

# Towards Constraining the Aerosol-Cloud Radiative Effect

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## Key Questions:

- Can top-down approaches to quantifying aerosol-cloud radiative effect be used in concurrence with bottom-up ACI metrics to constrain aerosol-cloud radiative forcing?
- How does the relative Cloud Radiative Effect-Albedo-Cloud fraction ( $rCRE-A-f_c$ ) relationship vary under different meteorological conditions and aerosol concentrations?
- What controls the  $rCRE-A-f_c$  relationship?
- Do  $rCRE-A-f_c$  relationships vary as a function of resolution or scale?

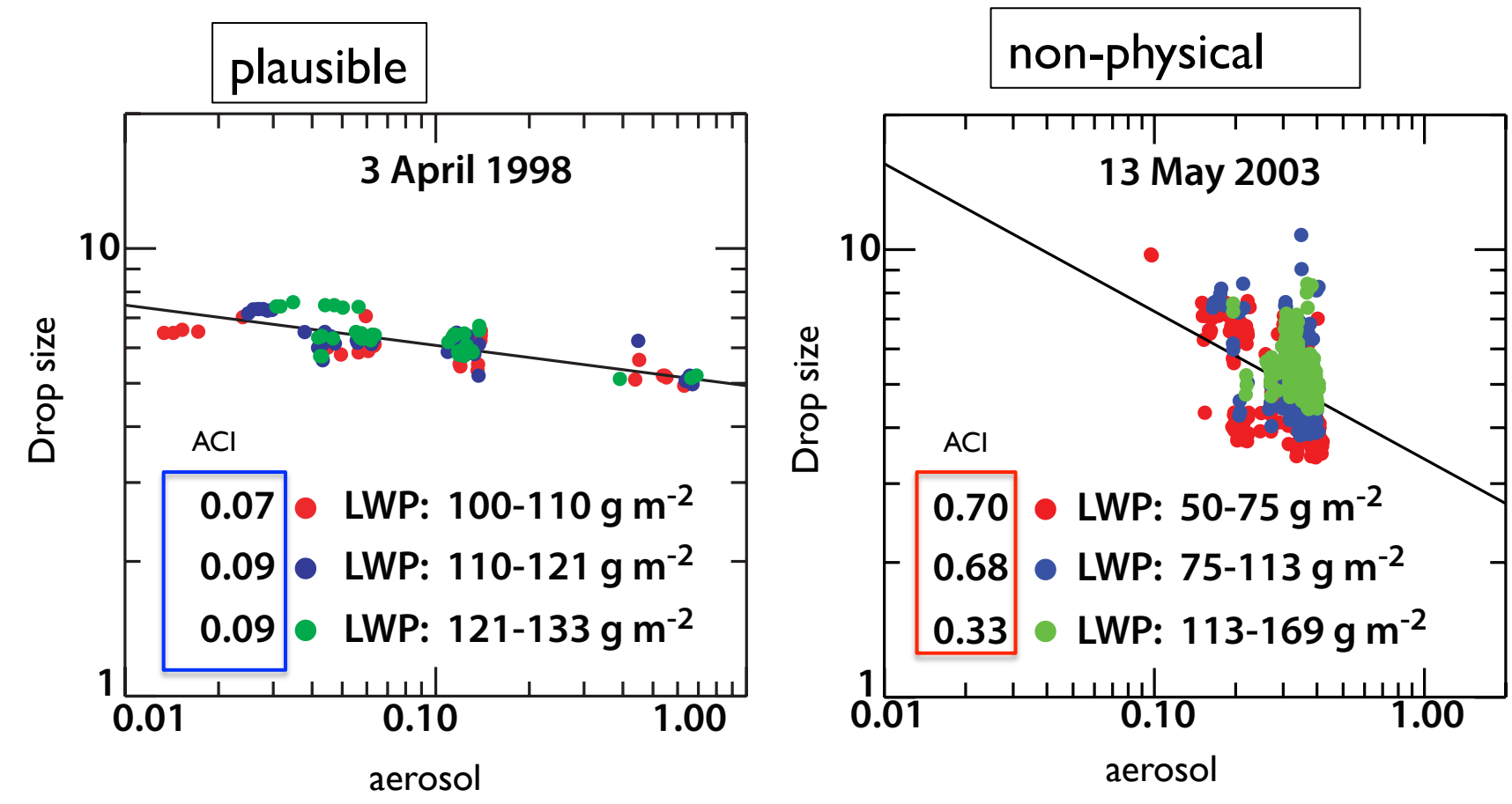
## Key Results:

- $rCRE-A-f_c$  relationships are remarkably robust;
- Co-variability of meteorology and aerosol influences detectability of aerosol effects on cloud and shape of  $A, f_c$  relationship!
- Strong case for routine LES accompanying routine observations (LASSO) for understanding interconnected aerosol and meteorological processes and their influence on  $rCRE$ .

