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Does precipitation susceptibility vary with increasing cloud thickness in marine stratocumulus?

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Abstract

The relationship between precipitation rate and accumulation mode aerosol concentration in marine stratocumulus-topped boundary layers is investigated by applying the precipitation susceptibility metric to aircraft data obtained during the VOCALS Regional Experiment. The mean precipitation rate R over a segment of the data is expressed as the product of a drizzle fraction f multiplied by a drizzle intensity I (mean rate for drizzling columns). The susceptibility S_x is then defined as the fractional decrease in precipitation variable $x = \{R, f, I\}$ per fractional increase in the concentration of aerosols with dry diameter $> 0.1 \mu\text{m}$, with cloud thickness h held fixed. The precipitation susceptibility S_R is calculated using data from both precipitating and non-precipitating cloudy columns to quantify how aerosol concentrations affect the mean precipitation rate of all clouds of a given h range and not just the mean precipitation of clouds that are precipitating. S_R systematically decreases with increasing h , and this is largely because S_f decreases with h while S_I is approximately independent of h . In a general sense, S_f can be thought of as the effect of aerosols on the probability of precipitation, while S_I can be thought of as the effect of aerosols on the intensity of precipitation. Since thicker clouds are likely to precipitate regardless of ambient aerosol concentration, we expect S_f to decrease with increasing h . The results are broadly insensitive to the choice of horizontal averaging scale. Similar susceptibilities are found for both cloud base and near-surface drizzle rates and when the analysis is repeated with cloud liquid water path held fixed instead of cloud thickness. Simple power law relationships relating precipitation rate to aerosol concentration or cloud droplet concentration do not capture this observed behavior.

1 Introduction

Drizzle with wide ranging intensities and areal extent is a common feature in stratocumulus-topped boundary layers, especially in remote marine environments (Brost et al., 1982; Nicholls and Leighton, 1986; Frisch et al., 1995; Vali et al., 1998; Yuter et al., 2000; Pawlowska and

Brenguier, 2003; Bretherton et al., 2004; Comstock et al., 2004; vanZanten and Stevens, 2005; Leon et al., 2008; Kubar et al., 2009). Over parts of the eastern subtropical/tropical oceans dominated by stratocumulus, including the southeastern Pacific, the intensity and frequency of drizzle tends to increase westwards from the coast (Leon et al., 2008; Kubar et al., 2009; Bretherton et al., 2010). The westward increase in drizzle coincides with changes in both aerosol concentrations and macrophysical properties of the stratocumulus deck (e.g., George and Wood, 2010; Bretherton et al., 2010; Allen et al., 2011) thereby raising the question: to what extent does the westward increase in drizzle reflect changes in cloud macrophysical properties, changes in cloud microphysical properties, and changes in aerosol (e.g., Wood et al., 2009).

Of the drizzle falling from stratocumulus, a significant fraction (70-80%) evaporates before reaching the surface (Comstock et al., 2004; Wood, 2005). Whether evaporating or not, drizzle can substantially change the boundary layer characteristics by vertically redistributing energy and water (Wang and Albrecht, 1986; Turton and Nicholls, 1987; Stevens et al., 1998; Wood, 2007; Savic-Jovicic and Stevens, 2008; Feingold and Siebert, 2009). Therefore, to improve the simulated structural and radiative properties of stratocumulus in general circulation models, we need a better understanding of the macrophysical and microphysical properties that determine the frequency and intensity of drizzle (Wyant et al., 2007).

Aerosol indirect effects (AIEs), broadly defined here as the effect of aerosols on the cloud albedo, are strongly sensitive to the amount of drizzle falling from warm clouds (Rotstayn and Liu, 2005) and to the amount evaporating below cloud (Wood, 2007; Chen et al., 2011). The AIEs have been difficult to quantify, in part because different cloud types respond differently to perturbations in aerosol concentration (Ackerman, 2004; Stevens and Feingold, 2009). A key effect that aerosols can have on clouds is the suppression of warm rain, which forms the basis of the second AIE (Albrecht, 1989). Part of the difficulty in quantifying the magnitude of the second AIE is that warm rain responses to perturbed aerosol concentrations are poorly understood.

Accumulation mode aerosols, which often are good proxies for cloud condensation nuclei (Martin et al., 1994), have a typical size range of 0.1 to 1 μm (Seinfeld and Pandis, 2006), and in many marine environments where accumulation mode aerosol concentrations are typically $<200\text{ cm}^{-3}$ increasing the accumulation mode aerosol number concentration increases the cloud droplet number concentration N_d in similar proportion (Ramanathan et al., 2001). In warm clouds, the increase in N_d reduces the average cloud droplet size, reduces the collision coalescence efficiency, and suppresses precipitation formation (Albrecht, 1989, inter alia). However, it has been difficult to assess the extent of this suppression in observational datasets, because aerosol concentrations tend to correlate well with meteorological factors, making it difficult to separate the aerosol effects from simultaneous macrophysical drivers (Mauger and Norris, 2007; George and Wood, 2010).

The precipitation susceptibility metric S_0 quantifies the suppression of precipitation by aerosols, while minimizing the confounding effects of macrophysics (Feingold and Siebert, 2009). It is defined as

$$S_0 = \left(-\frac{d\ln R}{d\ln N_d} \right)_{macro}, \quad (1)$$

where R represents the precipitation rate (at some altitude) and N_d represents the cloud droplet number concentration (Feingold and Siebert, 2009). In previous studies that used satellite measurements to estimate the susceptibility, the N_d in Eq. 1 was replaced with a cloud condensation nuclei proxy α (Sorooshian et al., 2009; Duong et al., 2011). This study follows the second formulation of susceptibility and uses the ambient accumulation mode aerosol concentration to obtain susceptibility estimates. The subscript “macro” indicates that the susceptibility is calculated with fixed cloud macrophysical properties, e.g. fixed cloud thickness or liquid water path (LWP). This is critical because significant correlations between the microphysical and meteorological properties could potentially cause nonzero precipitation susceptibility even if the aerosols are not responsible for the precipitation changes.

Precipitation susceptibility estimates for marine stratocumulus can be obtained from previ-

ous field studies. Field studies of precipitating stratocumulus (e.g., Pawlowska and Brenguier, 2003; Comstock et al., 2004; vanZanten and Stevens, 2005) identified that simple power law relationships display some skill in relating the cloud base precipitation rate R_{CB} to cloud thickness h (or, alternatively, LWP) and to cloud droplet number concentration N_d (Geoffroy et al., 2008). The relationships take the form

$$R_{CB} = Ah^\alpha N_d^{-\beta}. \quad (2)$$

For clouds with the same thickness h , the exponent β is identical to the susceptibility S_0 in Eq. 1. The β values reported in previous field studies range from 1 to 1.75 (Geoffroy et al., 2008). Variations in β with cloud thickness, however, have not been reported in these studies. This may be primarily due to sampling constraints in the previous studies, but it is important to point out that the single power law fits implicitly assume that the precipitation susceptibility does not vary with cloud thickness. Recent satellite observations (Kubar et al., 2009) and simple theoretical model results (Wood et al., 2009) provide observational evidence and theoretical justification that the precipitation susceptibility decreases as the cloud thickness increases in marine stratocumulus. Studies of warm cumulus clouds (Sorooshian et al., 2010; Duong et al., 2011), on the other hand, report that the susceptibility peaks at intermediate values of cloud LWP (~ 600 – 1300 g m^{-2}). Reconciling these various relationships into a coherent framework to explain precipitation susceptibility in warm clouds is likely to be a formidable challenge.

In this study we investigate the factors controlling precipitation susceptibility in warm, stratiform clouds using aircraft observations of precipitating and non-precipitating marine stratocumulus observed in the southeastern Pacific during the VOCALS Regional Experiment (REx, Wood et al., 2011b). The susceptibility metric is used to determine the extent to which precipitation susceptibility depends upon cloud thickness/LWP. The manuscript is organized as follows. Section 2 describes the data used in the analysis and the methods used to calculate the susceptibilities. Section 3 reports the susceptibilities that are calculated and their associated sensitivities to the averaging length, to the height of the drizzle, and to the different methods of calculating the susceptibility. How the results fit in with the existing literature and the precau-

tions that need to be taken when interpreting the susceptibility metric are discussed in Section 4.

2 Data and methods

We use a combination of in-situ and remotely sensed observations of aerosol and cloud properties from eleven research flights of the NSF/NCAR C-130 aircraft during VOCALS REx, which was conducted in the southeast Pacific during October and November 2008 (Wood et al., 2011b). Since aerosol measurements are missing from RF03 and cloud thickness measurements are missing from RF01 and RF09, data from only eleven of the total fourteen research flights are used. The data used in this study are all analyzed at 1 Hz time resolution, which translates to a spatial resolution of approximately 100 m.

