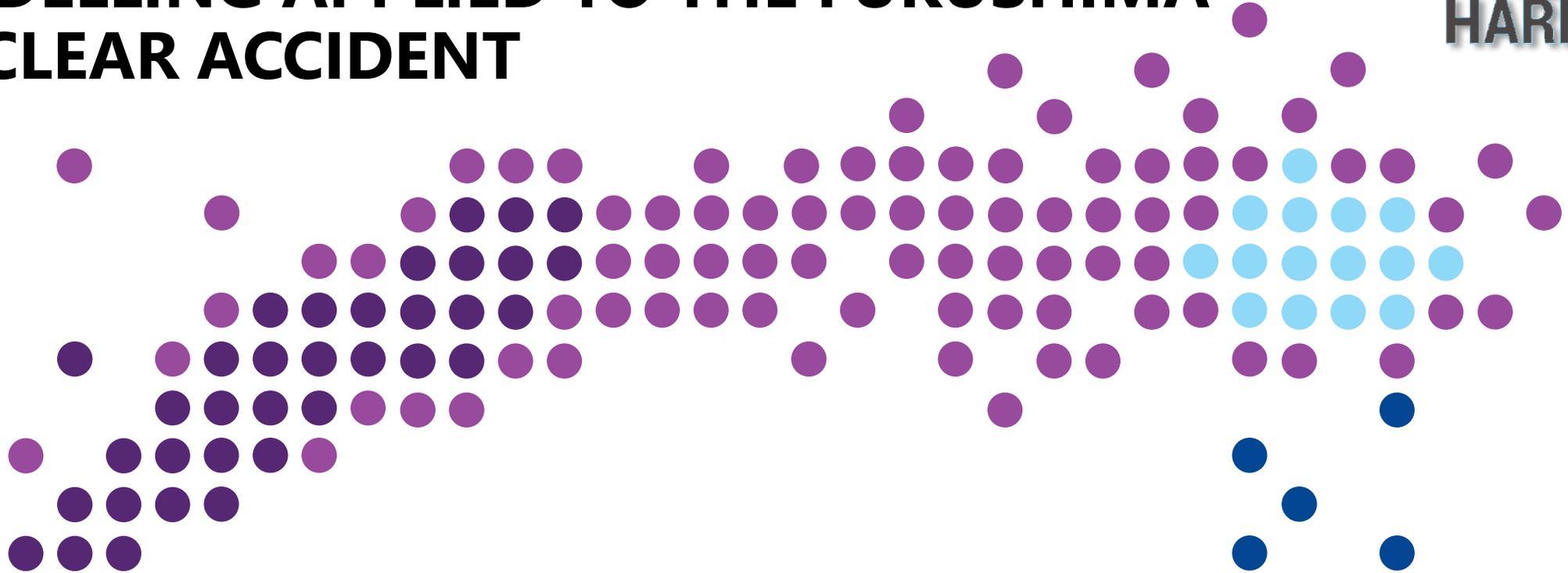


SENSITIVITIES IN WET DEPOSITION MODELLING APPLIED TO THE FUKUSHIMA NUCLEAR ACCIDENT



sck cen

Belgian Nuclear Research Centre



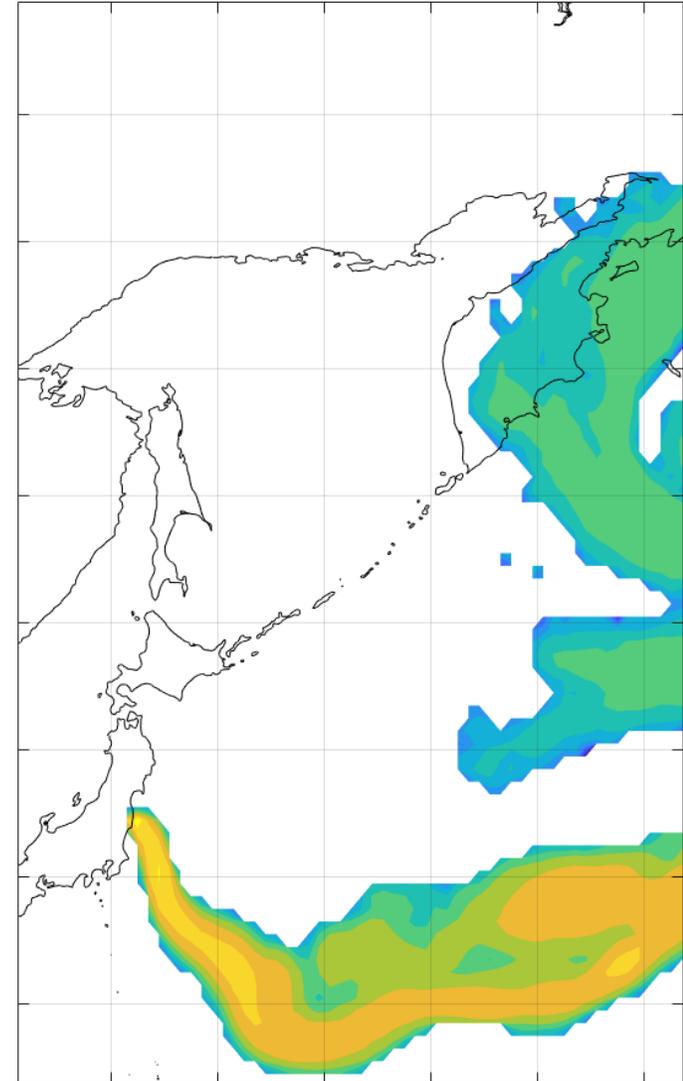
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Introduction

Wet deposition is main contributor to radioactive contamination

- Atmospheric transport modelling (ATM) = important for
 - emergency preparedness
 - emergency response
 } nuclear releases
- **Wet deposition** reduces air concentration by depositing radionuclides on the ground
→ contamination
- **BUT** difficult to model



Introduction

