

# Cloud Processing of Aerosol

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Measurements from two aircraft field projects;

Marine Stratus/Stratocumulus Experiment (MASE)

Ice in Clouds Experiment-Tropical (ICE-T)

polluted stratus off the central California coast in  
July 2005

small low altitude cumuli of the eastern Caribbean  
in July 2011

DOE Gulfstream 1 airplane

NCAR C-130 airplane

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DRI high resolution CCN spectrometers have revealed bimodal spectra caused by within cloud processes (Hoppel et al. 1985) by **chemical** or **physical** transformations (coalescence or Brownian capture of small interstitial material by cloud droplets).

These enlarge (decrease  $S_c$ ) CCN dissolved within activated cloud droplets while unactivated particles that did not produce cloud droplets remain the same size and  $S_c$ .

After evaporation of cloud water, which happens to most clouds, there is a size separation gap (**Hoppel minimum**) that defines a bimodal spectrum from a precloud unimodal aerosol.

The two modes are **processed** (accumulation) and **unprocessed** (Aitken) (H15).

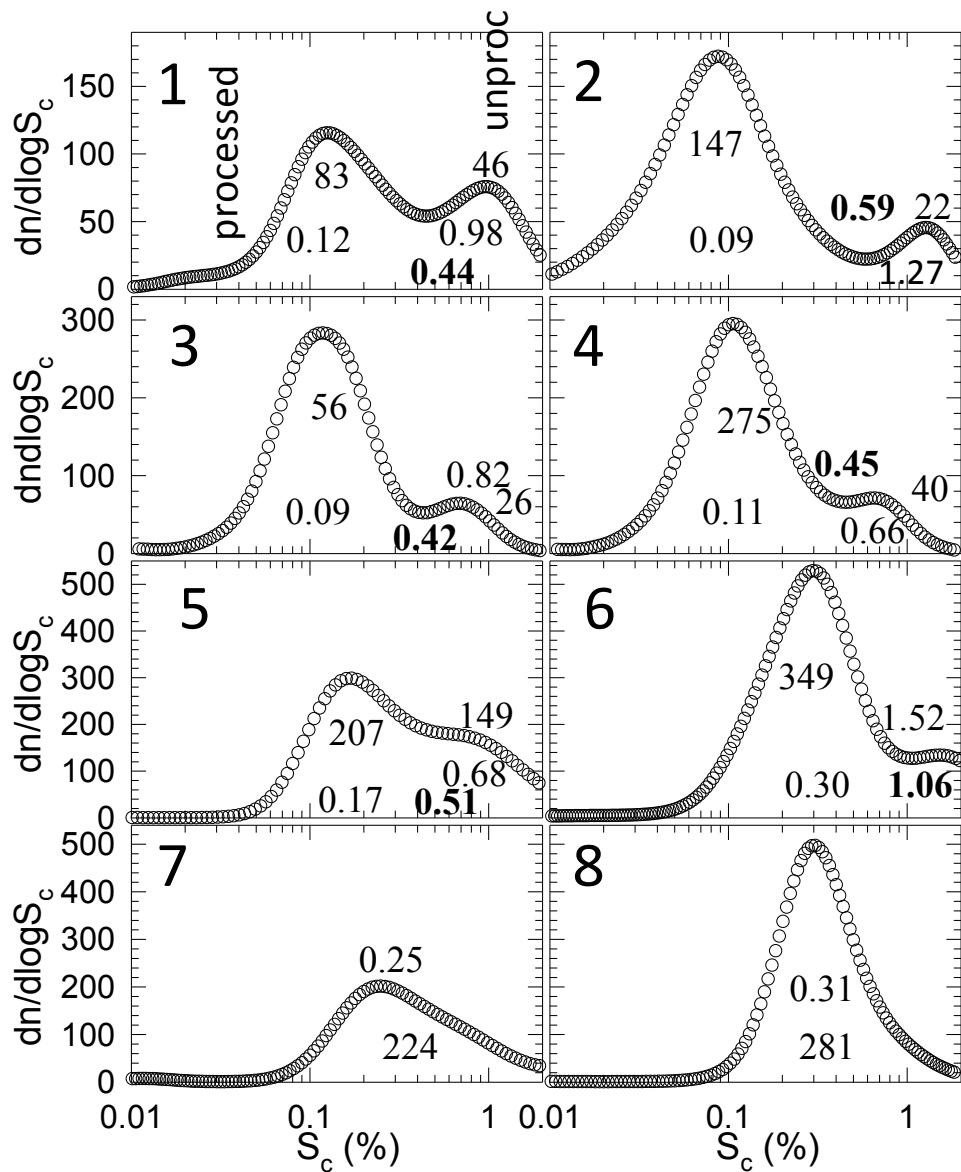
Hudson, J.G., S. Noble, and S. Tabor, 2015: Cloud supersaturations from CCN spectra Hoppel minima. *J. Geophys. Res., Atmos.*, 120, Issue 8, 27 April, 3436–3452, doi:10.1002/2014JD022669.

Variability of CCN bimodality was quantified on a 1-8 scale.