## I. Approaches to quantifying ACI in cloud systems

### The Bottom-up Approach:

Quantify cloud microphysical responses to aerosol perturbations



$$\Delta R = R \frac{\Delta \ln R}{R} \frac{\Delta \ln \tau}{\tau} \frac{\Delta \ln N_d}{N_d} \frac{\Delta \ln CCN}{CCN} \frac{\Delta \ln E}{E}$$

$$\frac{d \ln \tau}{d \ln N_d} = \frac{1}{3} \left[ 1 + 2 \frac{d \ln L}{d \ln N_d} + \frac{d \ln k}{d \ln N_d} + 3 \frac{d \ln H}{d \ln N_d} \right]$$

Caution!! Uncertainties compound

Quantifying these terms is a challenge

- Measurement errors, scale dependence, etc.
- Experience shows that ACI metrics do not constrain the aerosol-cloud radiative effect

### The Top-down Approach:

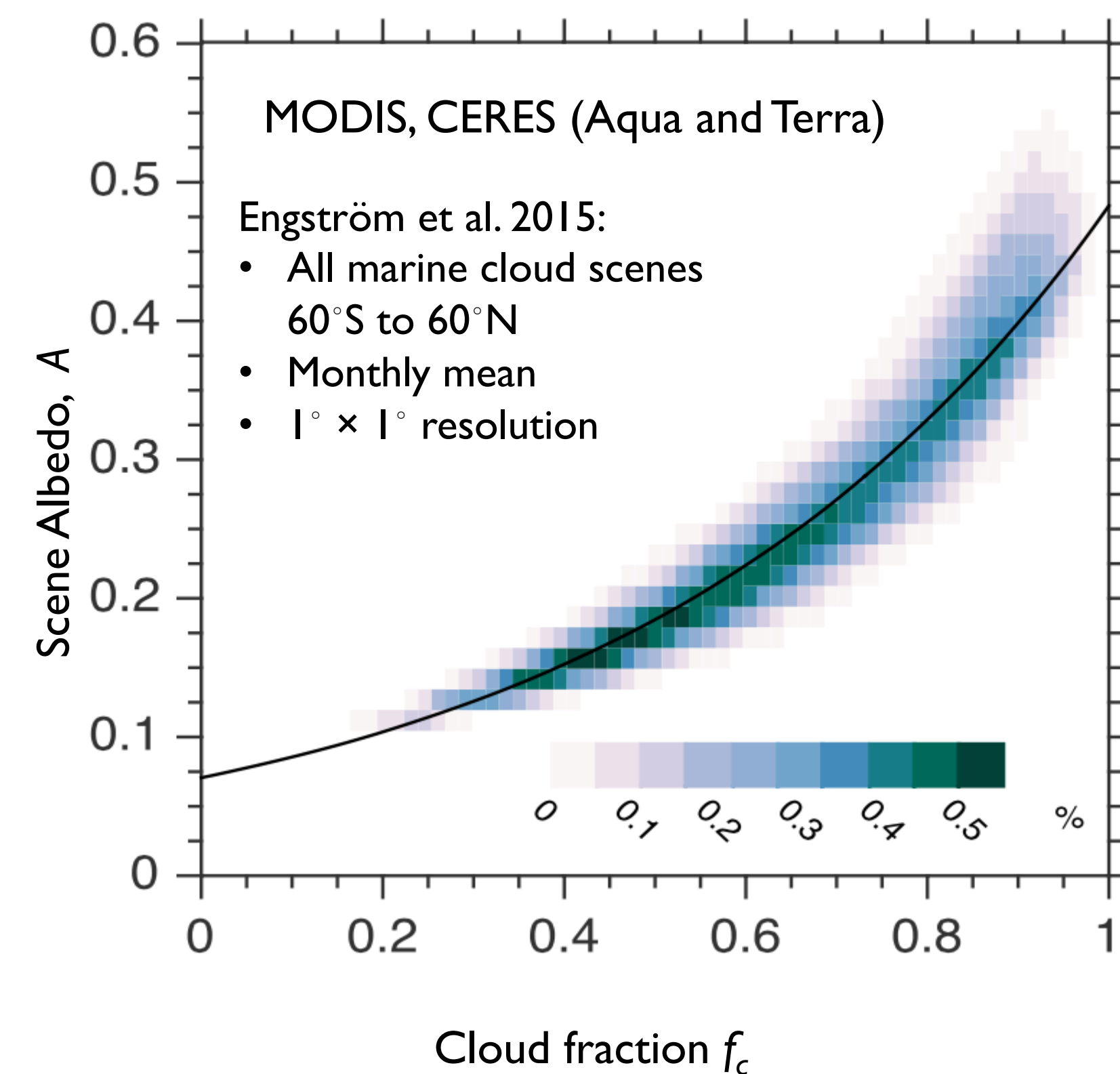
Quantify Radiative and cloud macroscale properties

$$A = A_c f_c + A_s (1 - f_c)$$

**Cloud field Properties:**  
 Cloud fraction,  $f_c$   
 Liquid water path,  $L$   
 Optical depth,  $\tau$   
 Cloud albedo,  $A_c$   
 Cloud depth,  $H$   
 Relative cloud radiative forcing,  $rCRE$

