

Using Gaussian Process Emulator to Explore the Source-Receptor Relationships of Black Carbon

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Source receptor relationship

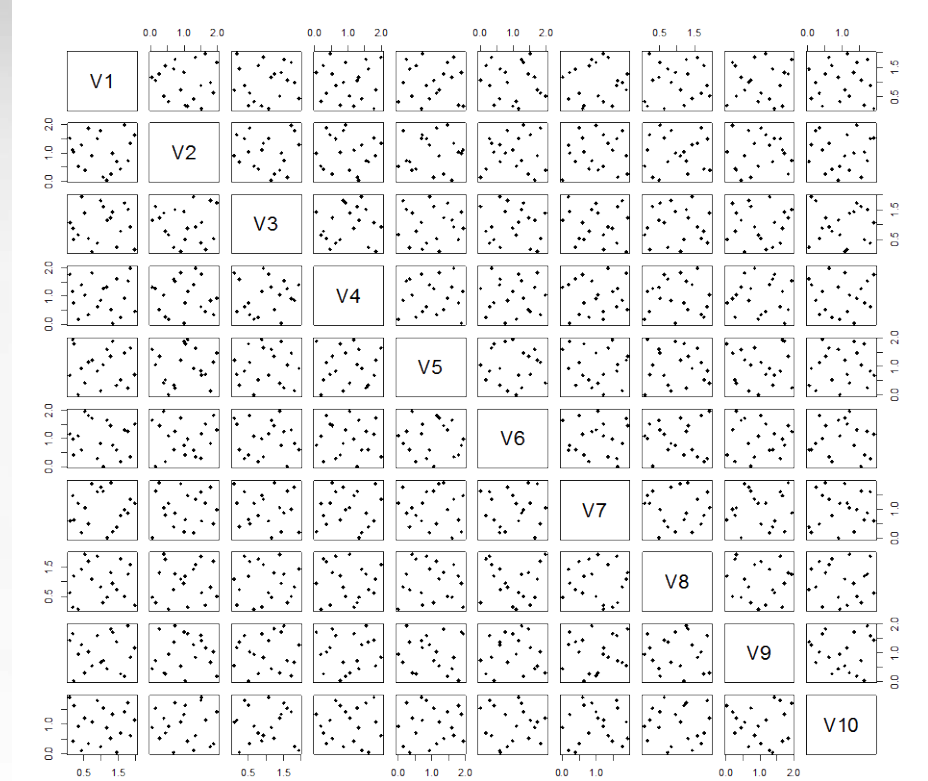
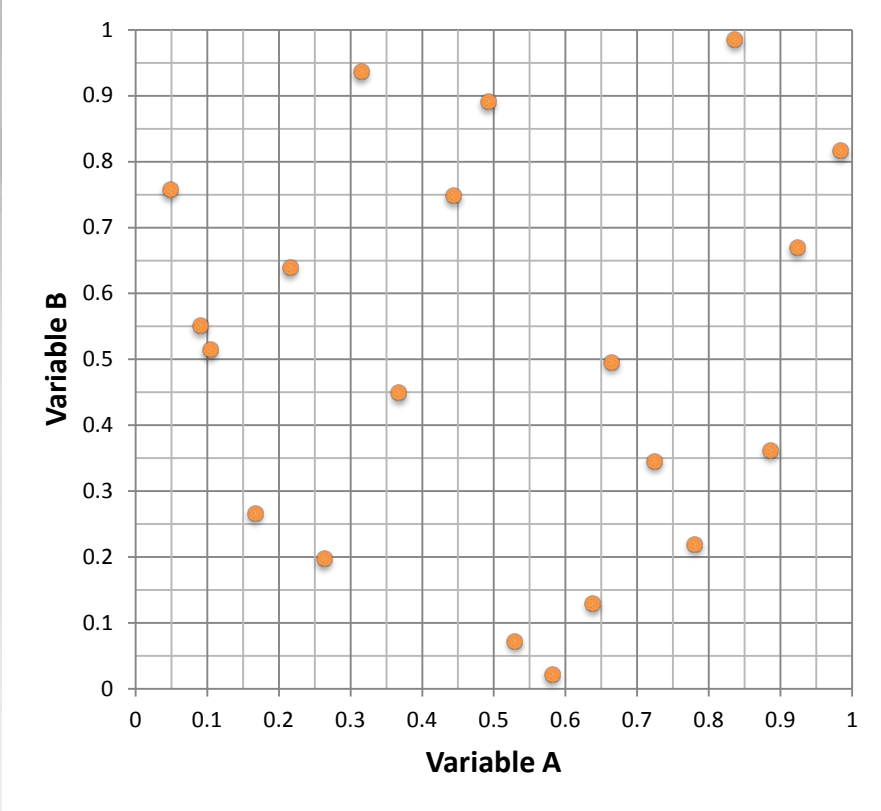
- simple sensitivity test
- tagging
- trajectory modeling
- chemical composition analysis
- inversion modeling
-

We are going to use a “UQ method” to explore this!