Unqualified bimodal spectra designated 1.

Strictly unimodal spectra designated 8.

**Intermediate designations for spectra with unequal or un-separated modes;**  
i.e., one mode a shoulder of another.

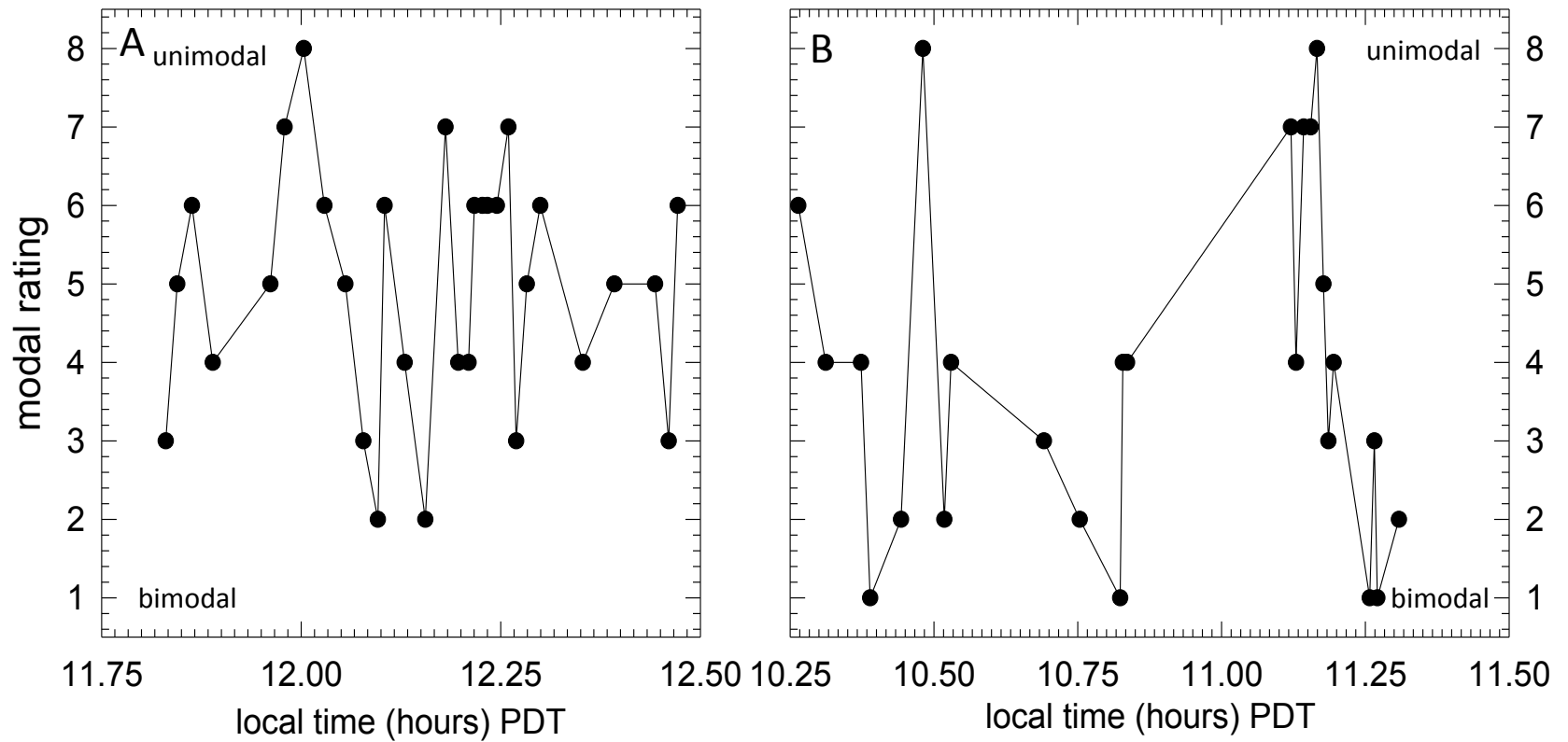


**Fig. 1.** Differential CCN concentrations (per  $\text{cm}^3$ ) against critical supersaturation ( $S_c$ ) for ICE-T below cloud examples of the 8 modal ratings.  $S$  in percent for Hoppel minima (**bold**), the modes of the processed distributions (lower  $S_c$ ) and the unprocessed distributions (higher  $S_c$ ) are quantified in 1-5. The unimodal modes ( $S_m$ ) are quantified in 6-8.  $N_{CCN}^P$  ( $N_p$ ),  $N_{CCN}^U$  ( $N_u$ ) and  $N_{CCN}^M$  ( $N_m$ ) in number per  $\text{cm}^3$  are also displayed. Revision of Fig. 3 of Hudson et al. (2015).

