

# Constraining the aerosol influence on liquid water path

E. Gryspeerdt <sup>1</sup>

T. Goren <sup>2</sup> J. Quaas <sup>2</sup> O. Sourdeval <sup>2</sup> A. Gettelman <sup>3</sup>

<sup>1</sup>Imperial College London <sup>2</sup>Universitt Leipzig <sup>3</sup>National Center for Atmospheric  
Research

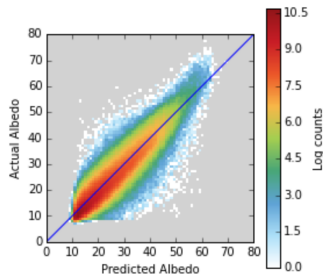
October 9, 2017

# Aerosols and liquid clouds

- ▶ Aerosol modify the properties of liquid clouds

The albedo of a liquid cloud scene is a function of:

- ▶ Cloud fraction (CF)
  - ▶ Liquid water path (LWP)
  - ▶ Droplet number concentration ( $N_d$ )
- 
- ▶ Fitting CERES scene albedo (at  $1^\circ \times 1^\circ$ ) as  $f(CF, LWP, N_d)$  can diagnose liquid cloud albedo from MODIS cloud properties
  - ▶ What is the aerosol effect/what is a good model constraint?

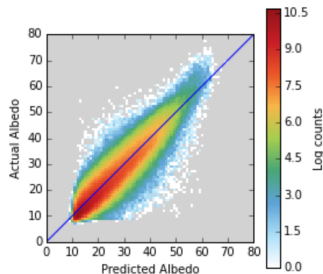


# Aerosols and liquid clouds

- ▶ Aerosol modify the properties of liquid clouds

The albedo of a liquid cloud scene is a function of:

- ▶ Cloud fraction (CF)
  - ▶ Liquid water path (LWP)
  - ▶ Droplet number concentration ( $N_d$ ) (✓)
- 
- ▶ Fitting CERES scene albedo (at  $1^\circ \times 1^\circ$ ) as  $f(CF, LWP, N_d)$  can diagnose liquid cloud albedo from MODIS cloud properties
  - ▶ What is the aerosol effect/what is a good model constraint?

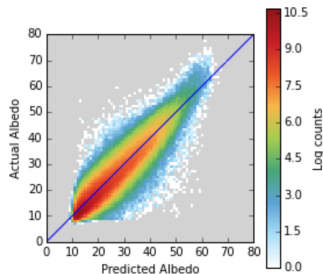


# Aerosols and liquid clouds

- ▶ Aerosol modify the properties of liquid clouds

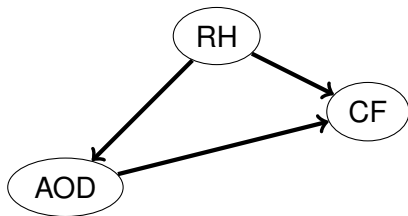
The albedo of a liquid cloud scene is a function of:

- ▶ Cloud fraction (CF) (✓)
  - ▶ Liquid water path (LWP)
  - ▶ Droplet number concentration ( $N_d$ ) (✓)
- 
- ▶ Fitting CERES scene albedo (at  $1^\circ \times 1^\circ$ ) as  $f(CF, LWP, N_d)$  can diagnose liquid cloud albedo from MODIS cloud properties
  - ▶ What is the aerosol effect/what is a good model constraint?



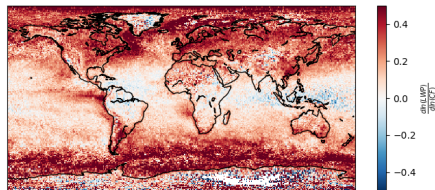
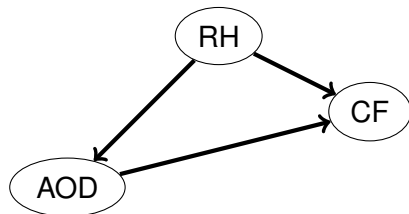
## Confounding variables

- ▶ The Aerosol optical depth (AOD)-CF relationship is controlled by humidity (RH)



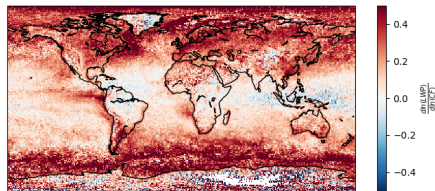
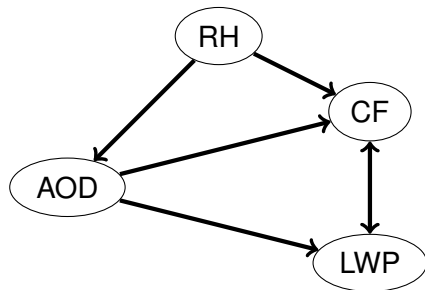
# Confounding variables

- ▶ The Aerosol optical depth (AOD)-CF relationship is controlled by humidity (RH)
- ▶ LWP is strongly correlated to CF (lower plot, data from MODIS Aqua)