2.1 Cloud macrophysical properties

Cloud macrophysical properties (thickness and LWP), drizzle properties (rates at cloud base and below cloud), and aerosol properties below cloud are obtained simultaneously from straight and level flight legs flown below the cloud at approximately 150 m altitude (subcloud legs). Cloud thickness, precipitation rate, and cloud LWP are estimated from measurements obtained from the Wyoming Cloud Lidar (WCL), the Wyoming Cloud Radar (WCR), and G-band Vapor Radiometer (GVR). The cloud base height is determined from the column maximum vertical gradient in the lidar backscatter. The cloud top height is determined from the maximum height at which the power return and the variance of the Doppler velocity exceed threshold values. An overview of cloud properties observed during VOCALS REX can be found in Bretherton et al. (2010). Details of the LWP retrievals from the GVR can be found in Zuidema et al. (2011). Cloud thicknesses are available from 51 % of all clouds detected with the WCL, because cloud top heights are not retrievable for clouds with low radar reflectivity; cloud thicknesses are available from 80 % of samples in which the column maximum reflectivity exceeds -20 dBZ.

On the other hand, the cloud LWP measurements are available from the GVR for 69 % of all clouds and 86 % of clouds with maximum reflectivity greater than -20 dBZ.

2.2 Precipitation rate estimates

Unless otherwise noted, precipitation rates R are obtained from radar reflectivities Z at cloud base using the Z - R relationship $R(\text{mm day}^{-1}) = 2.01 Z^{0.77}$, proposed by Comstock et al. (2004). A reflectivity threshold of -15 dBZ, corresponding to a drizzle rate of 0.14 mm day^{-1} , is used to distinguish between drizzling and non-drizzling clouds. Cloud base precipitation rates are used, because this is the most energetically-relevant level for stratocumulus drizzle and to facilitate comparison with previous studies of stratocumulus drizzle (Pawlowska and Brenguier, 2003; Comstock et al., 2004; vanZanten and Stevens, 2005; Wood, 2005). Though using different Z - R relationships has little effect on our results, for the precipitation rates at 250 m, we use the Z - R relationship for surface precipitation from Comstock et al. (2004), $R(\text{mm day}^{-1}) = 0.61 Z^{0.91}$ (consistent with Bretherton et al. (2010)). For the precipitation rate at 500 m, we use the same Z - R relationship that is used to calculate the cloud base precipitation rate (consistent with Wood et al. (2011a)). Different Z - R relationships for cloud base precipitation rates and surface precipitation rates are obtained by Comstock et al. (2004), because subcloud evaporation increases the mean drizzle drop radius with distance below cloud base (see also vanZanten and Stevens (2005)).

2.3 Aerosols

Aerosol concentrations in the subcloud layer are obtained from the Passive Cavity Aerosol Spectrometer Probe (PCASP), which measures aerosol number concentrations in the size range of 0.1 – $3 \mu\text{m}$. Most accumulation mode aerosols are found in this size range (Martin et al., 1994; Ramanathan et al., 2001). The relationship between the PCASP aerosol concentration at 150 m altitude and the cloud droplet concentration, N_d , as measured by the Cloud Droplet Probe (CDP, size range of 1 – $52 \mu\text{m}$) is analyzed using the data from profile legs flown during the research

flights. Profile legs are flight legs flown from 150 m to above the inversion or vice versa (with a rate of ascent/descent typically around 300 m min^{-1}). From 90 profiles, we compare mean PCASP aerosol concentrations, N , from below 400 m with mean in-cloud N_d to establish the relationship between them.

2.4 Averaging Scale

We choose to average the data over 10 km segments (100 seconds of data), because drizzle has been found to vary at this scale (vanZanten and Stevens, 2005), and the dominant scale of variability in marine stratocumulus clouds is usually that associated with mesoscale cellular structures with scales of this order (Wood and Hartmann, 2006). We find an e -folding length of 1.6 km for the cloud base precipitation rate field, suggesting that for the dataset as a whole, the shortest averaging length scale for independent data segments (Leith, 1973) is 3.2 km. Similarly, the e -folding length for the cloud thickness is 1.7 km. A given 10 km segment (100 seconds of data) from the subcloud leg is used in the analysis if estimates of cloud thickness (or cloud LWP) and aerosol concentration are available for some or for the entire segment. Means of precipitation and cloud thickness are averaged over only the regions of the 10 km segments where cloud thickness measurements are obtained. Since cloud thickness measurements could not be obtained for the thinnest of clouds, our analysis is restricted to the thicker clouds that were sampled by the research flights.

2.5 Susceptibility estimation

The 10 km segment-averaged cloud thickness h , PCASP aerosol concentration N , and precipitation rate R are plotted in Fig. 1. Along the x -axis towards thicker clouds, R increases substantially. To a lesser extent, towards higher N , R decreases. The aim of this study is to quantify this decrease in R with increasing N .

The high resolution aircraft data also allow us to determine the fraction and intensity of the drizzle in each segment. Segment-mean precipitation rates R from the 10 km segments are partitioned into the fraction of the cloudy columns that are drizzling f , and the mean drizzle rate in those columns, here termed drizzle intensity I . This relation can be written as

$$R = fI. \quad (3)$$

A data point in this study consists of estimates of the mean cloud thickness h (or mean cloud LWP), the mean PCASP concentration N , and the three precipitation variables R , f , and I . In non-drizzling segments, where $R=0$ and $f=0$, I is considered undefined. Hence, only drizzling segments are used to calculate the susceptibility for I . This is to reduce the covariance between S_f and S_I when we calculate the susceptibility.

We examine the susceptibility for R , f , and I separately, referring to these susceptibilities as S_R , S_f , and S_I respectively. The susceptibility is derived from the observational dataset for four bins of increasing cloud thickness h , each with the same number of data points (percentiles 0–25 %, 25–50 %, 50–75 %, 75–100 % of h). The range of h in each of the h bins are 14–207 m, 208–294 m, 294–379 m, and 381–792 m. By dividing the data into four bins and estimating the susceptibility separately for each bin, we (a) minimize the macrophysical effects that can be detrimental to our susceptibility estimates and (b) estimate how the susceptibility changes with h . The correlations between h and N within the four bins from thinnest to thickest h are 0.3, 0.0, -0.1 , and -0.3 respectively.

Wet scavenging of aerosols by precipitation has been found by Duong et al. (2011) to negatively bias the susceptibility values that they calculated from satellite measurements. Given the boundary layer conditions that were observed during VOCALS REx, we expect that wet scavenging would have a small effect on our susceptibility estimates, because the time scales for wet scavenging of aerosol are substantially longer than the time scales for drizzle formation and boundary layer mixing. Using the parameterization from Wood (2006) with a cloud base precipitation rate of 1 mm day^{-1} and a cloud droplet number concentration of 100 cm^{-3} , we calculate a wet scavenging rate of approximately $3 \text{ cm}^{-3} \text{ hr}^{-1}$. Given that the typical lifetime

of drizzle cells in the region is approximately two hours (Comstock et al., 2005) and that the mixing of the boundary layer, estimated from the convective velocity and depth of the boundary layer, is approximately one hour, we do not expect the aerosol concentrations to decrease substantially before precipitation rates adjust to the changes in aerosol concentrations. Therefore, we expect that wet scavenging would not statistically bias our susceptibility estimates. In making this argument we assume that the cloud thickness and aerosol concentration are the main controls of precipitation, and we cannot rule out the possibility of another control on precipitation that enhances precipitation and hence leads to an anti-correlation between N and R through cumulative wet scavenging events. However, given the nature of our dataset, we cannot test this issue, and hence, we do not filter the data for potential wet scavenging effects.

In contrast to previous studies (e.g., Jiang et al., 2010; Duong et al., 2011) that examined the drizzle susceptibility in regions determined to be precipitating, using some cutoff to distinguish precipitating and non-precipitating clouds, we examine the drizzle susceptibility of all clouds, precipitating or not, by investigating the how aerosols effect both the frequency and the intensity of drizzle in clouds of a given thickness. Thus, this analysis attempts to quantify the effect of aerosols on the mean precipitation rates of all clouds of a given thickness, which is different from quantifying the effect of aerosols on the mean precipitation rates of only precipitating clouds of a given thickness. This approach is appropriate in terms of understanding, because we include clouds for which precipitation is completely suppressed due to high ambient aerosol concentrations. Therefore, we account for how increased aerosol concentrations can decrease the likelihood that a cloud precipitates, which is important for weakly-precipitating clouds. For example, a cloud whose precipitation is completely suppressed by increased aerosol concentration is considered in our calculation of susceptibility, but would not be if only precipitating clouds are included in the calculation. By incorporating non-drizzling clouds, we can address two questions in our susceptibility estimates: (a) to what extent do aerosols perturb existing drizzle rates, and (b) to what extent do aerosol perturbations affect the likelihood or fraction of drizzle?