Subjective categorization; but we also have 2 objective categorizations:

$N_u - N_p$ ; high numbers unimodal, low or negative numbers bimodal

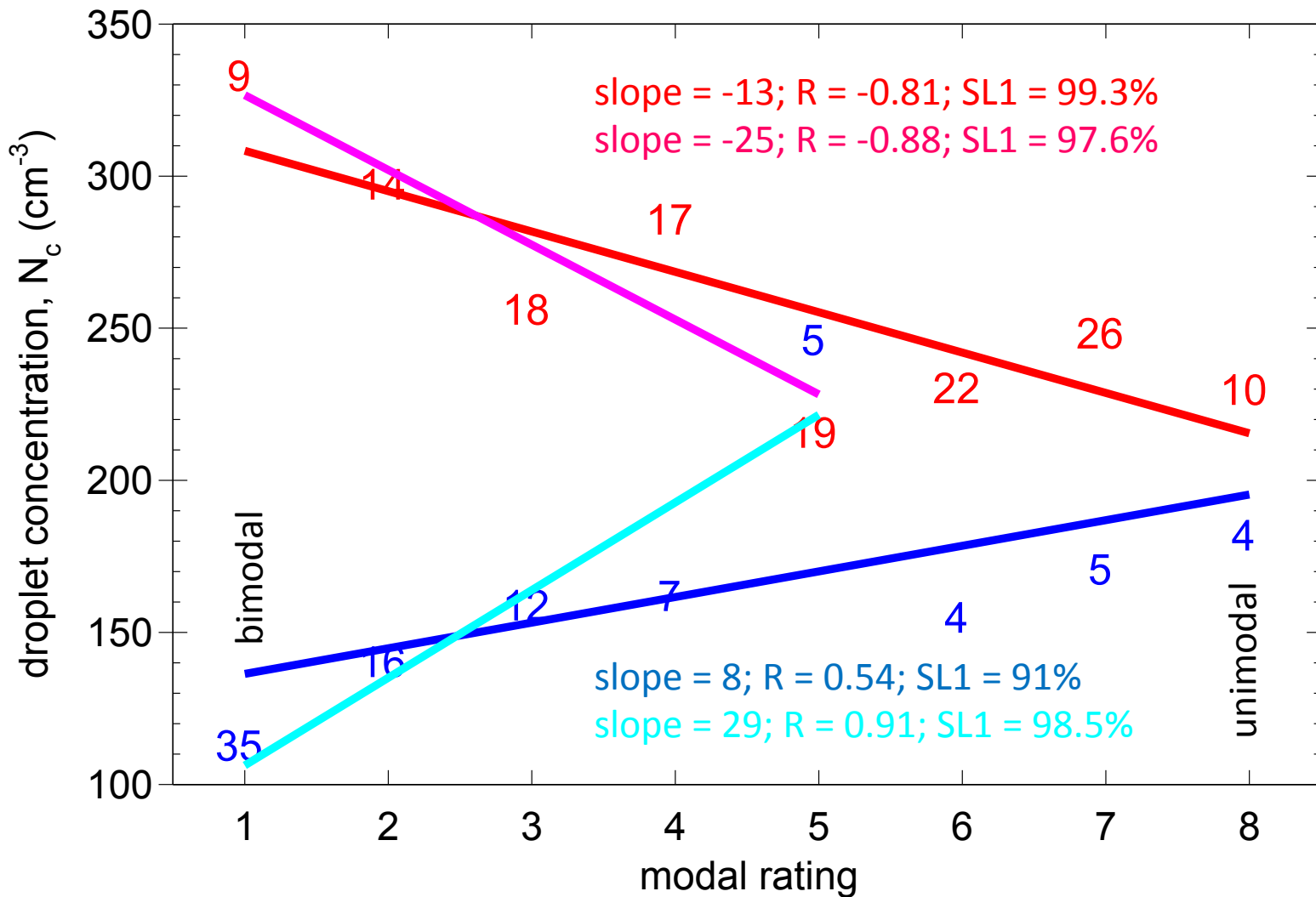
Classified  $N_u - N_p$ ; sort  $N_u - N_p$  into 8 groups of equal cases; lowest group 1 unimodal, highest group 8 bimodal; reduces extremes



**Fig. 2.** Time plots of CCN spectral modality under MASE stratus; (a) 18 July, (b) 23 July. Minor tick marks are minutes. From Hudson et al. (2015)

Aerosol/CCN modality varies over short time intervals.

Compared CCN  
spectral modality  
with microphysics of  
the closest clouds



**Fig. 3.** Mean cloud droplet concentrations,  $N_c$ , against closest CCN modal rating for **MASE** (red and pink; polluted stratus) and **ICE-T** (blue and cyan; Caribbean cumuli). Correlation coefficients (R) and one-tailed significance levels (SL1) are shown. **Pink** and **cyan** regressions consider only modes 1-5 that have Hoppel minima. Data points are plotted as numerals showing the numbers of cases. From Hudson et al. (2015).