$$rCRE = 1 - \frac{F_{sw,all}}{F_{sw,clr}}$$

$F_{sw}$  = downwelling shortwave flux



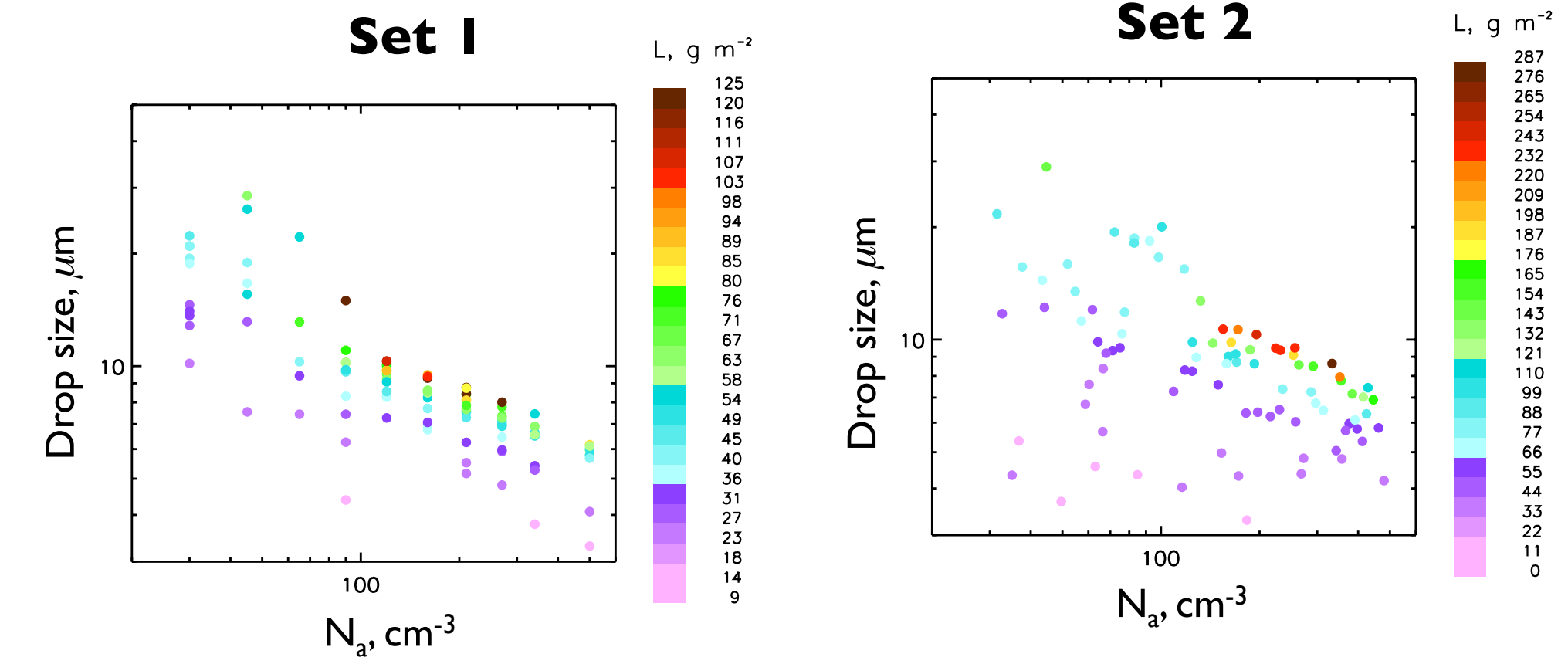
## 2. The importance of co-variability between meteorology and aerosol

**Example from LES:** Two sets of simulations (~100) differing only in co-variability of initial meteorology and aerosol conditions