Design of experiments (DOE) vs. One-at-a-time (OAT)

Latin Hypercube DOE (20 samples, 10 variables)

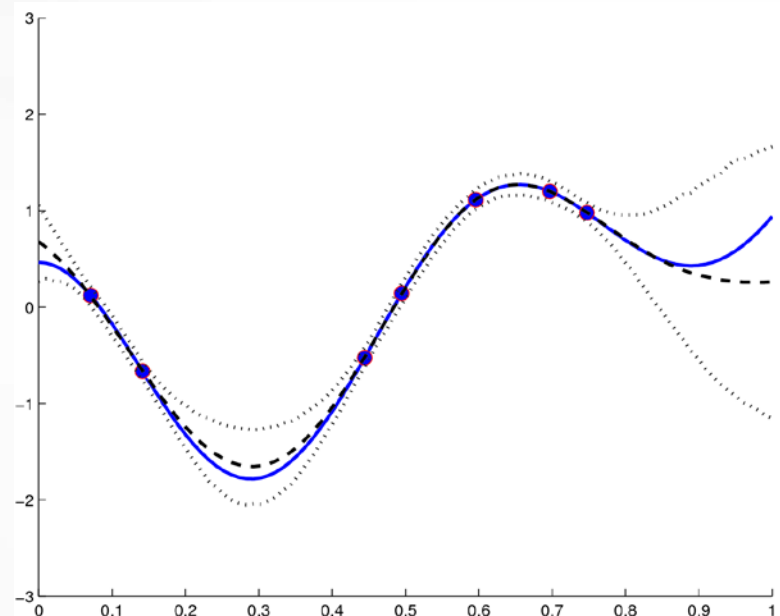
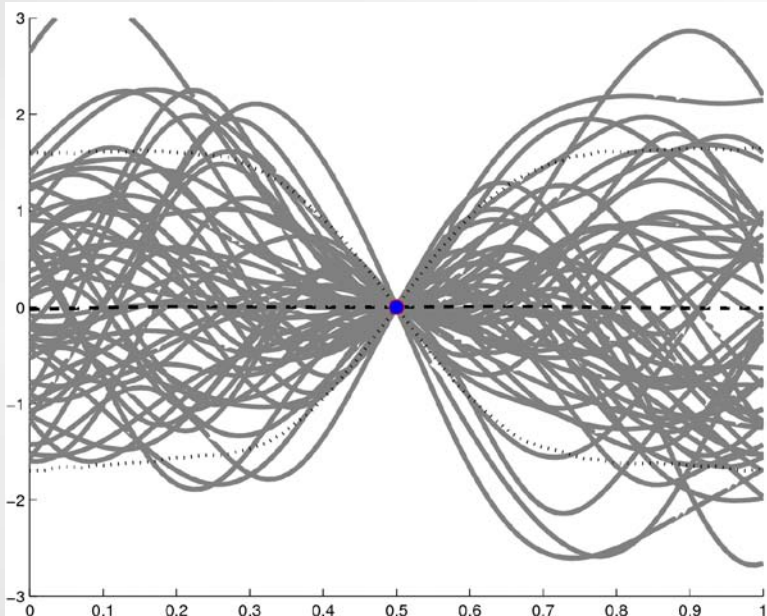
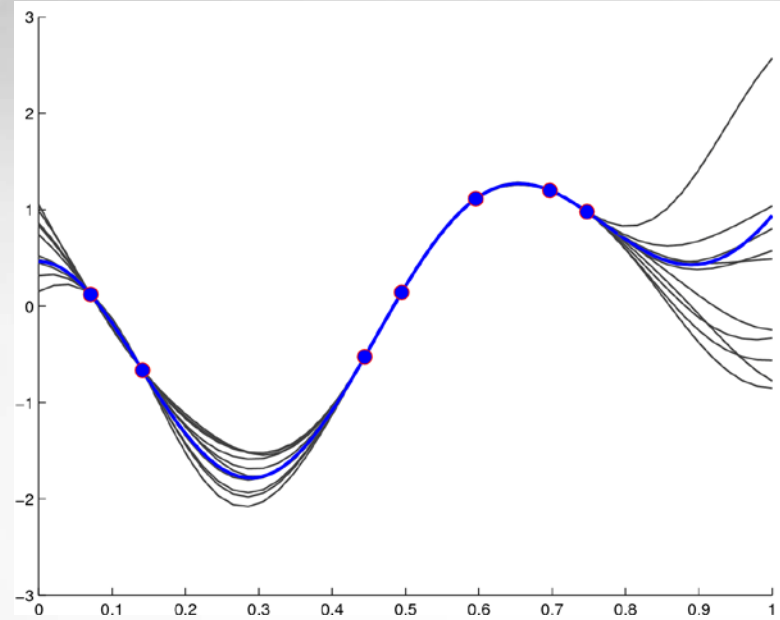
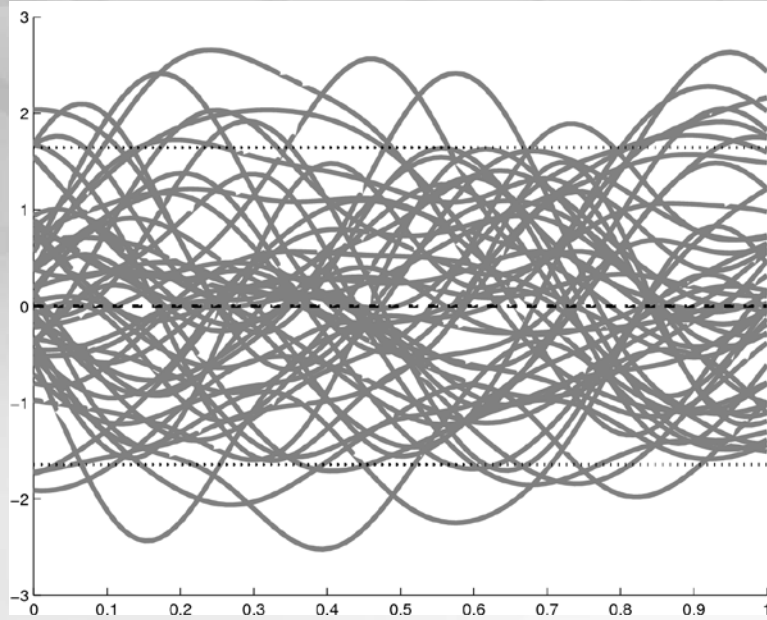


