

Explorer le potentiel de History Matching pour calibrer les paramètres d'ORCHIDEE

Simon Beylat, Nina Raoult, James M. Salter, Frédéric Hourdin, Vladislav Bastrikov, Catherine Ottlé, and Philippe Peylin

ORCHIDEE



- Interaction between land and atmosphere.
- Energy and Water Balance
- Carbon Budget

! A lot of empirical equation = Many parameters !

For example :

Vegetation describe using 15 plant functional types :

PFT	Climate	Vegetation type	Phenology class
1	Global	NA	Bare soil
2	Tropical	Woody	Broadleaf evergreen
3	Tropical	Woody	Broadleaf deciduous
4	Temperate	Woody	Needleleaf evergreen
5	Temperate	Woody	Broadleaf evergreen
6	Temperate	Woody	Broadleaf summer green
7	Boreal	Woody	Needleleaf evergreen
8	Boreal	Woody	Broadleaf summer green
9	Boreal	Woody	Needleleaf deciduous
10	Temperate	Herbaceous	Natural (C ₃)
11	Global	Herbaceous	Natural (C ₄)
12	Global	Herbaceous	Managed (C ₃)
13	Global	Herbaceous	Managed (C ₄)
14	Tropical	Herbaceous	Natural (C ₃)
15	Boreal	Herbaceous	Natural (C ₃)

ORCHIDEE use many free parameters to describe vegetation.

A parameter has a value for each PFTs -> Very large number of parameters.

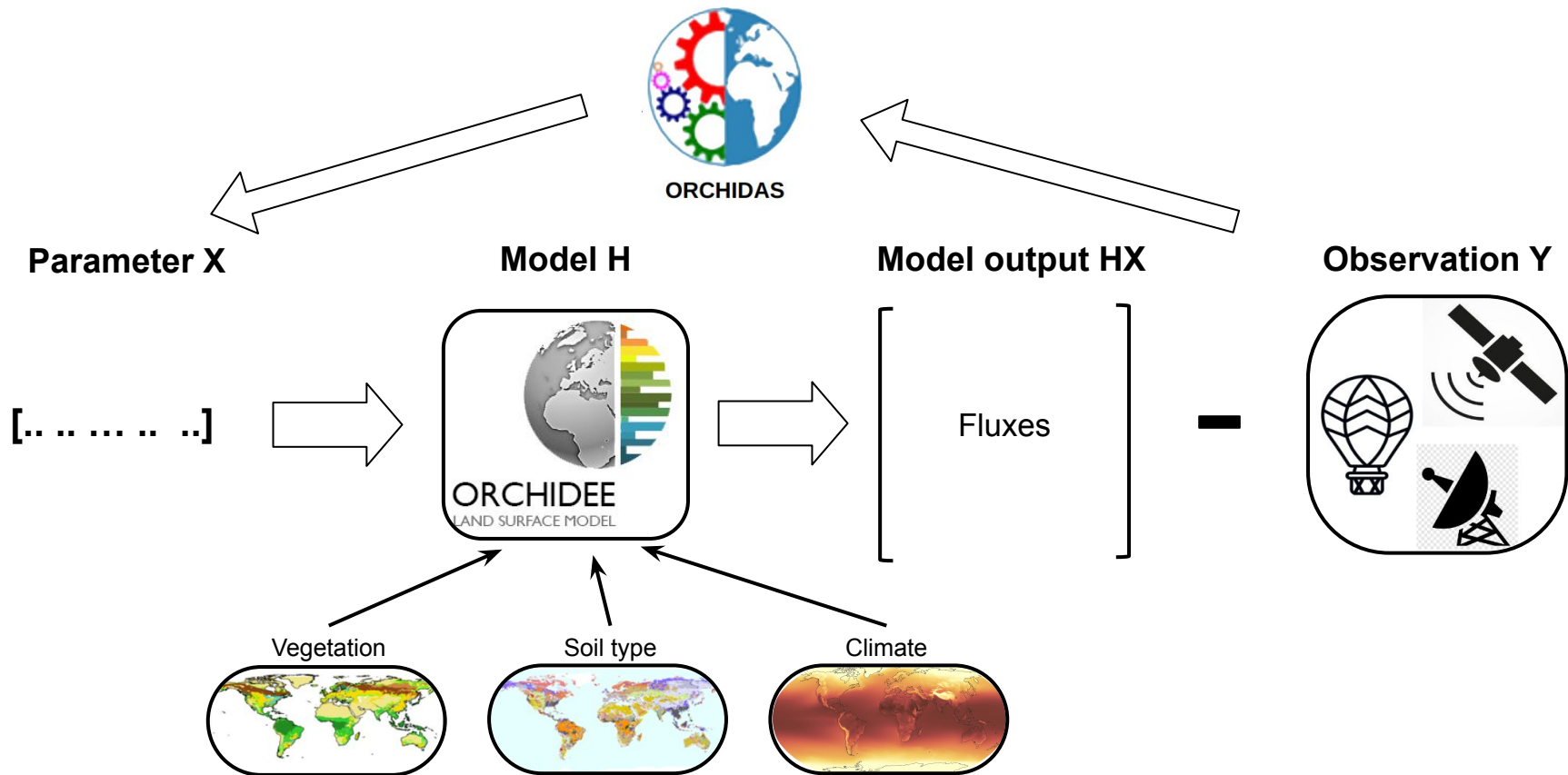
We need to calibrate them rigorously.

ORCHIDAS : ORCHIDEE Data Assimilation Systems

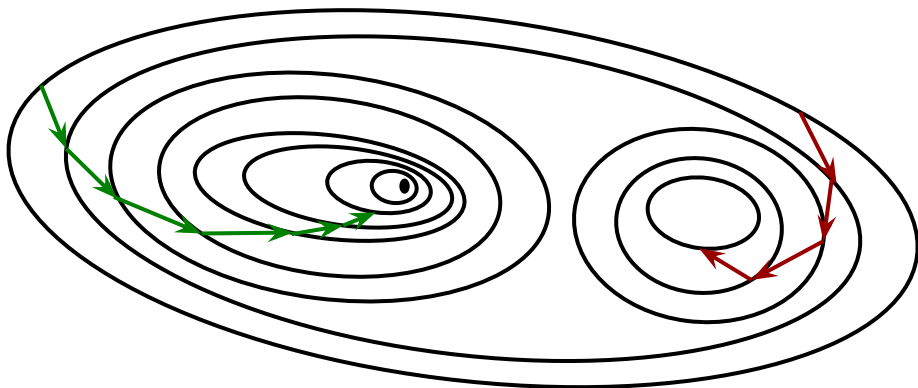
A Bayesian setup :

