condensation nuclei (CN) and refractory black carbon (rBC) are higher in the OLI area (Fig. 3). This is related to emissions associated with local oil and natural gas extraction activities. In contrast, concentrations of larger (diameter > 100 nm) particles are not elevated around OLI when compared to NSA.

2. In addition, we found (Fig. 5) that liquid clouds generally feature smaller \(r_{eff}\) at OLI (mean 9.8±4.0 µm) when compared with NSA (12.2±6.9 µm). Furthermore, collision-coalescence and precipitation rates are reduced by up to two orders of magnitude around OLI (Fig. 6). Only half of this reduction can be explained by the reduced \(r_{eff}\). As a consequence, the breadth of the size distribution of liquid droplets has to be smaller at OLI.
3. Multiple lines of evidence connect these changes in cloud properties to the observed local emissions. First, HYSPLIT simulations show that 62% of all cloud observations around OLI can be traced back to local emission sources (Fig. 7). Second, reduced $r_{\text{eff}}$ (between 9 and 12 $\mu$m) of OLI clouds correspond to increased CN and rBC concentrations (Figs. 8, 9). Third, the mean size of cloud-associated rBC particles is smaller at OLI which is consistent with freshly emitted, less aged particles (Fig. 11). Finally, while no enhanced concentrations of larger accumulation mode particles were observed for OLI (Fig. 3), clouds there were found to be frequently connected to enhanced CN and accumulation mode concentrations (Fig. 4).

4. Quantification of aerosol cloud interaction (ACI) is challenging due to the small data set. Having said this, the results from this study, based on evaluation of clouds impacted by both local emissions and forest fires, are consistent with previous studies of ACI in the Arctic environment (Fig. 12). While forest fire cases have typically higher aerosol concentrations and consequently droplet concentrations, their inclusion into the estimation of ACI does not substantially alter the found relationship.

Because only liquid clouds were observed during the majority of the flights, the impact of local pollution on mixed phase and pure ice clouds is not covered here. Moreover, the question whether the industrial activities at the North Slope of Alaska also lead to a change in local climate (e.g. due to cloud radiative forcing, precipitation impacts, or cloud life cycle), cannot be answered with in-situ aircraft measurements alone. These questions can likely better be answered using ground- and satellite-based remote sensing data from OLI and NSA by identifying differences between the sites in cloud cover, liquid water path, emissivity, effective droplet size, and precipitation occurrence. Nevertheless, based on this limited in-situ data set we can conclude that local emissions form industrial facilities in Alaska do influence local cloud properties while the overall spatial extent of these influences has yet to be evaluated. Given the observed cloud modifications, the effects of anthropogenic pollution on local climate should be considered when developing industrial infrastructure in an already fragile and warming Arctic environment.

8 Data availability

The surface observations from OLI and NSA as well as the ACME-V data set are available at the ARM archive www.arm.gov/data (Holdridge and Kyrouac, 1993; Biraud et al., 2016), the phase classification of the cloud probes is available form the corresponding author on request.

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