

# A statistical framework to quantify model uncertainty

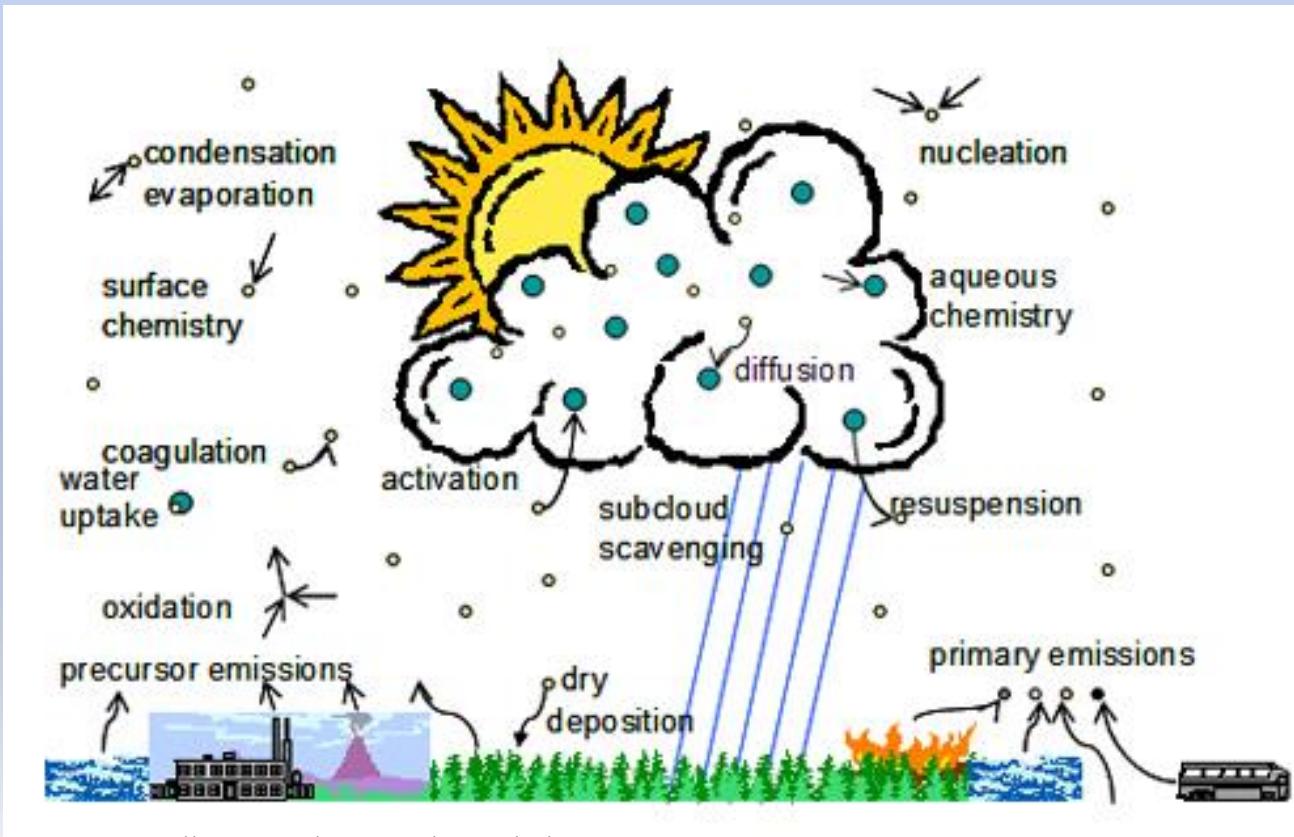
Lindsay Lee

[l.a.lee@leeds.ac.uk](mailto:l.a.lee@leeds.ac.uk)

Ken Carslaw, Carly Reddington, Kirsty Pringle,  
Graham Mann, Dominick Spracklen

# 1. GLOMAP

- We use the global aerosol model **GLOMAP** (Mann et al. 2010)
- A microphysical modal model simulating the evolution of global aerosol including sulphate, sea-salt, dust and black carbon





## 2. Uncertainty in modelling

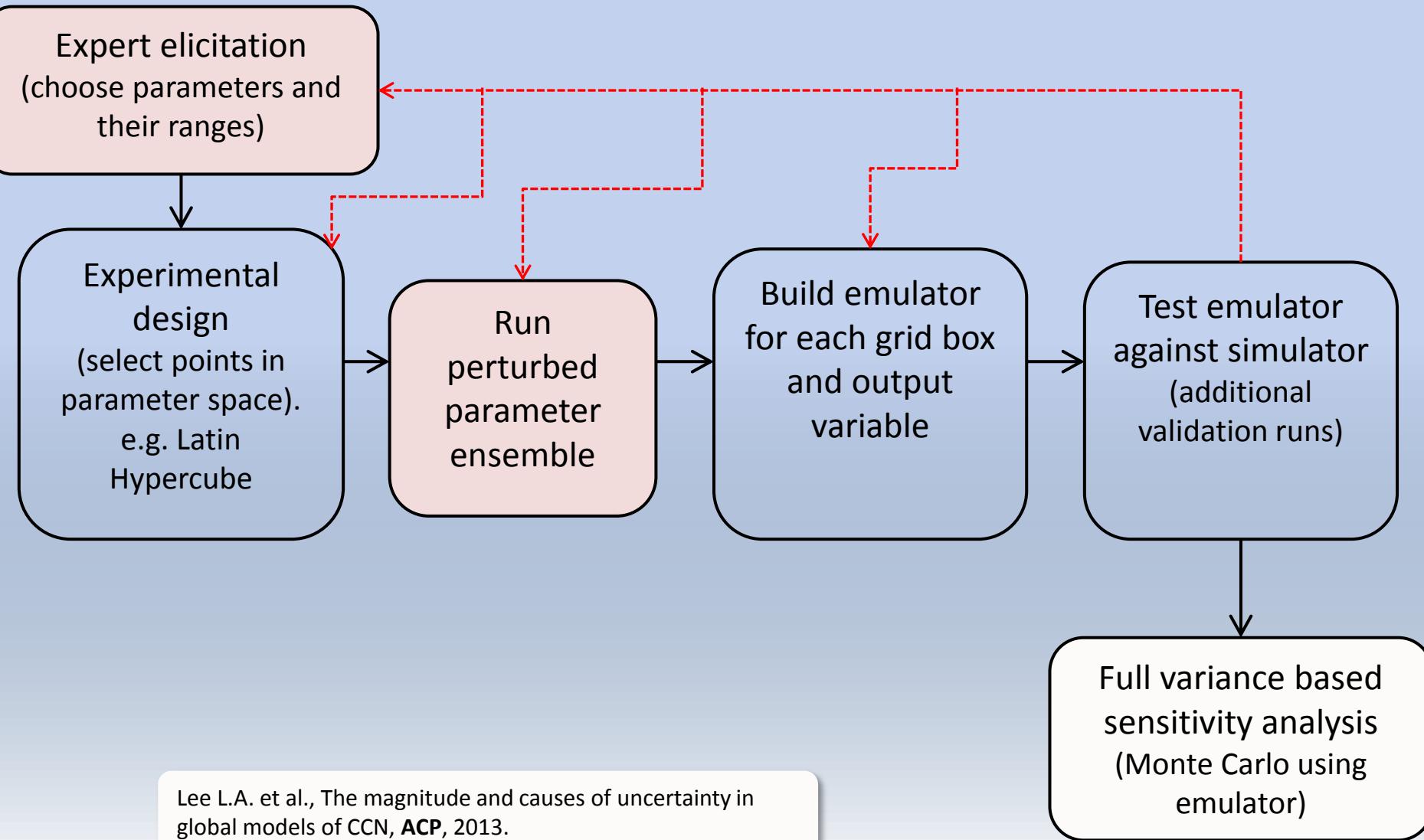
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- Structural uncertainty
  - Compare multiple models of the same type
  - Compare model to observations
- Parametric uncertainty
  - Compare runs of the same model with perturbed parameters
  - Compare model to observations

AEROCOM

AEROS

### 3. The procedure





## 4. Expert elicitation

- Discussion
- Variety of views
- Broaden the range of any single expert
- Potentially already too anchored?