(Tarantola 1987/2005)

$$J(\mathbf{X}) = \frac{1}{2}(\mathbf{H}\mathbf{X}-\mathbf{Y})^T\mathbf{R}^{-1}(\mathbf{H}\mathbf{X}-\mathbf{Y}) + \frac{1}{2}(\mathbf{X}_b-\mathbf{X})^T\mathbf{B}^{-1}(\mathbf{X}_b-\mathbf{X})$$



Gradient Descent : L-BFGS-B



Pros :

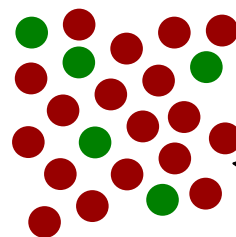
- Fast
- A direct path leading to the minimum

Cons:

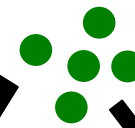
- Can be stuck in local minima
- Need to calculate the Jacobian and Hessian

Genetic Algorithm

Random Population



Selection of the Best Candidates



Generation of a New Population (Reproduction and Mutation)



Pros :

- Less chance to be stuck in local minima
- Exploring a higher proportion of the loss function

Cons:

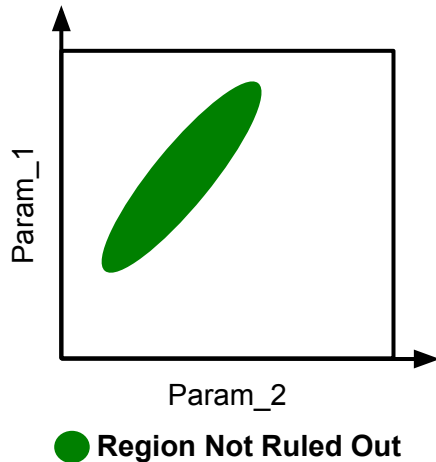
- Expensive and slow

In the end, you have one set of parameters and reduced posterior uncertainty!

History Matching

“**History matching** uses **emulators** to rule out regions of the parameter space of the climate model that are inconsistent with physical observations” (Williamson et al. 2013)

Output of History Matching :



What could we do with this subset of parameters?



Several possibilities :

- **Exploration** : Having more information about the parameter space.
- **Uncertainty** : Simulation of a set of parameters to quantify uncertainty.
- **Optimisation** :
 - Using optimisation (minimisation) techniques on the subset to find the optimum.
 - Reduces the possibility of falling into local minima.
 - **Reduces the computing time.**
 - Continuing the iteration with History Matching.

To investigate the potential of History Matching, we first set up a twin experiment

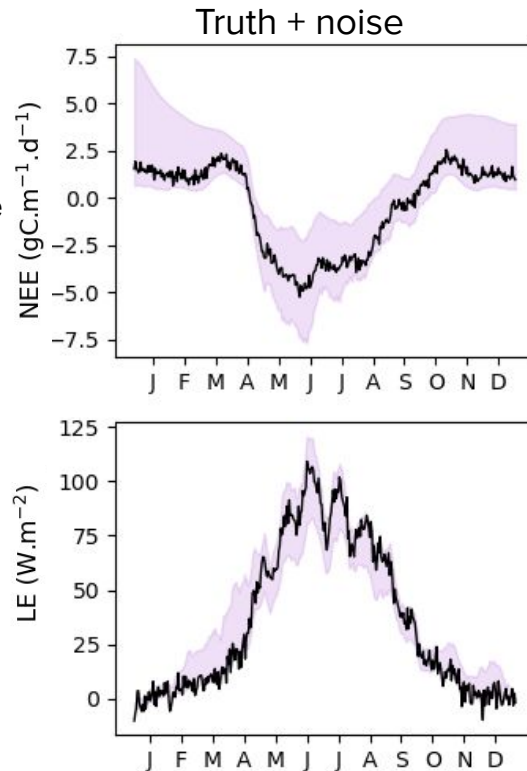
Twin experiment :
Pseudo data created by
ORCHIDEE.
We know the true
parameter

Site: FR-Fon (2005)



NEE :
Net Ecosystem Exchange
Net Carbon flux

LE : Latent heat flux



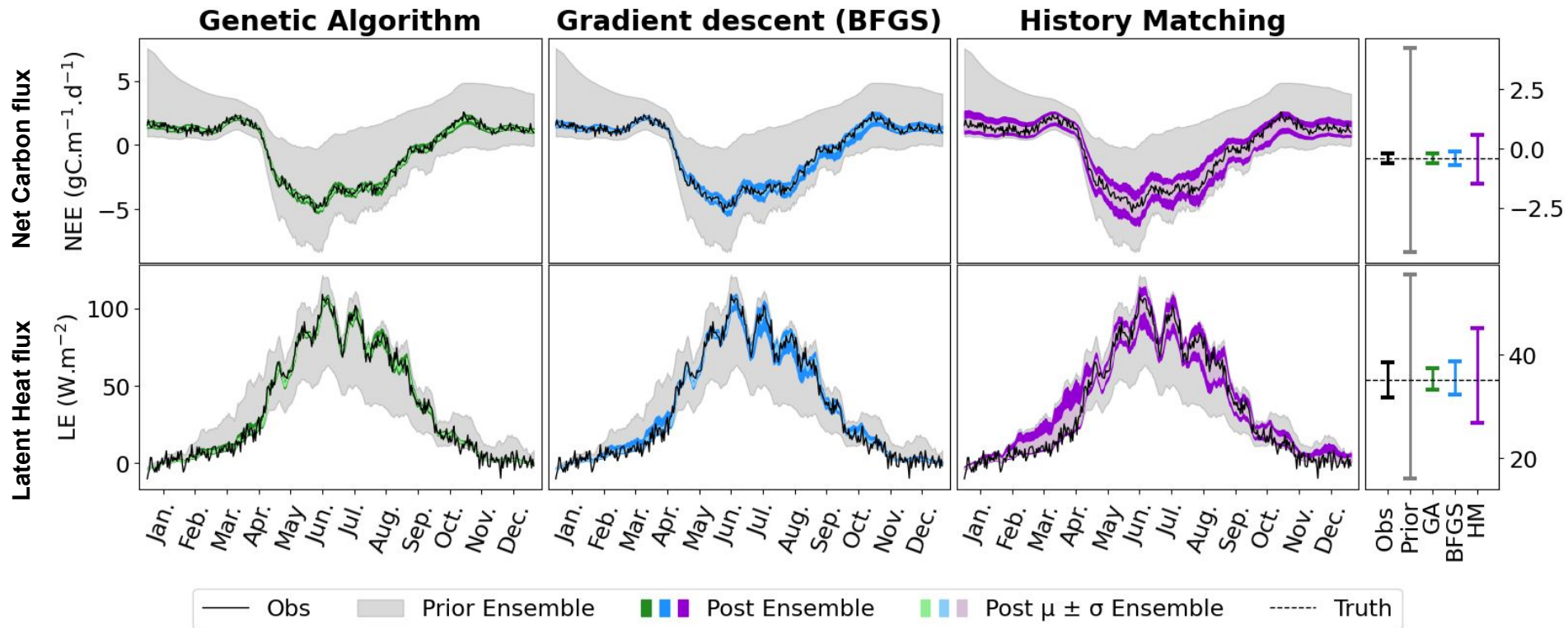
6 parameters used :