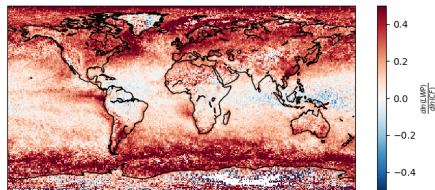
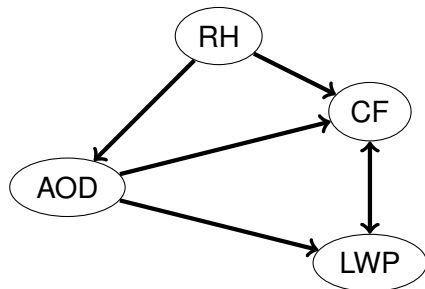
# Confounding variables

- ▶ The Aerosol optical depth (AOD)-CF relationship is controlled by humidity (RH)
- ▶ LWP is strongly correlated to CF (lower plot, data from MODIS Aqua)
- ▶ The AOD-LWP relationship is also confounded by humidity



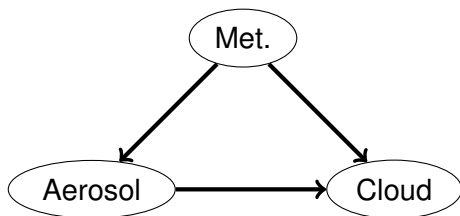
# Confounding variables

- ▶ The Aerosol optical depth (AOD)-CF relationship is controlled by humidity (RH)
- ▶ LWP is strongly correlated to CF (lower plot, data from MODIS Aqua)
- ▶ The AOD-LWP relationship is also confounded by humidity
- ▶ How do we separate the aerosol effect?





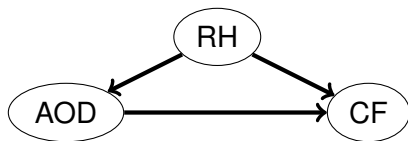
## How do we identify aerosol effects?



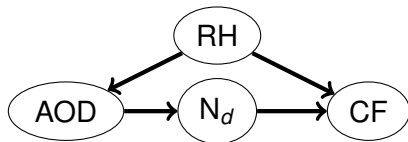
A number of different methods have been used:

- ▶ Controlling for confounders
- ▶ Exogenous perturbations (experiments)
- ▶ Mediating variables

## Mediating variables

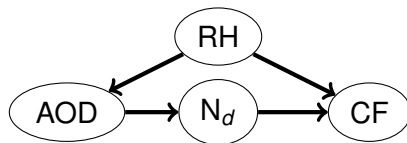


## Mediating variables

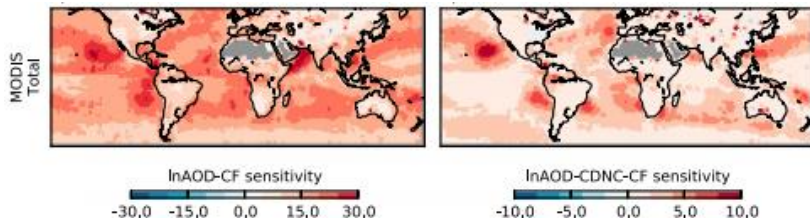


- ▶ Requires knowledge of the confounders (but not measurement!)

# Mediating variables



- Requires knowledge of the confounders (but not measurement!)



## Testing causality

Can we use this method for identifying the aerosol-LWP relationship?

$$\left. \frac{dLWP}{dAI} \right|_{causal} = \frac{dLWP}{dN_d} \times \frac{dN_d}{dAI}$$

## Testing causality

Can we use this method for identifying the aerosol-LWP relationship?

$$\left. \frac{dLWP}{dAI} \right|_{causal} = \frac{dLWP}{dN_d} \times \frac{dN_d}{dAI}$$

Note that we can't test this by setting  $N_d$  constant

$$\frac{dN_d}{dAI} = 0 \rightarrow \left. \frac{dLWP}{dAI} \right|_{causal} = 0$$

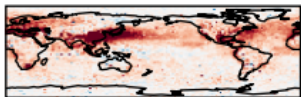
But it should also be able to “predict” the anthropogenic LWP change ( $\Delta LWP_{actual}$ )

# Testing causality

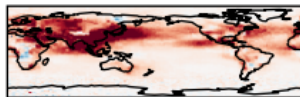
$$\Delta LWP_{diag} = \left. \frac{dLWP}{dAI} \right|_{PD} \times \Delta AI$$

ECHAM6  
HAM

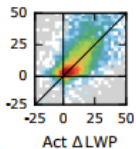
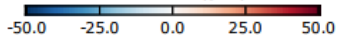
Actual  $\Delta LWP$



Diag.  $\Delta LWP$  (AI)

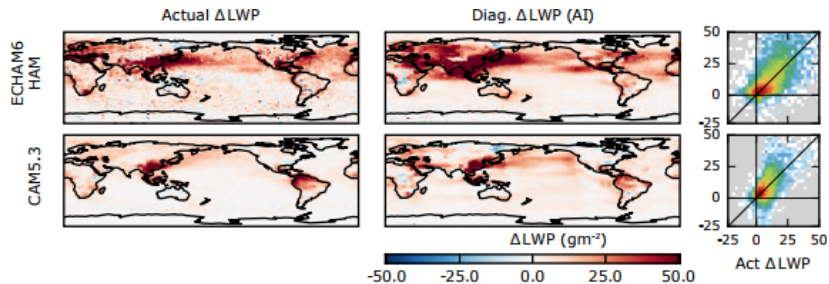


$\Delta LWP$  ( $\text{gm}^{-2}$ )