Advantage of DOE over OAT (Czitrom, 1999):

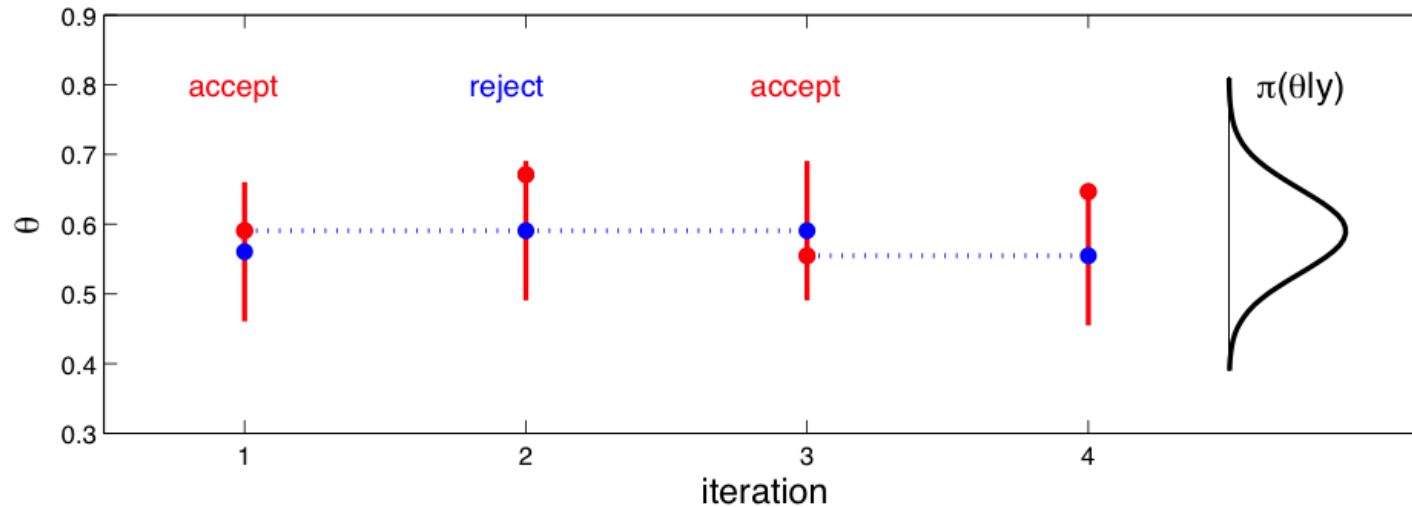
- Greatly **reduce** the number of experiments (**20** vs. 2×10 , 20×10 , or 20^{10} experiments)
- Provide more **precise** estimates of the effect of each parameter
- Give accurate estimates of the effect of the **interactions** between two factors
- Cover a **larger portion** of the parameter space