	Description
VC_{max}	Maximum carboxylation rate ($\mu\text{molm}^{-1}\text{s}^{-1}$)
SLA	Specific leaf area (m^2)
$L_{agecrit}$	Critical leaf age for starting leaf senescence (days)
$Evap_{res}$	Factor controlling bare soil resistance to evapotranspiration (-)
$Root_{prof}$	Root profile parameter of an exponential function that describes the decrease of root density as a function of depth (m)
Q_{10}	Parameter determining the temperature dependency of the heterotrophic respiration (-)

Need to compare Ensemble :
DA : 200 optimisations using random priors
1 ensemble of HM

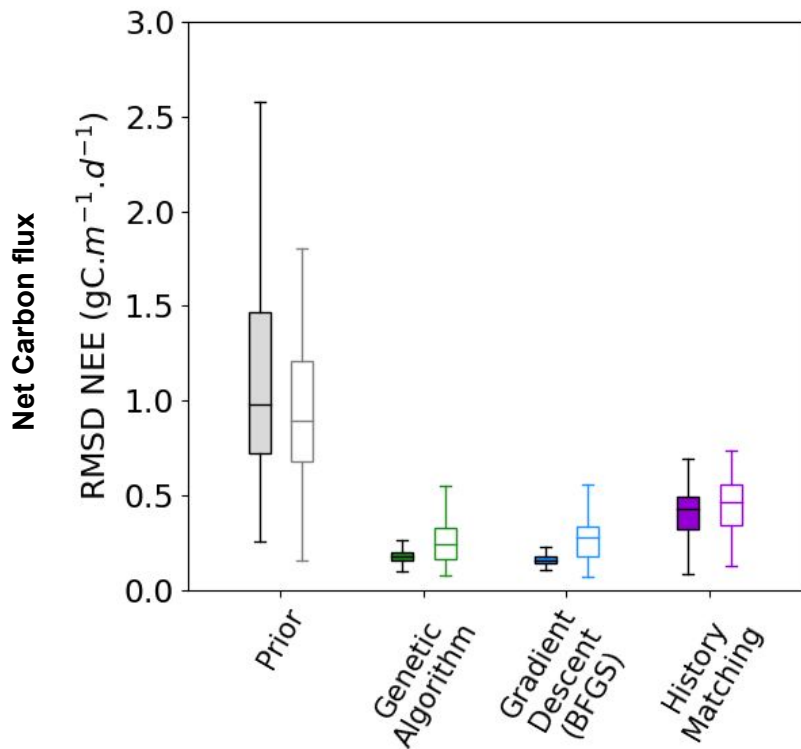
Results : Model-data fit

Metric: RMSD

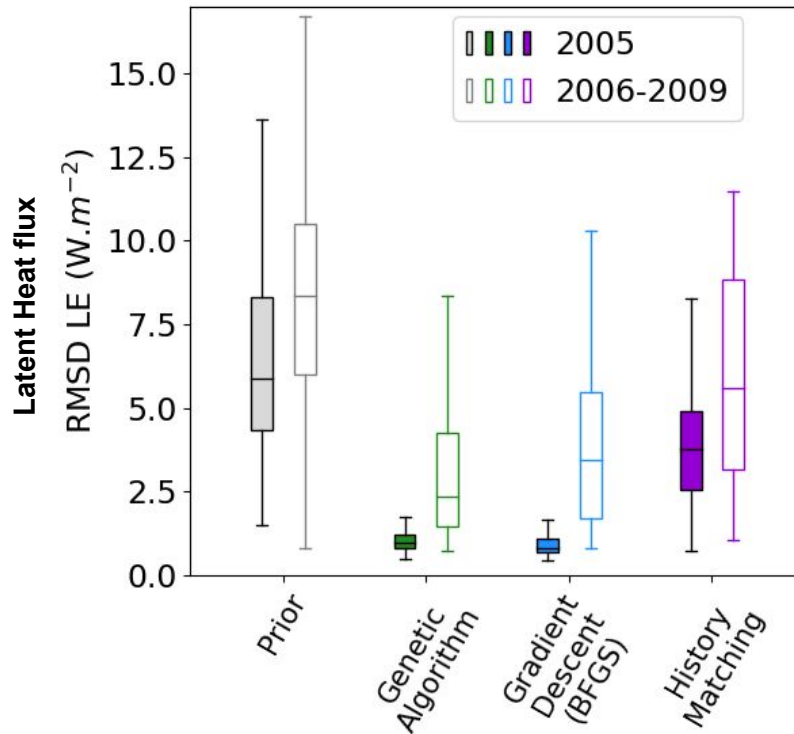


Results : Model-data fit

HM overfit less than DA

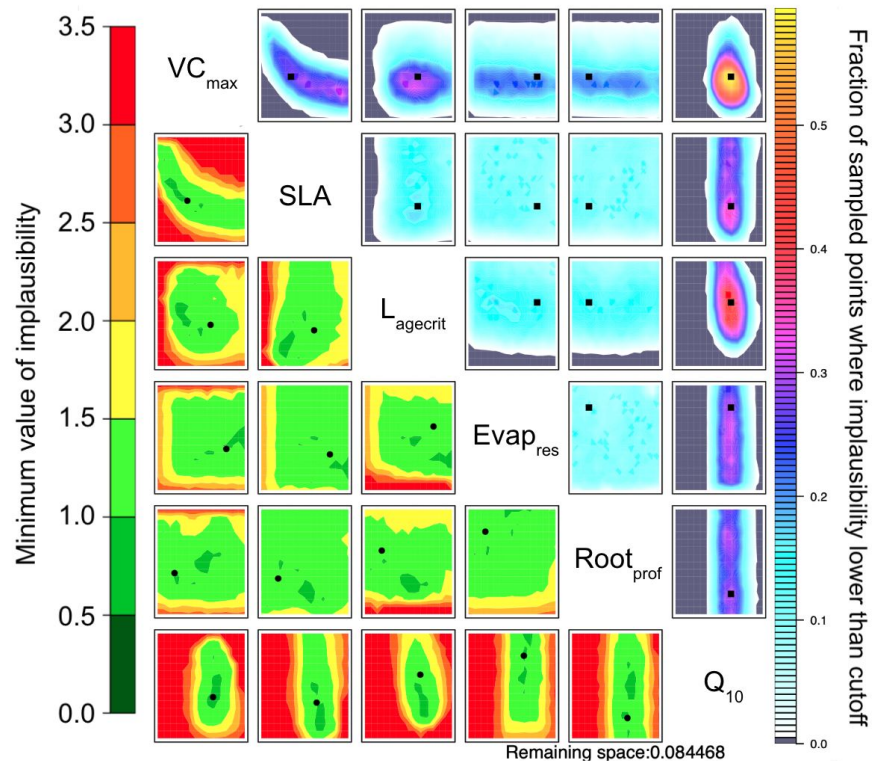
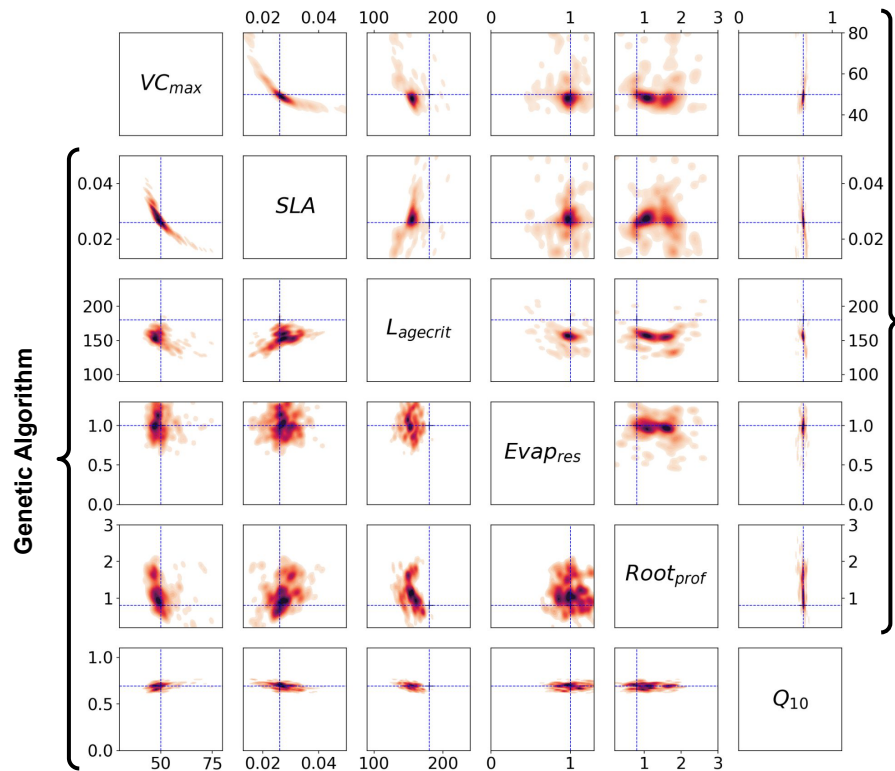


