
Principle and example of parameter optimisation with ORCHIDEE : the application to Tibetan grassland ?

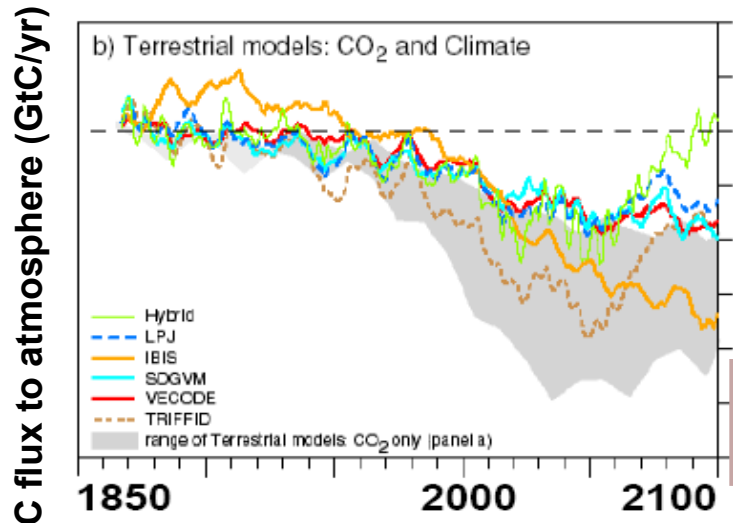
Philippe Peylin, Hui Yang, Sushi Peng, Natasha MacBean, Cédric Bacour, Vladislav Bastrikov, Tea Thum, Fabienne Maignan, Sylvain Kuppel, Frédéric Chevallier, Philippe Ciais, Shilong Piao

Laboratoire des Sciences du Climat et de l'Environnement (LSCE), France



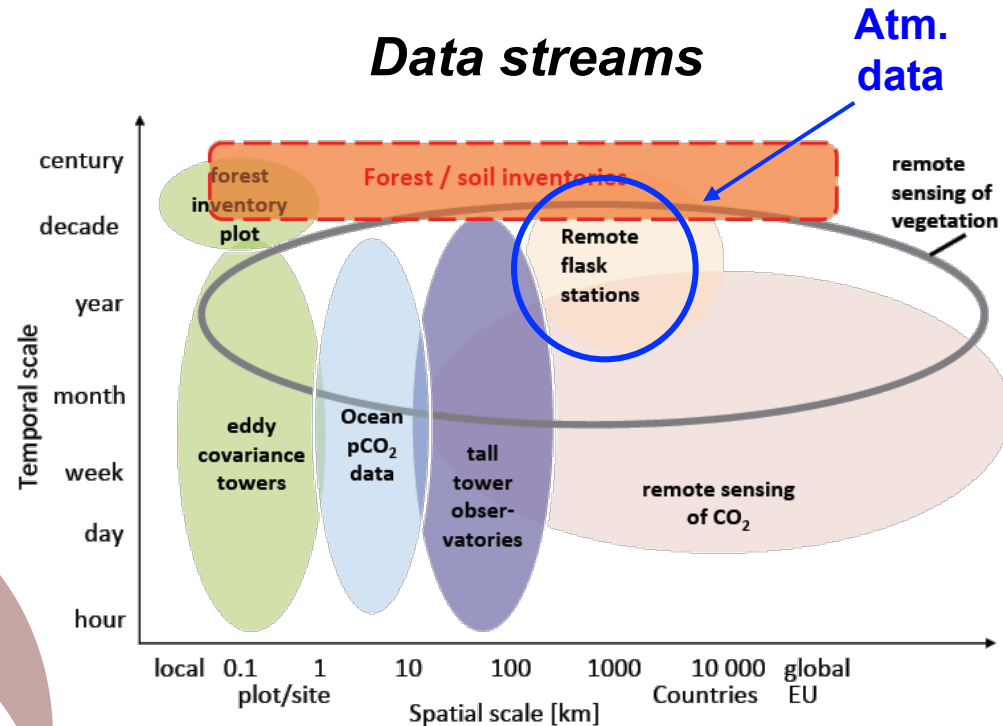
Needs for a Carbon Cycle Data Assimilation System

Large uncertainty from land to predict global C-balance (C4MIP)