# Testing causality

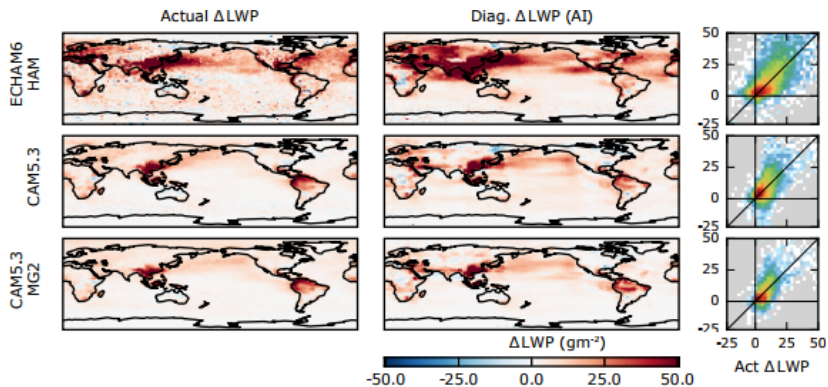
$$\Delta LWP_{diag} = \left. \frac{dLWP}{dAI} \right|_{PD} \times \Delta AI$$





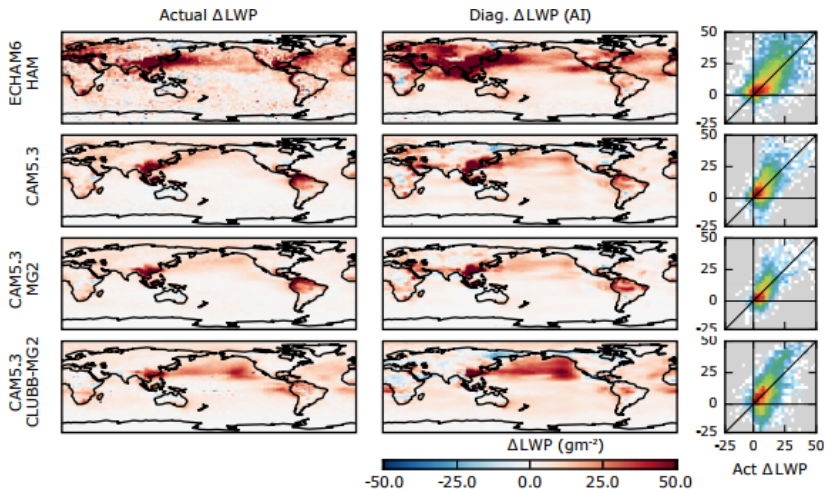
# Testing causality

$$\Delta LWP_{diag} = \left. \frac{dLWP}{dAI} \right|_{PD} \times \Delta AI$$

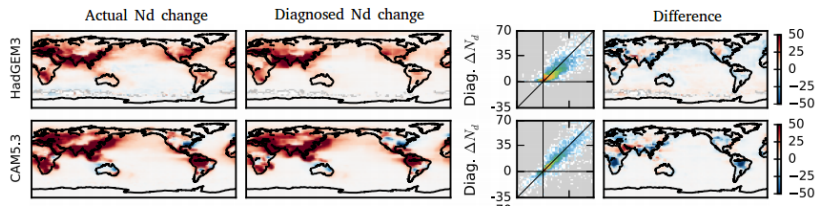


# Testing causality

$$\Delta LWP_{diag} = \left. \frac{dLWP}{dAI} \right|_{PD} \times \Delta AI$$



# Calculating $\Delta N_d$

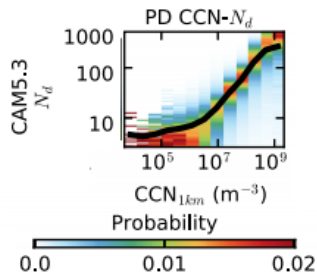


- ▶ Anthropogenic  $\Delta N_d$  can be accurately calculated from the present day CCN- $N_d$  relationship

$$\Delta N_d = \left. \frac{dN_d}{dCCN} \right|_{PD} \times \Delta CCN$$

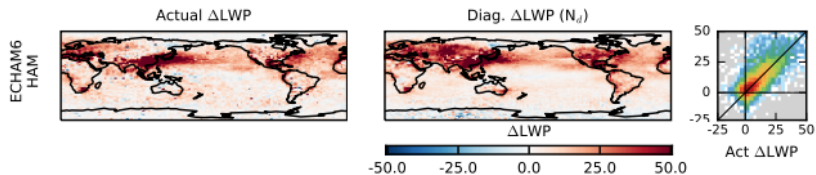
- ▶ Use conditional probabilities to conserve the non-linearity (e.g.  $P(N_d|CCN)$ )