‘We think these are the uncertain parameters and their values are very unlikely to fall outside of these ranges’

## 5. Biogenic SOA and BL nucleation



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- Design an experiment to look at just two parameters and their relationship to CCN
- Uncertainty scaled with usual values
- BIO\_SOA: 0.2 – 20  
$$\text{EMTERP}(I,K) = \text{EMTERP}(I,K) * \text{AEROS\_BIO\_SOA}$$
- BL\_NUC: 0.01 – 100  
$$\text{JBLN} = \text{JBLN} * \text{AEROS\_BLN\_RATE}$$



## 6. Experimental design

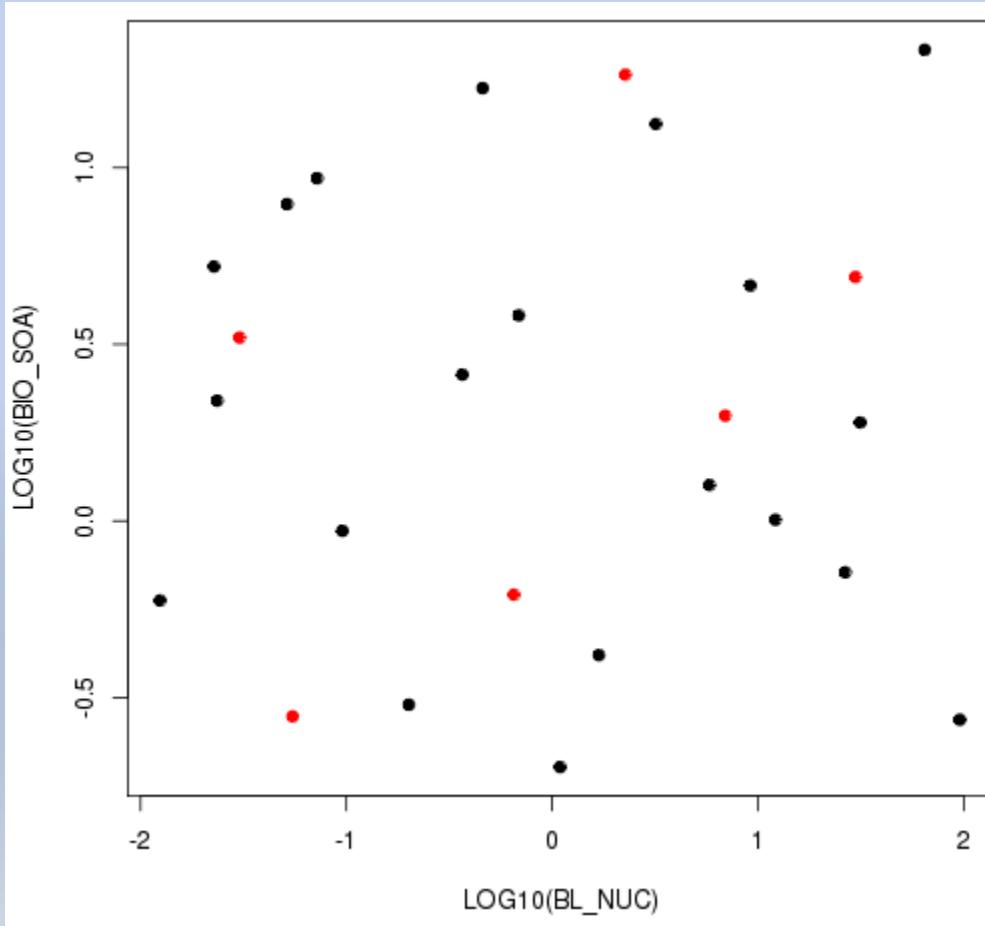
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- Need **maximum information** in **fewest runs**
- Maximin Latin Hypercube used
  - good marginal coverage
  - good space-filling properties
- Number of runs?
  - depends on ‘active’ parameters and function smoothness
- Validation crucial to highlight design issues
  - may need to sequentially design
  - could identify discontinuities/regime shifts



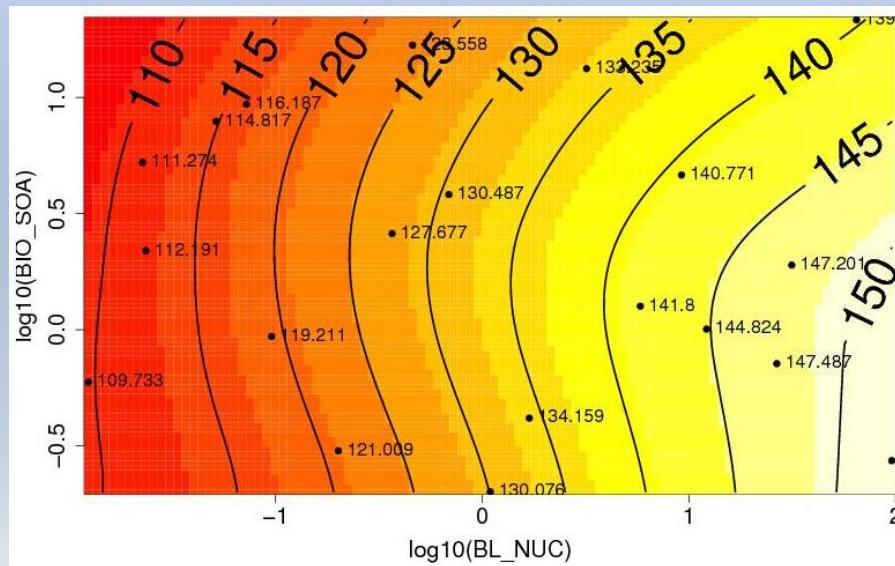
## 7. BIO\_SOA / BL\_NUC design

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## 8. Filling in the gaps - emulation

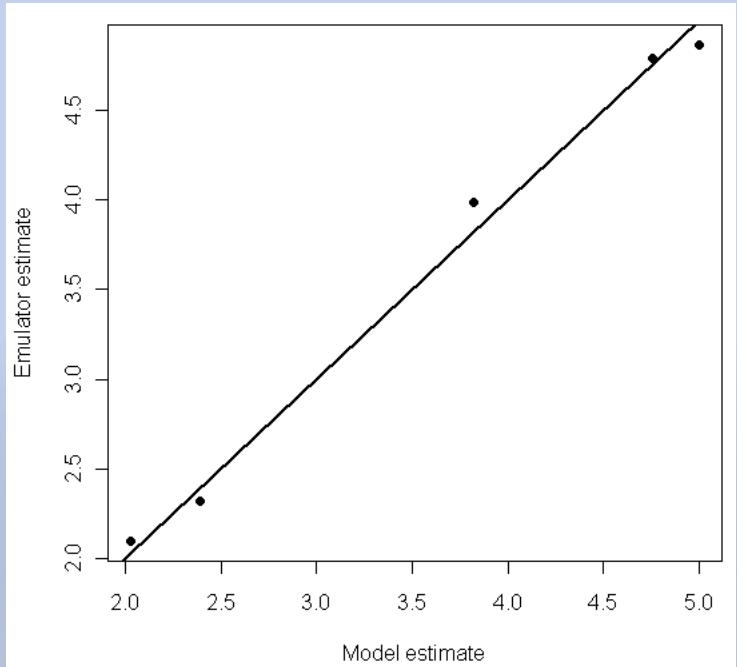
- Interpolate well-space model runs to estimate at untried points
- Bayesian method exploiting conditional probability
- Non-parametric