Calibration over 2005
Evaluation over 2006-2009

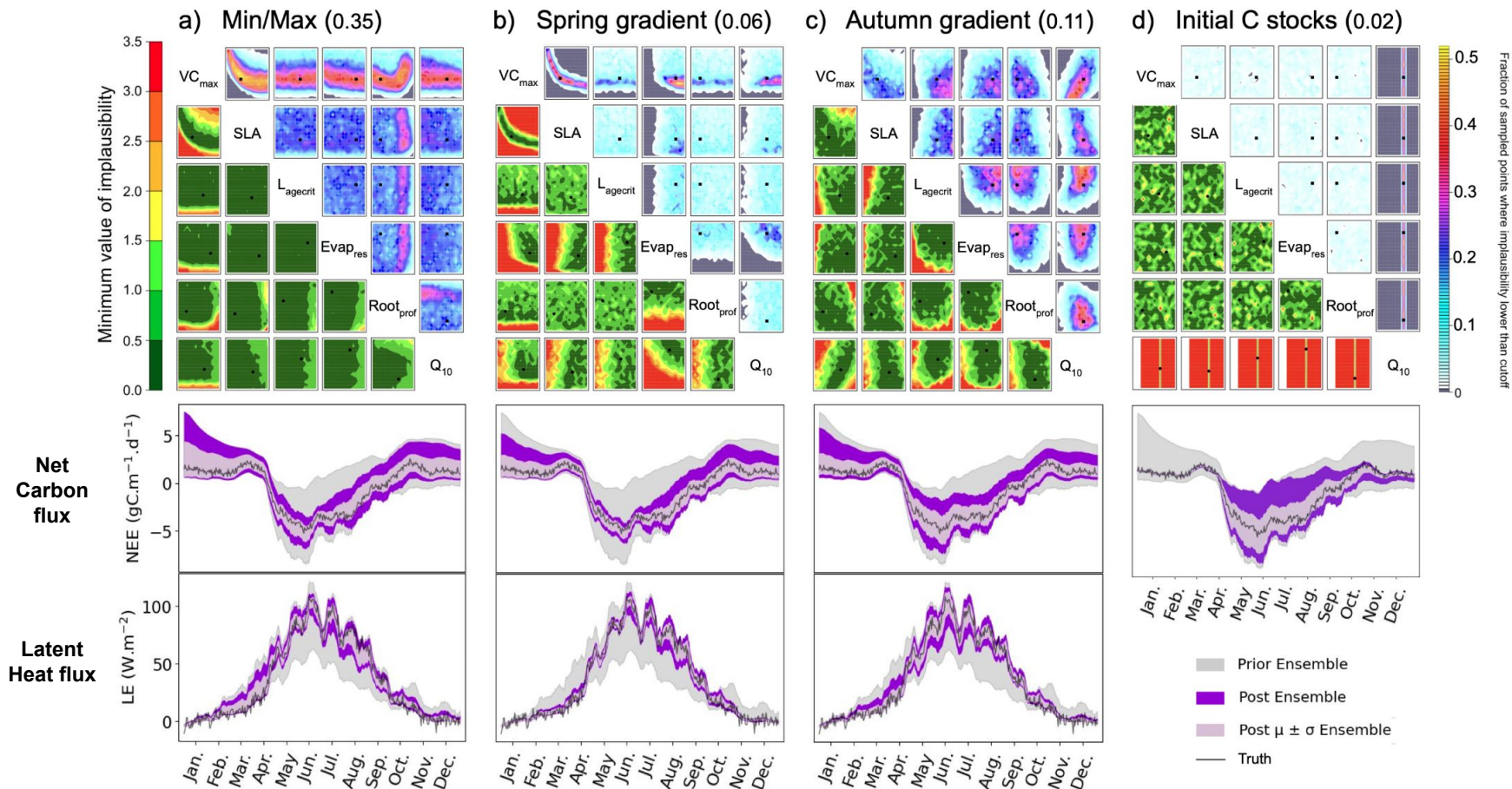


Results : Parameter uncertainty

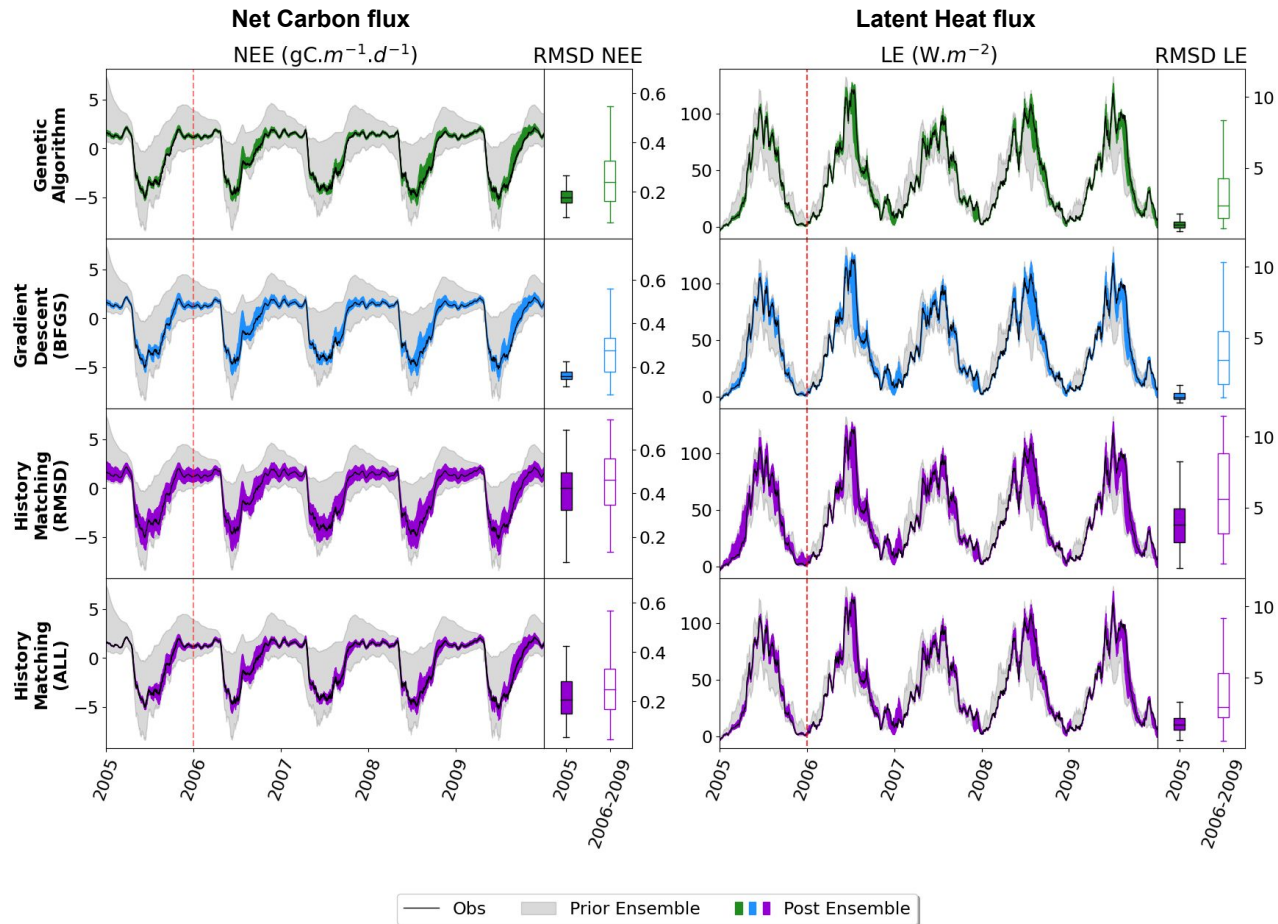
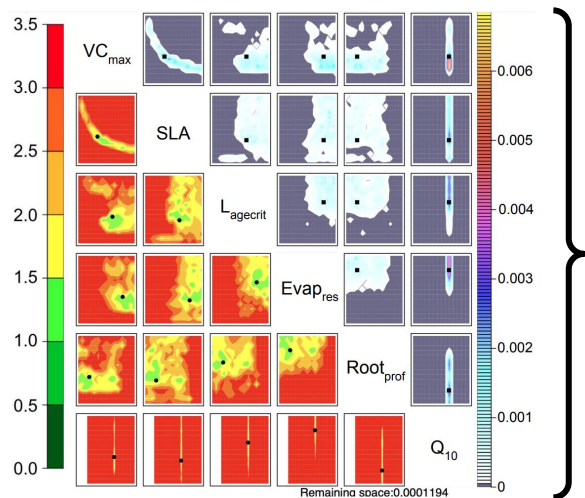
Spread of 200 optimisations



Results : Multi metrics



Results : Multi metrics



Summary

- **HM provides more information about parameter uncertainties**
- **HM does not overfit as much as DA**
- **Multi metric is very powerful when using HM**

All these results are in the article of Nina Raoult currently under review in GMD :