Statistical surrogate model, or “emulator”: Gaussian process model



Metropolis recipe for MCMC

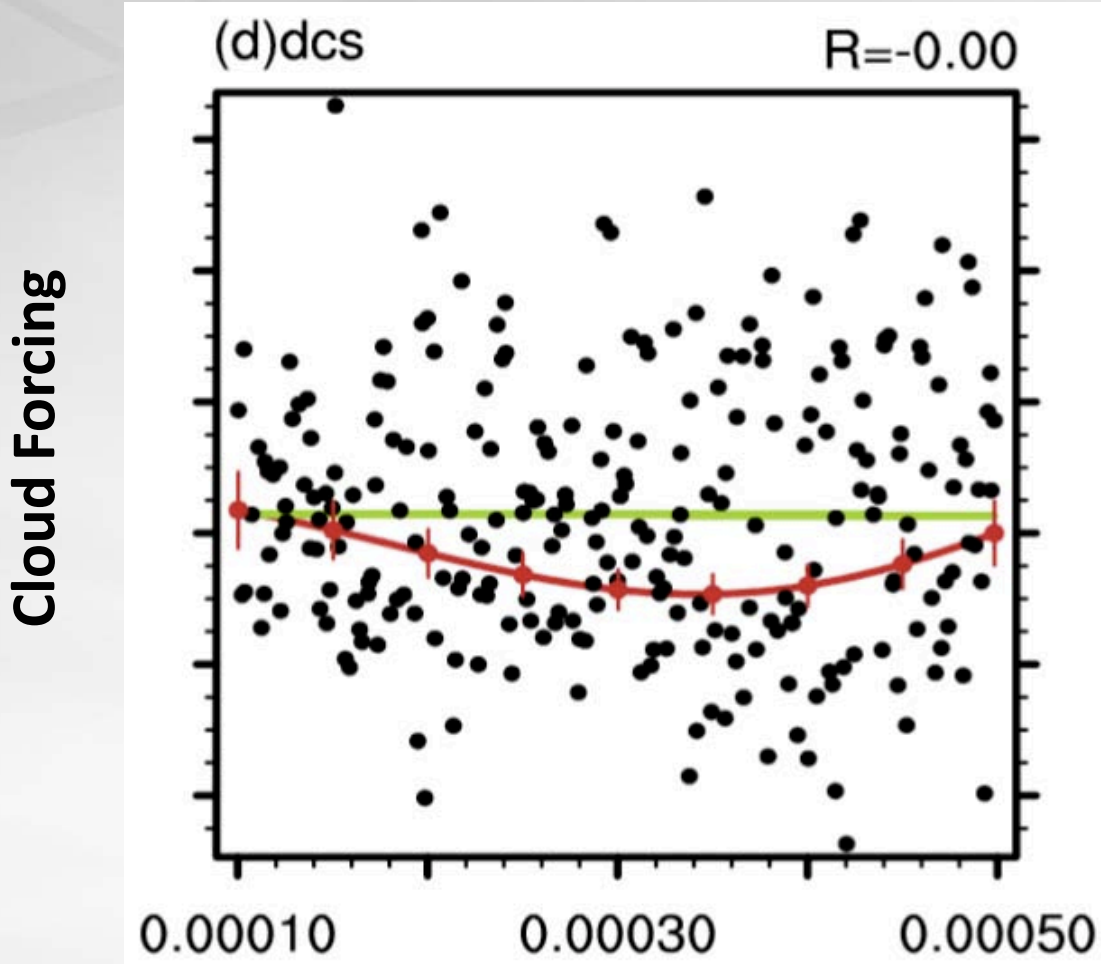


Initialize chain at θ^0

1. Given current realization θ^t , generate θ^* from a symmetric kernel $q(\theta^t \rightarrow \theta^*)$
i.e. $q(\theta^t \rightarrow \theta^*) = q(\theta^* \rightarrow \theta^t)$
2. Compute acceptance probability $\alpha = \min \left\{ 1, \frac{\pi(\theta^*|y)}{\pi(\theta^t|y)} \right\}$
3. Set $\theta^{t+1} = \theta^*$ with probability α , otherwise $\theta^{t+1} = \theta^t$
4. Iterate steps 1 – 3



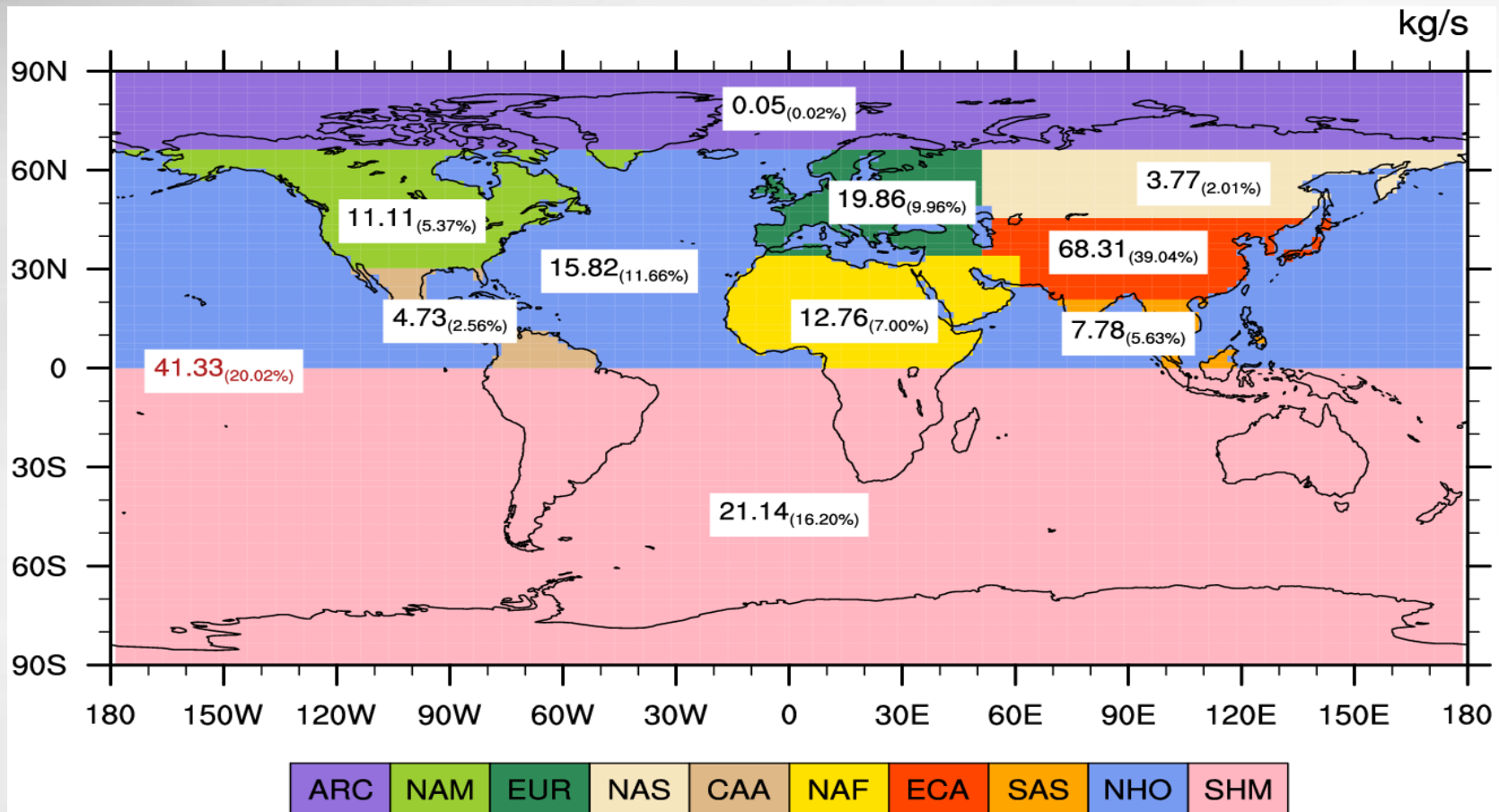
Example (global cloud forcing vs. dcs)