Susceptibilities of each precipitation variable $x=\{R,f,I\}$ are estimated for each of the four cloud thickness bins using the form,

$$S_x = -\frac{\ln(x_+/x_-)}{\ln(N_+/N_-)}, \quad (4)$$

where the plus and minus subscripts indicate ensemble means of a particular variable over the population corresponding to the top and bottom 33 % of N respectively. The uncertainty in S_x is calculated using bootstrap resampling (Efron and Gong, 1983). For each cloud thickness bin, the data are resampled with replacement and the resulting susceptibility is calculated 10,000 times to obtain a distribution of susceptibility estimates for that cloud thickness bin. The middle 95th percentile (2.5 %,97.5 %) of this distribution is used to denote the 95 % confidence interval of S_x .

Mean properties of N , h , and R in the four cloud thickness bins are summarized in Table 1. One of the noticeable changes across the bins is the marked increase in mean R . Another is that the range of N observed differs across the bins. Since in the VOCALS region the thickest clouds are found west of 80 W, where aerosol concentrations are typically below 200 cm^{-3} (Allen et al., 2011), this dataset does not include data points with high aerosol concentrations and high cloud thicknesses.

3 Results

3.1 Aerosol vs. cloud droplet number concentration

The physical argument that increased aerosol concentrations leads to reduced precipitation is dependent on the assumption that PCASP aerosol concentrations N at 150 m correlate well with cloud droplet concentrations N_d in the overlying cloud. To test this assumption, we take N and N_d from the profile legs and plot the data in Fig. 2. Each point represents mean aerosol

and cloud droplet concentrations from one profile. The square of the linear correlation (r^2) between N and N_d is 0.74. As in previous studies (McComiskey and Feingold, 2008; Duong et al., 2011), a power law relationship of the form $N_d = A N^b$ is used to characterize the relationship between cloud droplet and aerosol number concentration. The exponent b , referred to as ACI_N in the literature (McComiskey and Feingold, 2008), can be thought of as the cloud droplet concentration susceptibility: the fractional increase in cloud droplet concentration in response to the fractional increase in accumulation mode aerosol concentration.

To obtain the b value, b is increased from 0 to 1.5 in increments of 0.05 and the squared correlation (r^2) of N_d and N^b is calculated at each increment. The highest r^2 value of 0.80 is found for $b = 0.55$. The b value that we obtain lies within the range of previously reported values (0.26–0.85 from McComiskey and Feingold (2008)). A study in the same marine stratocumulus region in the southeast Pacific that used N_d estimated from MODIS retrievals and N from ship-based measurements of accumulation mode aerosols found a b value of 0.56 (Painemal and Zuidema, 2010). This agreement is surprising, given that this simple relationship only uses aerosol concentrations in a single size range and does not take into account the aerosol size distribution, aerosol hygroscopicity, or incloud supersaturation. The b value will be necessary when comparing the S_R values in this study, which quantify precipitation responses to aerosol, with previously reported β values from Eq. 2, which quantify precipitation responses to cloud droplets.

3.2 Drizzle susceptibility

The susceptibility S_x of each of the three cloud base precipitation variables $x = \{R, f, I\}$ for each of the cloud thickness bins (h bins) is shown in Fig. 3. The susceptibility S_R of the mean drizzle rate is clearly positive for all h bins. For each bin there is lower than a 2.5 % chance that S_R is less than zero. This is consistent with the hypothesis that increasing aerosols decreases precipitation, as has been noted in numerous previous studies (Pawlowska and Brenguier, 2003; Comstock et al., 2004; vanZanten and Stevens, 2005). The new finding in this study is that S_R

decreases strongly with increasing cloud thickness. The h -dependence is quite marked, with S_R decreasing monotonically by a factor of four from approximately 3 for the h bin with thinnest clouds (mean $h \approx 150$ m) to approximately 0.75 for the h bin with the thickest clouds (mean $h \approx 500$ m).

Further, we find that the susceptibility S_f of drizzle fraction and susceptibility S_I of drizzle intensity exhibit very different dependences upon cloud thickness (Fig. 3). To first order, since R is a product of f and I , S_f and S_I add linearly to give the mean precipitation rate susceptibility, i.e. $S_R = S_f + S_I$. More explicitly, if we expand our susceptibility estimate in Eq. 4, then S_R reduces to the sum between S_f , S_I , and a covariance term

$$S_R = S_f + S_I - \frac{\ln \left[1 + \frac{\overline{I'_+ f'_+}}{I'_+ f'_+} \right] - \ln \left[1 + \frac{\overline{I'_- f'_-}}{I'_- f'_-} \right]}{\ln[N_+/N_-]}, \quad (5)$$

where $\overline{I'f'}$ is the covariance between I and f in each population of h bin. We find that the covariance term is small in our dataset such that we can approximate that $S_R \approx S_f + S_I$.

Like S_R , S_f decreases with increasing h and explains the monotonic decrease in S_R with h . Much of this trend can be attributed to the preponderance of non-drizzling segments (85% are non-drizzling) in the bin with thinnest clouds, compared to the near absence of non-drizzling segments (4% are non-drizzling) in the bin with thickest clouds. Aerosols can, therefore, have a larger effect in determining the drizzle fraction in thinner clouds, compared to the drizzle fraction in thicker clouds.

On the other hand, we find that the susceptibility S_I of drizzle intensity does not significantly change with increasing cloud thickness and has a near-constant value $S_I \approx 0.6$. Dividing S_R between the S_f and S_I and comparing their relative contributions demonstrates that drizzle fraction is important to consider when calculating susceptibility in thinner stratocumulus clouds. Finally, we note that the mean S_R of the four h bins is 1.6, which is in the upper part of the range of β exponents (Eq. 2) found in previous studies (Geoffroy et al., 2008). We are,

however, examining the relationship between ambient aerosol concentration and precipitation, while previous studies examined the relationship between the cloud droplet concentration and precipitation. Based on the b value of 0.55 that we report in Section 3.1, we expect that the S_R would give an underestimate of the relationship between the cloud droplet concentration and precipitation, but we are unable to confirm this from our dataset.

The susceptibility estimates are only useful indicators of the role that aerosols play in modulating precipitation, provided that cloud macrophysical and meteorological properties are held constant. Even when we bin the data by cloud thickness, correlations between macrophysics and microphysics exist. The correlations between h and N in the four cloud thickness bins from thinnest to thickest bin are 0.3, 0.0, -0.1 , and -0.3 . We argue, however, that the decrease in correlation between N and h with increasing cloud thickness is not the reason why we observe the decrease in S_R with increasing cloud thickness in Fig. 3. Since there is a strong positive correlation between precipitation and cloud thickness (Pawlowska and Brenguier, 2003; vanZanten and Stevens, 2005), if correlations had a significant effect on the behavior of the susceptibility, we would expect the susceptibility to increase with cloud thickness, instead of decrease with cloud thickness.

To examine whether the behavior of susceptibility is sensitive to the choice of the number of cloud thickness bins, we calculate the susceptibility after binning the data into greater number of bins (up to ten). The susceptibility values do not show a general increase or decrease with increasing number of bins (not shown). We do observe an increase in the uncertainty (95 % confidence intervals) of the susceptibility estimates, suggesting an increase in sampling errors with the reduced number of points per bin. While we observe the same monotonic decrease in susceptibility when we bin the data into five bins, when we bin the data into more bins, a peak in the susceptibility value is found in the second cloud thickness bin from which there is a monotonic decrease in susceptibility with increasing cloud thickness. While there may be a physical explanation for the peak value in the second bin, given the large uncertainty for the susceptibility of the bin with the thinnest clouds in Fig. 3, we cannot establish this with high

confidence. The main conclusion that we can make is that the susceptibility value of the thinnest clouds is the least robust to changes in the binning procedure. There is a relatively low number of drizzling clouds, and the inclusion or exclusion of the data points has a large impact in determining the susceptibility.

3.3 $-dR/dN$

While the susceptibility metric provides one way of quantifying the decrease in precipitation R due to an increase in aerosol concentration N , we can also calculate the linear change of R due to a change in N , i.e. $-dR/dN$, using the same averaging method (Eq. 4) used to calculate S_R . This may be a more useful metric to quantify the effect of aerosols on precipitation if the interest is in the absolute suppression of precipitation due to an increase in ambient aerosol concentration. In Fig. 4a, we can see that unlike S_R , $-dR/dN$ increases with increasing cloud thickness and that the increase is almost fifty fold from the bin with the thinnest clouds to the bin with the thickest clouds. This is the same order of magnitude increase as the increase in mean R across the four bins (see Table 1).