OPTIMISATION OF PARAMETERS
→ Data Assimilation

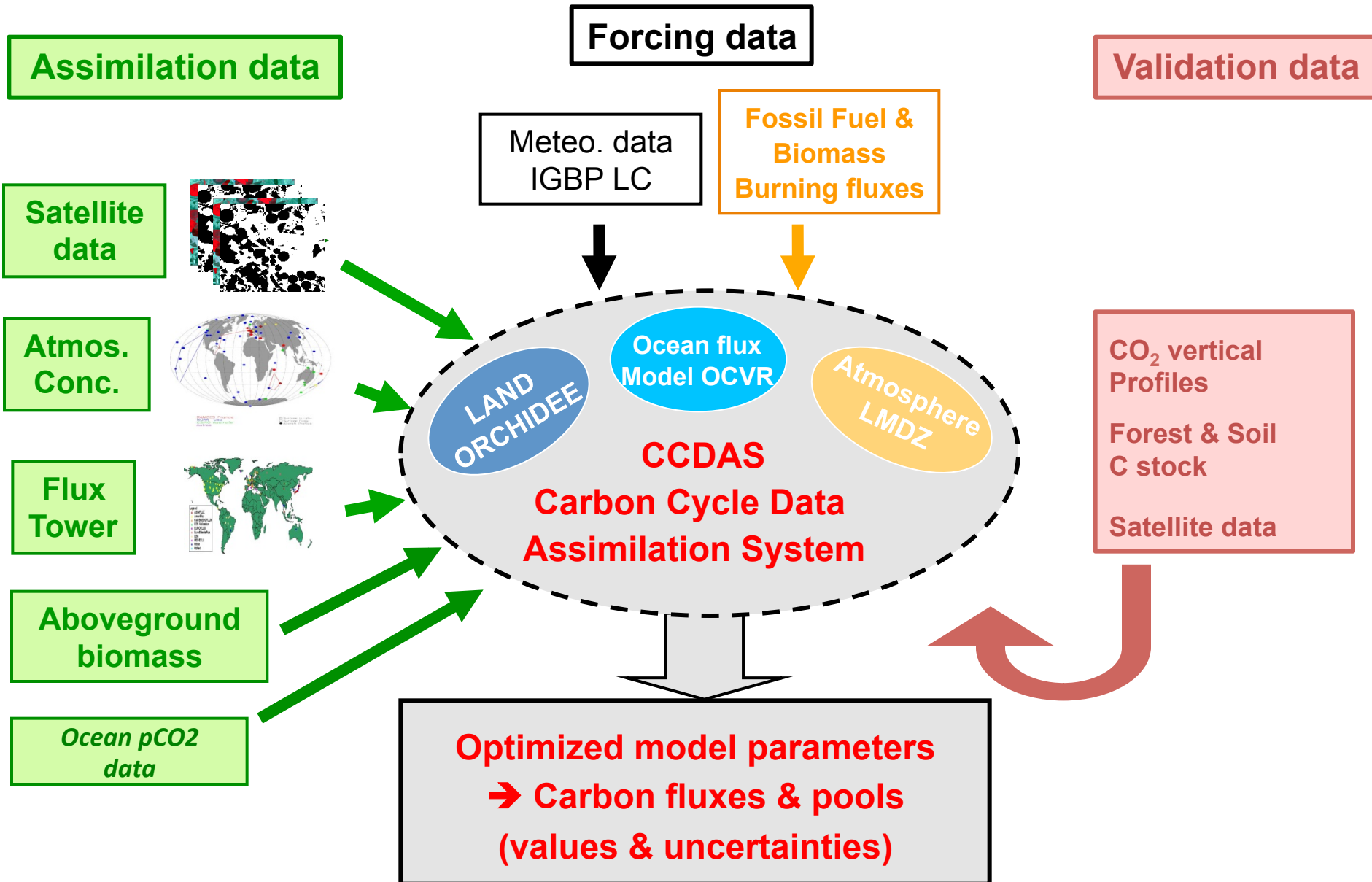
Optimized ecosystem models
→ reduce the spread ?



Improve:

- Uncertainty estimates
- C land budget estimates
- Future climate predictions
- Process understanding

Structure of the LSCE CCDAS



ORCHIDEE data assimilation systems...

- Several systems have been developed at LSCE....
- Effort around CARBONES project lead to a system that :
 - Can optimize many ORCHIDEE parameters
 - Work at local scale or global scale
 - Assimilation of many data stream separately or together: FluxNet, satellite NDVI, [CO2], Biomass,..
 - Use 2 different approaches to minimize a cost function:

$$J(\mathbf{x}) = (\mathbf{x} - \mathbf{x}_b)^T \mathbf{B}^{-1} (\mathbf{x} - \mathbf{x}_b) + (\mathbf{H}\mathbf{x} - \mathbf{y})^T \mathbf{R}^{-1} (\mathbf{H}\mathbf{x} - \mathbf{y})$$

- Variational approach based on BFGS algorithm
- Monte Carlo approach based on Genetic algorithm



1) Example of optimisation with different data streams

Optimisation with biomass and in situ flux data

Complementarity between FluxNet and NDVI data

Information brought by atmospheric [CO₂] data

2) Preliminary results for Tibetan grassland sites

Single site optimisations

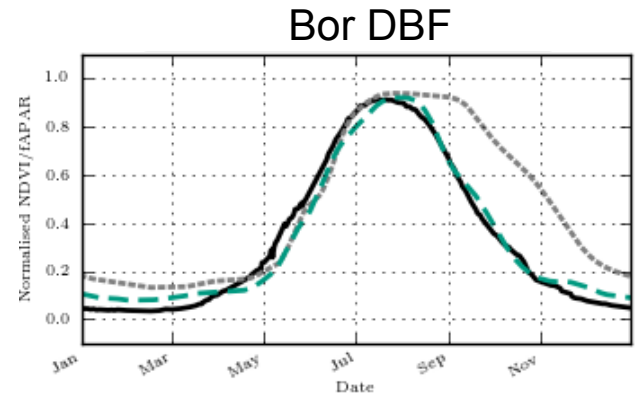
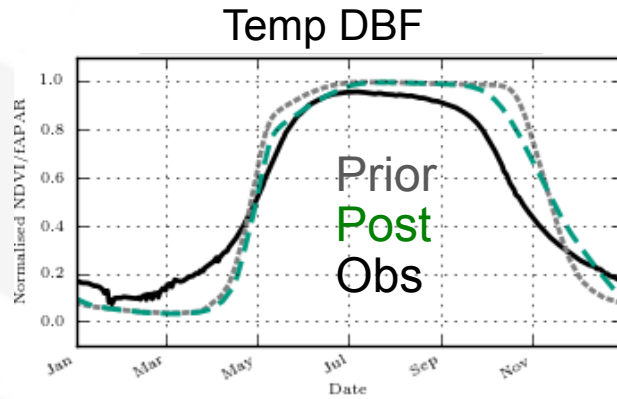
Multi-site optimisations



Stepwise approach (20 yr): a compromise!