Exploring the Potential of History Matching for Land Surface Model Calibration

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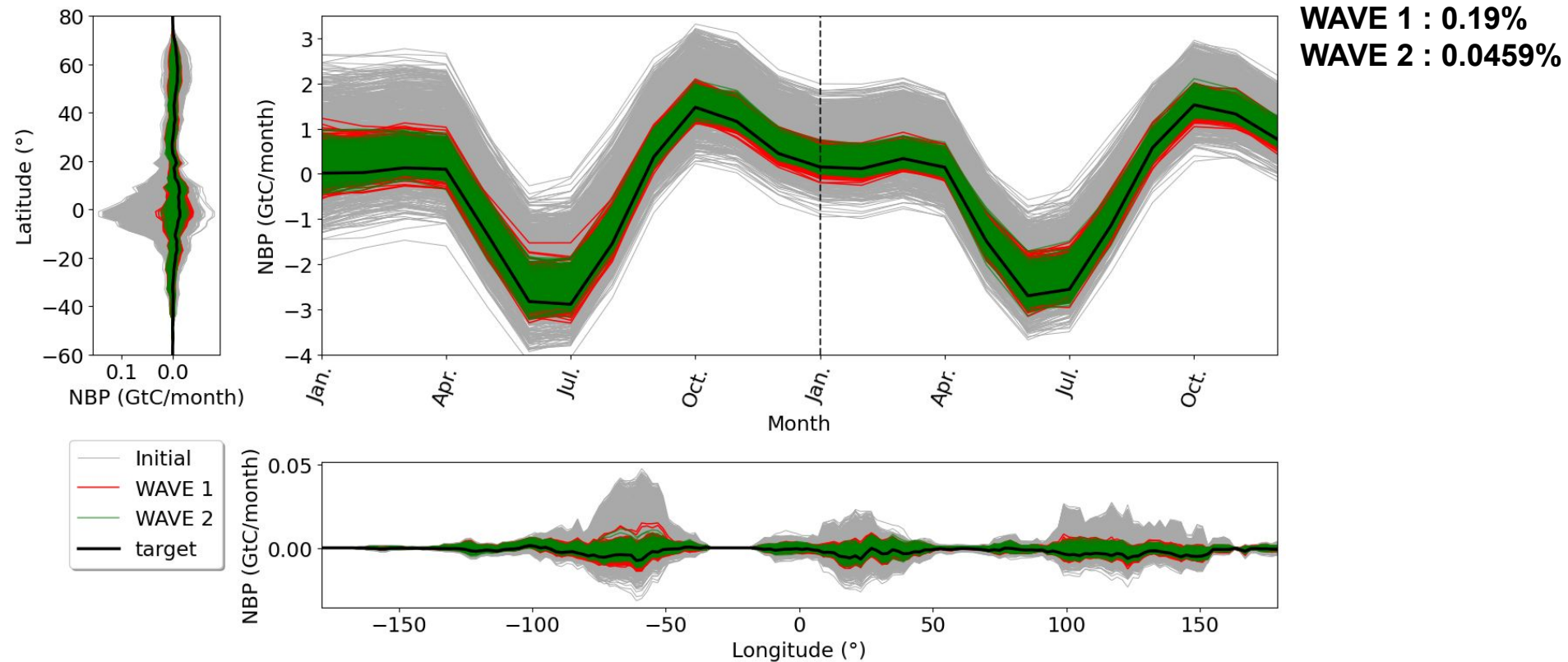
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work in progress : adjust the global carbon balance of ORCHIDEE

First results : first wave with 120 parameters

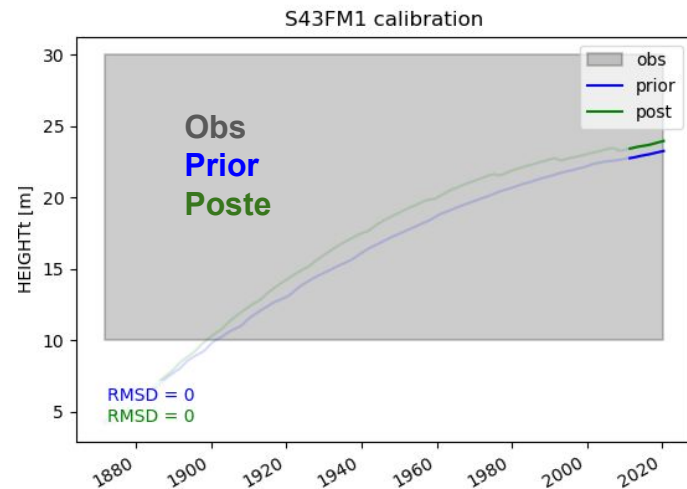
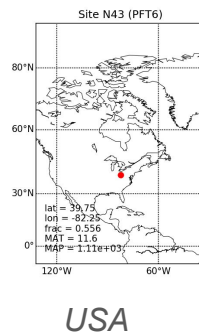
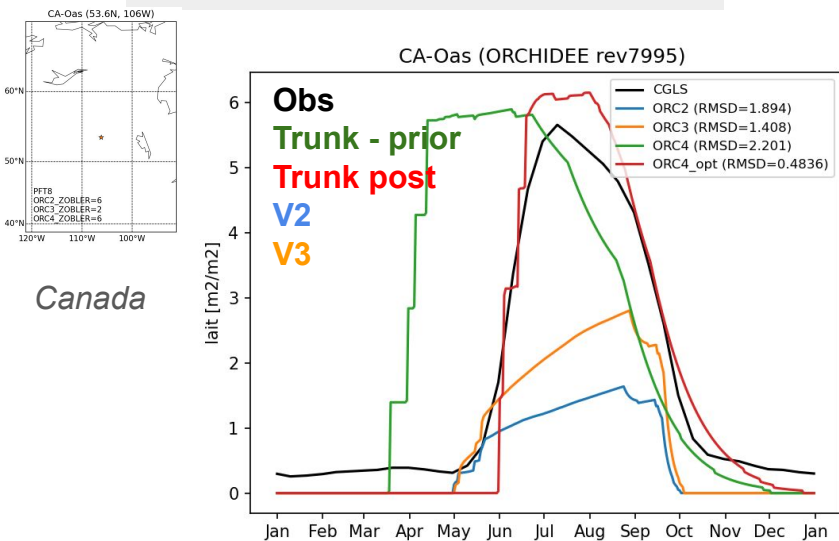


Offline Calibration des paramètres ORCHIDEE-Trunk pour CMIP7

- Optimisation sur sites (FluxNet) : Multi-site calibration par PFT
- Obs: Flux C / LE / SH / ; Hauteur arbres ; LAI satellite ; Ratio biomasse C ; ...
- Paramètres: Photosynthèse, Allocation C, Mortalité,...
- Approche standard avec Algorithme Génétique !