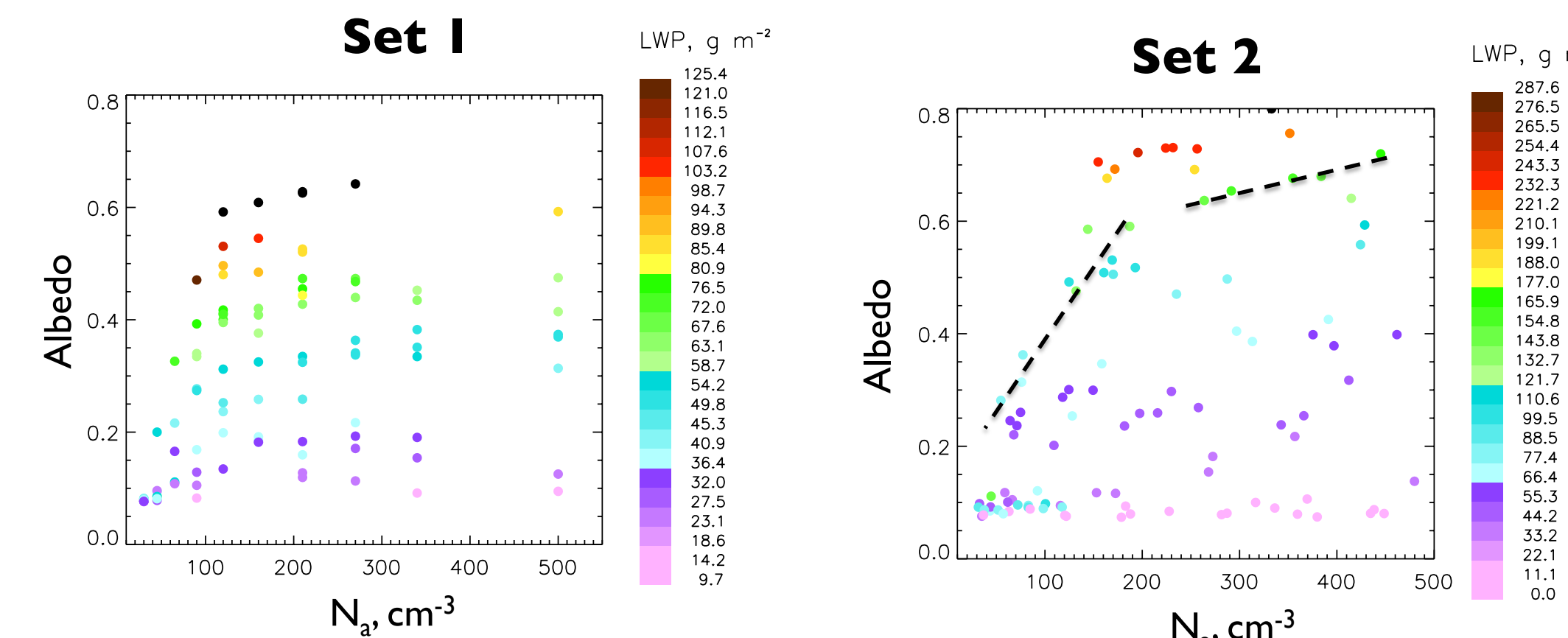
**Set 1:** regular grid spacings in 6-D parameter space; many runs vary N for fixed met conditions

**Set 2:** Latin Hypercube Sampling of 6-D parameter space; maximizes minimum distance between parameters for optimal coverage