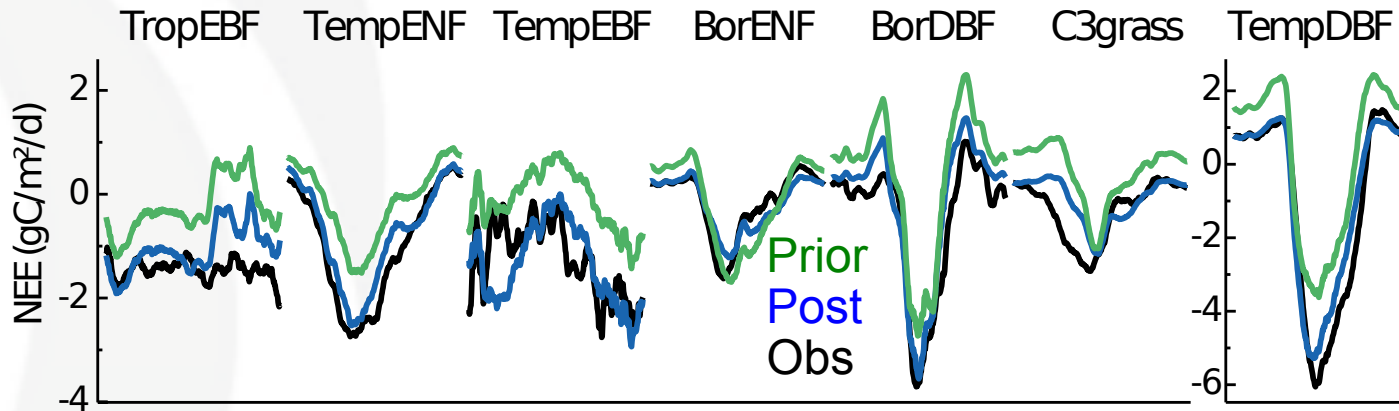
MODIS-NDVI

4 params /PFT



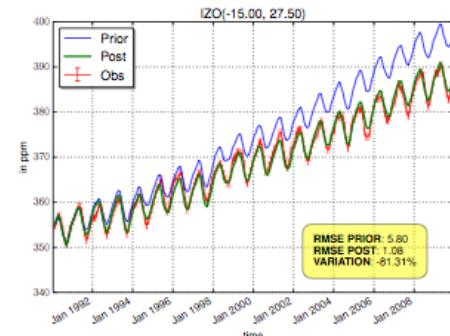
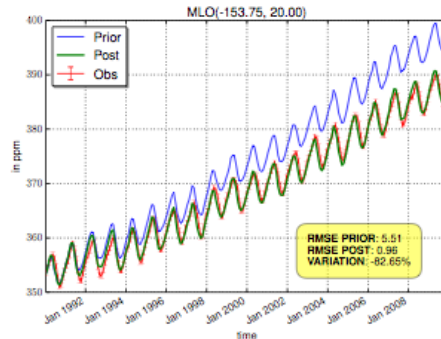
75 fluxnet data

≈ 20 params /PFT



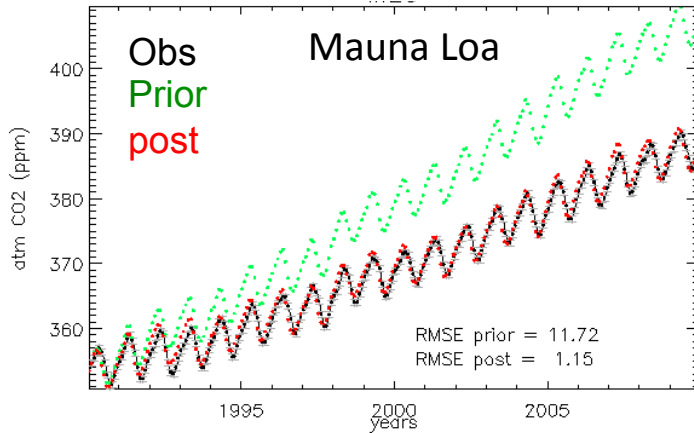
Atmospheric data

≈ 100 params total



Assimilation of atmospheric [CO₂] data

Optimization of the CO₂ trend



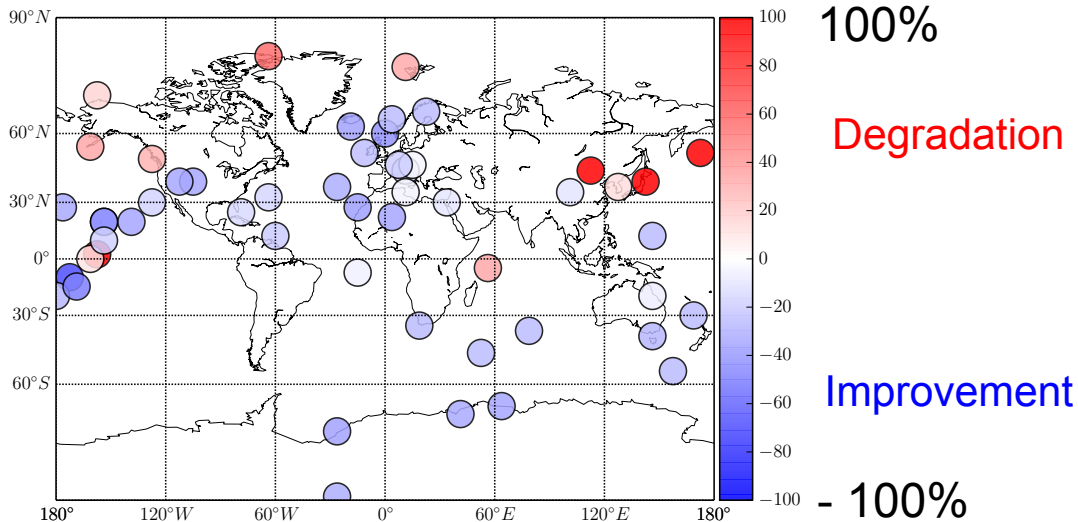
Signal decomposition:

- Amplitude : max – min
- Phase : CPU

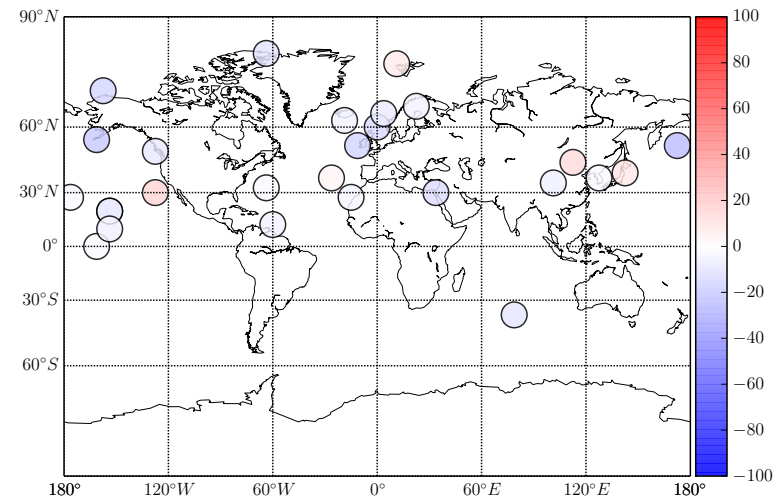
$$(1 - RMSE_{post} / RMSE_{prior})$$



Seasonal amplitude



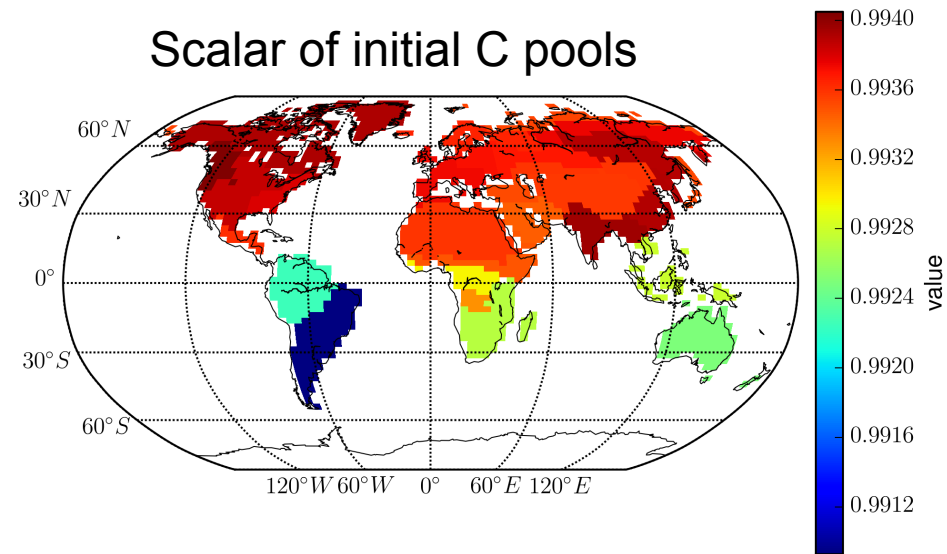
Carbon uptake period length



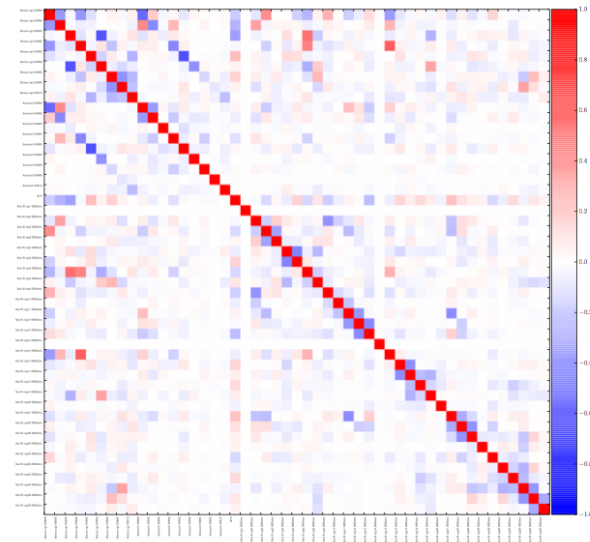
Assimilation of atmospheric [CO₂] data

→ Primary constraint on:

- Soil initial carbon pools..