## 9. Emulator validation

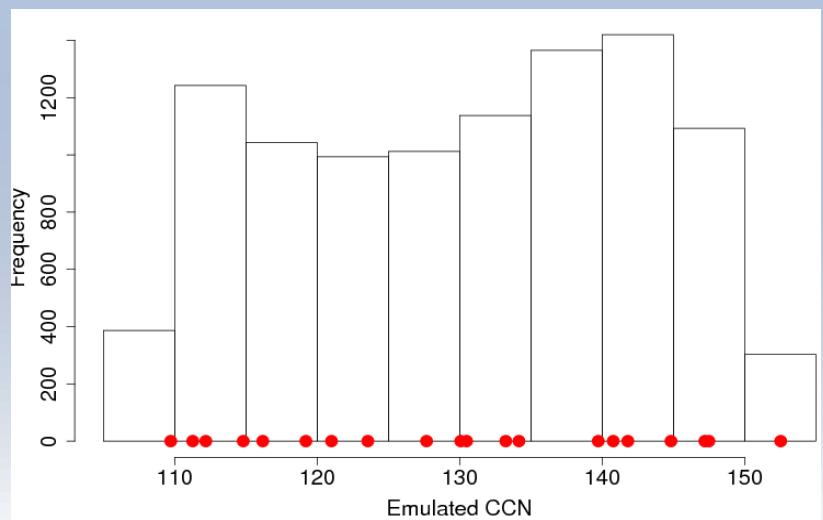
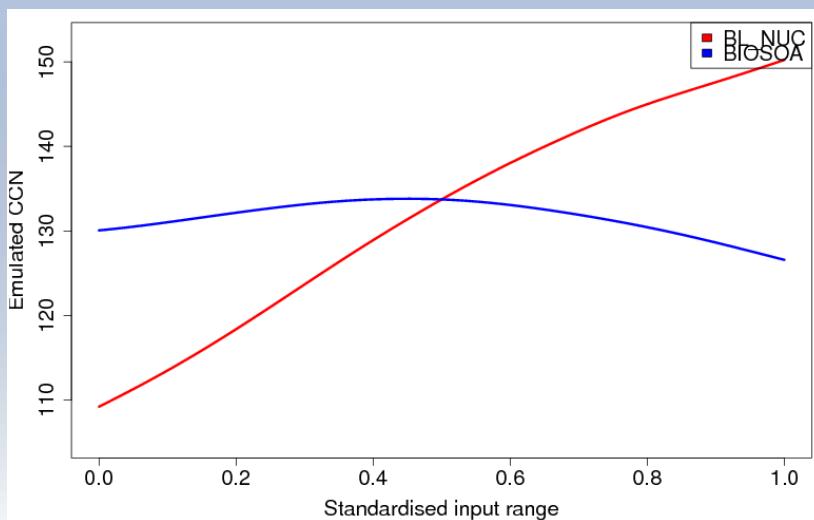
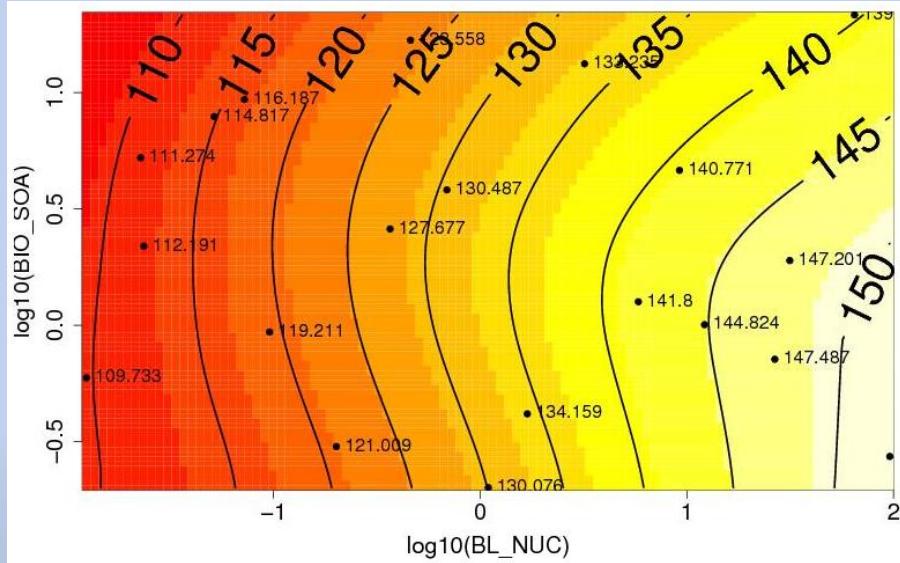
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Is the emulator output a good  
approximate of the model  
output?

YES – use the emulator mean  
instead

# 10. Filling in the gaps - emulation



# 11. Variance-based sensitivity analysis



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**Variance decomposition:**

$$Var(Y) = \sum_{i=1} V_i + \sum_{i < j} V_{ij} + \cdots + V_{12\cdots p}$$

**Variance due to each parameter:**

$$V_i = Var_{X_i}(E(Y | X_i)), V_{ij} = Var_{X_{ij}}(E(Y | X_{ij}))$$

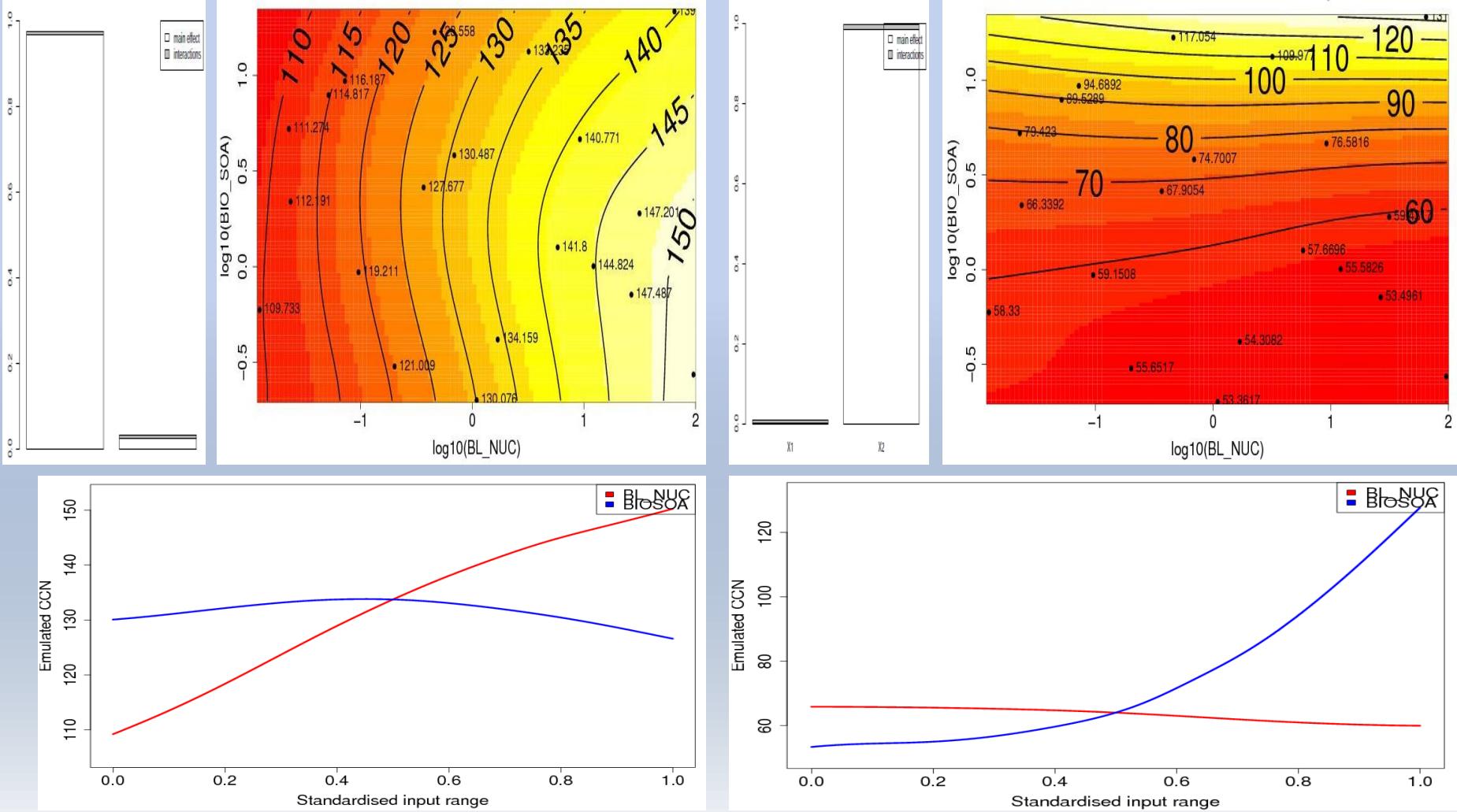
**Main effect sensitivity:**  $S_i = \frac{V_i}{Var(Y)}$