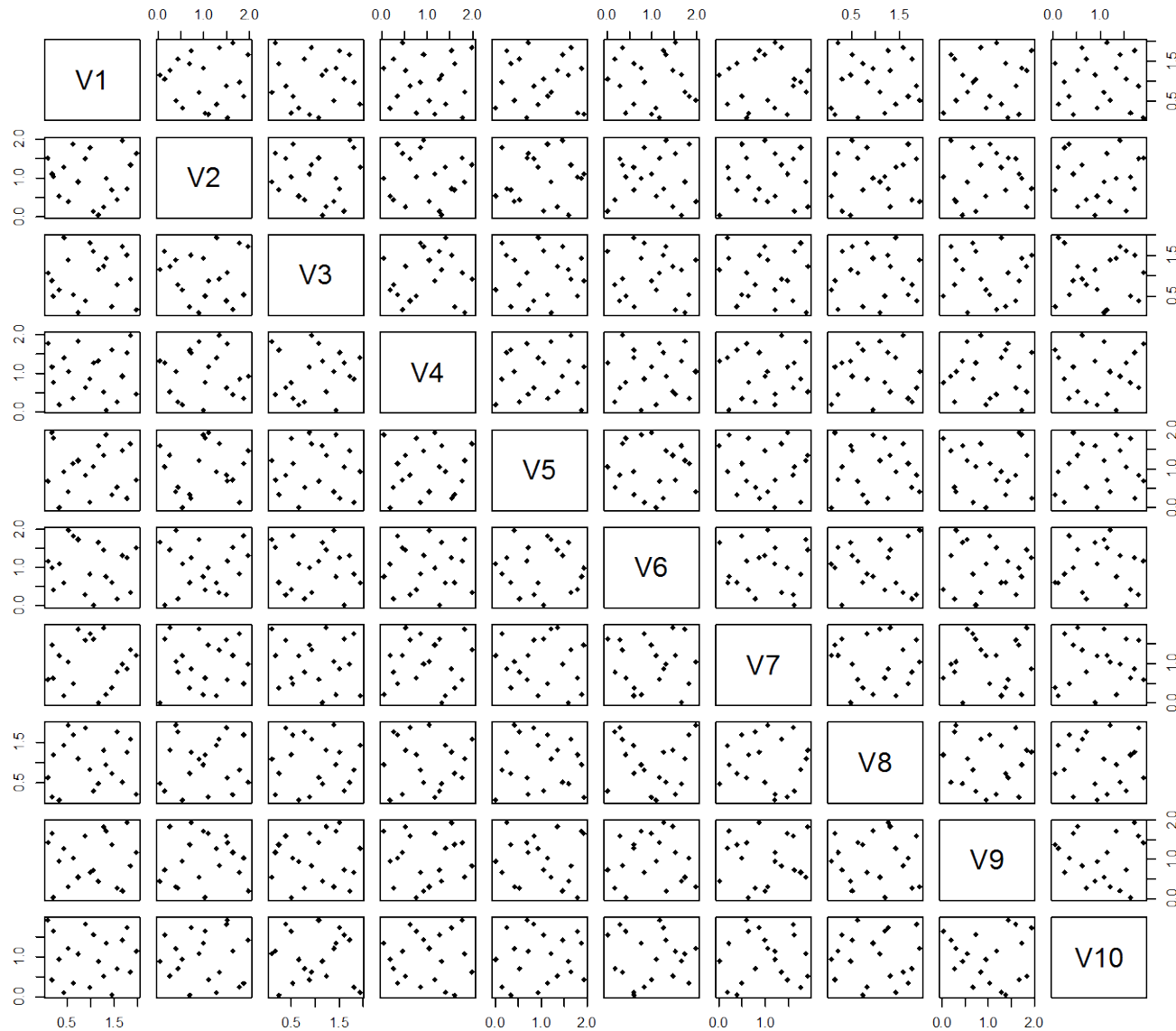


Design of Experiment

- Latin Hypercube DOE of 20 simulations
- Regional emission perturbation 0-10% of global emission
- Only anthropogenic emissions are perturbed
- CAM-driven and ERAI-driven simulations