### Bottom-up Approach



As expected, higher N → smaller drops at constant L

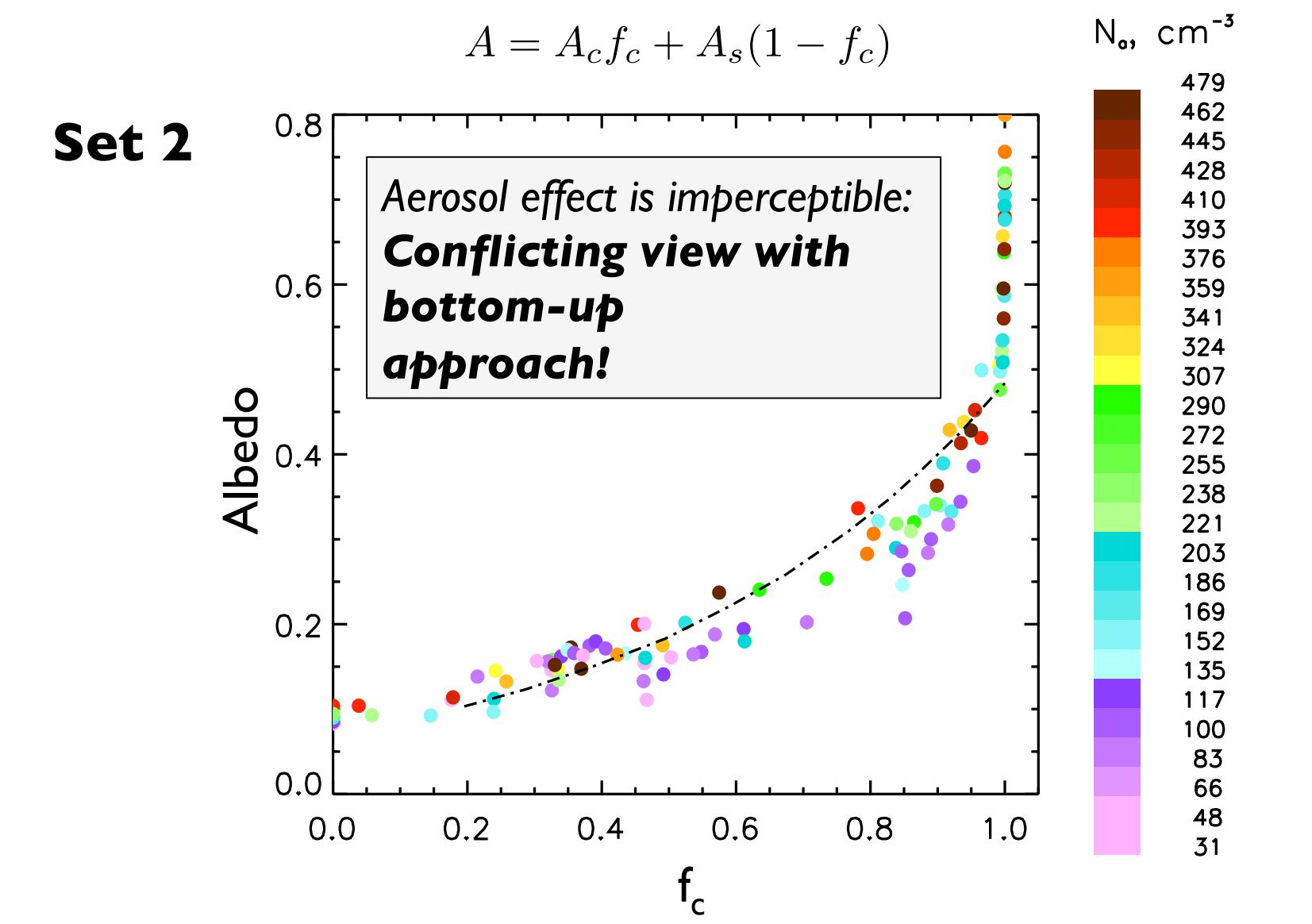
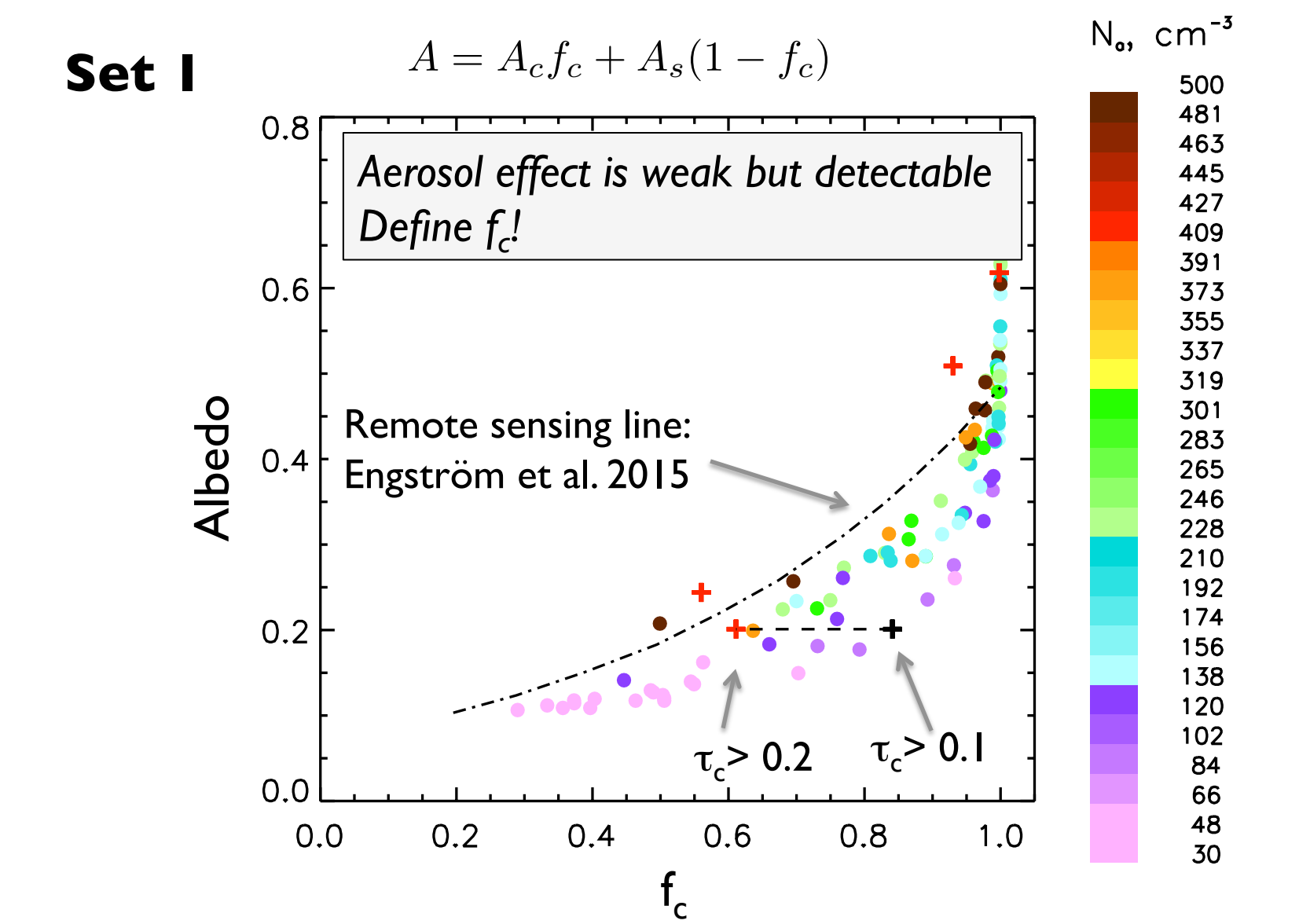