Gryspeerd et al, PNAS, 2017



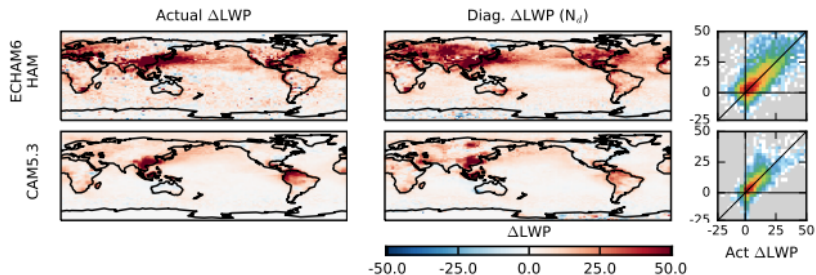
# Testing causality

$$\Delta LWP_{diag} = \left. \frac{dLWP}{dN_d} \right|_{PD} \times \left( \left. \frac{dN_d}{dAI} \right|_{PD} \times \Delta AI \right) = \left. \frac{dLWP}{dN_d} \right|_{PD} \times \Delta N_d$$



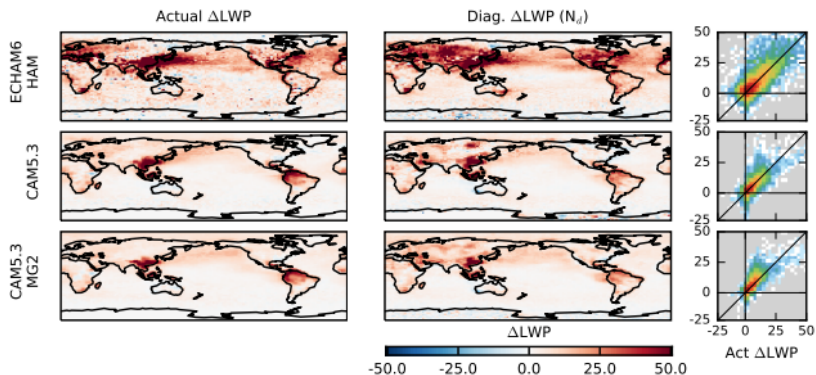
# Testing causality

$$\Delta LWP_{diag} = \left. \frac{dLWP}{dN_d} \right|_{PD} \times \left( \left. \frac{dN_d}{dAI} \right|_{PD} \times \Delta AI \right) = \left. \frac{dLWP}{dN_d} \right|_{PD} \times \Delta N_d$$



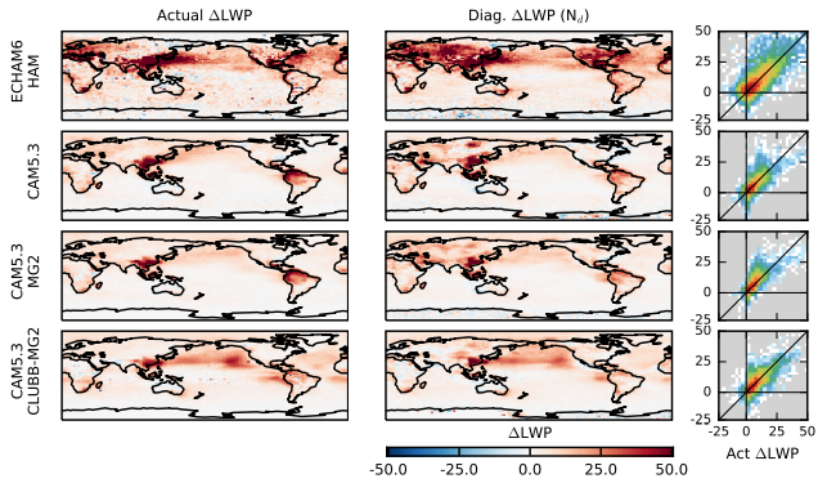
# Testing causality

$$\Delta LWP_{diag} = \left. \frac{dLWP}{dN_d} \right|_{PD} \times \left( \left. \frac{dN_d}{dAI} \right|_{PD} \times \Delta AI \right) = \left. \frac{dLWP}{dN_d} \right|_{PD} \times \Delta N_d$$



# Testing causality

$$\Delta LWP_{diag} = \left. \frac{dLWP}{dN_d} \right|_{PD} \times \left( \left. \frac{dN_d}{dAI} \right|_{PD} \times \Delta AI \right) = \left. \frac{dLWP}{dN_d} \right|_{PD} \times \Delta N_d$$