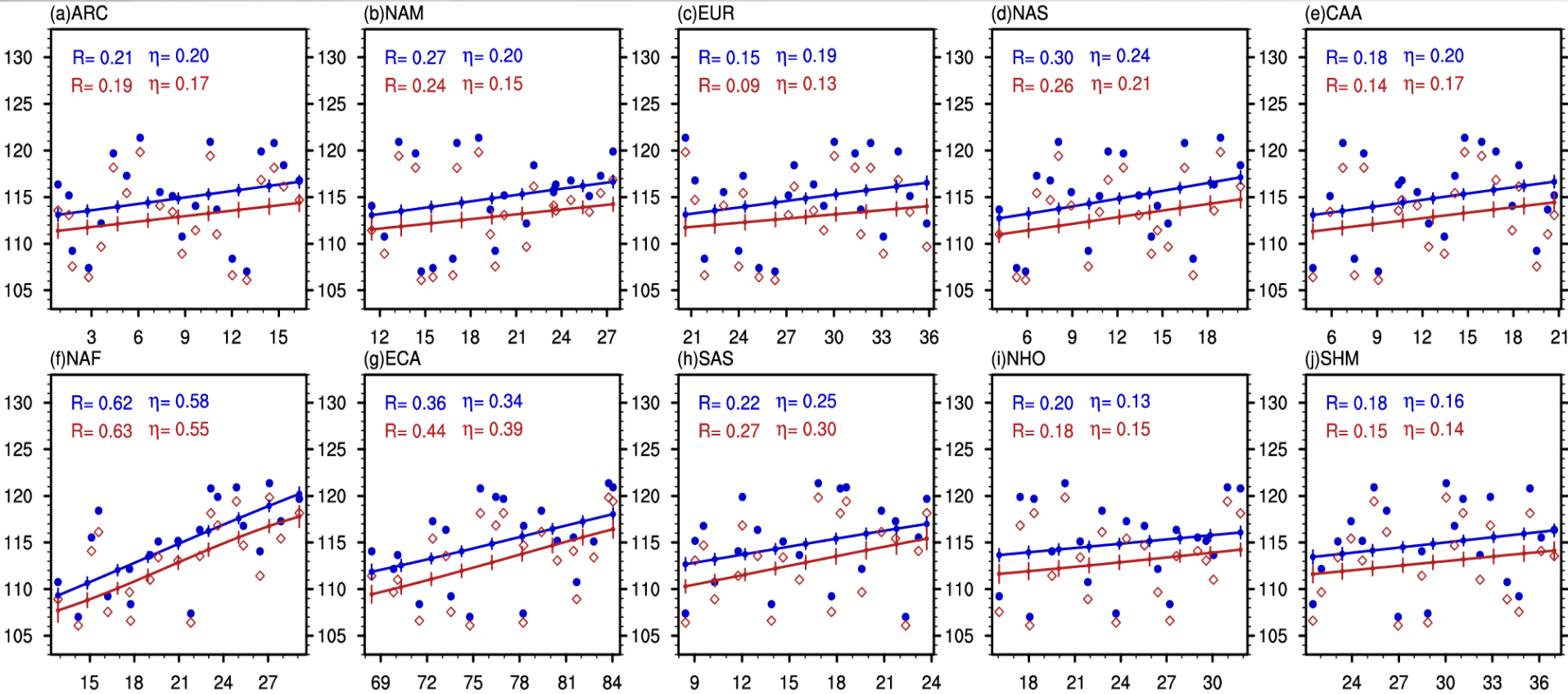


Latin Hypercube DOE (20 samples, 10 variables)