Both sets show the expected Albedo Susceptibility response  
 Higher susceptibility at smaller N

Feingold, McComiskey, Yamaguchi et al. PNAS, 2016

$$S_a = \frac{\partial A}{\partial N_d} = \frac{A(1-A)}{3N_d}$$

### Top-Down Approach

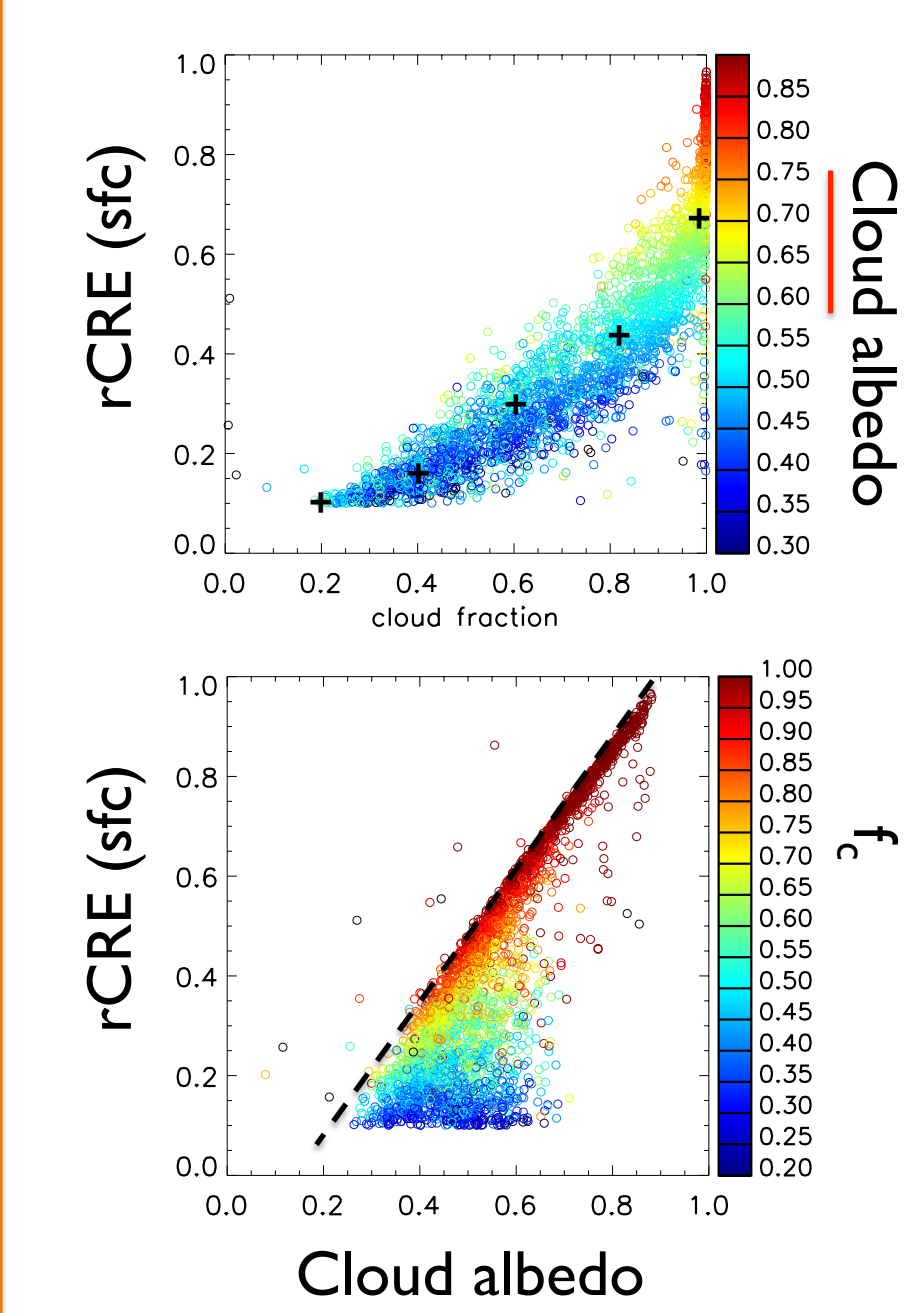


- Co-variability of inputs influences detectability of aerosol effects; Strong Case for Routine LES accompanying Routine Observations (LASSO)
- Understand what controls shape of  $rCRE-A-f_c$  relationship!
- Carefully define  $f_c$ !