Furthermore, we can estimate the susceptibility S_R from $-dR/dN$ by approximating that

$$S_R \approx -\frac{\overline{N}}{\overline{R}} \frac{d \ln R}{d \ln N}, \quad (6)$$

where \overline{R} is the bin-mean precipitation rate and \overline{N} is the bin-mean PCASP aerosol concentration in each h bin. When the initial susceptibility estimate is plotted with this new estimate, the two estimates are consistent with each other within the sampling uncertainty (Fig. 4b). This demonstrates that the susceptibility S_{rmR} can be used to estimate both absolute and fractional changes in precipitation in response to increases in N .

3.4 Susceptibility of drizzle at varying heights

We now examine whether the precipitation susceptibility is sensitive to the altitude at which drizzle is measured (i.e. whether drizzle is measured at cloud base, at the surface, or in an intermediate altitude). Evaporation-sedimentation models and observations have shown that in a number of cases of drizzling marine stratocumulus, much of the drizzle evaporates within 250 m of the cloud base (Wood, 2005) and that the fraction evaporating by a given level below cloud base is highly sensitive to the mean drizzle drop size. Studies of southeastern Pacific stratocumulus have found that most of the drizzle evaporates before reaching the surface (Comstock et al., 2004; Bretherton et al., 2010). Since the degree of evaporation is important for understanding precipitation impacts on marine stratocumulus (e.g. Wood, 2007; Savic-Jovicic and Stevens, 2008), it is important to establish whether the precipitation susceptibility is markedly different at the surface from that at the cloud base.

We use precipitation rate estimates from the WCR at altitudes of 500 m and 250 m (see Section 2.2). Radar-derived drizzle rates at the surface are not estimable from the subcloud runs used here, but since the 250 m altitude is considerably lower than the typical cloud base height of ~ 1000 m (Bretherton et al., 2010), the 250 m level is somewhat representative of conditions near the surface. The susceptibility as a function of height is shown in Fig. 5. Quantitative measures of susceptibility of the precipitation at 500 m and 250 m are not obtained in the lower cloud thickness bins, because precipitation is detected in less than 10 % of the segments in those bins. From Fig. 5, there is no evidence that the susceptibility is a strong function of the height below cloud base. Therefore, use of cloud base vs. surface precipitation rates is unlikely to result in significant changes in the susceptibility estimate. Since we do not have a clear physical justification for why the susceptibility should remain constant with height below cloud base, further studies are needed to establish whether there is indeed a physical explanation for what is observed.

We find that our results are insensitive to the Z - R relationship that is used (see Section 2.2).

Switching between the two Z - R relationships has less than a 15 % effect on the susceptibility. Since the susceptibility takes the form $-d\ln R/d\ln N$, from the difference in the Z dependence of R alone, we would expect a difference of 18 % in the susceptibility.

3.5 Susceptibility for different averaging length

An averaging length of 10 km is used for the preceding analyses, but now we explore whether different averaging lengths affect the susceptibility estimates. Duong et al. (2011) found that when the averaging area of the LES model output was decreased, the susceptibility maximum shifted to higher LWP values. To address this possible issue in the data, in Fig. 6 we examine the extent to which the precipitation susceptibility is changed by changing the averaging length (i.e. the segment length). The general result that S_R decreases with cloud thickness holds for all different averaging lengths examined here (5–20 km). The greatest spread is found for the thinnest h bin, but here the sensitivity to averaging length is not monotonic, so the spread is likely sampling noise. The susceptibility of drizzle fraction S_f contributes to the decreasing trend for all three averaging length scales, while the susceptibility of drizzle intensity S_I also remains relatively constant with h for the three averaging length scales and lies between 0.2 and 0.9.

3.6 Susceptibility when binned by cloud liquid water path

So far we have addressed the effect of cloud thickness on the susceptibility. While some studies have examined the effect of cloud thickness on precipitation (Pawlowska and Brenguier, 2003; vanZanten and Stevens, 2005), others have examined cloud liquid water path (LWP) as the macrophysical control of precipitation (Comstock et al., 2005; Sorooshian et al., 2009). The reason for this has been largely driven by the availability of instrumentation. However, in the VOCALS dataset we have estimates of both cloud thickness and LWP. Cloud thickness is typically well correlated with cloud LWP for marine stratocumulus, where the assumption

of adiabatic LWP is a good approximation (Albrecht et al., 1990; Zuidema et al., 2005, 2011). In some areas where there are scuds (cumulus under stratocumulus), the assumption of adiabatic LWP breaks down. A breakdown is also observed for the thick, strongly precipitating stratocumulus clouds with high values of LWP, although precisely how thick the clouds need to be before this occurs appears to differ between studies and likely depends upon the vigor of the turbulent water resupply (Wood, 2005). This motivates our examination of whether susceptibility trends are different if we bin the data by cloud LWP instead of cloud thickness. The GVR obtained cloud LWP for a greater percentage of clouds (69 %), compared to the WCR and WCL, which obtained cloud thickness values for 51 % of all clouds. Therefore, even if the clouds measured in this region are completely adiabatic, we can expect differences in the susceptibilities from the different sampling.

The susceptibilities estimated when we bin by cloud LWP are shown in Fig. 7. As in previous sections, the LWP data are divided into four bins. However, only three susceptibility estimates are shown for each of the three averaging lengths, because in the bin with the lowest LWP (mean LWP of approximately 15 g m^{-2}) precipitation is detected in less than 10 % of the segments. The lack of data in the bin with the lowest quartile of LWP value shows again the difficulty in calculating the susceptibility for the thinnest clouds.

Susceptibilities binned by LWP (Fig. 7) are similar to those by cloud thickness (Fig. 6), but the decrease with increasing LWP is not as marked as it is with cloud thickness, and the trend is dependent upon the averaging length. Specifically, when the data are averaged over 5 km segments, the susceptibility S_R increases from the second to the third LWP bin. This increase in S_R from the second to the third LWP bin is due to an increase in S_I , which counteracts decreases in S_f from the second to third LWP bin (not shown). To investigate the robustness of the maximum in S_R in the third LWP bin, we explore the sensitivity of the 5 km averaged LWP data to the number of bins (Fig. 8). Since the number of data points used for each susceptibility estimate decreases as the number of bins is increased, we use lighter shading for estimates with more bins. The susceptibility generally decreases with increasing LWP, but there is a recognizable

dip in susceptibility at 50 g m^{-2} . The dip in susceptibility at 50 g m^{-2} appears to be what causes the second LWP bin to have a low susceptibility value, resulting in the susceptibility maximum in the third LWP bin when the data is binned into four bins. Since we have no prior reason to expect a dip in susceptibility at 50 g m^{-2} , we expect this dip to be an artifact of the sampling (indeed, the large sampling errors are suggestive of statistical frailty in this result), but further studies with other datasets are necessary to establish whether this dip has a physical explanation.

3.7 Susceptibility with integrated cloud thickness data

One of the main points of this paper is to stress the importance of including all clouds when estimating the precipitation susceptibility. Using the Wyoming Cloud Lidar (WCL) and the Wyoming Cloud Radar (WCR) retrievals, we are only able to retrieve cloud thickness measurements for 51 % of the clouds. Thus, our susceptibility estimates are based on those 51 % of the clouds. When we use the LWP data, we calculate the precipitation susceptibility estimates based on 69 % of the clouds. If we combine the LWP and the cloud thickness, we would expect to incorporate a larger fraction of clouds to calculate the precipitation susceptibility. We, therefore, supplement the WCR and WCL-derived cloud thickness measurements with LWP-derived cloud thickness estimates to create an integrated cloud thickness dataset. When both cloud thickness and LWP measurements are available, the cloud thickness measurements from the WCR and WCL are used.

To convert the LWP retrievals into cloud thickness values we must assume that the cloud liquid water profiles follow the moist adiabat. This assumption is typically valid in marine stratocumulus (Albrecht et al., 1990; Zuidema et al., 2005, 2011), and the approximation is especially good for the thinner, non-precipitating clouds for which the WCR is unable to detect a cloud top height (Zuidema et al., 2005). We estimate the cloud thickness h values from the

cloud LWP values using the expression

$$h = \left(\frac{2 \times LWP}{C_w} \right)^{\frac{1}{2}}, \quad (7)$$

where C_w is a weak function of temperature and pressure, here $C_w = 2 \times 10^{-6} \text{ kg m}^{-4}$ (Pawlowska and Brenguier, 2003). With the combined cloud thickness measurements, we can estimate cloud thickness values for 75 % of all clouds detected by the WCL. When we calculate the susceptibility for averaging lengths of 10, 20, and 5 km, and plot it alongside the previous estimates, as in Fig. 9, we note that in general, the mean susceptibility values are similar to those calculated using the cloud thickness measurements from just the WCL and WCR. The uncertainty in the susceptibility estimates, however, substantially increases in the second h bin for this dataset, when compared to the susceptibility estimates in Fig. 6. This demonstrates the difficulty in constraining the susceptibility values using observations. Nonetheless, our overall conclusion that the susceptibility decreases with increasing cloud thickness still holds.