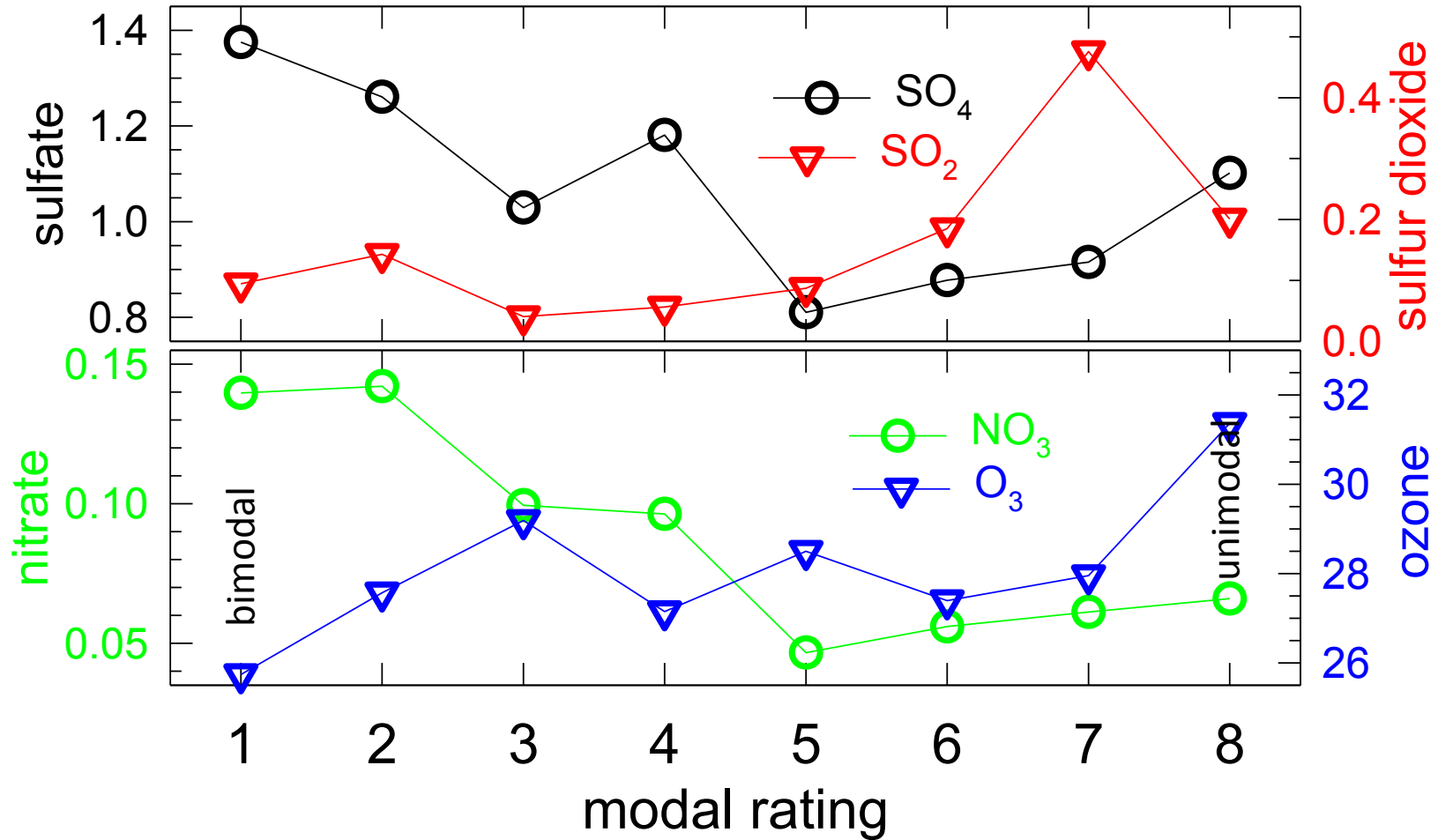
More **chemical processing** by gas-to-particle conversion within droplets (especially sulfate but also nitrate) and **Brownian capture in MASE polluted stratus**.

Lower cloud supersaturations ( $S$ ) because of lower vertical wind,  $W$ , and higher  $N_{\text{CCN}}$  leave many interstitial particles that can be Brownian captured by cloud droplets.