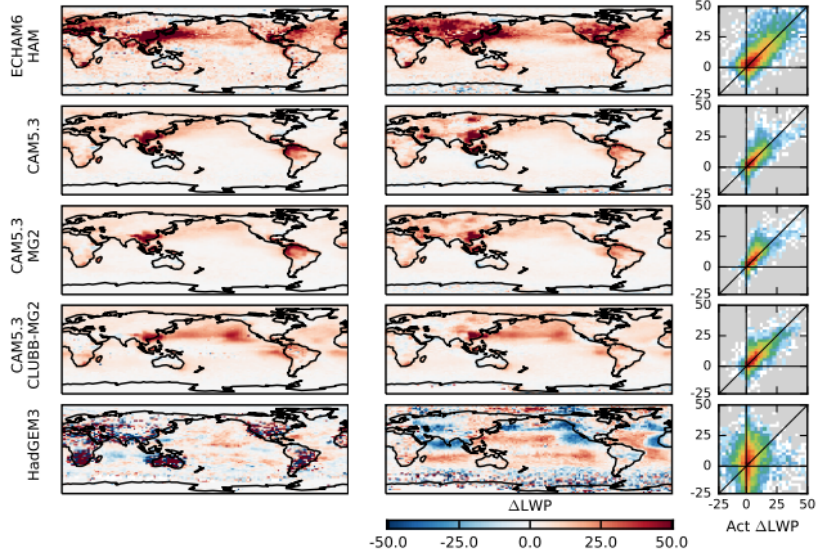


# Testing causality

$$\Delta LWP_{diag} = \left. \frac{dLWP}{dN_d} \right|_{PN} \times \left( \left. \frac{dN_d}{dAI} \right|_{PN} \times \Delta AI \right) = \left. \frac{dLWP}{dN_d} \right|_{PN} \times \Delta N_d$$

Actual  $\Delta LWP$

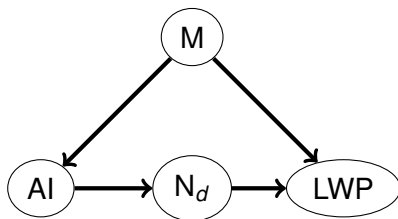
Diag.  $\Delta LWP (N_d)$





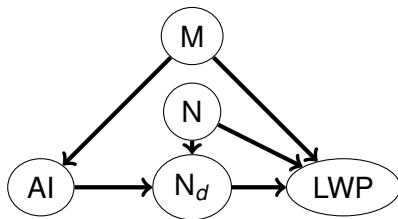
## The CDNC-LWP relationship

Most of these models suggest a causal network of this form:



## The CDNC-LWP relationship

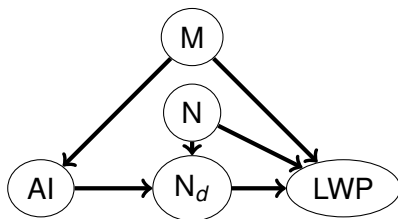
Most of these models suggest a causal network of this form:



Confounding effects may exist in observations that are not simulated (N):

## The CDNC-LWP relationship

Most of these models suggest a causal network of this form:

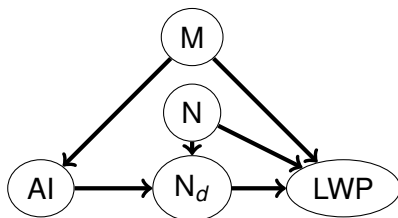


Confounding effects may exist in observations that are not simulated (N):

- ▶ Retrieval errors
  - ▶ Correlated errors in  $N_d$  and LWP
  - ▶ Systematic errors in cloud retrievals (e.g. sub-adiabaticity)

# The CDNC-LWP relationship

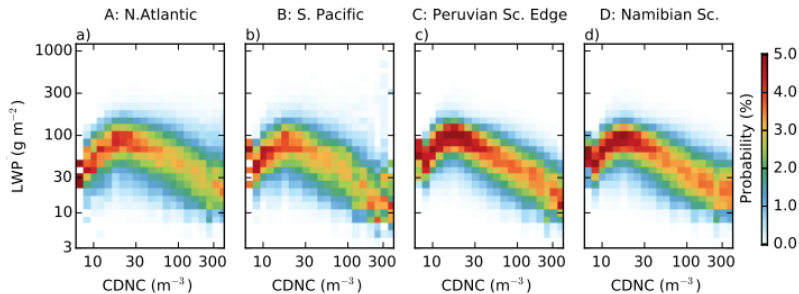
Most of these models suggest a causal network of this form:



Confounding effects may exist in observations that are not simulated (N):

- ▶ Retrieval errors
  - ▶ Correlated errors in  $N_d$  and LWP
  - ▶ Systematic errors in cloud retrievals (e.g. sub-adiabaticity)
- ▶ Feedbacks
  - ▶ Aerosol dependent entrainment of dry air
  - ▶ (Aerosol processing)

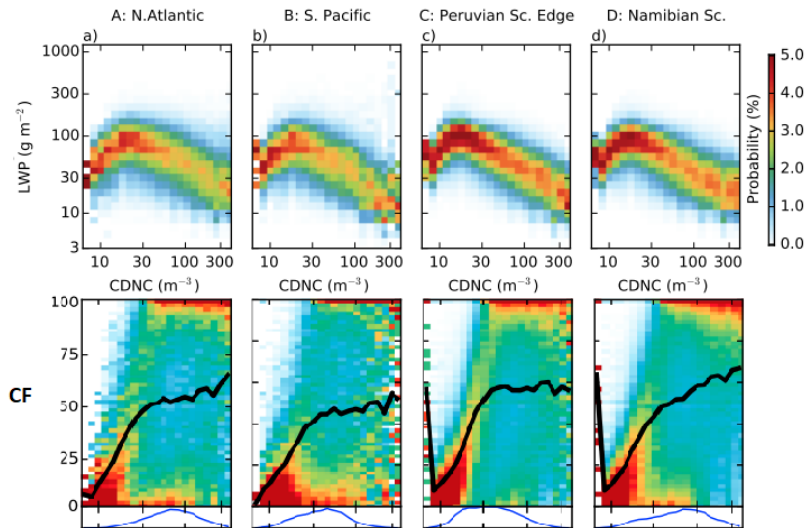
# Satellite relationships



The  $N_d$ -LWP relationship in MODIS data is strongly non-linear

- ▶ Increase in LWP with increasing  $N_d$  at low  $N_d$
- ▶ Decrease in LWP at high  $N_d$