Ex: optimisation phenologie (LAI)

Ex: optimisation hauteur arbres



Offline Calibration des paramètres ORCHIDEE-Trunk pour CMIP7

- Optimisation Albedo (MODIS) et Couverture de Neige (ESA-CCI)
- Paramètres: 2 paramètres pour couverture neige, 2 albedo neige, 1 albedo veg

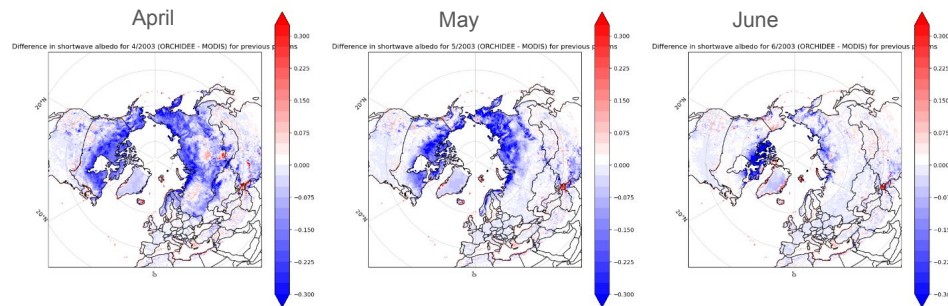
$$albedo = frac_{veg}[(1 - frac_{snow,veg})alb_{veg} + frac_{snow,veg}alb_{snow,veg}] + frac_{nobio}[(1 - frac_{snow,nobio})alb_{nobio} + frac_{snow,nobio}alb_{snow,nobio}]$$

$$alb_{snow,veg} = \frac{\sum_{pff=1}^{15} frac_{max,pff} \times alb_{snow,pff}}{\sum_{pff=1}^{15} frac_{max,pff}}$$

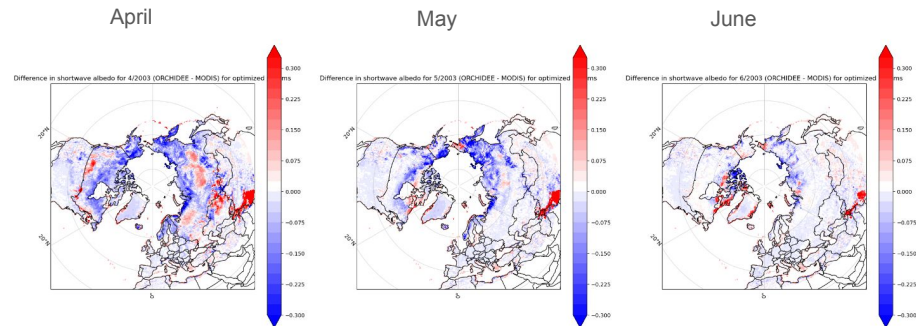
$$alb_{snow,pff} = alb_{snow,aged,pff} + alb_{snow,dec,pff} \times e^{-age_{snow}/tcs_{snow}}$$

- Approche standard avec Algorithme Génétique !

Albedo : Prior difference with MODIS



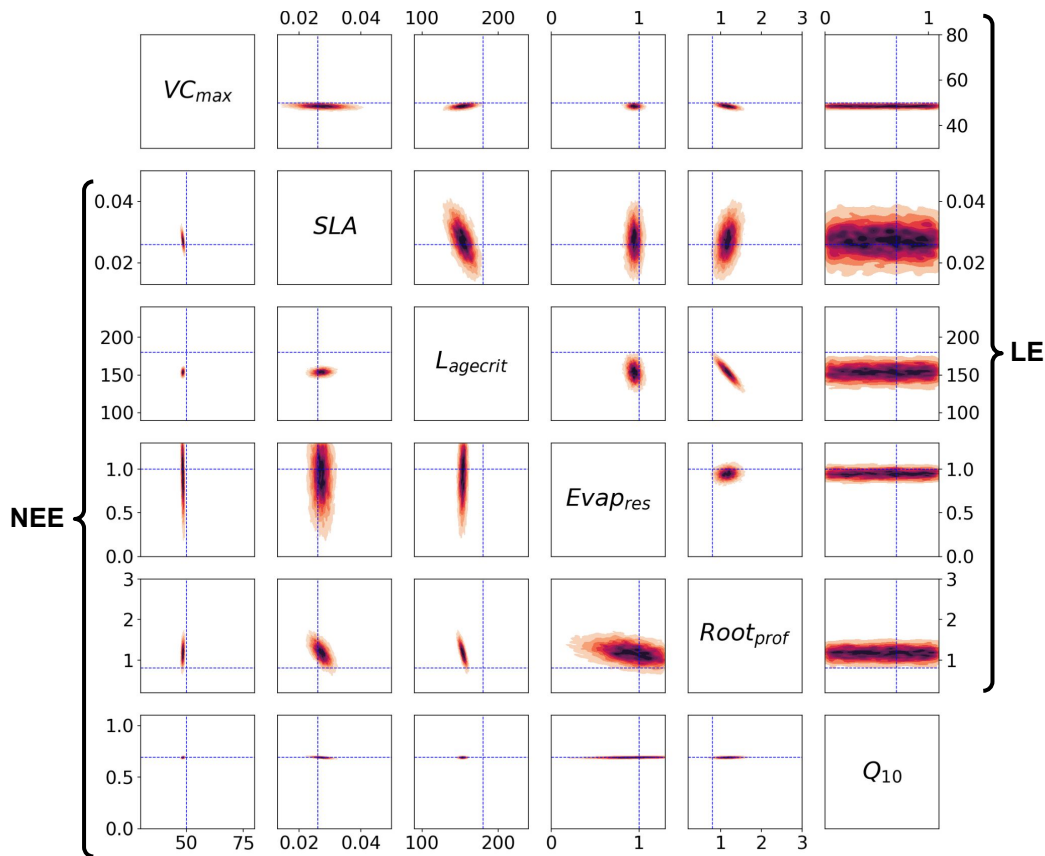
Albedo : Posterior difference with MODIS



Thank you for your attention !

Results : Parameter uncertainty

Using Bpost matrix



Posterior Covariance operator :

$$B_{post} = [H^T R^{-1} H + B^{-1}]^{-1}$$

B_{post} diagonal elements are variances, the root square of diagonal elements can be seen as “**uncertainty bars**”.

B_{post} off-diagonal elements (which represent the covariances), it is easier to check correlations:

$$Corr^{\alpha,\beta} = B_{post}^{\alpha,\beta} / (\sigma^\alpha \sigma^\beta)$$

If $Corr^{\alpha,\beta}$ close to 0, the uncertainty of m^α and m^β is uncorrelated.
If $Corr^{\alpha,\beta}$ close to ± 1 , the uncertainty of m^α and m^β is correlated.

To compute uncertainty, let's generate pseudorandom gaussian sample of m (set of parameters) using m_{post} as mean and B_{post} as covariance matrix.