**Main effects**  
+

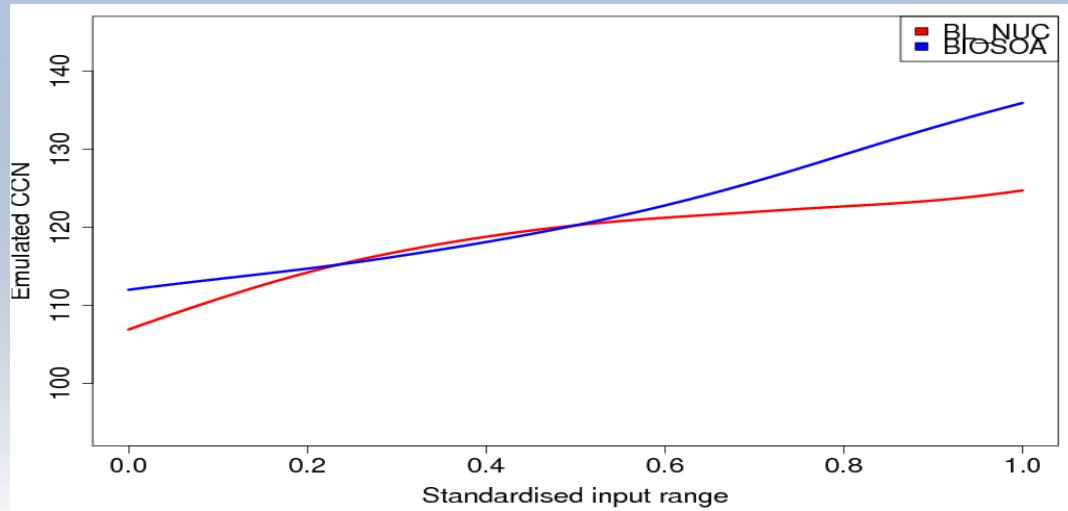
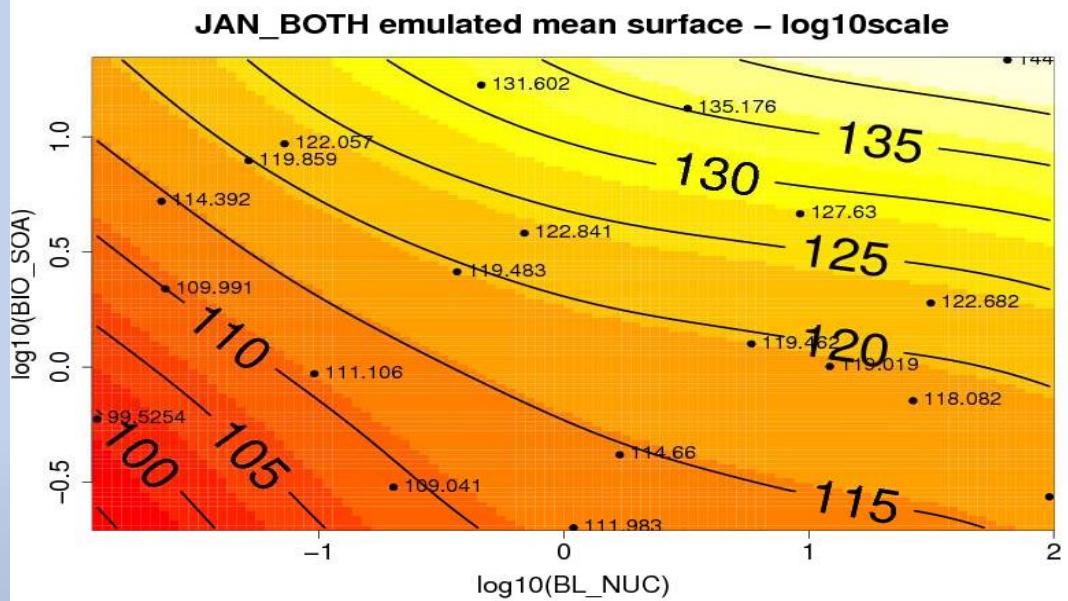
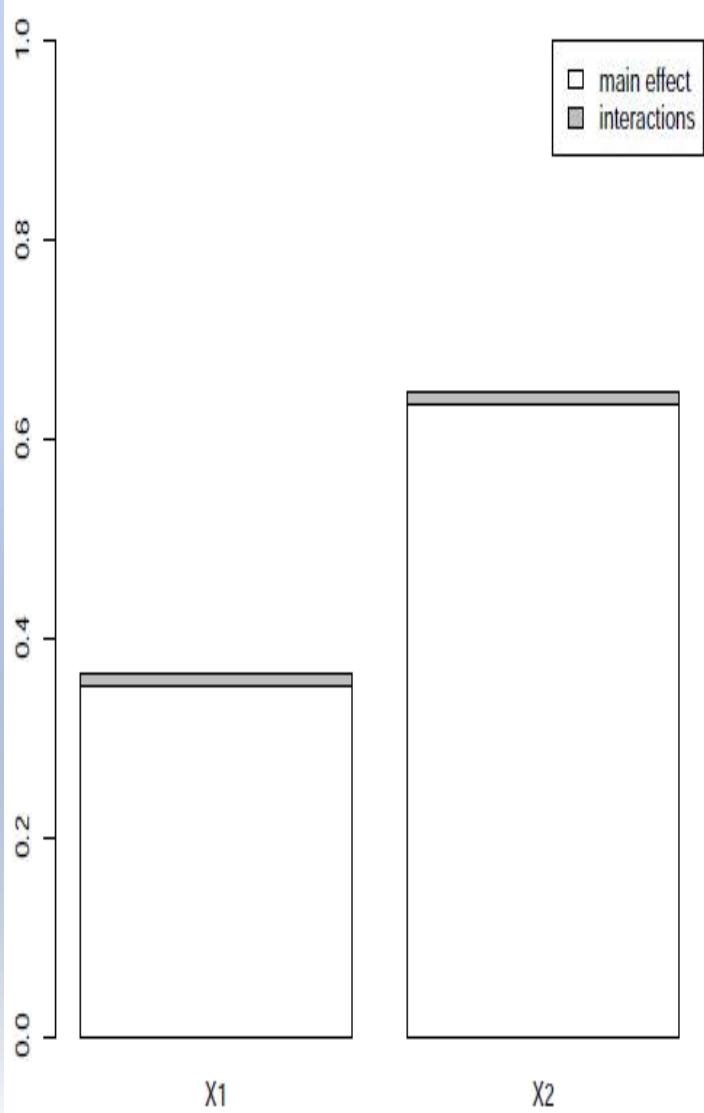
$$\sum_{i=1} S_i + \sum_{i < j} S_{ij} + \cdots + S_{12\cdots p} = 1$$

**Interactions:**

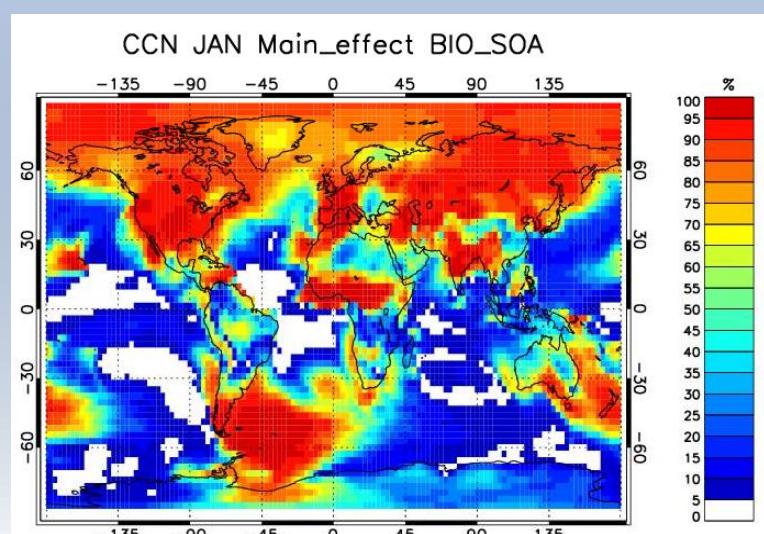
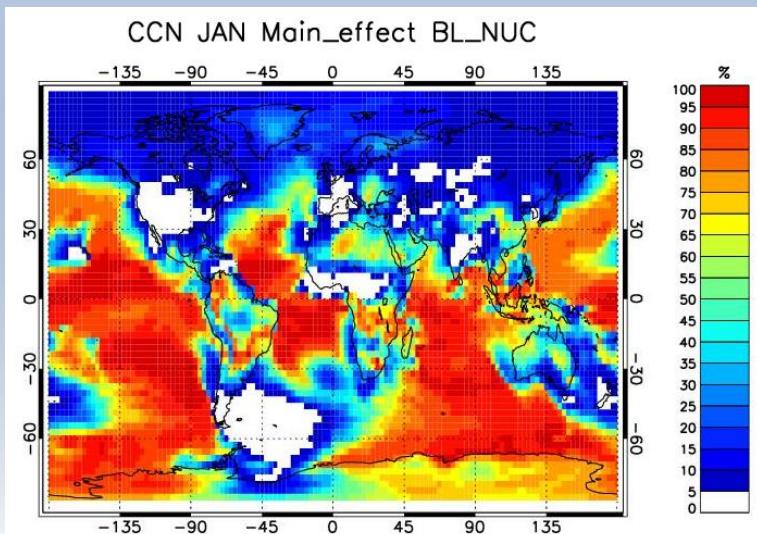
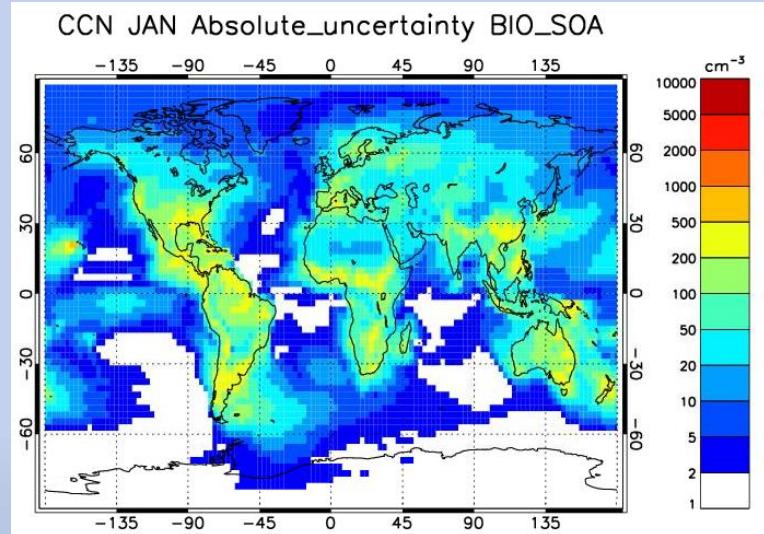
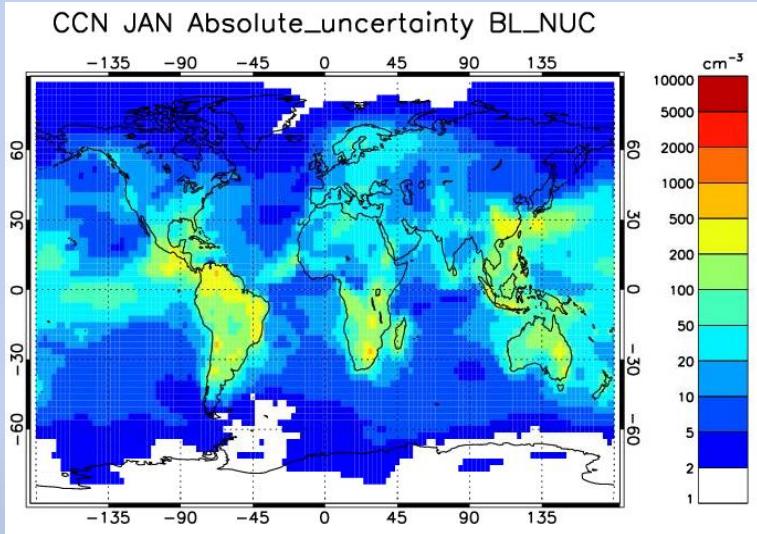
# 12. Main effects