### Top-Down Approach: Observations (SGP) vs. Modeling (Ocean)

#### Observations:

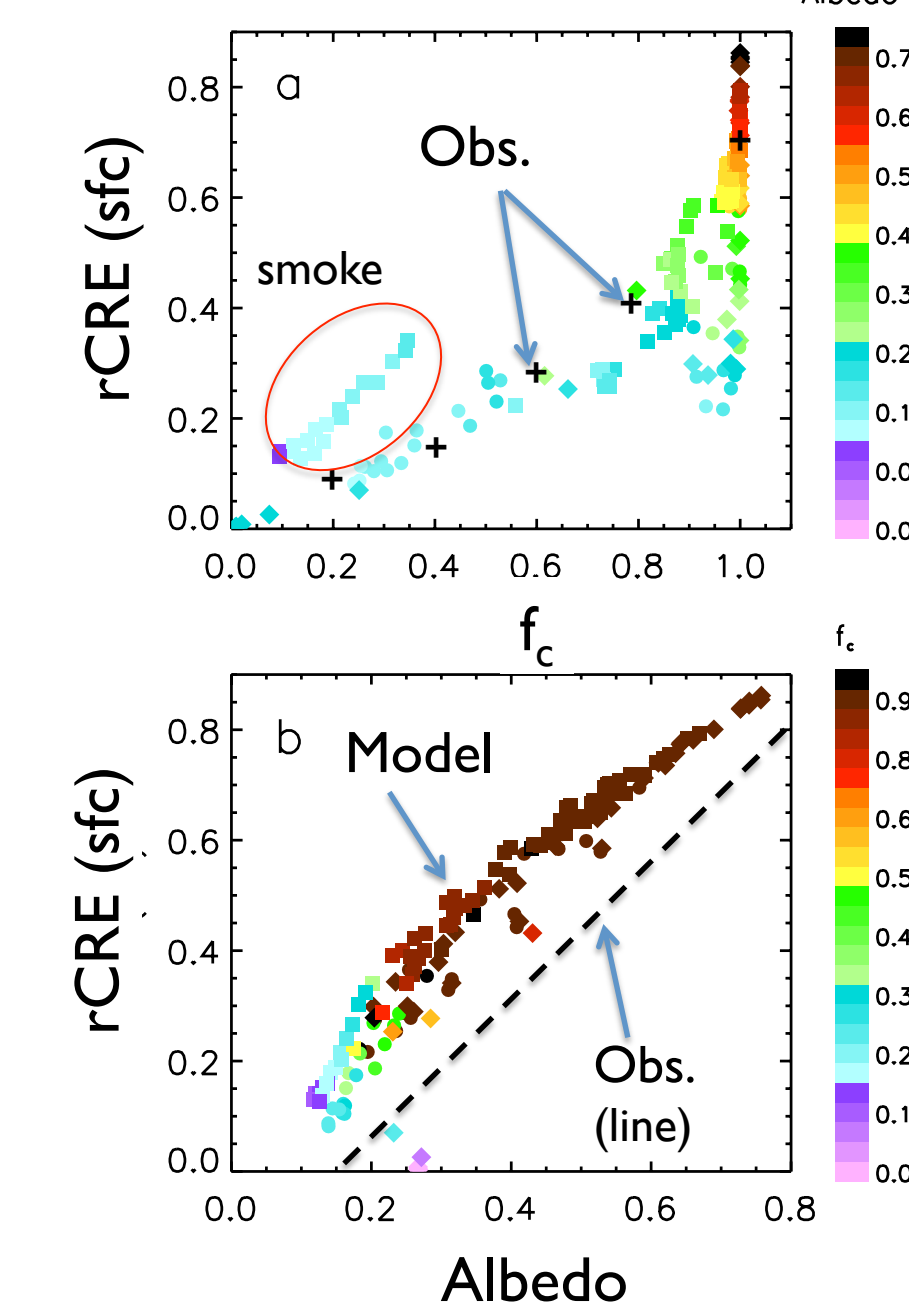
14 years of warm clouds at SGP (McComiskey et al.)