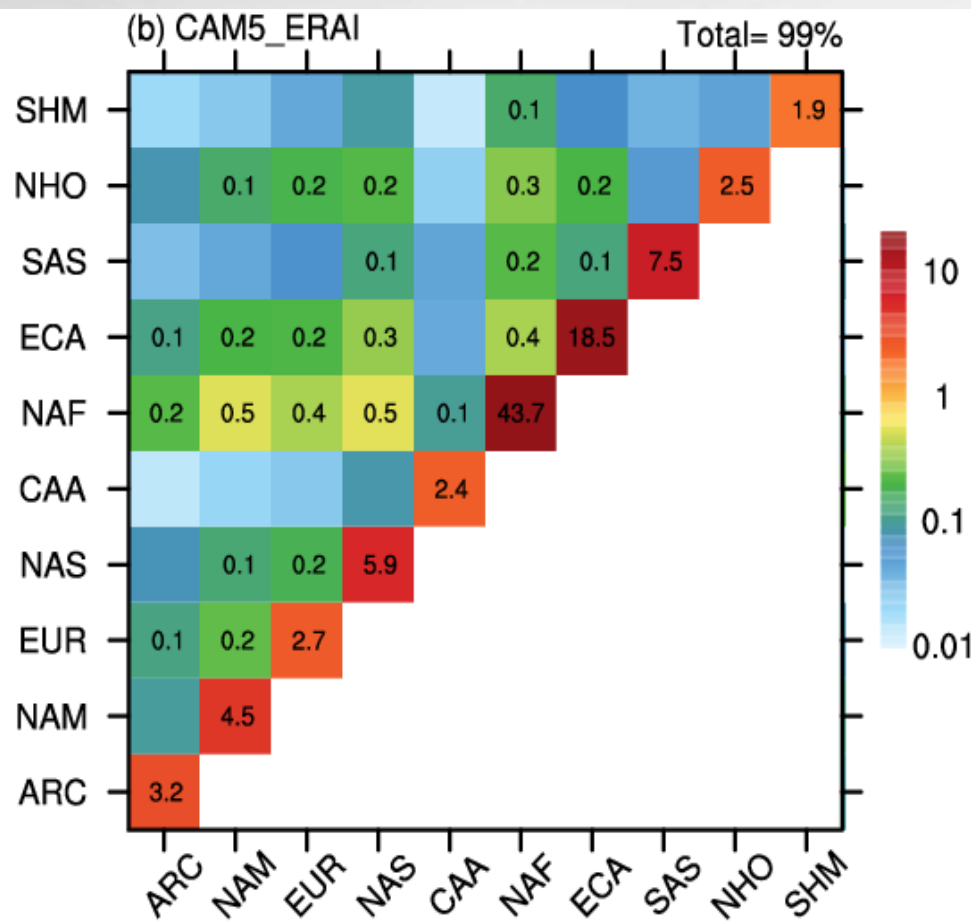
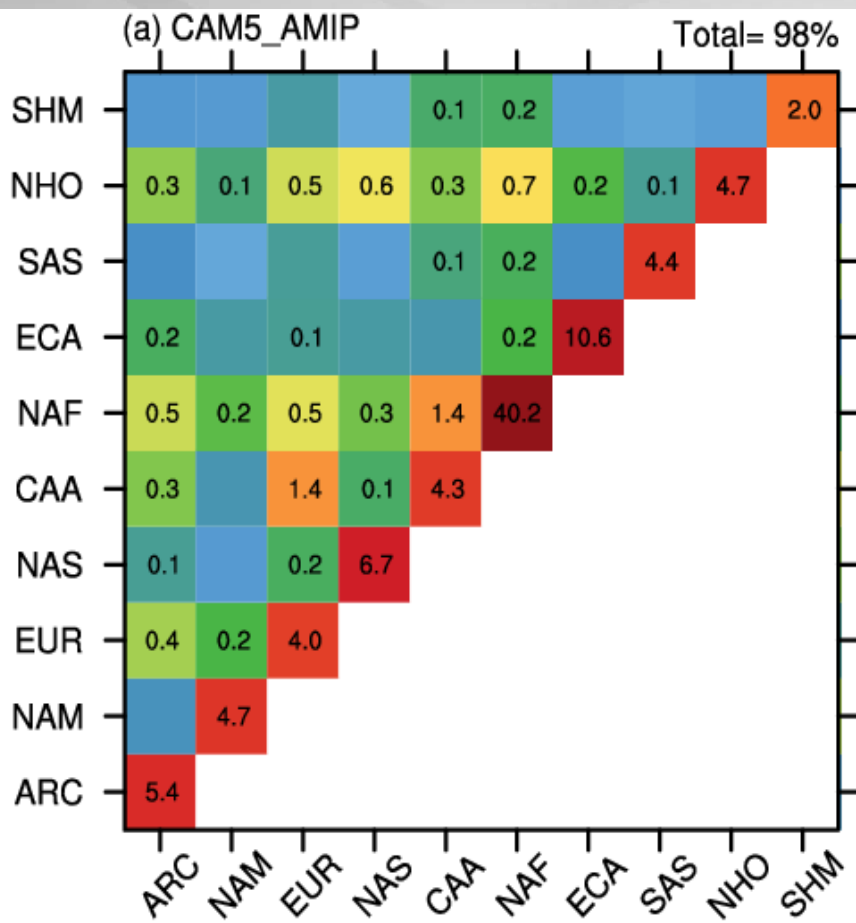