# Satellite relationships



## Correlated errors?

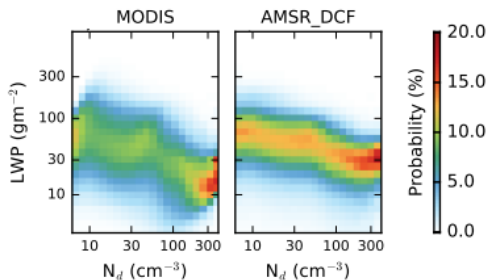
Both  $N_d (\propto \tau_c^{0.5} r_e^{-2.5})$  and LWP ( $\propto \tau_c r_e$ ) are calculated from  $r_e$  and  $\tau_c$

- ▶ Errors in  $r_e$  or  $\tau_c$  generate correlated errors in  $N_d$  and LWP.

## Correlated errors?

Both  $N_d$  ( $\propto \tau_c^{0.5} r_e^{-2.5}$ ) and LWP ( $\propto \tau_c r_e$ ) are calculated from  $r_e$  and  $\tau_c$

- ▶ Errors in  $r_e$  or  $\tau_c$  generate correlated errors in  $N_d$  and LWP.



Similar results using MODIS and AMSR-E (microwave) LWP

- ▶ Retrieval errors don't dominate the relationship
- ▶ Although they still play a role



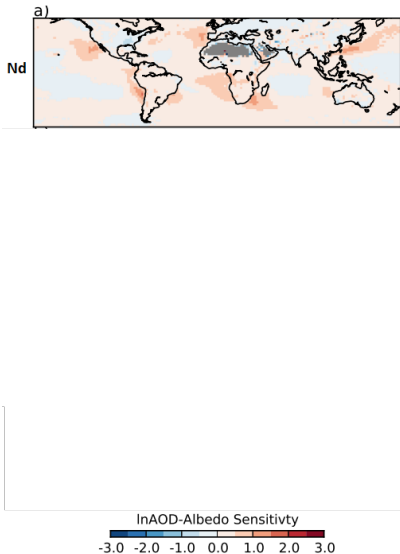
## Implied forcing

What radiative forcing does this suggest?

# Implied forcing

What radiative forcing does this suggest?

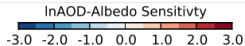
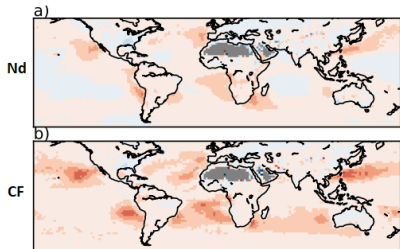
- ▶ Changes in  $N_d$  only (RFaci)
  - ▶  $\approx -0.30 \text{ Wm}^{-2}$



# Implied forcing

What radiative forcing does this suggest?

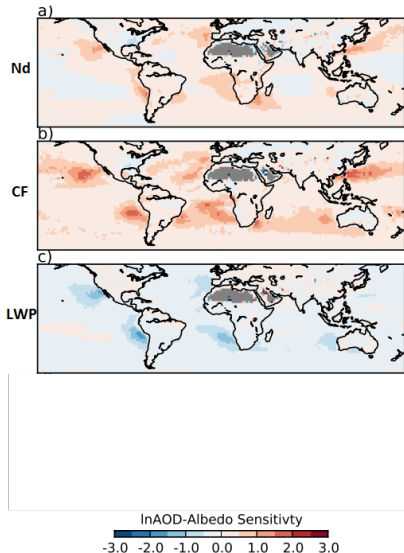
- ▶ Changes in  $N_d$  only (RF<sub>aci</sub>)
  - ▶  $\approx -0.30 \text{ Wm}^{-2}$
- ▶ Changes in CF (mediated by  $N_d$ )
  - ▶  $\approx -0.50 \text{ Wm}^{-2}$(Gryspeerd et al., 2016)



# Implied forcing

What radiative forcing does this suggest?

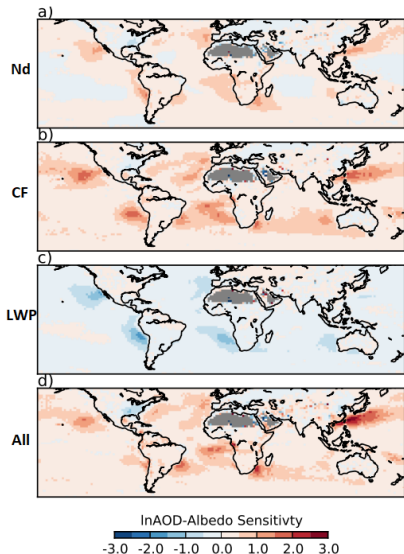
- ▶ Changes in  $N_d$  only (RFaci)
  - ▶  $\approx -0.30 \text{ Wm}^{-2}$
- ▶ Changes in CF (mediated by  $N_d$ )
  - ▶  $\approx -0.50 \text{ Wm}^{-2}$  (Gryspeerd et al., 2016)
- ▶ Changes in LWP (mediated by  $N_d$ )
  - ▶  $\approx +0.27 \text{ Wm}^{-2}$  (upper bound)



# Implied forcing

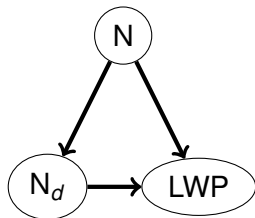
What radiative forcing does this suggest?