4 Discussion and conclusions

conclusions

We find that the precipitation susceptibility in marine stratocumulus clouds decreases with increasing cloud thickness. The susceptibility of the mean precipitation rate at cloud base is found to decrease by approximately a factor of two to four from the thinnest to the thickest clouds. Because precipitation rate increases strongly with cloud thickness, this result is particularly pertinent to our understanding of aerosol indirect effects. It confirms the findings from both satellite data and simple heuristic models (Kubar et al., 2009; Wood et al., 2009) showing that the ability of aerosols to suppress precipitation is stronger in clouds that produce only weak precipitation. On the other hand, it is important to note that with the exception of a couple cases, the precipitation susceptibility S_R , as derived in this study, is significantly positive even in the quartile with the thickest clouds. Since the clouds in our dataset span the range of cloud

thicknesses typically found in marine stratocumulus, this perhaps indicates a role for aerosols to suppress precipitation in most marine stratocumulus, albeit to a degree that weakens considerably for the thickest clouds.

This result is qualitatively consistent with an analysis of satellite data (Fig. 12 from Kubar et al. (2009)), which found that the frequency of drizzle decreases dramatically with a decrease in effective cloud droplet number concentration for clouds with low liquid water path, while there is less of a decrease in the frequency for clouds with higher liquid water path. In their study of precipitating cumulus clouds, Sorooshian et al. (2009) found little change in susceptibility at liquid water paths comparable to the liquid water path of the stratocumulus in our study. We attribute the difference to whether or not non-precipitating clouds are included in calculating the susceptibility. When we repeat our analysis, but remove all of the segments with mean drizzle rates less than 0.01 mm day^{-1} , we find that there is little change in the susceptibility with increasing cloud thickness. This difference in behavior is related to results from Jiang et al. (2010) and Duong et al. (2011) that found that susceptibility values are sensitive to the minimum threshold for precipitation.

Another key finding in this study is that the decrease in S_R with increasing cloud thickness is due to a decrease in the susceptibility S_f of drizzle fraction rather than the susceptibility S_I of intensity. In other words, S_R decreases, because for thicker clouds aerosol concentrations have a smaller impact on whether clouds drizzle. However, since the S_I is positive in all cloud thickness bins, S_R remains positive even for the thickest of clouds. Regardless of cloud thickness, increasing aerosol concentrations has the same effect in decreasing the relative intensity of drizzle. Sensitivity tests show that although the averaging lengths used to calculate average drizzle rates change the absolute values of the susceptibility, they do not systematically change the response of the susceptibility with cloud thickness.

Care must be taken, however, if we consider the absolute suppression of precipitation in stratocumulus, since the precipitation susceptibility quantifies only the fractional change in pre-

precipitation due to a change in aerosols. Although higher susceptibilities are reported at lower cloud thicknesses, more precipitation is not necessarily suppressed by aerosols at the lower cloud thicknesses. The mean precipitation rate at cloud base of the quartile with thinnest clouds and the quartile with thickest clouds are 0.04 mm day^{-1} and 3.69 mm day^{-1} respectively. A factor of four decrease of the susceptibility is actually quite small compared to the more than ninety fold increase in mean precipitation rate.

It should also be noted that while this analysis may encompass the range of cloud thicknesses commonly observed in marine stratocumulus, it has only explored a particular subspace of cloud thicknesses and aerosol concentrations occurring in other marine boundary layer clouds. In particular, these results do not inform us about deeper marine boundary layer clouds such as precipitating trade wind cumuli, nor do they help with clouds that are substantially more polluted than the drizzling clouds observed over the southeast Pacific. As noted earlier, this study also excludes more than 20% of the thinnest clouds for which cloud liquid water path or cloud thickness estimates could not be obtained. Satellite observations from Kubar et al. (2009) also imply that the precipitation susceptibility is a function not only of cloud thickness, but also of the aerosol concentration. Exploring these relationships across different datasets in future studies will help us further constrain the effect of aerosols on precipitation susceptibility.

Finally, we stress that the precipitation susceptibility construct is a correlative rather than a causal one. Establishing the causal nature of the observed susceptibilities is critical for their credibility as evidence of aerosol influence on precipitation. Model results such as those by Feingold and Siebert (2009), Sorooshian et al. (2009), and Wood et al. (2009) are helping to do just that, but there remain significant differences in the magnitudes of the precipitation susceptibilities found in these studies. Further, there are still major gaps in our understanding about what these measurements are telling us about how precipitation susceptibility changes with cloud macrophysical properties. Most notably, studies of relatively thick cumulus clouds (liquid water paths greater than 500 g m^{-2}) such as those by Sorooshian et al. (2009) and Jiang et al. (2010) show increases in susceptibility, while this study of much thinner stra-

tocumulus clouds appears to show a monotonic decrease. For thinner clouds, the reduction in precipitation susceptibility with increasing cloud thickness can be understood as a transition from autoconversion-dominated precipitation to accretion-dominated precipitation (Wood et al., 2009). As Jiang et al. (2010) point out, liquid water path is a useful proxy for the ratio of accretion A_{acc} to autoconversion A_{auto} . However, while the results here support the theoretical arguments of Wood et al. (2009) implying that precipitation susceptibility should decrease monotonically with $A_{\text{acc}}/A_{\text{auto}}$, the cloud resolving model results by Jiang et al. (2010) show a maximum in susceptibility at intermediate values of $A_{\text{acc}}/A_{\text{auto}}$ suggesting that this ratio is not a unique predictor of precipitation susceptibility and that other factors may play a role. Seifert and Stevens (2010) suggest that the cloud lifetime (i.e. the time allowed for precipitation to develop) may be one such factor, but we leave this to future studies.

Acknowledgements. TEXT

References

references

- Ackerman, A. S., Kirkpatrick, M. P., Stevens, D. E., and Toon, O. B.: The impact of humidity above stratiform clouds on indirect aerosol climate forcing, *Nature*, 432, 1014–1017, doi:10.1038/nature03174, 2004.
- Albrecht, B. A.: Aerosols, cloud microphysics, and fractional cloudiness, *Science*, 245, 1227–1230, doi:10.1126/science.245.4923.1227, 1989.
- Albrecht, B. A., Fairall, C., Thomson, D., White, A., and Snider, J.: Surface-based remote sensing of the observed and the adiabatic liquid water content, *Geophys. Res. Lett.*, 17, 89–92, doi:10.1029/GL017i001p00089, 1990.
- Allen, G., Coe, H., Clarke, A., Bretherton, C., Wood, R., Abel, S. J., Barrett, P., Brown, P., George, R., Freitag, S., McNaughton, C., Howell, S., Shank, L., Kapustin, V., Brekhovskikh, V., Kleinman, L., Lee, Y.-N., Springston, S., Toniazzo, T., Krejci, R., Fochesatto, J., Shaw, G., Krecl, P., Brooks, B., McMeeking, G., Bower, K. N., Williams, P. I., Crosier, J., Crawford, I., Connolly, P., Allan, J. D., Covert, D., Bandy, A. R., Russell, L. M., Trembath, J., Bart, M., McQuaid, J. B., Wang, J., and