→ But significant error correlations btw parameters

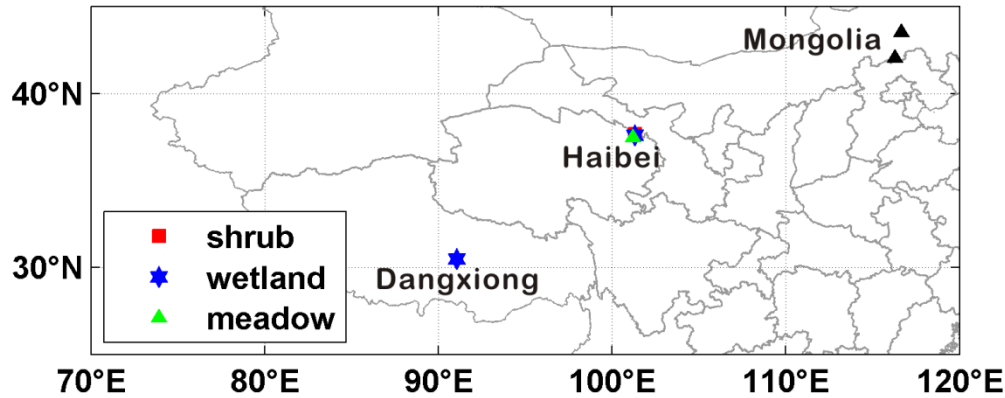


Grassland in ORCHIDEE....

→ One PFT (unique set of params) for very different ecosystems...



Parameter Optimization in the Tibetan Plateau



- 11 Grassland sites
- Meteorological data (Temperature, Precipitation, Radiation, Wind, Humidity)
- Flux data (NEE, GPP, LE, H)

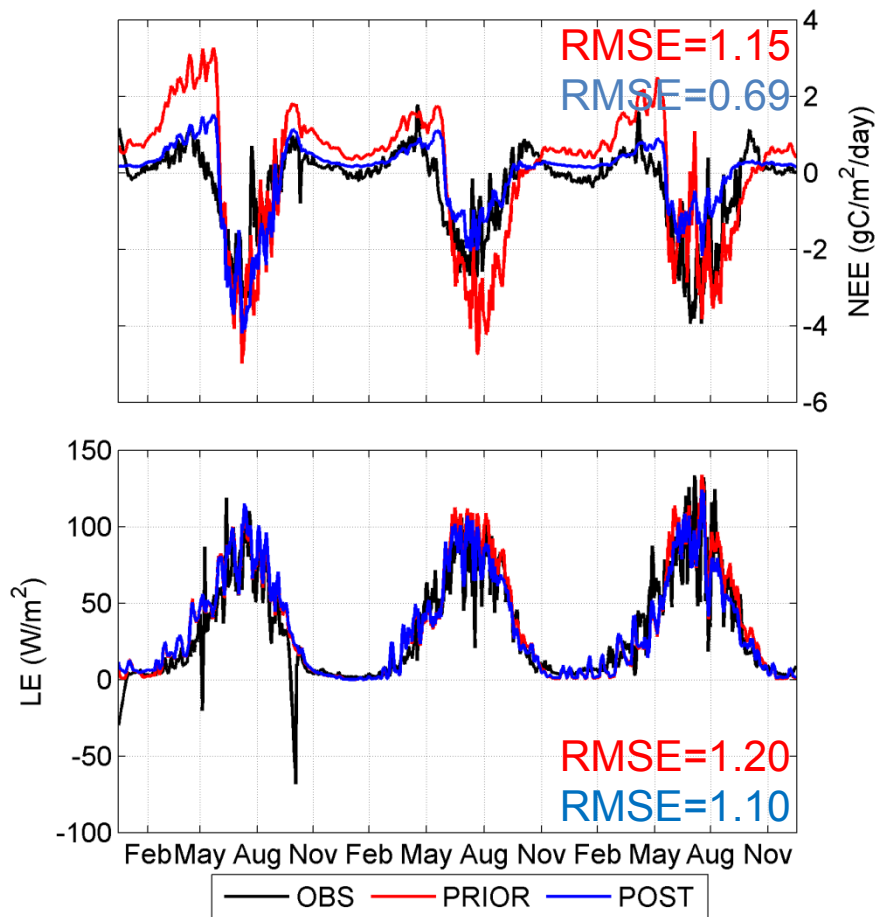
Site	Location	Altitude (m)	Vegetation type	Period
HB-meadow	37.610°N, 101.310°E	3148	Kobresia meadow	2002-2004
HB-shrub	37.670°N, 101.330°E	3293	Alpine shrub meadow	2004-2005
HB-SW	37.610°N, 101.330°E	3160	Swamp meadow	2004
DX-SW	30.470°N, 91.060°E	4286	Swamp meadow	2010-2011