$$rCRE = 1 - \frac{F_{sw,all}}{F_{sw,clr}}$$

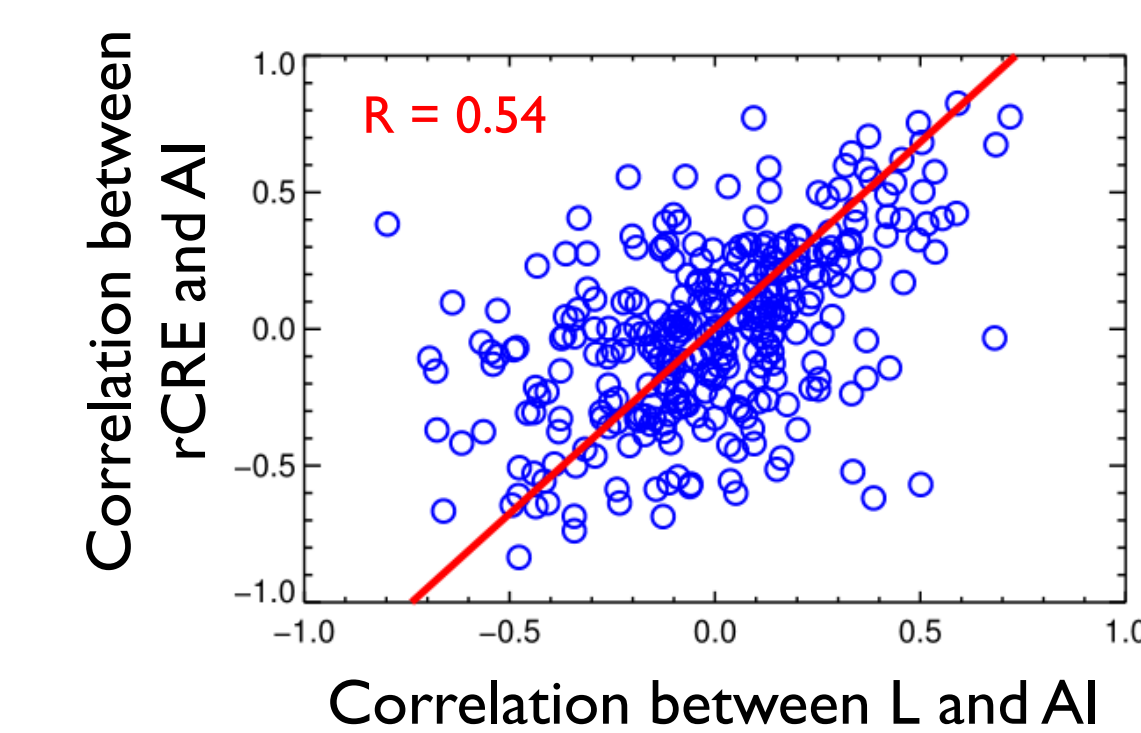
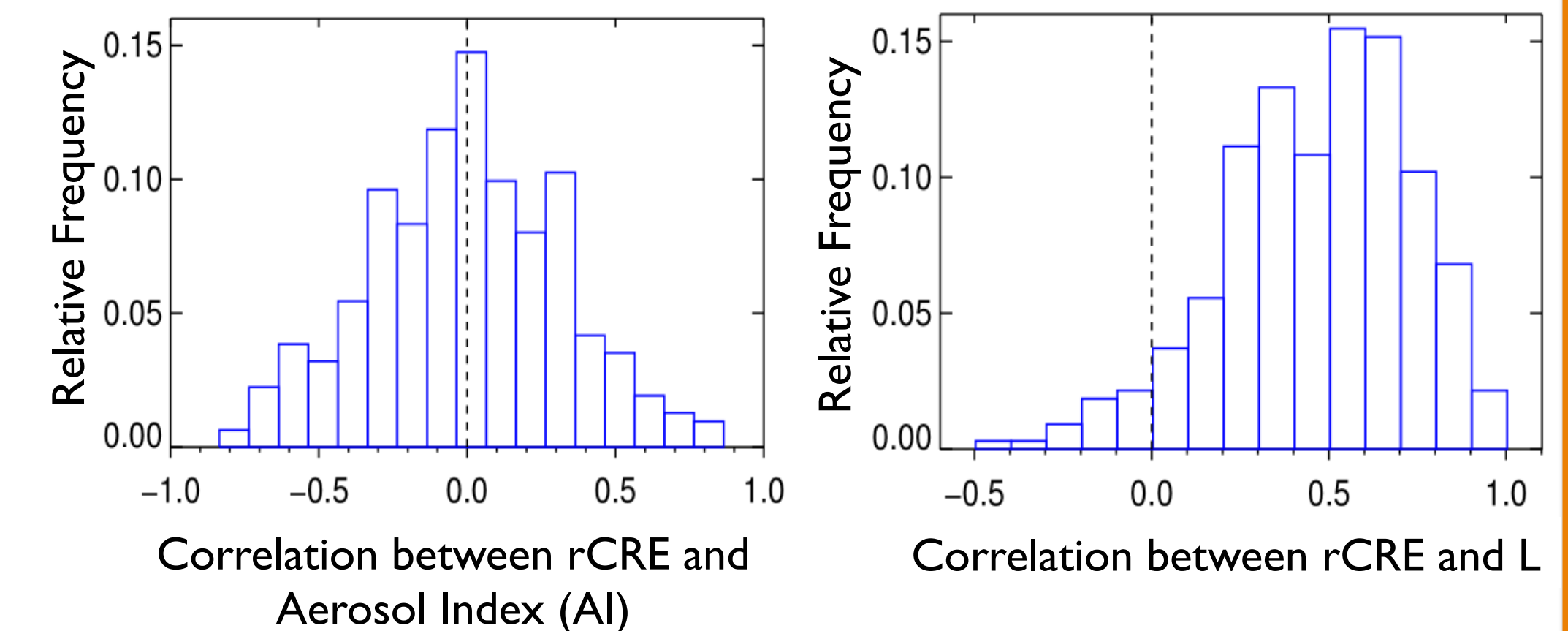
#### LES/CRM Modeling

Ocean: SCu (closed → open) SCu → Cu (+ smoke)



Feingold, McComiskey, Yamaguchi et al. PNAS, 2016

### Top-Down Approach: Observations at SGP



14 years of warm clouds at SGP:  $rCRE$  is driven by liquid water path, not aerosol

Sena et al. 2016 ACPD