Global BC burden change due to regional emission perturbation



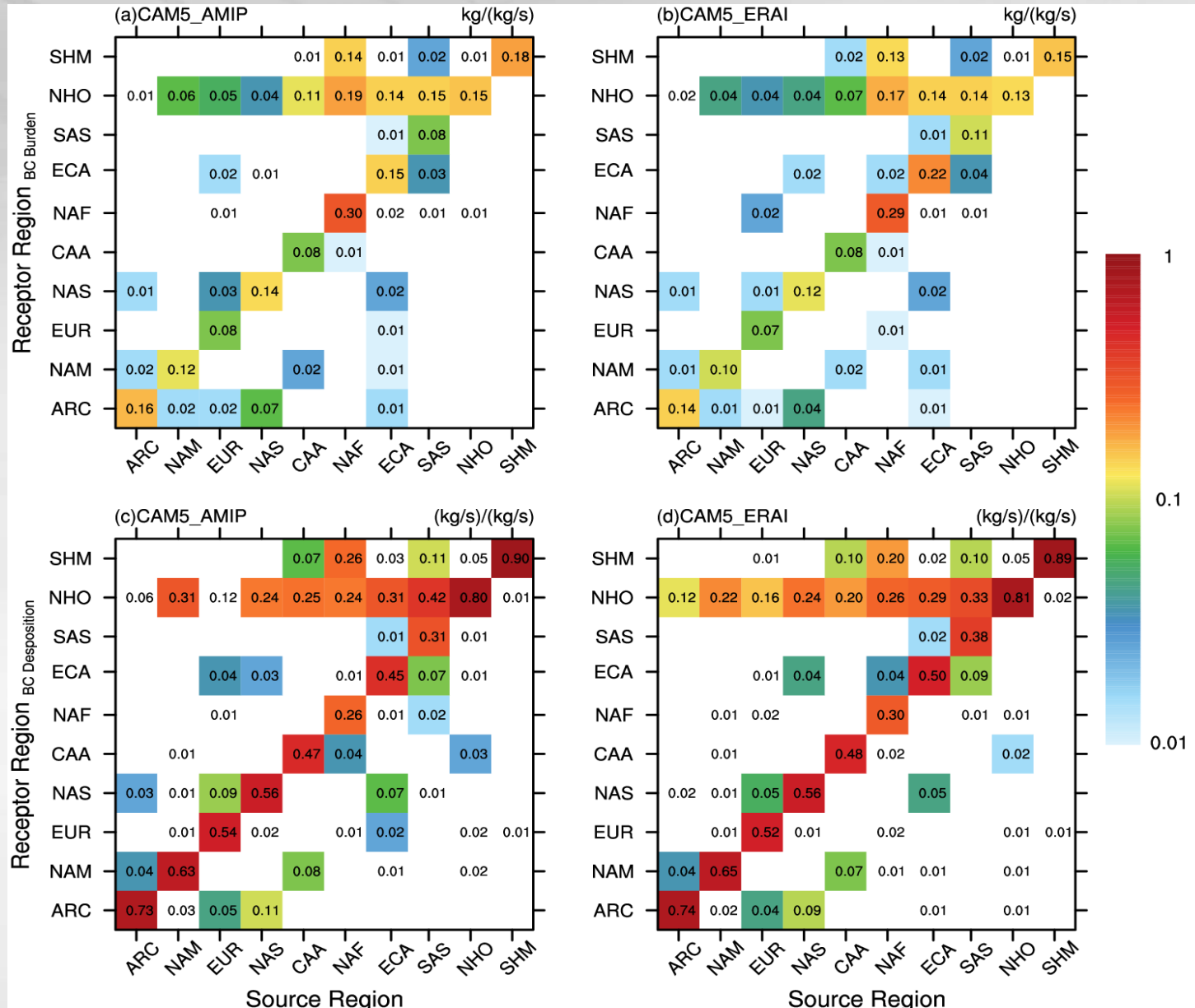


Variance of Global BC burden explained by main and interactive (2-variable) effects



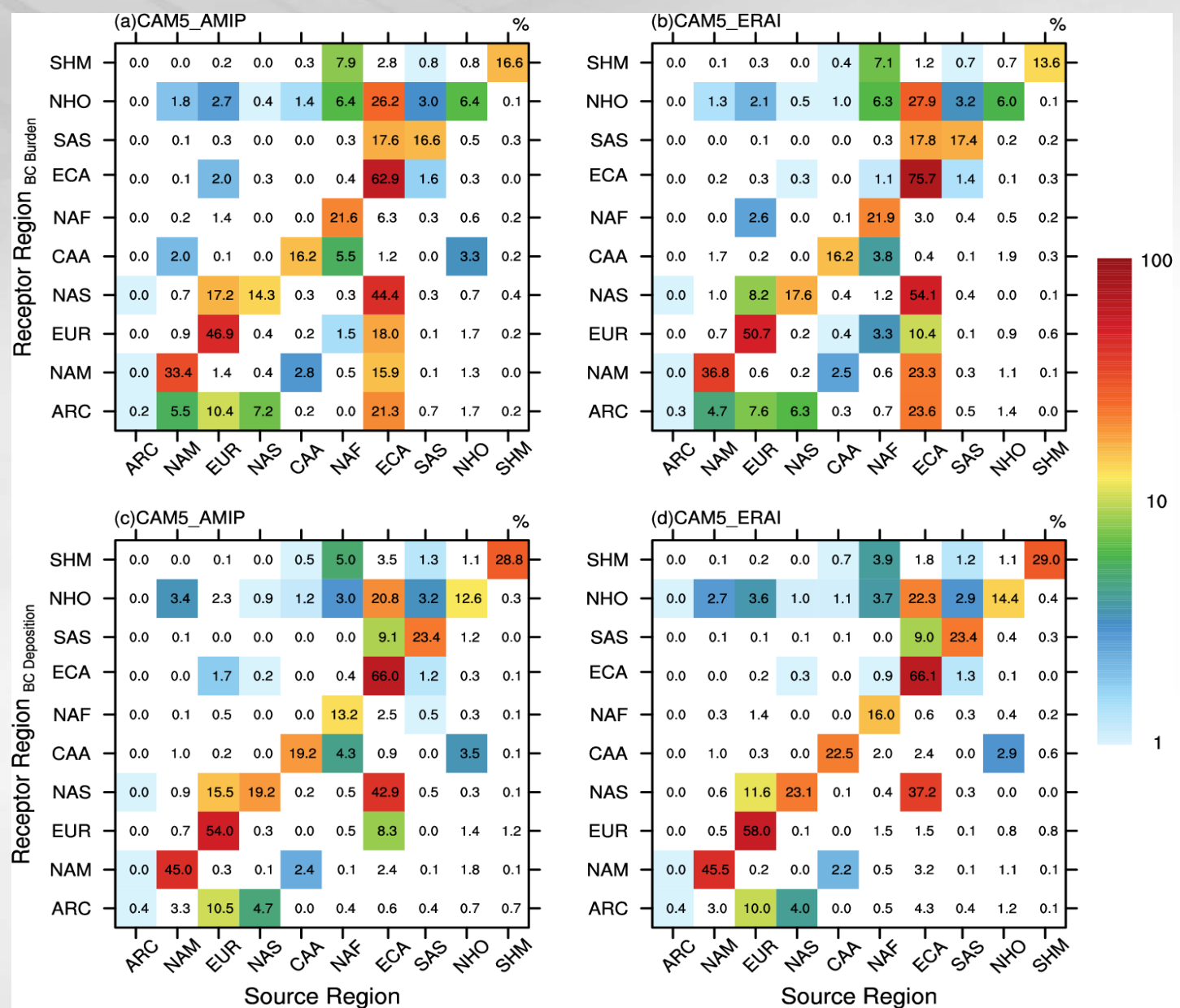


Contribution Efficiency





Source-Receptor Relationship





Concluding Remarks

- Owing to the complexity and nonlinearity of the climate system, it is **computationally expensive to systematically identify the cause-and-effect** of multiple factors in climate models. With an appropriate design of experiment and emulations, **sensitivity studies** can be achieved with relatively **little computational cost**.
- In this study, we have demonstrated that the **free-running** model CAM5 and the reanalysis-driven **offline** model produce **similar results**.
- BC source-receptor relationship is found **approximately linear**.
- The burden and surface deposition rates in every region are the **most vulnerable to the increase of local emissions**.
- Due to the **large emission**, **Asia** contributes a significant portion of BC burden and deposition over many regions in the North Hemisphere.