➔ Selection of 4 sites

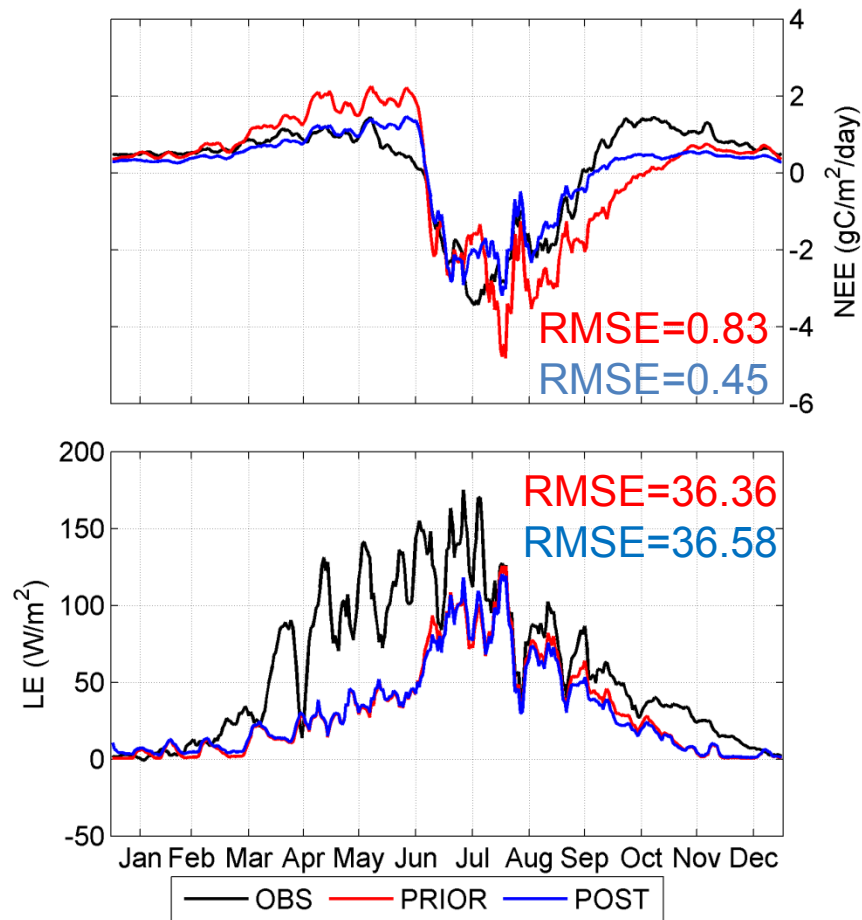
➔ Optimisation of 29 parameters

Part 1. Single Site Optimization

HB Meadow



HB Swamp

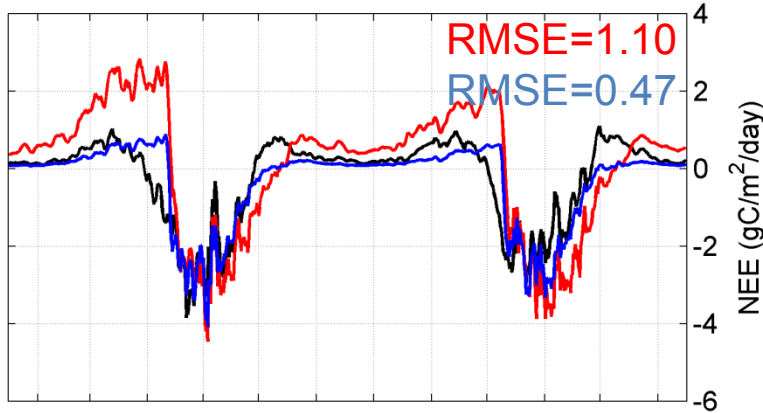


1. Recurrent overestimation of the winter carbon release
2. Overestimation of the summer carbon uptake
3. Significant phase shift at some sites (onset and senescence)
4. Recurrent underestimation of the latent heat

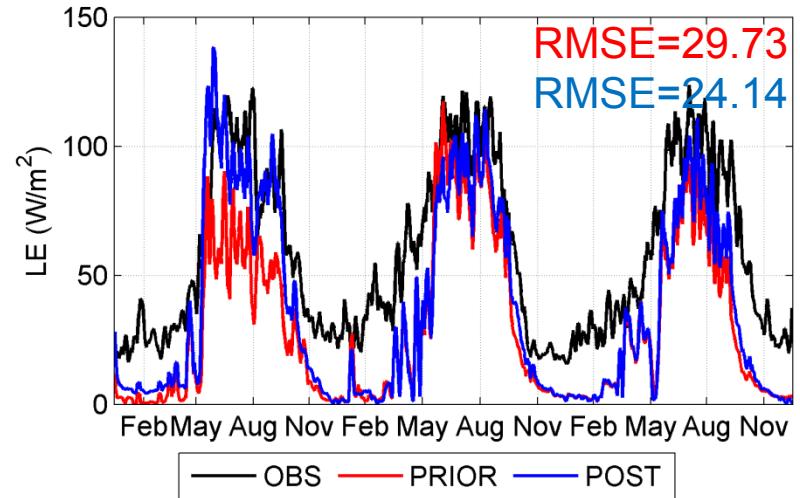
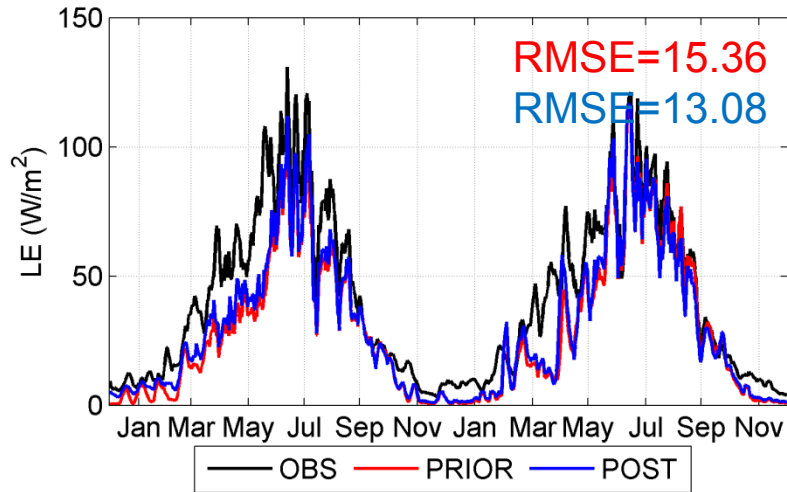
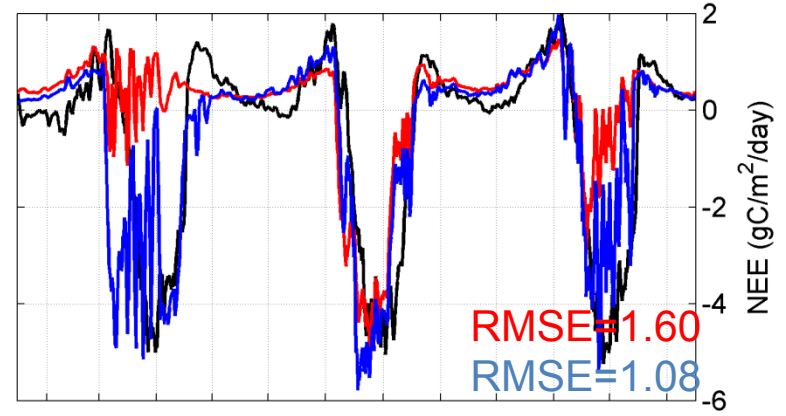
Obs
Prior
Post