We propose a novel method to improve wet deposition modelling

- Wet deposition is difficult to model
 - Physical properties difficult to measure
 - Simulation parameters are possibly case-dependent
- **Solutions**
 - Brute force: simulate whole parameter space (possibly 1000's of simulations)
 - **This presentation**: novel method to improve wet deposition simulations through optimisation scheme

Introduction

Fukushima as case study with Flexpart



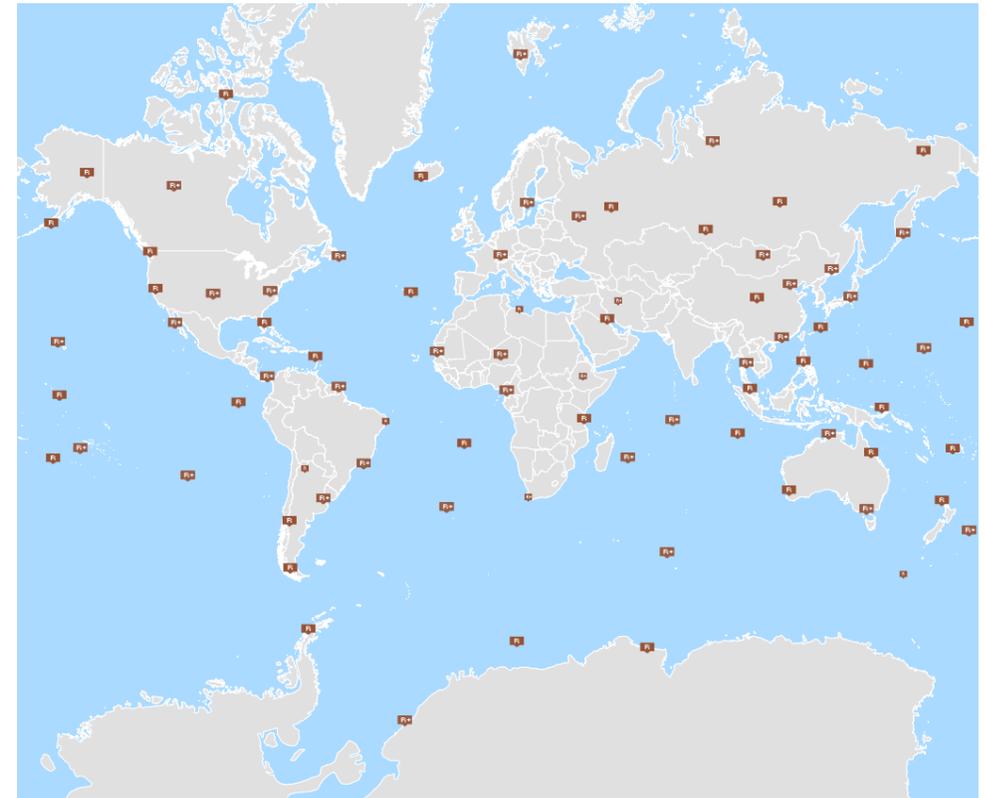
- **Case study**
 - ^{137}Cs transport + deposition following **Fukushima** incident (2011)
- **Flexpart**
 - Stochastic **Lagrangian particle** model
 - Meteodata from ECMWF