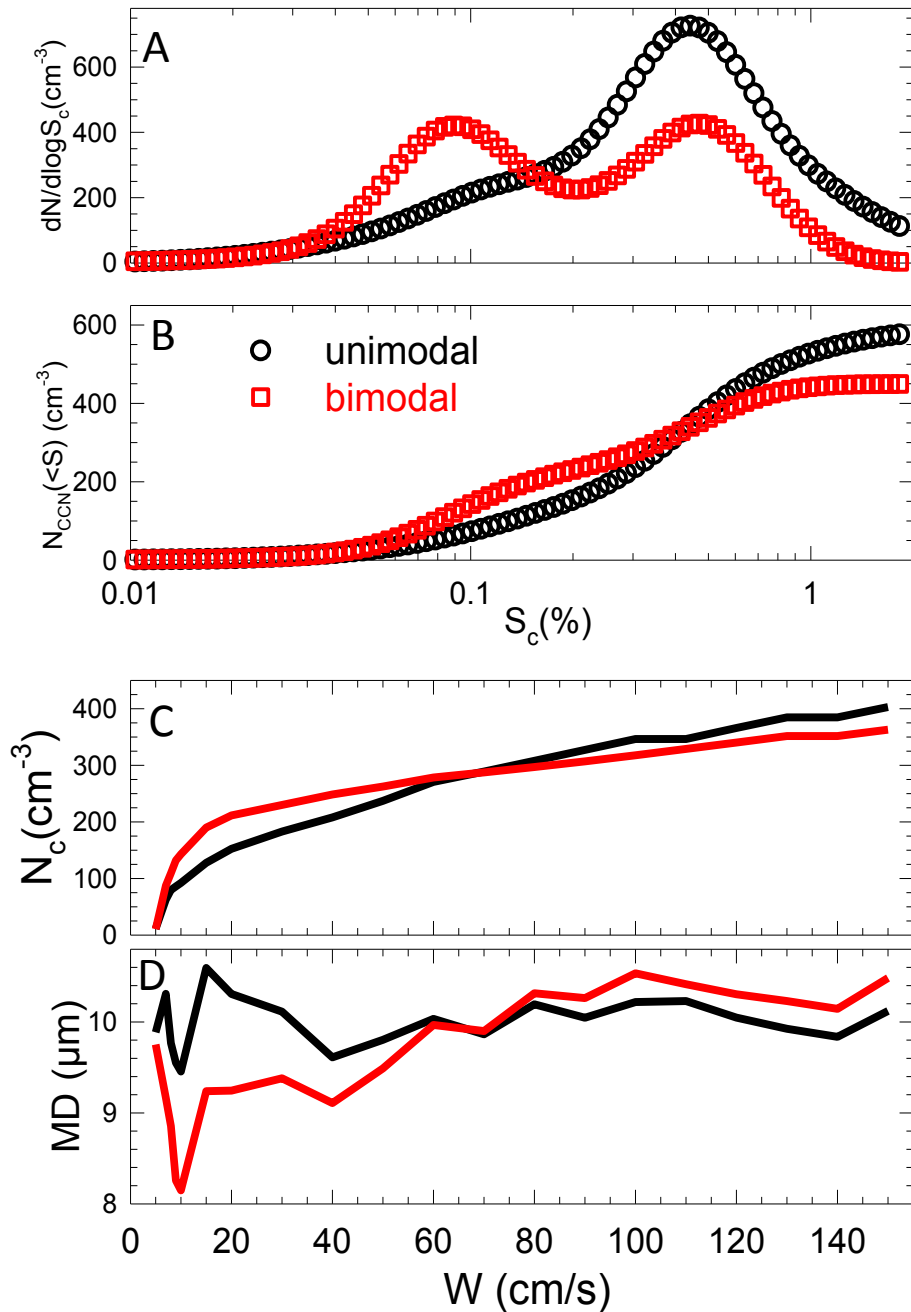
More **coalescence** processing in thicker higher LWC ICE-T cumuli, which have larger droplets.  
Higher  $W$  and  $S$  leave few interstitial CCN.

Hudson et al. (2015)