- Chand, D.: South East Pacific atmospheric composition and variability sampled along 20S during VOCALS-REX, *Atmos. Chem. Phys.*, 11, 5237–5262, doi:10.5194/acp-11-5237-2011, 2011.
- Bretherton, C. S., Uttal, T., Fairall, C. W., Yuter, S., Weller, R., Baumgardner, D., Comstock, K., and Wood, R.: The EPIC 2001 stratocumulus study, *B. Am. Meteorol. Soc.*, 85, 967–977, doi:10.1175/BAMS-85-7-967, 2004.
- Bretherton, C. S., Wood, R., George, R. C., Leon, D., Allen, G., and Zheng, X.: Southeast Pacific stratocumulus clouds, precipitation and boundary layer structure sampled along 20S during VOCALS-REX, *Atmos. Chem. Phys.*, 10, 10639–10654, doi:10.5194/acp-10-10639-2010, 2010.
- Brost, R. A., Wyngaard, J. C. and Lenschow, D. H.: Marine Stratocumulus Layers. Part II: Turbulence Budgets, *J. Atmos. Sci.*, 39(4), 818–836, doi:10.1175/1520-0469(1982)039<0818:MSLPIT>2.0.CO;2, 1982.
- Chen, Y.-C., Xue, L., Lebo, Z. J., Wang, H., Rasmussen, R. M. and Seinfeld, J. H.: A comprehensive numerical study of aerosol-cloud-precipitation interactions in marine stratocumulus, *Atmos. Chem. Phys.*, 11(18), 9749–9769, doi:10.5194/acp-11-9749-2011, 2011.
- Comstock, K. K., Wood, R., Yuter, S. E. and Bretherton, C. S.: Reflectivity and rain rate in and below drizzling stratocumulus, *Q. J. Roy. Meteor. Soc.*, 130(603), 2891–2918, doi:10.1256/qj.03.187, 2004.
- Comstock, K. K., Bretherton, C. S., and Yuter, S. E.: Mesoscale Variability and Drizzle in Southeast Pacific Stratocumulus, *J. Atmos. Sci.*, 62(10), 3792–3807, doi:10.1175/JAS3567.1, 2005.
- Duong, H. T., Sorooshian, a and Feingold, G.: Investigating potential biases in observed and modeled metrics of aerosol-cloud-precipitation interactions, *Atmos. Chem. Phys.*, 11(9), 4027–4037, doi:10.5194/acp-11-4027-2011, 2011.
- Efron, B. and Gong, G.: A Leisurely Look at the Bootstrap, the Jackknife, and Cross-Validation, *Am. Stat.*, 37(1), 36, doi:10.2307/2685844, 1983.
- Feingold, G. and Siebert, H.: Cloud - Aerosol Interactions from the Micro to the Cloud Scale, from the Strüngmann Forum Report, *Clouds in the Perturbed Climate System: Their Relationship to Energy Balance, Atmospheric Dynamics, and Precipitation*, edited by: Heintzenberg, J. and Charlson, R. J., MIT Press, ISBN 978-0-262-01287-4, 2009.
- Frisch, A. S., Fairall, C. W. and Snider, J. B.: Measurement of stratus cloud and drizzle parameters in ASTEX with a K_a -band Doppler radar and a microwave radiometer, *J. Atmos. Sci.*, 52(16), 2788–2788, doi:10.1175/1520-0469(1995)052<2788:MOSCAD>2.0.CO;2, 1995.
- Geoffroy, O., Brenguier, J.-L. and Sandu, I.: Relationship between drizzle rate, liquid water path and droplet concentration at the scale of a stratocumulus cloud system, *Atmos. Chem. Phys.*, 8, 4641–4654, doi:10.5194/acp-8-4641-2008, 2008.

- George, R. C. and Wood, R.: Subseasonal variability of low cloud radiative properties over the southeast Pacific Ocean, *Atmos. Chem. Phys.*, 10(8), 4047–4063, doi:10.5194/acp-10-4047-2010, 2010.
- Jiang, H., Feingold, G. and Sorooshian, A.: Effect of Aerosol on the Susceptibility and Efficiency of Precipitation in Warm Trade Cumulus Clouds, *J. Atmos. Sci.*, 67(11), 3525–3540, doi:10.1175/2010JAS3484.1, 2010.
- Kubar, T. L., Hartmann, D. L. and Wood, R.: Understanding the Importance of Microphysics and Macrophysics for Warm Rain in Marine Low Clouds. Part I: Satellite Observations, *J. Atmos. Sci.*, 66(10), 2953–2972, doi:10.1175/2009JAS3071.1, 2009.
- Leith, C. E.: The Standard Error of Time-Average Estimates of Climatic Means, *J. Appl. Meteorol.*, 12(6), 1066–1069, doi:10.1175/1520-0450(1973)012<1066:TSEOTA>2.0.CO;2, 1973.
- Leon, D. C., Wang, Z. and Liu, D.: Climatology of drizzle in marine boundary layer clouds based on 1 year of data from CloudSat and Cloud-Aerosol Lidar and Infrared Pathfinder Satellite Observations (CALIPSO), *J. Geophys. Res.*, 113, doi:10.1029/2008JD009835, 2008.
- Martin, G., Johnson, D. and Spice, A.: The measurement and parameterization of effective radius of droplets in warm stratocumulus clouds., *J. Atmos. Sci.*, 51, 1823–1842, doi:10.1175/1520-0469(1994)051<1823:TMAPOE>2.0.CO;2, 1994.
- Mauger, G. S. and Norris, J. R.: Meteorological bias in satellite estimates of aerosol-cloud relationships, *Geophys. Res. Lett.*, 34(16), 1–5, doi:10.1029/2007GL029952, 2007.
- McComiskey, A. and Feingold, G.: Quantifying error in the radiative forcing of the first aerosol indirect effect, *Geophys. Res. Lett.*, 35(2), 1–6, doi:10.1029/2007GL032667, 2008.
- Nicholls, S. and Leighton, J.: An observational study of the structure of stratiform cloud sheets: Part I. Structure, *Q. J. Roy. Meteor. Soc.*, 112(472), 431–460, doi:10.1063/1.96562, 1986.
- Painemal, D. and Zuidema, P.: Microphysical variability in southeast Pacific Stratocumulus clouds: synoptic conditions and radiative response, *Atmos. Chem. Phys.*, 10, 6255–6269, doi:10.5194/acp-10-6255-2010, 2010.
- Pawlowska, H. and Brenguier, J.-L.: An observational study of drizzle formation in stratocumulus clouds for general circulation model (GCM) parameterizations, *J. Geophys. Res.*, 108(D15), 1–13, doi:10.1029/2002JD002679, 2003.
- Ramanathan, V., Crutzen, P. J., Kiehl, J. T. and Rosenfeld, D.: Aerosols, climate, and the hydrological cycle., *Science*, 294(5549), 2119–24, doi:10.1126/science.1064034, 2001.
- Rotstayn, L. D. and Liu, Y.: A smaller global estimate of the second indirect aerosol effect, *Geophys. Res. Lett.*, 32(5), 4–7, doi:10.1029/2004GL021922, 2005.
- Savic-Jovicic, V. and Stevens, B.: The structure and mesoscale organization of precipitating stratocumu-

- lus, *J. Atmos. Sci.*, 65, 1587–1605, doi:10.1175/2007JAS2456.1, 2008.
- Seifert, A. and Stevens, B.: Microphysical Scaling Relations in a Kinematic Model of Isolated Shallow Cumulus Clouds, *J. Atmos. Sci.*, 67(5), 1575–1590, doi:10.1175/2009JAS3319.1, 2010.
- Seinfeld, J. H. and Pandis, S. N.: *Atmospheric Chemistry and Physics - From Air Pollution to Climate Change*, John Wiley and Sons, Hoboken, New Jersey, 2006.
- Sorooshian, A., Feingold, G., Lebsock, M. D., Jiang, H. and Stephens, G. L.: On the precipitation susceptibility of clouds to aerosol perturbations, *Geophys. Res. Lett.*, 36(13), 1–5, doi:10.1029/2009GL038993, 2009.
- Sorooshian, A., Feingold, G., Lebsock, M. D., Jiang, H. and Stephens, G. L.: Deconstructing the precipitation susceptibility construct: Improving methodology for aerosol-cloud precipitation studies, *J. Geophys. Res.*, 115, D17201, doi:10.1029/2009JD013426, 2010.
- Stevens, B. and Feingold, G.: Untangling aerosol effects on clouds and precipitation in a buffered system., *Nature*, 461(7264), 607–13, doi:10.1038/nature08281, 2009.
- Stevens, B., Cotton, W. R., Feingold, G. and Moeng, C.-H.: Large-Eddy Simulations of Strongly Precipitating, Shallow, Stratocumulus-Topped Boundary Layers, *J. Atmos. Sci.*, 55(24), 3616–3638, doi:10.1175/1520-0469(1998)055<3616:LESOSP>2.0.CO;2, 1998.
- Turton, J. and Nicholls, S.: A study of the diurnal variation of stratocumulus using a multiple mixed layer model, *Q. J. Roy. Meteor. Soc.*, 113(477), 969–1009, doi:10.1256/smsqj.47710, 1987.
- Vali, G., Kelly, R. D., French, J., Haimov, S., Leon, D., McIntosh, R. E. and Pazmany, A.: Finescale Structure and Microphysics of Coastal Stratus, *J. Atmos. Sci.*, 55(24), 3540–3564, doi:10.1175/1520-0469(1998)055<3540:FSAMOC>2.0.CO;2, 1998.
- vanZanten, M. C. and Stevens, B.: Observations of drizzle in nocturnal marine stratocumulus, *J. Atmos. Sci.*, 62, 88–106, doi:10.1175/JAS-3355.1, 2005.
- Wang, S. and Albrecht, B. A.: A stratocumulus model with an internal circulation., *J. Atmos. Sci.*, 43, 2374–2391, doi:10.1175/1520-0469(1986)043<2374:ASMWAI>2.0.CO;2, 1986.
- Wood, R.: Drizzle in Stratiform Boundary Layer Clouds. Part I: Vertical and Horizontal Structure, *J. Atmos. Sci.*, 62(9), 3011–3033, doi:10.1175/JAS3529.1, 2005.
- Wood, R.: Rate of loss of cloud droplets by coalescence in warm clouds, *J. Geophys. Res.*, 111, D21205, doi:10.1029/2006JD007553, 2006.
- Wood, R.: Cancellation of Aerosol Indirect Effects in Marine Stratocumulus through Cloud Thinning, *J. Climate*, 19(9), 1748–1764, doi: 10.1175/JAS3942.1, 2007.
- Wood, R. and Hartmann, D. L.: Spatial Variability of Liquid Water Path in Marine Low Cloud: The Importance of Mesoscale Cellular Convection, *J. Climate*, 19(9), 1748–1764, doi:10.1175/JCLI3702.1,

2006.