Introduction

Concentration measurements provided by CTBTO

- **IMS** radionuclide stations from **CTBTO**
 - International **M**onitoring **S**ystem
 - **C**omprehensive Nuclear-**T**est-**B**an **T**reaty **O**rganization
- 80 stations worldwide
 - 20 with highest measurements as 'receptors' in Flexpart



www.ctbto.org/map

Introduction

How is wet deposition modelled?

- **Scavenging:** exponential decay process

- $c(t + \Delta t) = c(t)\exp(-\Lambda\Delta t)$

- c : concentration [Bq m^{-3}]

- Λ : scavenging coefficient [s^{-1}]

- Δt : timestep [s]

- **Deposition:**

- $F = \int dz \Delta c$ [Bq m^{-2}]

$$\Lambda = \Lambda(I, d_p)$$

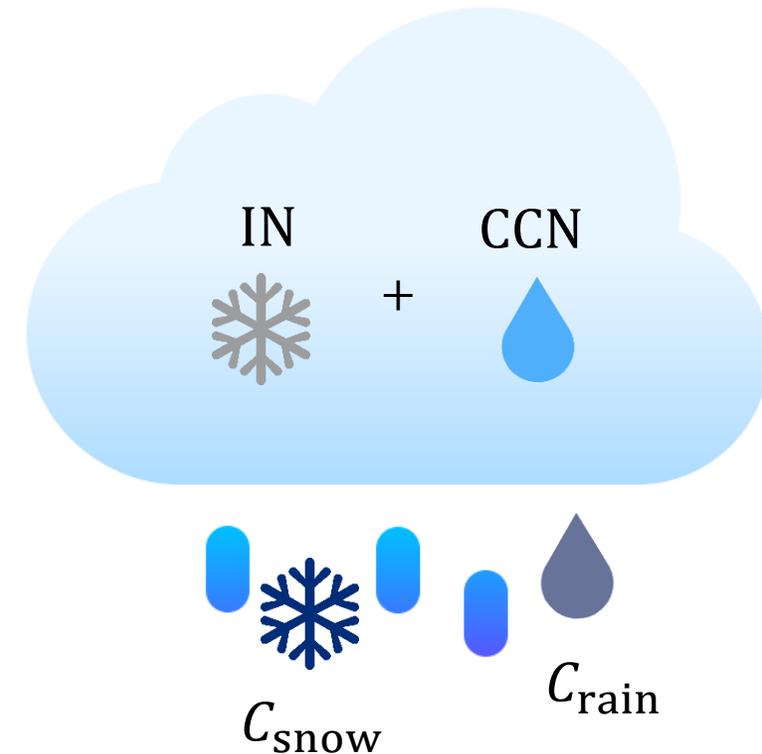
- I : rain intensity [mm h^{-1}]

- d_p : particle diameter (distribution) [μm]

Introduction

Different processes are represented in Flexpart 10.4 (aerosols)