# MASE



**Fig. 4.** Sulfur dioxide conversion to sulfate; nitrogen oxides to nitrate; ozone aides reactions

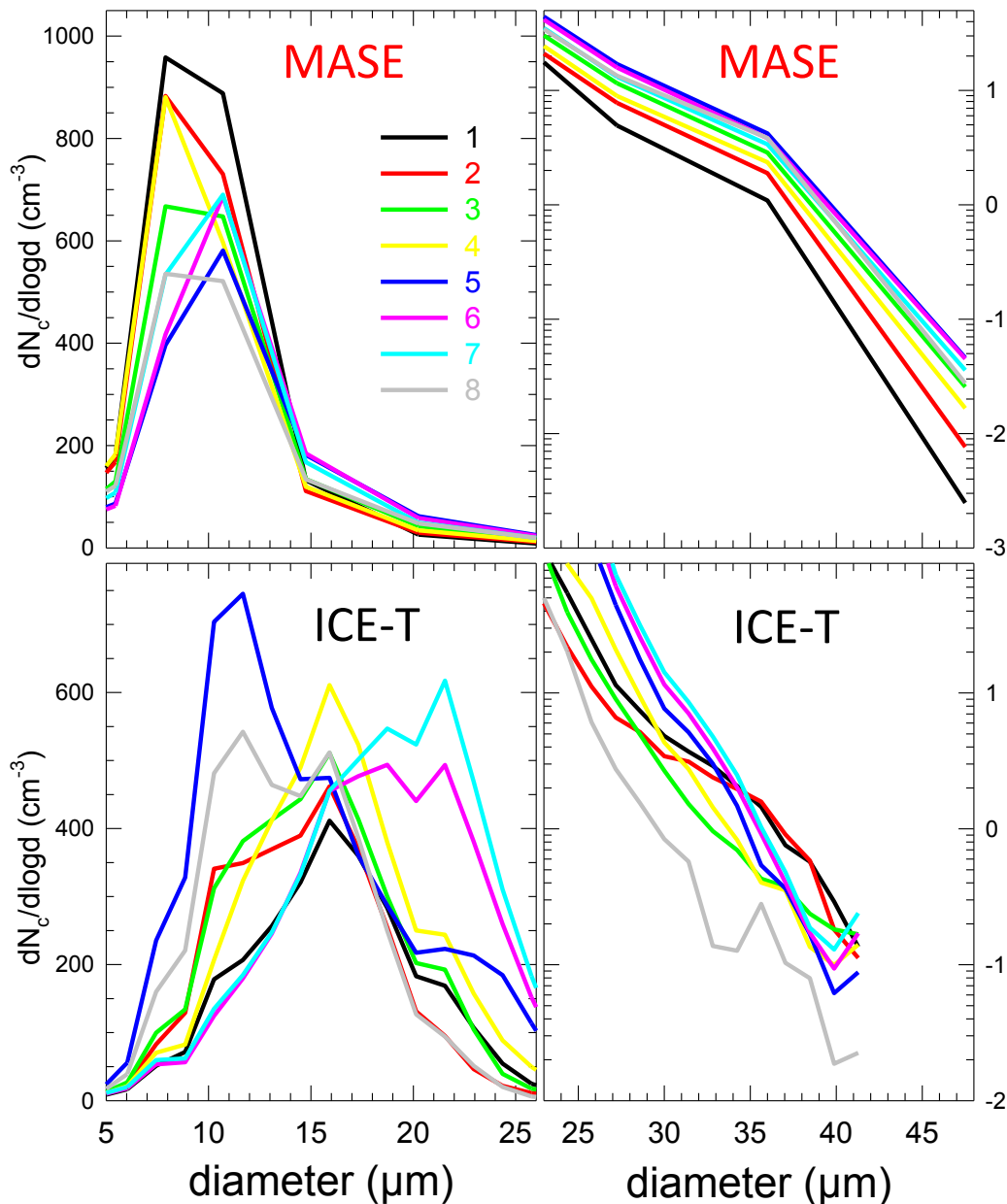


**Fig. 5.** (a) and (b) are measured CCN spectra during MASE. (a) is differential and (b) is cumulative presentation. (c and d) are cloud microphysics predictions by an adiabatic model (Robinson 1984) that applies various  $W$  to these two CCN spectra. (c) Shows cloud droplet concentrations. ( $N_c$ ) (d) is the mean diameter of the cloud droplet spectra at the same distances above cloud base for clouds grown on each CCN spectrum.

At low  $W$  typical of **stratus**, droplet concentration,  $N_c$ , is greater for **clouds grown on bimodal CCN** and droplet mean diameter (MD) is smaller.

At high  $W$  typical of **cumuli**  $N_c$  is greater for **clouds grown on unimodal CCN** and MD is smaller.

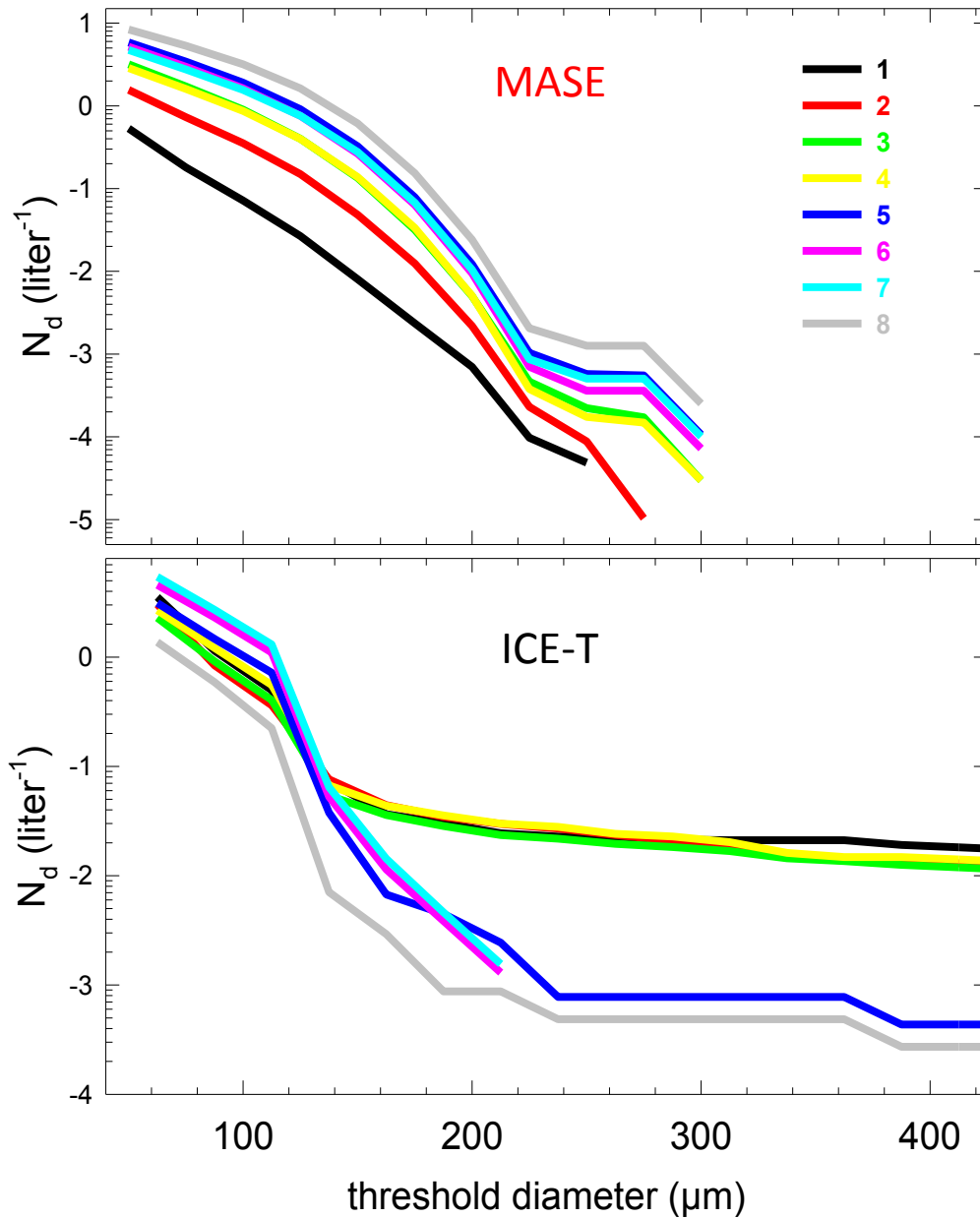
Similar results are found for ICE-T CCN spectra.