Part 1. Single Site Optimization

HB Shrub



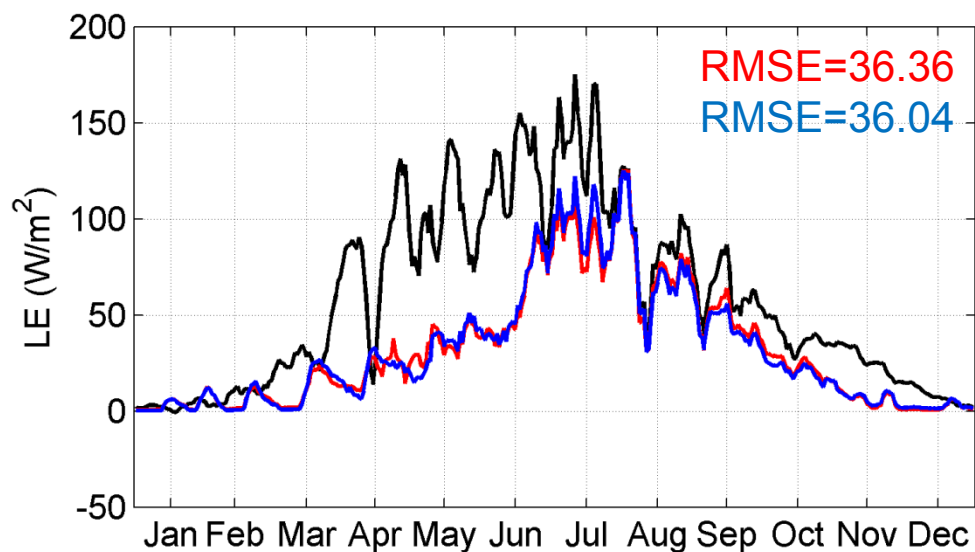
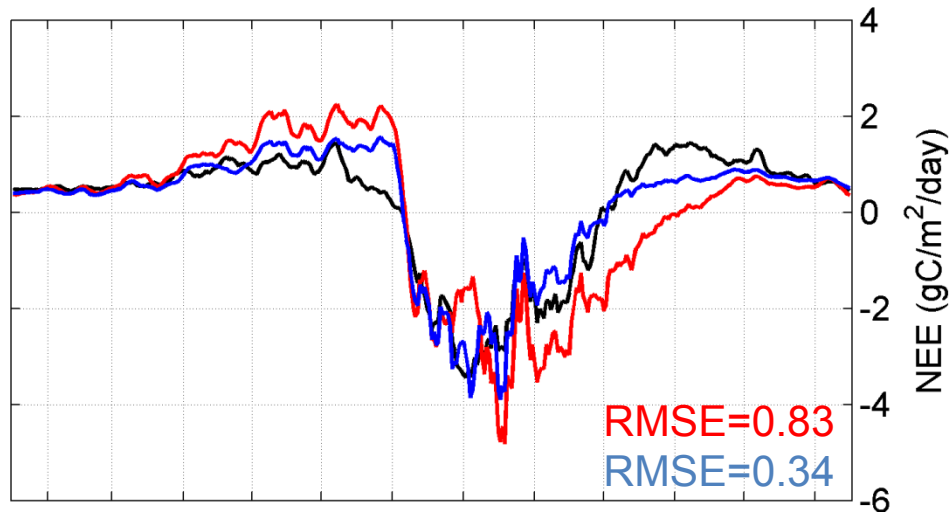
DX Swamp



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Prior
Post

HB Swamp

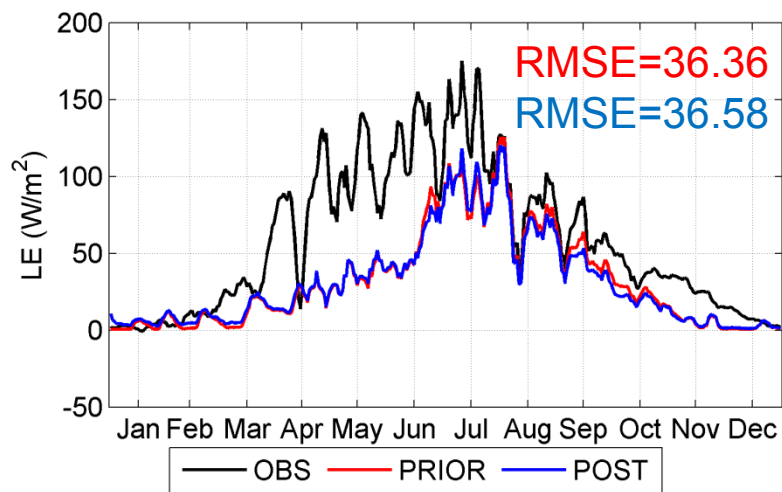
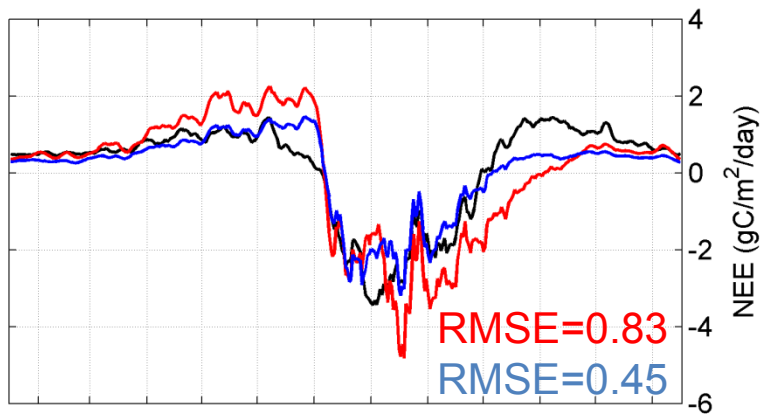