# 13. Main effects



# 14. Global main effects



# 15. 28 parameter elicitation

Parameter	Lower	Upper
BCOC mass emission rate (fossil fuel)	0.5	2.0
BCOC mass emission rate (biomass burning)	0.25	4.0
BCOC mass emission rate (biofuel)	0.25	4.0
Sea spray mass flux (coarse/acc)	0.2x	5.0x
SO2 emission flux (anthropogenic)	0.6x	1.5x
SO2 emission flux (volcanic)	0.5x	2.0x
Biogenic monoterpane production of SOA	5 Tg/a	360Tg/a
Anthropogenic VOC production of SOA	3Tg/a	160Tg/a

Mass emission rates

DMS mass flux	0.5x
BCOC mode diameter (fossil fuel)	30 nm
BCOC mode diameter (biomass burning)	50 nm
BCOC mode diameter (biofuel)	50 nm
Subgrid conversion of SO2 to SO4 ("primary SO4")	0%
Mode diameter of "primary SO4"	20 nm

Microphysics

Model properties

Parameter	Lower	Upper
BL nucleation rate k[H <sub>2</sub> SO <sub>4</sub> ]	1E-10	2E-04
FT nucleation rate (BHN)	x0.01	X10
Ageing "rate" from insol to sol (monolayer)	0.3	5
Modal width (accumulation)	1.2	1.8
Modal width (Aitken)	1.2	1.8
Mode separation diameter (nucleation/Aitken)	9nm	20nm
Mode separation diameter (Aitken/accumulation)	x1.5	x3

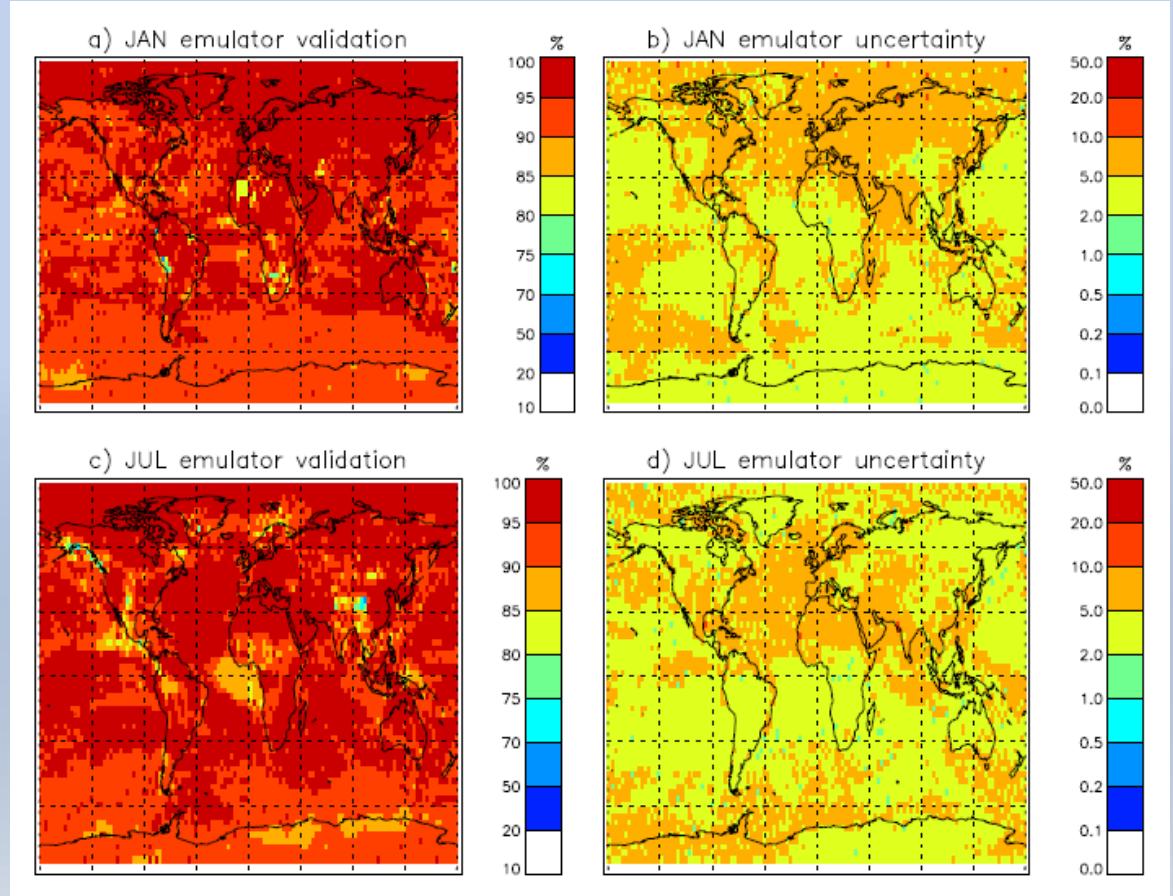
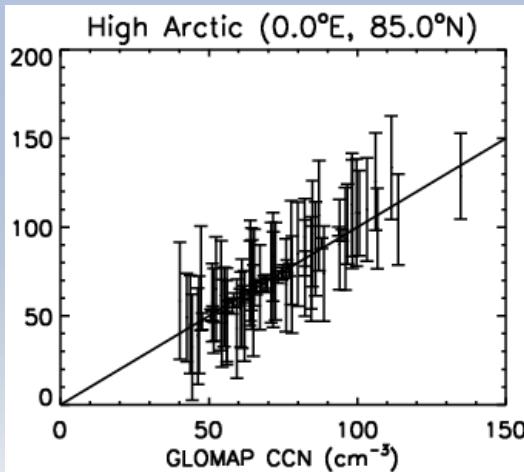
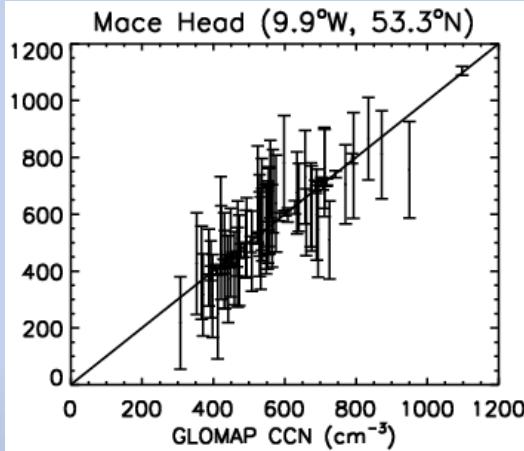
Parameter	Lower	Upper
Cloud drop activation dry diameter	30	100
Reaction SO <sub>2</sub> + O <sub>3</sub> in cloud water (clean)	pH=4	pH=6.5
Reaction SO <sub>2</sub> + O <sub>3</sub> in cloud water (polluted)	pH=3.5	pH=5

Cloud processing

Dry and wet deposition

Parameter	Lower	Upper
Nucleation scavenging dry D (above activation)	0	100
Nucleation scavenging fraction (T < -15C)	0.05	0.75
Dry deposition velocity (Aitken)	x0.5	X2.0
Dry deposition velocity (accumulation)	X0.1	X10.0
Dry deposition velocity (SO <sub>2</sub> )	X0.5	X2.0