- ▶ Changes in  $N_d$  only (RFaci)
  - ▶  $\approx -0.30 \text{ Wm}^{-2}$
- ▶ Changes in CF (mediated by  $N_d$ )
  - ▶  $\approx -0.50 \text{ Wm}^{-2}$  (Gryspeerd et al., 2016)
- ▶ Changes in LWP (mediated by  $N_d$ )
  - ▶  $\approx +0.27 \text{ Wm}^{-2}$  (upper bound)
- ▶ Total (liquid clouds only)
  - ▶  $\approx -0.51 \text{ Wm}^{-2}$



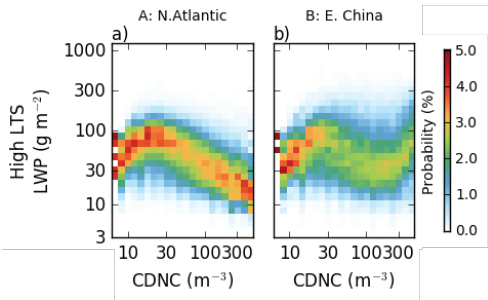
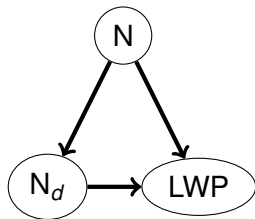
# Summary

## Exogenous perturbations



If the  $N_d$ - $LWP$  relationship is causal, it should hold under all conditions

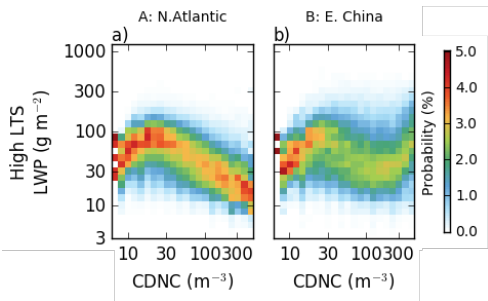
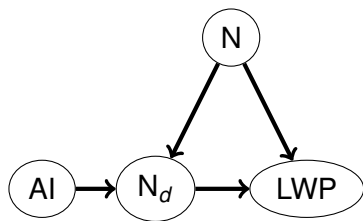
# Exogenous perturbations



If the  $N_d$ -LWP relationship is causal, it should hold under all conditions



# Exogenous perturbations

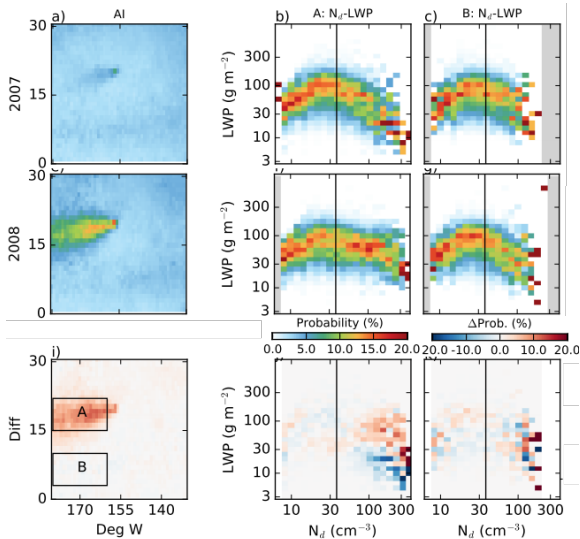


If the  $N_d$ -LWP relationship is causal, it should hold under all conditions

Conditions where the  $N_d$  is varied exogenously provide an opportunity to test this

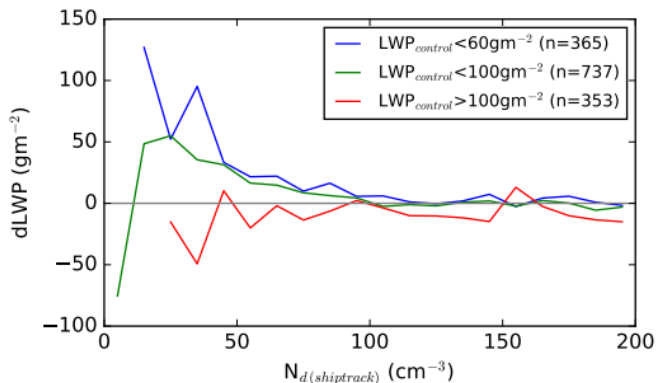
- ▶ Volcanoes
- ▶ Shiptracks

# Volcanic emissions



- ▶ In 2007,  $N_d$ -LWP similar in regions A and B
- ▶ Region A shows weaker LWP decrease with  $N_d$  in 2008
- ▶ Suggests little LWP change at high  $N_d$