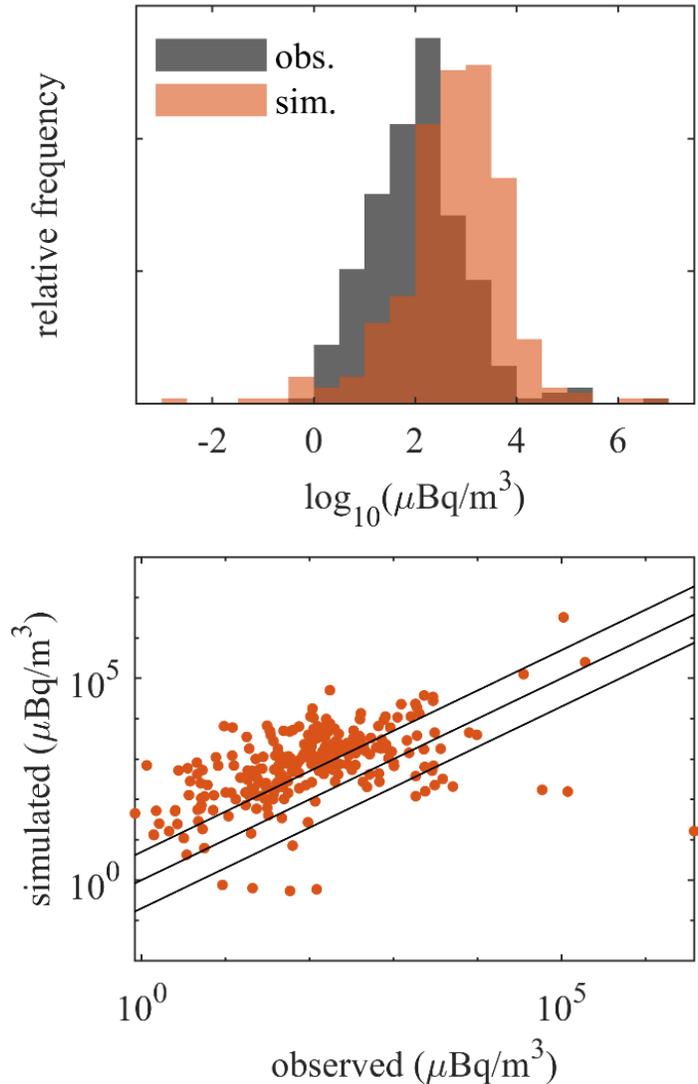
- **In-cloud** scavenging by
 - **cloud condensation nucleation** (CCN)
 - **ice nucleation** (IN)
- **Below-cloud** scavenging by
 - **rain** collision (C_{rain})
 - **snow** collision (C_{snow})
- **Total** scavenging: $\Lambda = \sum_i \Lambda_i$



Introduction

Default simulations show overestimation of concentration

- Default (CCN, IN, C_{rain} , C_{snow}) values
- Concentrations **too high** (x10)
 1. Source term? 
 - Literature: only factor x2-4 uncertainty
 2. Wind fields? 
 - Xe simulations show no bias
 - Xe: noble gas \rightarrow no deposition
 3. Wet scavenging too weak 



Methodology

We propose a 2-step method to improve wet deposition modelling

1. **Extract** scavenging contribution of each process from single Flexpart simulation

(CCN, IN, C_{rain} , C_{snow})

2. **Optimisation scheme**: scale individual scavenging contributions and fit remaining concentration to observations

Methodology

STEP 1 – extract scavenging contributions

- Concentration left over at location \vec{x} & after time T since release

$$c(\vec{x}, T) = c_0(\vec{x}, T) - \sum_i \Delta c_i(\vec{x}, T)$$

- c_0 : concentration without scavenging
- $\Delta c_i = \int_0^T \Lambda_i c dt$: contribution of each scavenging process (CCN, IN, C_{rain} , C_{snow}) between release and time T
→ extract from Flexpart: altering source code

Methodology

$$c = c_0 - \sum_i \Delta c_i$$

STEP 2 – optimise scavenging: scaling

- How to scale Δc_i 's?
 - Simply scaling $x_i \Delta c_i$ can produce negative c
→ **not physical!**
- **Solution:** introduce scaling factor A_i for every Δc_i so that
 - $\Delta c_i = (c + \Delta c_i) A_i$
 - A_i acts on part of the concentration that is 'available' to process i