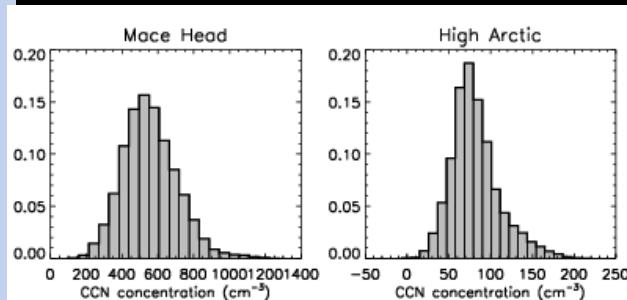
# 16. 28 parameter validation



# 17. Estimated CCN and its uncertainty concentration in every surface grid box

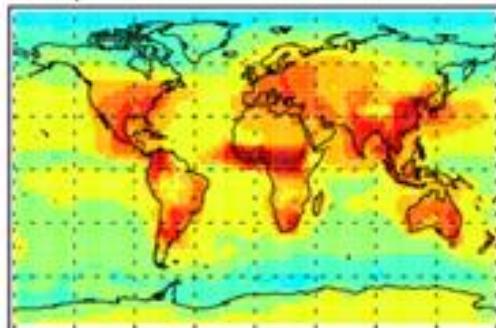


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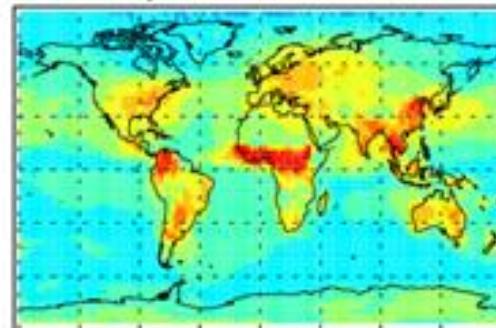


January

a) Emulator mean CCN

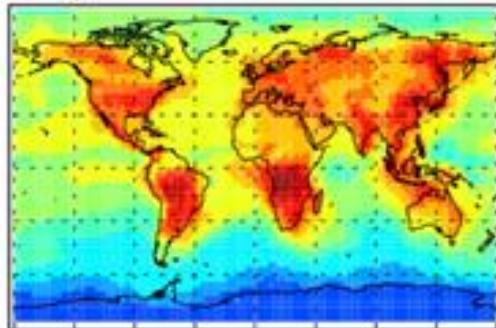


b) Emulator SD

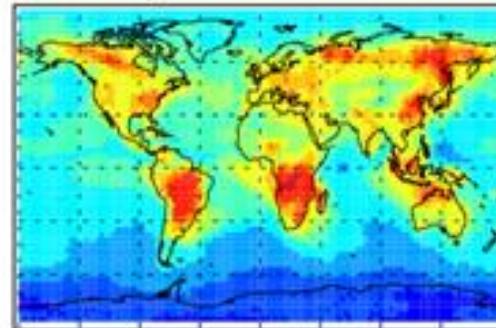


July

d) Emulator mean CCN



e) Emulator SD



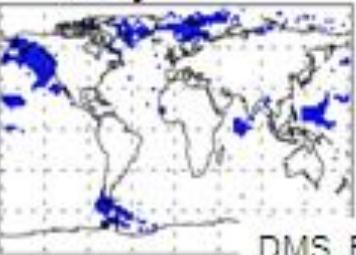
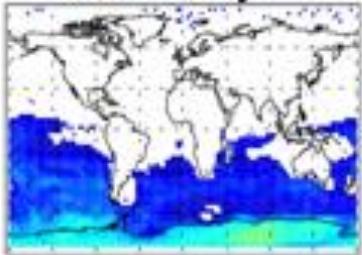
# 18. Surface parameter sensitivities



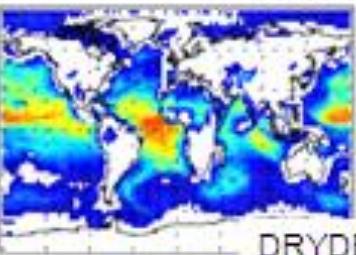
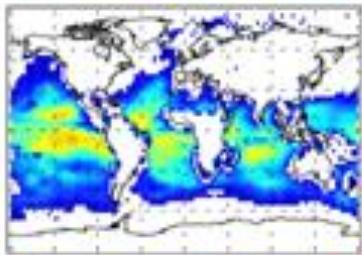
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January

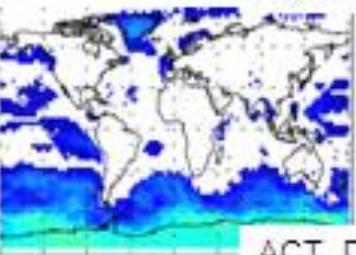
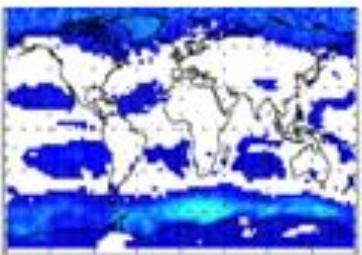
July



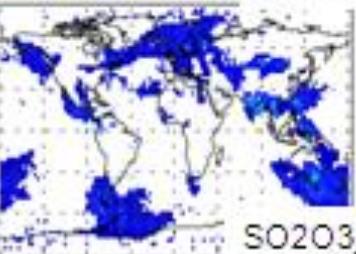
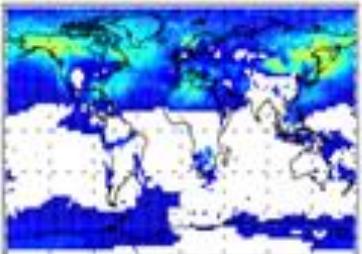
DMS\_FLUX



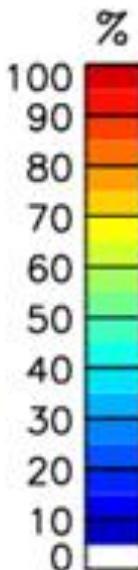
DRYDEP\_ACC



ACT\_DIAM

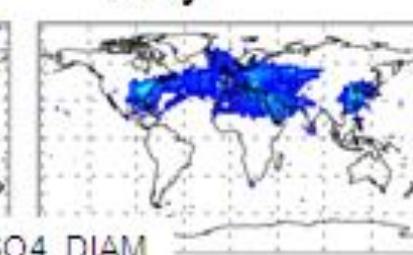
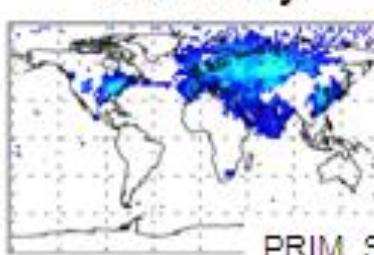


SO2O3\_CLEAN

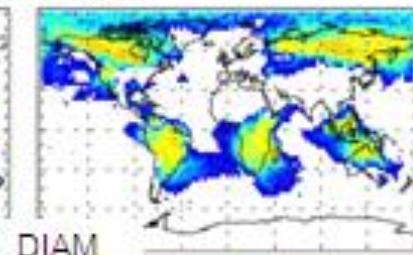
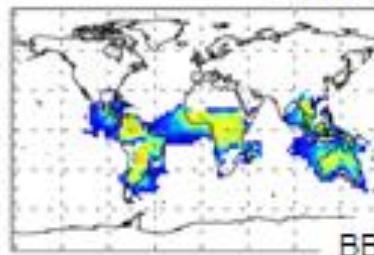


January

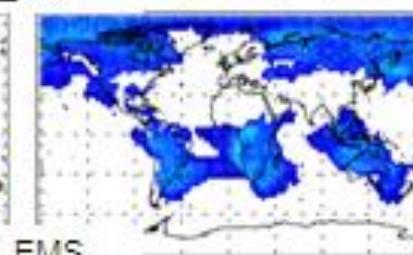
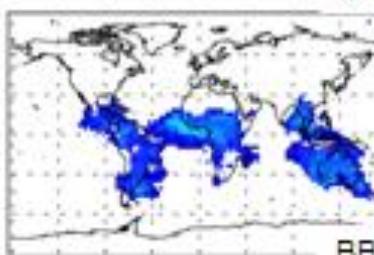
July



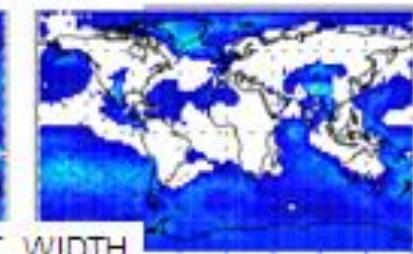
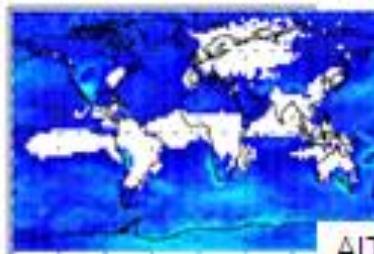
PRIM\_SO4\_DIAM