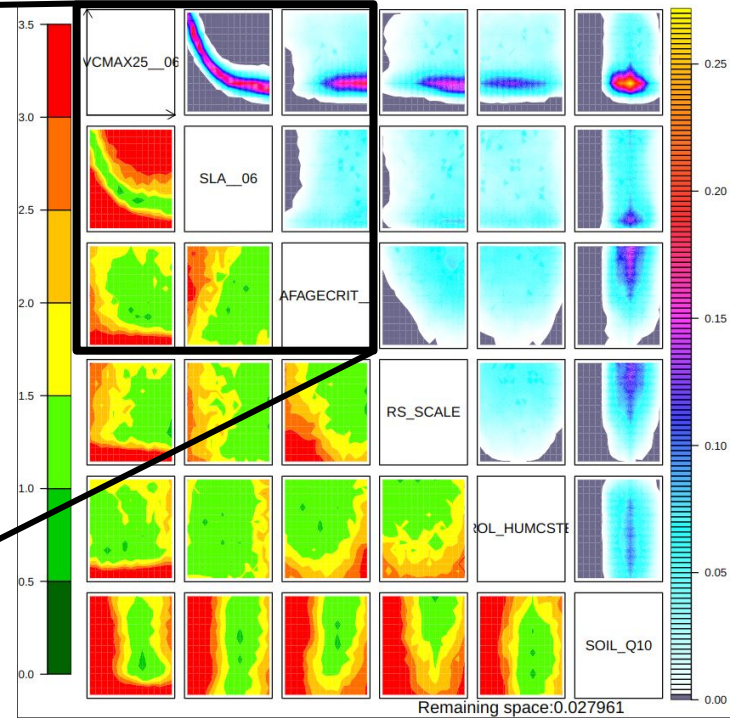
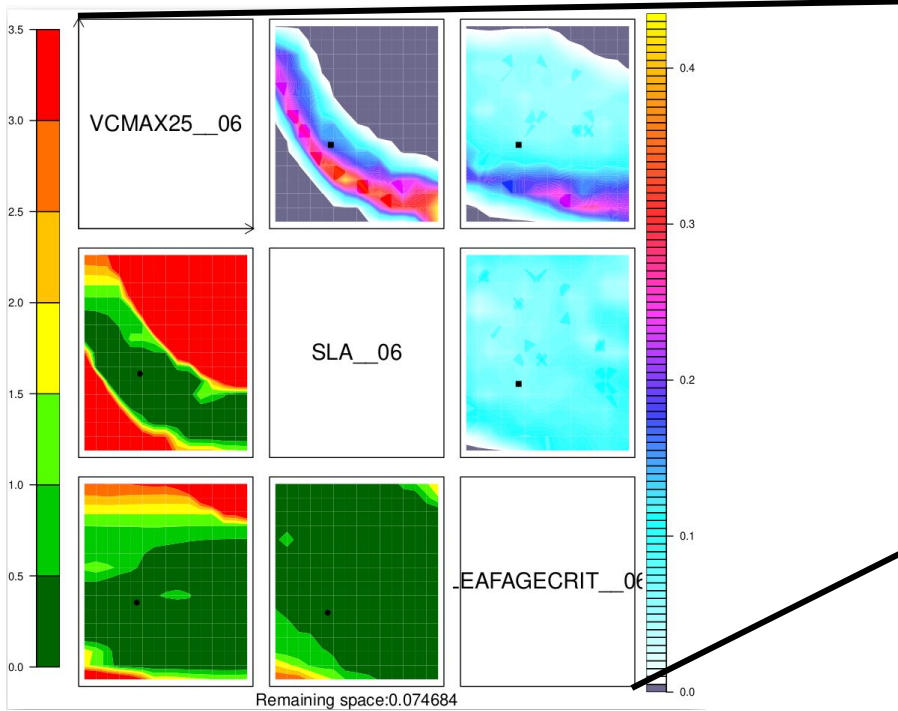
Bpost decomposition:

$$B_{post} = \left[\sum_{i=0}^D H_i^T \sigma_i^{-1} H_i + B^{-1} \right]^{-1} = [H_0^T \sigma_0^{-1} H_0 + H_1^T \sigma_1^{-1} H_1 + \dots + H_D^T \sigma_D^{-1} H_D + B^{-1}]^{-1}$$

Final word : Sensitive analysis + Structure of the model

Global - metrics Global net flux

FR-FON - metrics (9 metrics used)



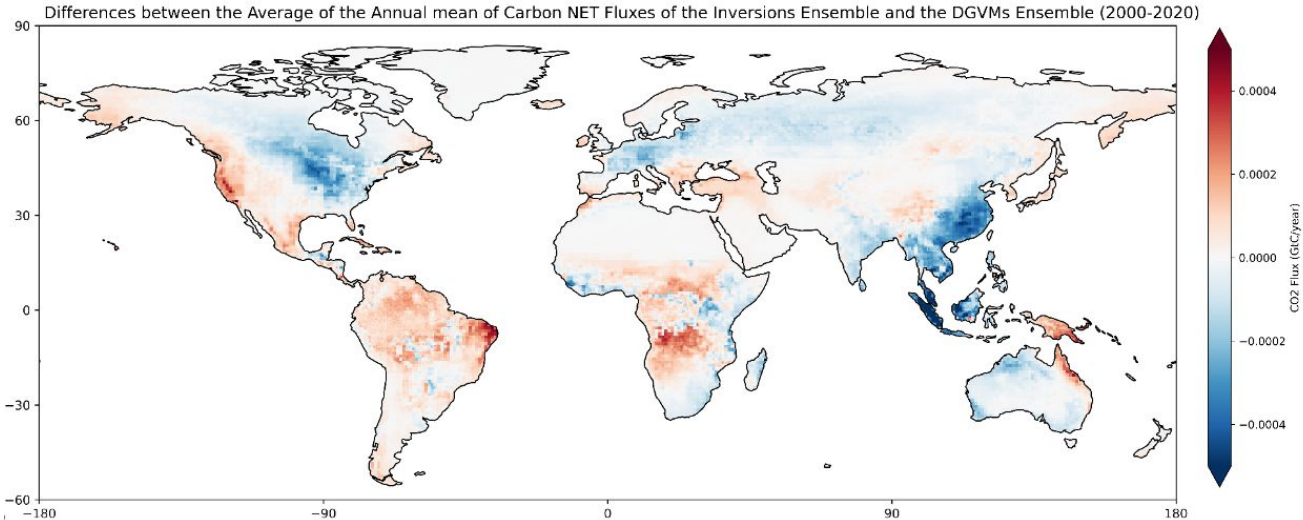
work in progress : adjust the carbon balance.

Understand the Land Carbon sink:

Top down : Atmospheric Inversion

Bottom up : Land Surface model

Using HM to calibrate ORCHIDEE off-line on Atmospheric Inversion first. *(and then used LMDz to transport CO₂ concentration)*



Data collected
from TrendyLand
V10 and GCP2021

Difficulty :

**Calibrating all the pft means
a large number of
parameters (between 100
and 200).**

Final word : Sensitive analysis

Heterotrophe Respiration