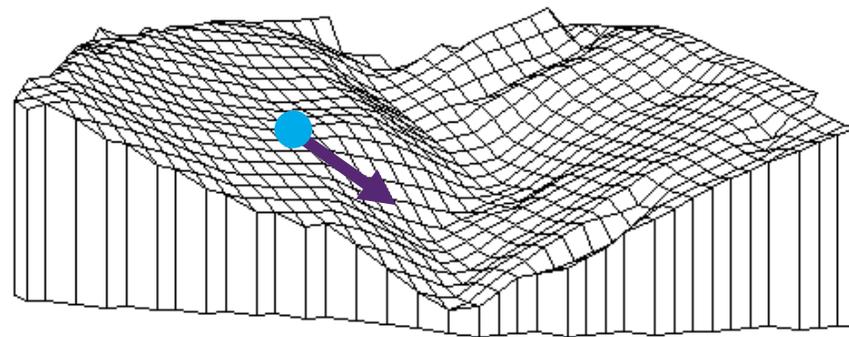
Methodology

STEP 2 – optimise scavenging: fitting

$$c = c_0 - \sum_i \Delta c_i$$

$$\Delta c_i = (c + \Delta c_i) A_i$$

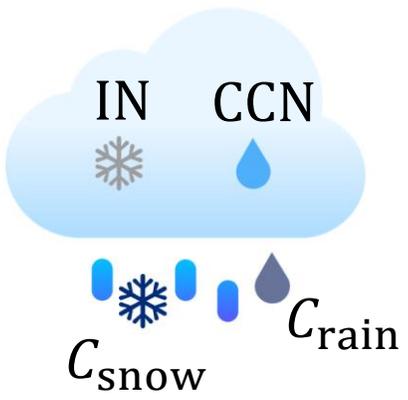
- Minimising cost function by varying A_i 's
 - $F(c, c_{\text{obs}}; A_i) = (\log_{10} c(A_i) - \log_{10} c_{\text{obs}})^2$
- Minimisation in 4-dimensional parameter space
 - $(\text{CCN}, \text{IN}, C_{\text{rain}}, C_{\text{snow}})$



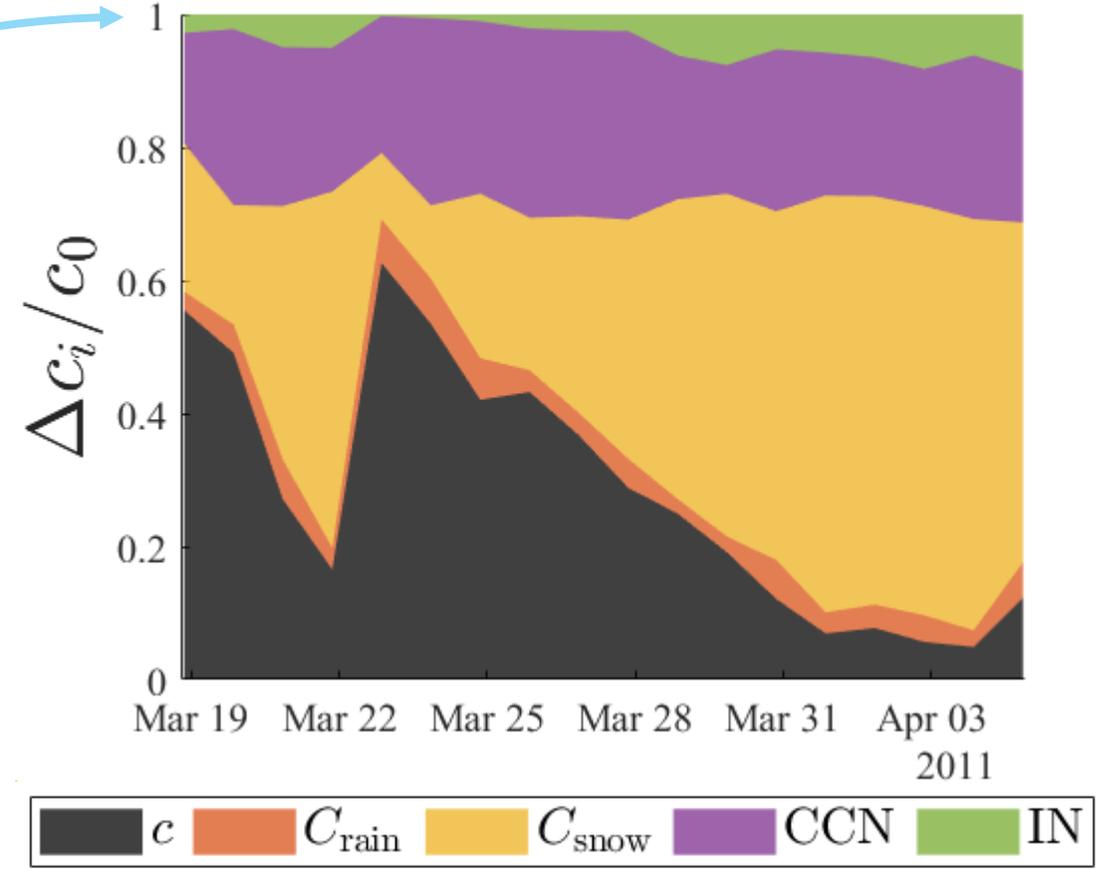
Results

Sanity check: step 1 works!