- Wood, R., Kubar, T. L. and Hartmann, D. L.: Understanding the Importance of Microphysics and Macro-physics for Warm Rain in Marine Low Clouds. Part II: Heuristic Models of Rain Formation, *J. Atmos. Sci.*, 66(10), 2973–2990, doi:10.1175/2009JAS3072.1, 2009.
- Wood, R., Bretherton, C. S., Leon, D., Clarke, A. D., Zuidema, P., Allen, G., and Coe, H.: An air-craft case study of the spatial transition from closed to open mesoscale cellular convection over the Southeast Pacific, *Atmos. Chem. Phys.*, 11, 2341–2370, doi:10.5194/acp-11-2341-2011, 2011.
- Wood, R., Mechoso, C. R., Bretherton, C. S., Weller, R. A., Huebert, B., Straneo, F., Albrecht, B. A., Coe, H., Allen, G., Vaughan, G., Daum, P., Fairall, C., Chand, D., Gallardo Klenner, L., Garreaud, R., Grados, C., Covert, D. S., Bates, T. S., Krejci, R., Russell, L. M., de Szoeke, S., Brewer, A., Yuter, S. E., Springston, S. R., Chaigneau, A., Toniazzo, T., Minnis, P., Palikonda, R., Abel, S. J., Brown, W. O. J., Williams, S., Fochesatto, J., Brioude, J., and Bower, K. N.: The VAMOS Ocean-Cloud-Atmosphere-Land Study Regional Experiment (VOCALS-REx): goals, platforms, and field operations, *Atmos. Chem. Phys.*, 11, 627–654, doi:10.5194/acp-11-627-2011, 2011.
- Wyant, M. C., Bretherton, C. S., Chlond, A., Griffin, B. M., Kitagawa, H., Lappen, C.-L., Larson, V. E., Lock, A., Park, S., de Roode, S. R., Uchida, J., Zhao, M., and Ackerman, A. S.: A single-column model intercomparison of a heavily drizzling stratocumulus-topped boundary layer, *J. Geophys. Res.*, 112(D24), doi:10.1029/2007JD008536, 2007.
- Yuter, S., Serra, Y., and Houze, R. A., Jr.: The 1997 Pan American climate studies tropical eastern Pacific process study. Part II: Stratocumulus region, *B. Am. Meteorol. Soc.*, 81(3), 483–490, 2000.
- Zuidema, P., Edgeworth, E. R., Fairall, C., and Hazen, D.: Ship-based liquid water path estimates in marine stratocumulus, *J. Geophys. Res.*, 110(D20), doi:10.1029/2005JD005833, 2005.
- Zuidema, P., Leon, D., Pazmany, A., and Cadeddu, M.: Aircraft millimeter-wave retrievals of cloud liquid water path during VOCALS-REx, *Atmos. Chem. Phys. Discuss.*, 11(7), 19581-19616, doi:10.5194/acpd-11-19581-2011, 2011.

Table 1. Mean values of cloud thickness h , PCASP aerosol concentration N , and precipitation rate R in the four cloud thickness bins (h_1, h_2, h_3, h_4 from bin with thinnest to thickest clouds). Total # indicates the total number of points (10 km averages) in each bin, and drizzling # indicates the number of those points where $f > 0$. x_- and x_+ indicate the mean values of x in the population with top and bottom 33 % values of N in each bin.

table

	h_1	h_2	h_3	h_4
Total #	93	93	94	93
Drizzling #	14	50	80	89
mean R (mm day ⁻¹)	0.04	0.12	0.30	3.69
mean h (m)	153	248	335	505
Min/max h (m)	14 / 207	208 / 294	294 / 379	381 / 792
N_- (cm ⁻³)	69	85	77	39
N_+ (cm ⁻³)	357	419	411	266
R_- (mm day ⁻¹)	0.12	0.28	0.52	7.1
R_+ (mm day ⁻¹)	0.00068	0.049	0.11	1.7
Drizzling # ₋	10	21	29	30
Drizzling # ₊	3	12	23	29

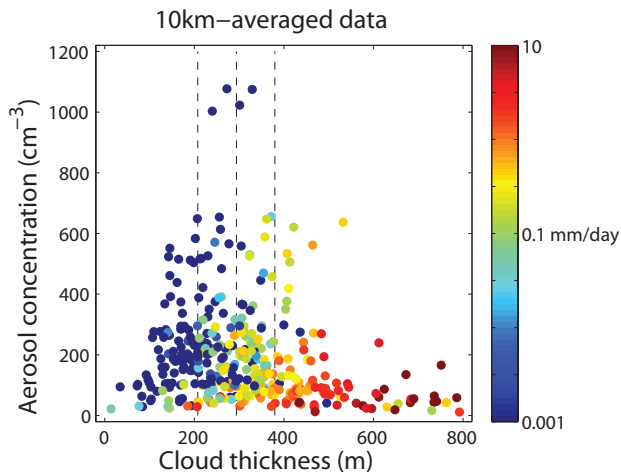


Fig. 1. Distribution of 10 km-averaged aerosol concentration and cloud thickness measurements taken during twelve research flights. Colors indicate cloud base precipitation rates derived from the cloud base reflectivities that were measured by the Wyoming Cloud Radar. Dotted lines indicate the boundaries of the cloud thickness bins described in the text.

figure

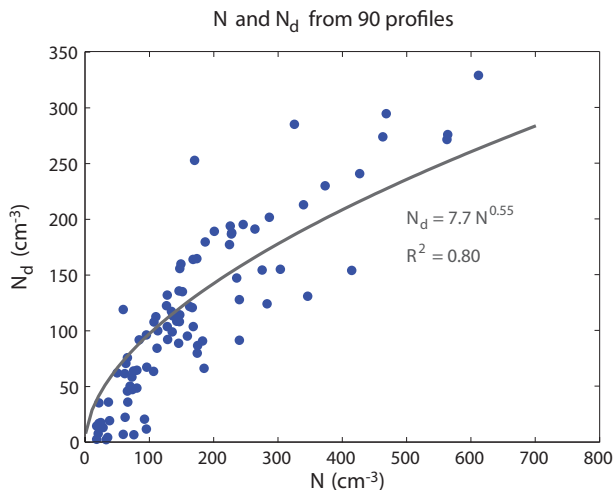


Fig. 2. PCASP measured aerosol concentration N and CDP measured cloud droplet concentration N_d from 90 profiles. N is the mean PCASP concentration at altitudes below 400 m and N_d is the mean CDP concentration (filtered for incloud). The gray line indicates the line of best fit, determined by the method described in Section 3.2.

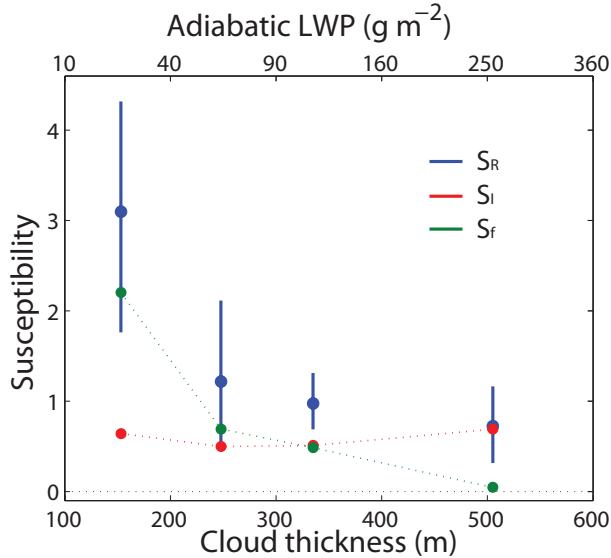


Fig. 3. The precipitation susceptibility (blue), the susceptibility of drizzle intensity (red), and the susceptibility of drizzle fraction (green) are calculated in each of the four equally weighted cloud thickness bins. Error bars on the precipitation susceptibility indicate the 95 % confidence interval calculated using bootstrap resampling technique. The dotted lines exist only to identify points that lie on top of each other. The precipitation susceptibility (blue) is approximately the sum of the susceptibility of drizzle fraction (green) and the susceptibility of drizzle intensity (red). The adiabatic liquid water path axis on the top is estimated from cloud thickness values, using the relationship $LWP = \frac{1}{2} C_w h^2$, where $C_w = 2 \times 10^{-6} \text{ kg m}^{-4}$ (Pawlowska and Brenguier, 2003).