**Fig. 6.** Differential  $N_c$  against diameter for the various modal ratings noted by the legend colors 1-8. Right side labels show only exponents of the common-log scale.

In MASE small droplets are more numerous in clouds grown on bimodal CCN (lower ratings; black, red, green), but larger droplets are more numerous in clouds grown on unimodal CCN (higher ratings; gray, cyan, pink).

In ICE-T small droplets are more numerous in clouds grown on unimodal CCN (higher ratings; gray, cyan, pink), but larger droplets are more numerous in clouds grown on bimodal CCN (lower ratings; black, red, green).



**Fig. 7.** Mean cumulative drizzle drop concentrations ( $N_d$ ) larger than various threshold diameters for various modal ratings denoted by the legend colors 1-8.

In MASE stratus there is more drizzle in clouds grown on unimodal CCN (high ratings; gray, cyan, pink, blue).

In ICE-T cumuli there is more drizzle in clouds grown on bimodal CCN (low ratings; black, red, green, yellow).

Similar results for drizzle LWC

Opposite results in these two projects

CCN modality has opposite cloud microphysics relationships in two different cloud systems.

Thus, cloud processing enhances both components of the indirect aerosol effect (IAE) in polluted stratus,

Cloud processing reduces both IAE in cumuli.

If cloud processing continues to show importance then trace gas concentrations (sulfur dioxide, ozone, hydrogen peroxide, etc.), emissions and predictions will be as necessary as CCN measurements and predictions for assessing IAE.

Therefore, it would be necessary to learn more about natural versus relevant anthropogenic trace gases and their distributions relative to cloud fields that are most significant for climate; i.e., mainly marine stratus that largely control IAE, and where chemical processing dominates.

This would have to include better assessments of the relative anthropogenic/natural ratios of the relevant trace gases. Not just anthropogenic/natural ratios of CCN

These cloud processing effects are above and beyond the effects of different CCN concentrations ( $N_{\text{CCN}}$ ) caused for instance by anthropogenic sources that cause the original indirect aerosol effect (IAE).

However, further cloud processing investigations might reveal the opposite;

**Cloud processing makes the best CCN**, lowest  $S_c$ .  
Cloud processing makes most of the accumulation mode.

$S$  is lower in stratus due to lower vertical wind.

Polluted status  $S$  is suppressed by high  $N_{\text{CCN}}$ .

Thus, **only a small subset of CCN** (lowest  $S_c$ ) are activated and thus cloud processed.



In subsequent cloud cycles slightly higher  $S_c$  CCN thus are made irrelevant.