- $c = c_0 - \sum_i \Delta c_i$
 $\rightarrow \frac{c}{c_0} + \sum_i \frac{\Delta c_i}{c_0} = 1$
- Default (CCN, IN, C_{rain} , C_{snow}) values in Flexpart

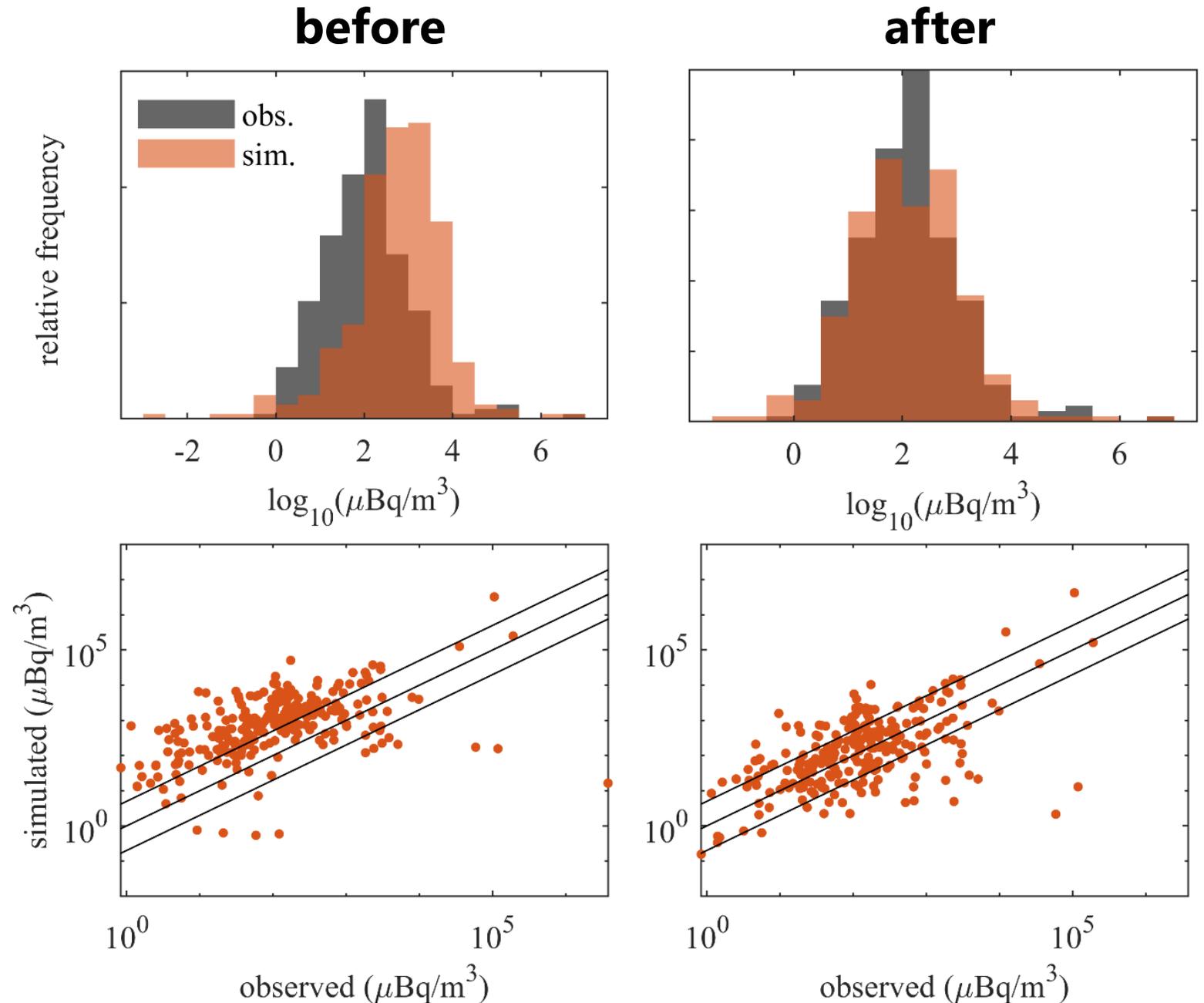


IMS station RN71



Results

After step 2:
optimisation
process makes big
improvement

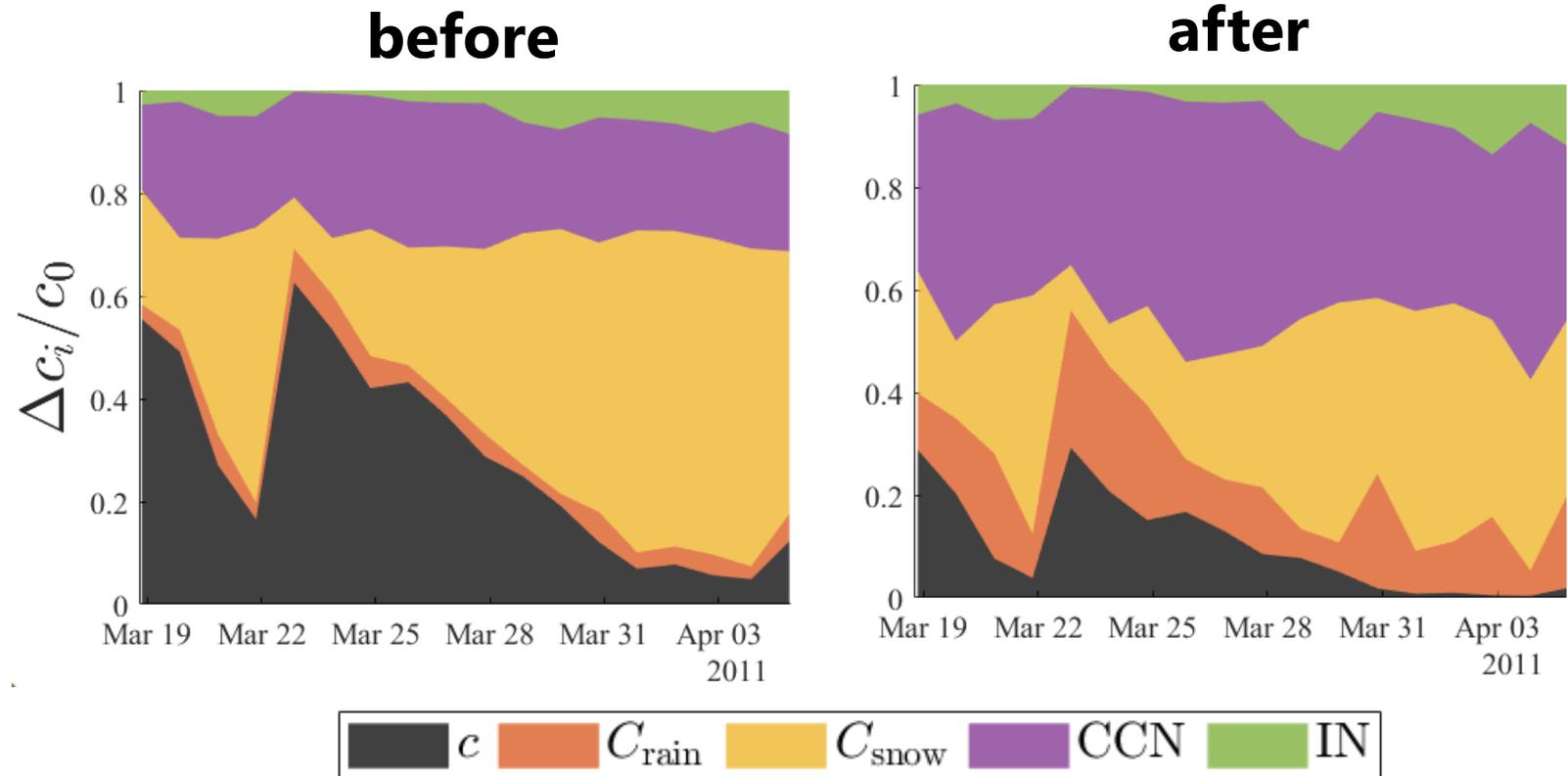


Results

What happened behind the scenes...



- Scaling factor of all scavenging processes increased
- Greatest increase for **CCN** and **C_{rain}**
- Due to compensating effects, relative contribution of some processes can decrease (e.g. **C_{snow}**)



Conclusions

- ✓ The proposed optimisation scheme is able to improve simulation-observation correspondence by scaling the wet scavenging contributions of different scavenging processes
- ✓ This method is more efficient than a brute force method, as it in principle only requires 1 simulation



Questions?