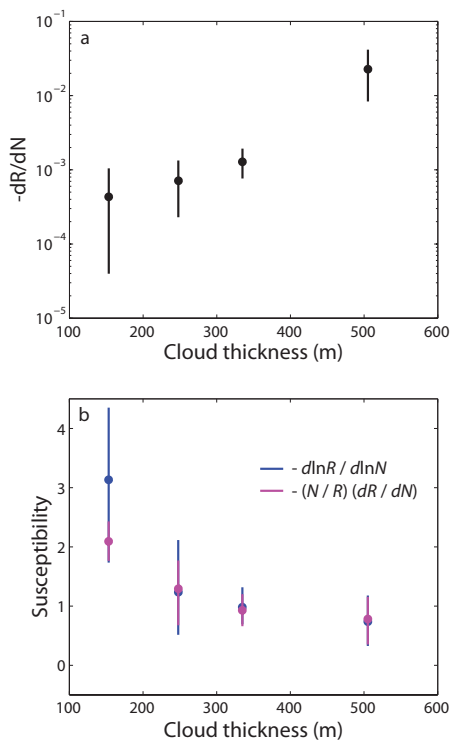


Fig. 4. a) $-dR/dN$ calculated in the four cloud thickness bins, as described in the Sec. 3.3. b) The susceptibility calculated using the original formulation of susceptibility (blue) and the susceptibility calculated using the formulation in Eq. 6 (magenta).

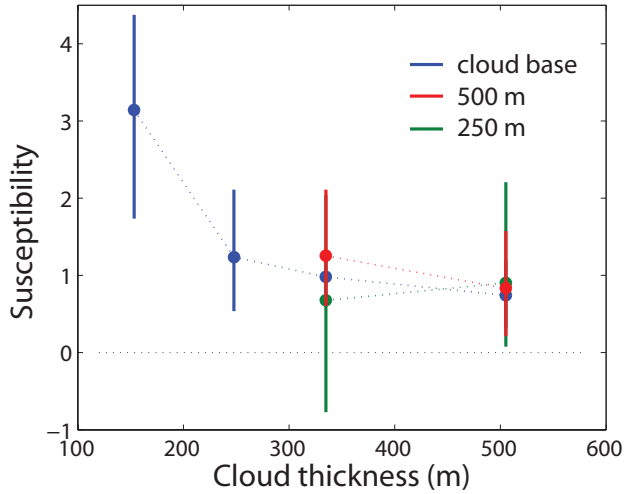


Fig. 5. The precipitation susceptibilities of drizzle at cloud base (blue), at 500 m (red), and at 250 m (green) are calculated. Precipitation rates are averaged over 10 km segments. While the same Z - R relationship ($Z = 1.04R^{1.3}$) from (Comstock et al., 2004) is used for cloud base precipitation rates and 500 m precipitation rates, a different Z - R relationship ($Z = 2.38R^{1.1}$) is used for the 250 m precipitation rates.

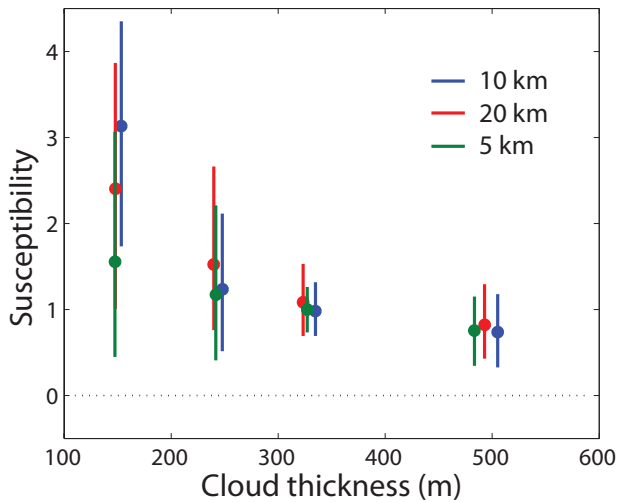


Fig. 6. The precipitation susceptibility are calculated in four equally weighted cloud thickness bins for three different averaging length scales: 10 km (blue), 20 km (red), 5 km (green).

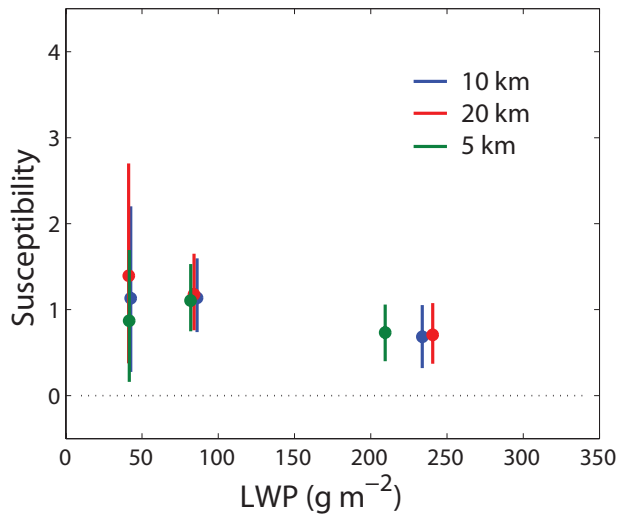


Fig. 7. The precipitation susceptibility is calculated in four equally weighted cloud liquid water path (LWP) bins for three different averaging length scales: 10 km (blue), 20 km (red), 5 km (green).

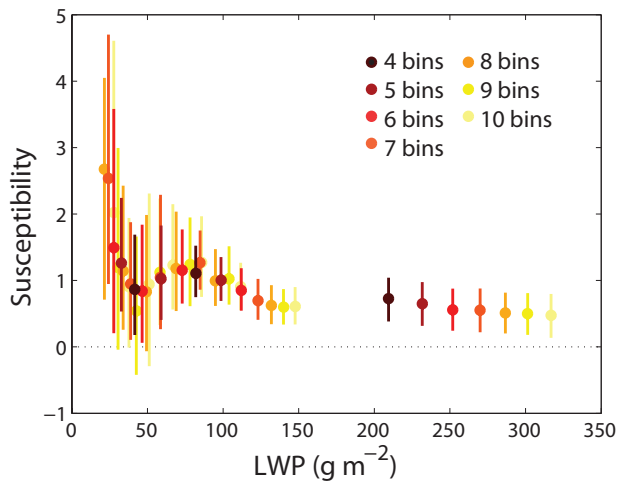


Fig. 8. The precipitation susceptibility is calculated in equally weighted cloud liquid water path (LWP) bins when the data are averaged over 5 km segments. Different shades denote the number of bins.

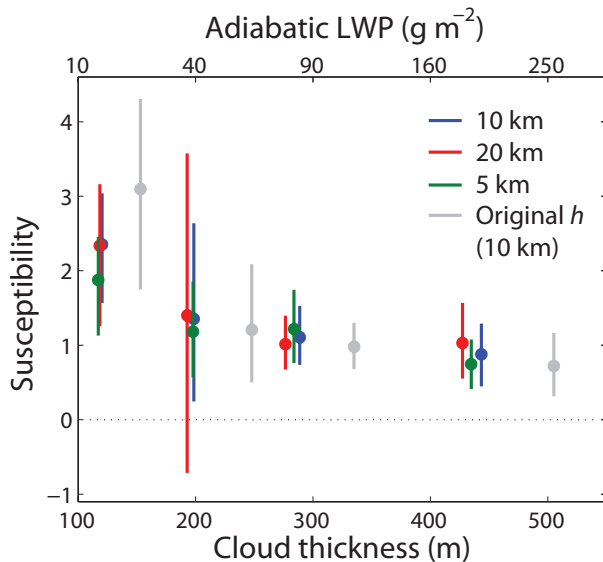


Fig. 9. After combining LWP retrievals from the GVR and the cloud thickness measurements from the WCL and WCR into an integrated cloud thickness dataset, the precipitation susceptibility are calculated in four equally weighted cloud thickness bins for three different averaging length scales: 10 km (blue), 20 km (red), 5 km (green). The original precipitation susceptibility using cloud thickness measurements derived from the WCL/WCR is shown in gray for comparison.