BB\_DIAM



BB\_EMS

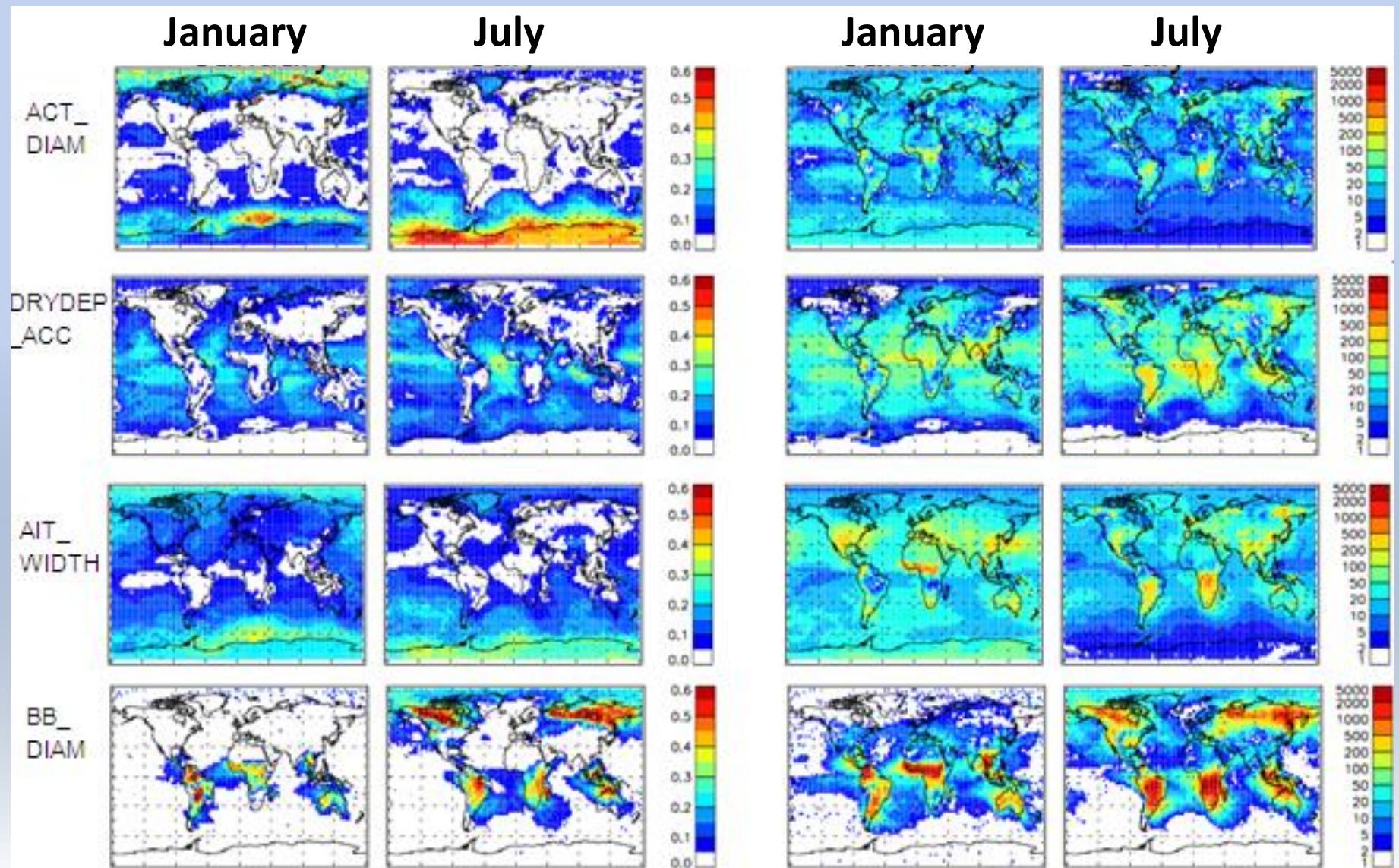


AIT\_WIDTH

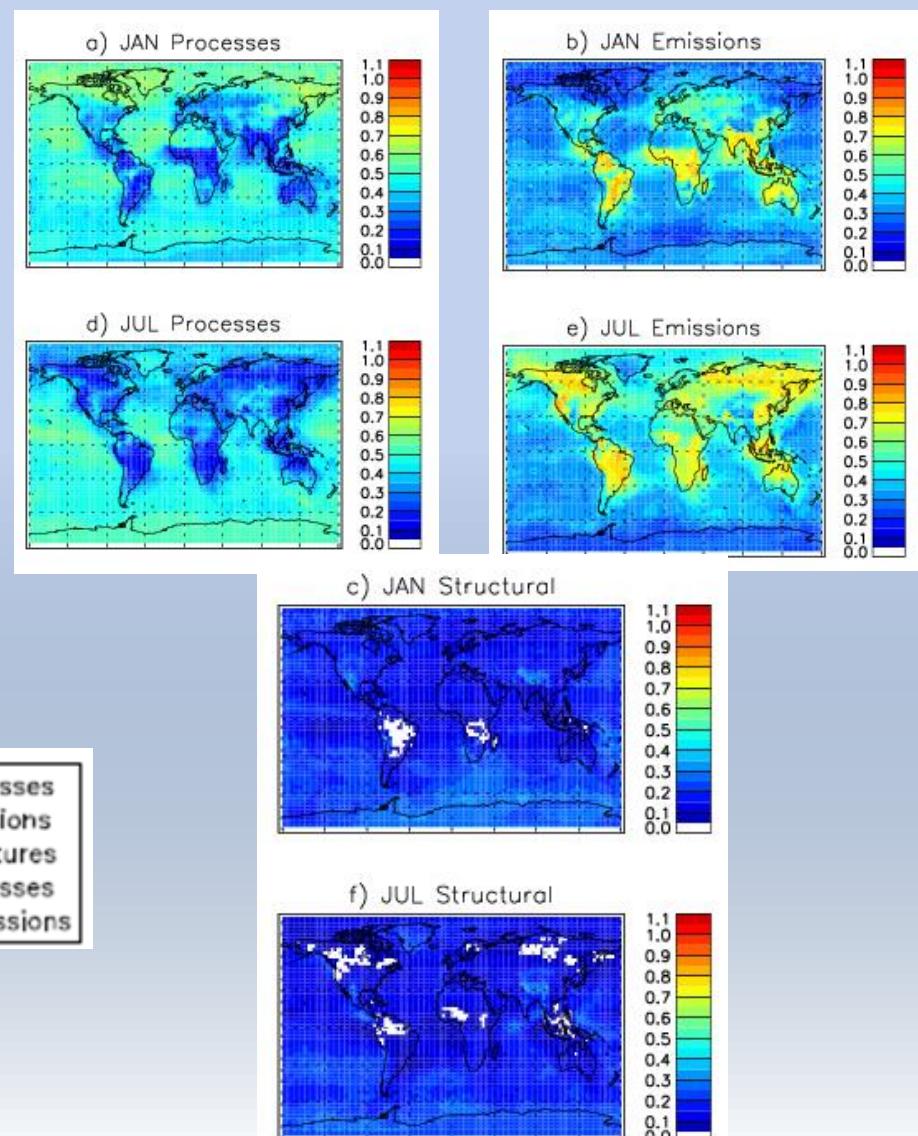
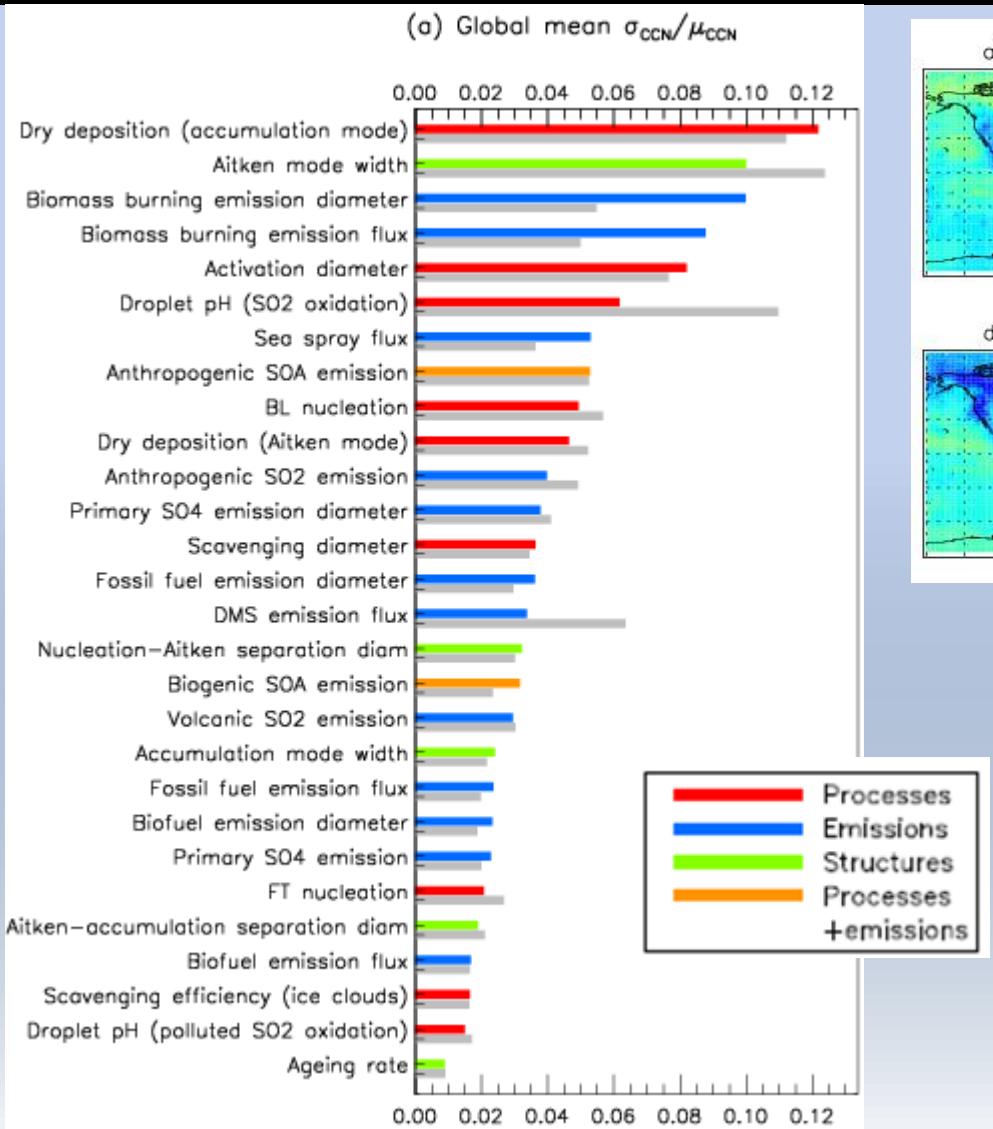
# 19. LHS: $\sigma_{CCN}/\mu_{CCN}$ and RHS: absolute $\sigma_{CCN}$