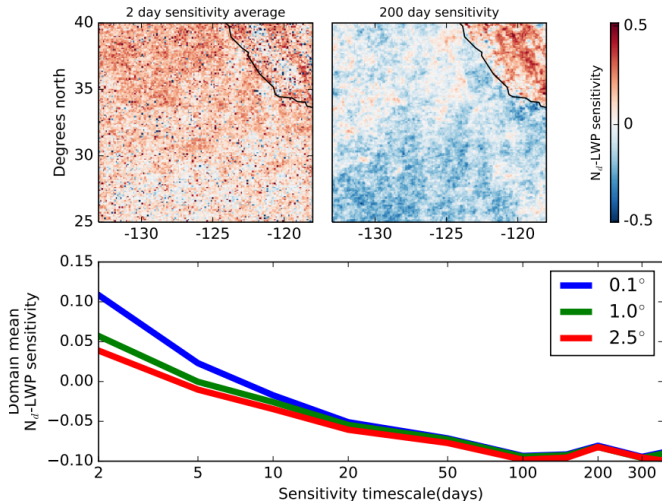
# Shiptracks



- ▶ A similar effect is observed in shiptracks
- ▶ LWP increases in shiptracks where the “shiptrack”  $N_d$  is low
- ▶ No change in LWP at high “shiptrack”  $N_d$

Thanks to Matt Christensen for shiptrack data!

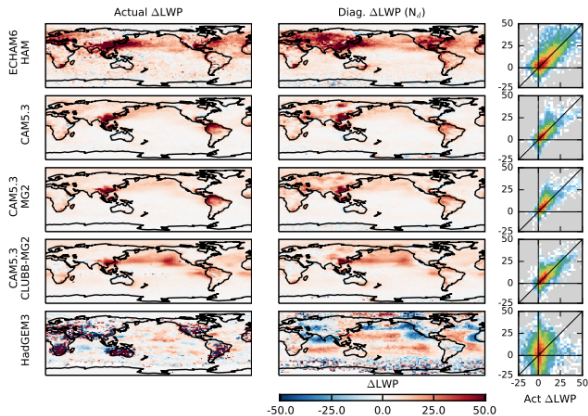
# Scale dependence



- ▶ The  $N_d$ -LWP relationship is scale dependent
- ▶ Must reach temporal/spatial scale where  $N_d$  is varied by aerosol

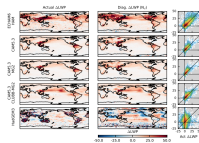
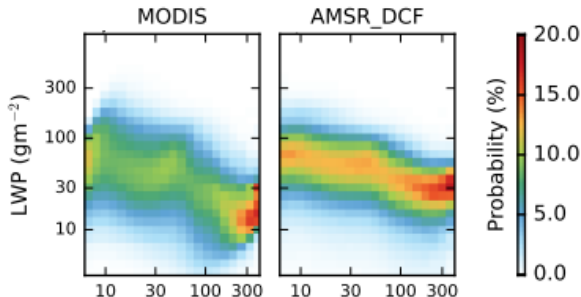
# Summary

- ▶ Model output confirms that the AI-LWP relationship overestimates the aerosol effect
  - ▶ The (plain) AI-LWP relationship overestimates the role of aerosols
  - ▶ The  $N_d$ -LWP seems a good choice (with caveats!)



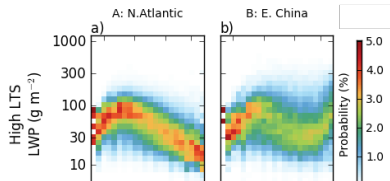
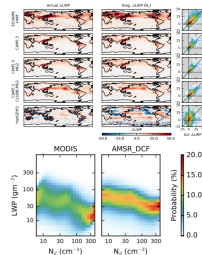
# Summary

- ▶ Model output confirms that the AI-LWP relationship overestimates the aerosol effect
  - ▶ The (plain) AI-LWP relationship overestimates the role of aerosols
  - ▶ The  $N_d$ -LWP seems a good choice (with caveats!)
- ▶ Satellite observations suggest increases and decreases in LWP with  $N_d$ 
  - ▶ Similar results with microwave LWP suggest not a retrieval issue



# Summary

- ▶ Model output confirms that the AI-LWP relationship overestimates the aerosol effect
  - ▶ The (plain) AI-LWP relationship overestimates the role of aerosols
  - ▶ The  $N_d$ -LWP seems a good choice (with caveats!)
- ▶ Satellite observations suggest increases and decreases in LWP with  $N_d$ 
  - ▶ Similar results with microwave LWP suggest not a retrieval issue
- ▶ Feedbacks/other confounders play a role
  - ▶ May obscure a small causal LWP change at high  $N_d$



# Summary

- ▶ Model output confirms that the AI-LWP relationship overestimates the aerosol effect
  - ▶ The (plain) AI-LWP relationship overestimates the role of aerosols
  - ▶ The  $N_d$ -LWP seems a good choice (with caveats!)
- ▶ Satellite observations suggest increases and decreases in LWP with  $N_d$ 
  - ▶ Similar results with microwave LWP suggest not a retrieval issue
- ▶ Feedbacks/other confounders play a role
  - ▶ May obscure a small causal LWP change at high  $N_d$
- ▶ What about ice processes... ?