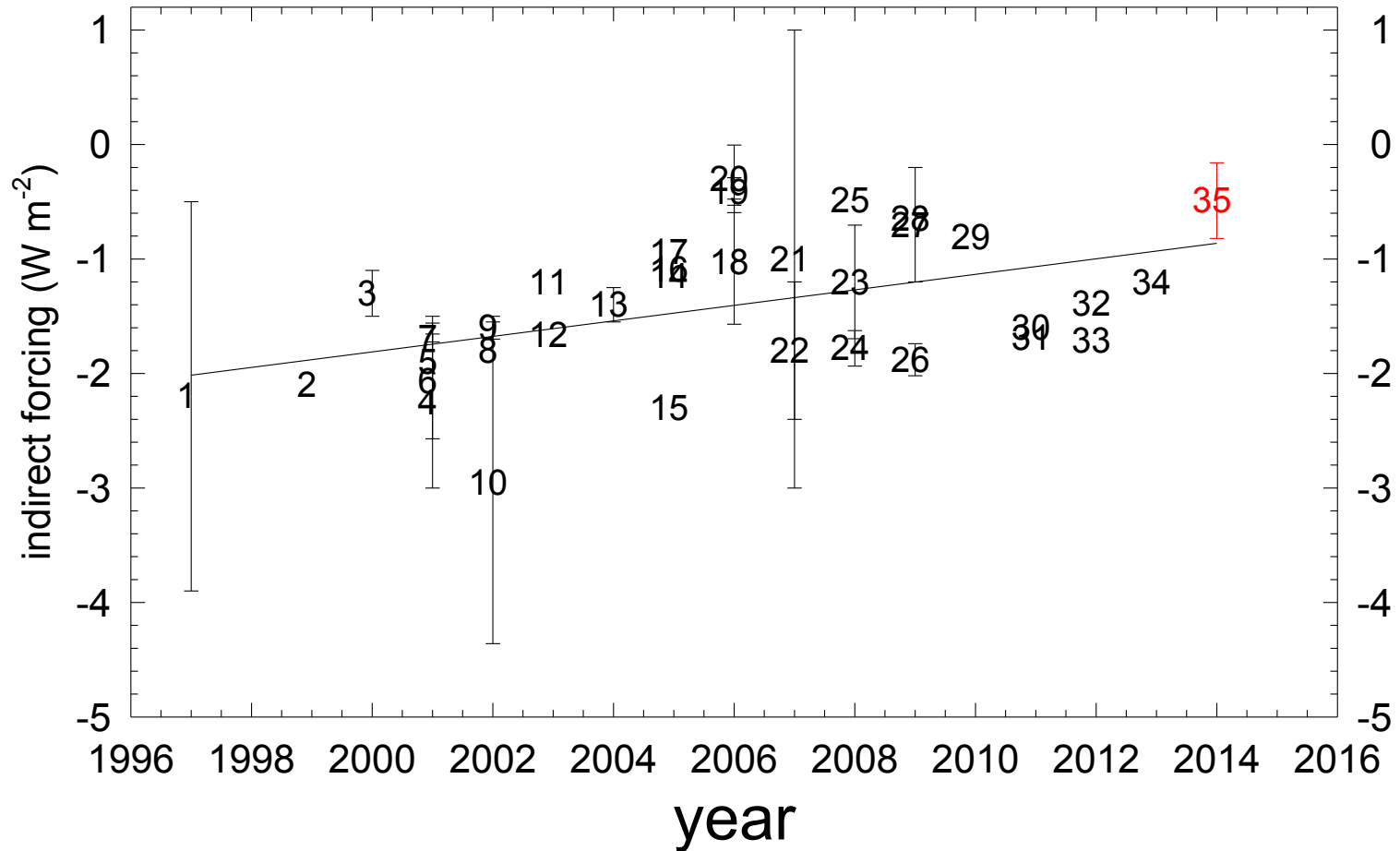
With the cloud processing improvement, in the next cycle not even all of the previously processed CCN make droplets. This further suppresses cloud  $S$ ; and this makes irrelevant even more higher  $S_c$  CCN.

Anthropogenic CCN tend to have high  $S_c$ .

Many high  $S_c$  CCN are lost to Brownian scavenging by cloud droplets. “Clouds are the vacuum cleaners of the atmosphere” (Volker Mohnen, circa 1985)

Cloud processing then tends to reduce IAE.

Cloud processing could therefore constitute IAE buffering (Feingold and Stephens, 2006).



**Fig. 8.** IAE estimates mostly from the 5<sup>th</sup> IPCC report. 1. Lohmann and Feichter (1997); 2. Rotstayn (1999); 3. Lohmann et al. (2000); 4. Ghan et al. (2001); 5. Jones et al. (2001); 6. Rotstayn and Penner (2001); 7. Williams et al. (2001); 8. Kristjansson (2002); 9. Lohmann (2002); 10. Menon et al. (2002); 11. Peng and Lohmann (2003); 12. Penner et al. (2003); 13. Easter et al. (2004); 14. Kristjansson et al. (2005); 15. Ming et al. (2005); 16. Rotstayn and Liu (2005); 17. Takemura et al. (2005); 18. Penner et al. (2006); 19. Quaas et al. (2006); 20. Storelvmo et al. (2006); 21. Menon and Del Genio (2007); 22. Ming et al. (2007); 23. Kirkevåg et al. (2008); 24. Seland et al. (2008); 25. Storelvmo et al. (2008); 26. Hoose et al. (2009); 27. Quaas et al. (2009); 28. Rotstayn and Liu (2009); 29. Storelvmo et al. (2010); 30. Ghan et al. (2011); 31. Penner et al. (2011); 32. Makkonen et al. (2012); 33. Takemura (2012); 34. Kirkevåg et al. (2013); 35. Chen et al. (2014) satellite.

IAE estimates have decreased over the years

Satellite measurements and other observations do not see as much IAE as predicted by models.

Stevens, B., 2013: Aerosols: Uncertain then, irrelevant now. *Nature*, 503, 47.

**“Uncertainty in estimates of the effects of aerosols on climate stems from poor knowledge of the past, pristine atmosphere — so getting a better understanding of these effects might not be as useful as was thought.”**

Clouds make the  
best CCN