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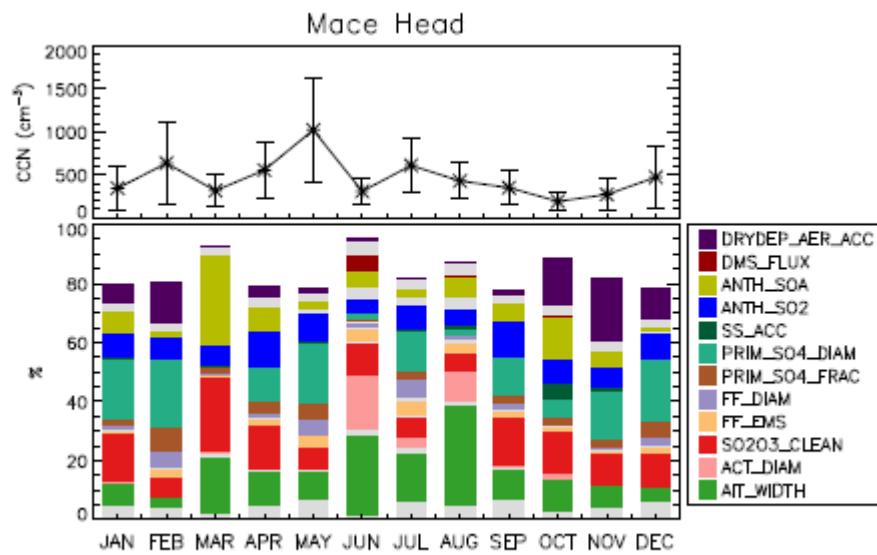
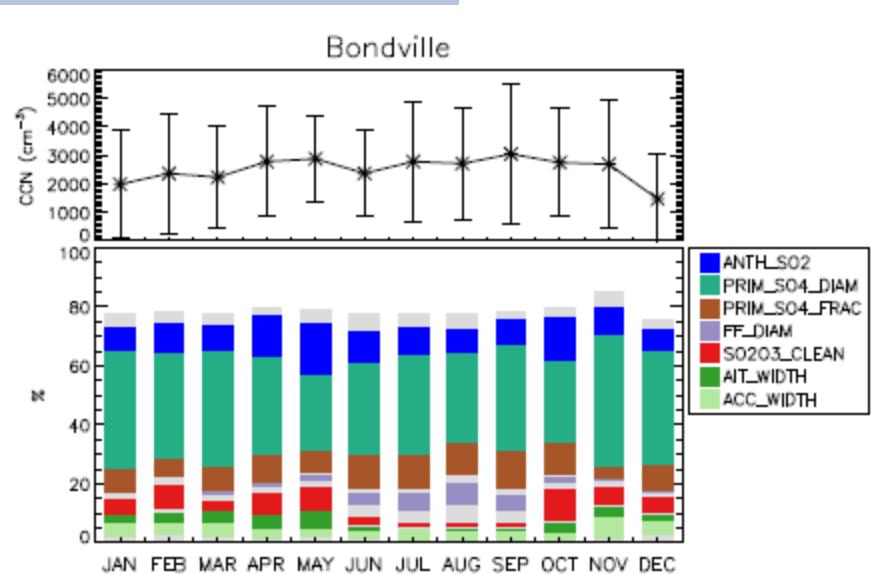
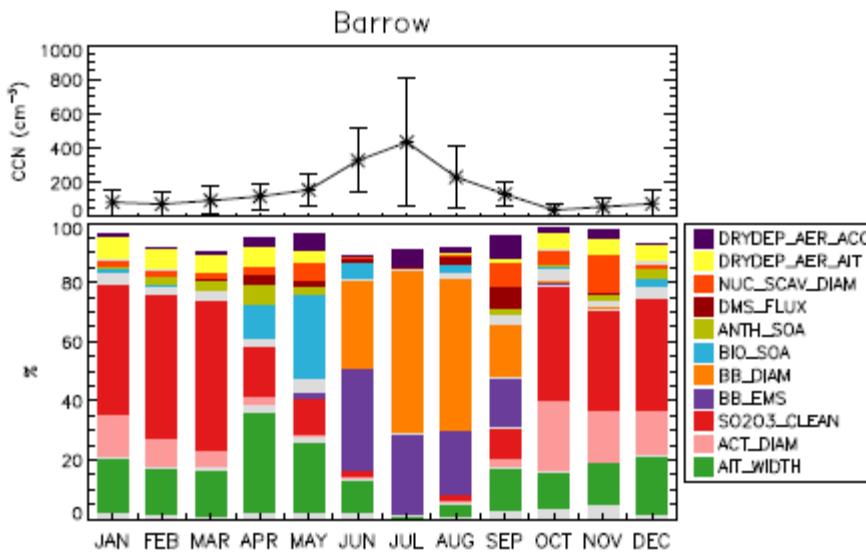
# 20. Summarising global maps



# 21. Seasonal parameter sensitivities

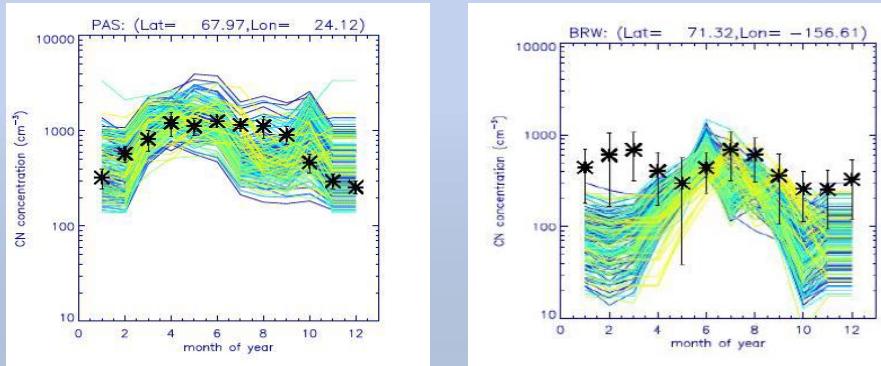


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## 22. Structural uncertainty

- Structural uncertainty
  - Considering the model uncertainty is any model configuration near to observations



- How do sensitivities in different models compare?
  - AEROCOM joint proposed experiments (presented by Ken Carslaw)



# References

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