Obs
Prior
Post



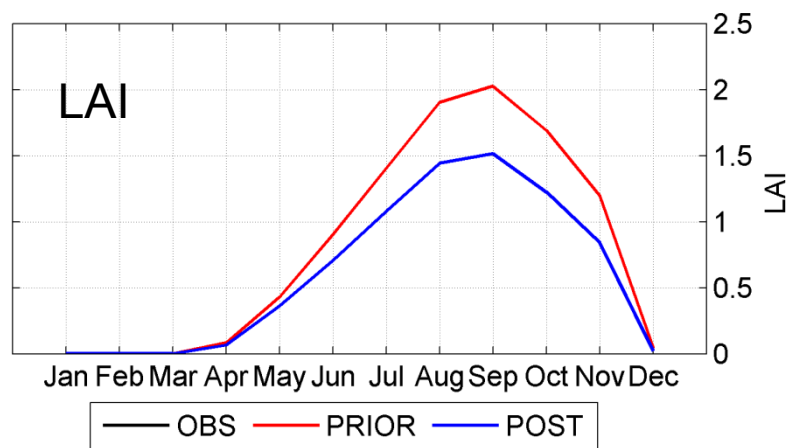
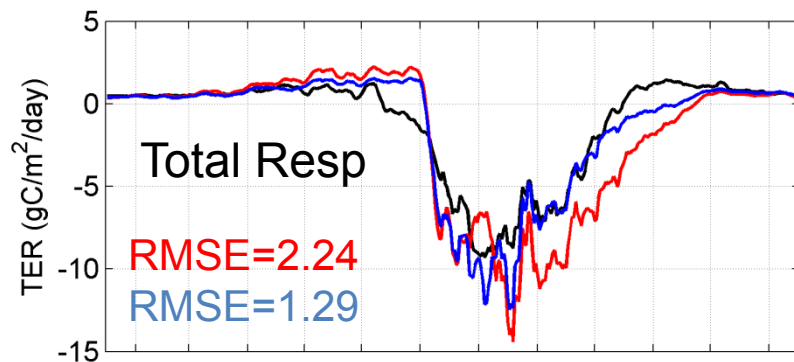
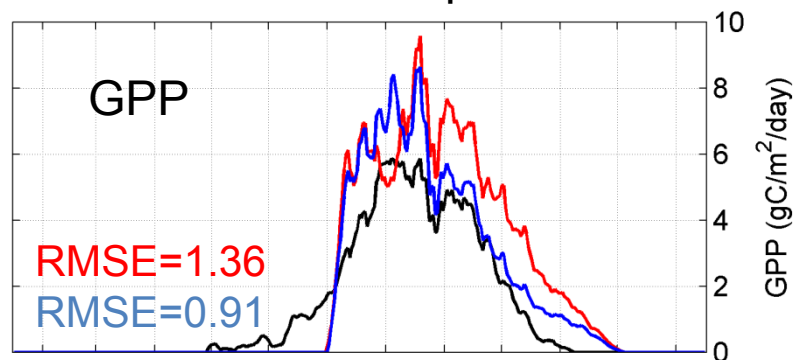
Variables	PRIOR	POST
Vcmax_opt-PFT10	70.0	75.93
Humcste-PFT10	4.0	1.0
Tphoto_min_c-PFT10	-4.0	-4.0
Tphoto_max_c-PFT10	55.0	55.0
Senescence_temp_c-PFT10	-1.375	-1.375
LAI_MAX-PFT10	2.5	2.77
SLA-PFT10	0.026	0.021
Leafagecrit-PFT10	120.0	88.43
Klaihappy-PFT10	0.5	0.56
Tau_leafinit-PFT10	11.0	10.07
Z0_over_height	0.0625	0.048
So_capa_dry	1.8	2.07
So_capa_wet	3.03	3.47
Q10	1.99	1.93
Moistcont_a	-1.1	-1.19
Moistcont_b	2.4	2.31
Moistcont_c	-0.29	-0.38
Z_decomp	0.2	0.71
Hcrit_litter	0.08	0.10
Maint_resp_c	1.4	1.81
Maint_resp_slope_c-PFT10	0.16	0.18
Frac_growthresp-PFT10	0.28	0.30
Dpu_cste	4.0	4.38
Residence_time-PFT10	0	30.0
R0	0.3	0.43
S0	0.3	0.40
Rsol_cste	3.3	1.75
Qsintcst	0.1	0.08
Defc_mult	1.5	1.5

HB Swamp

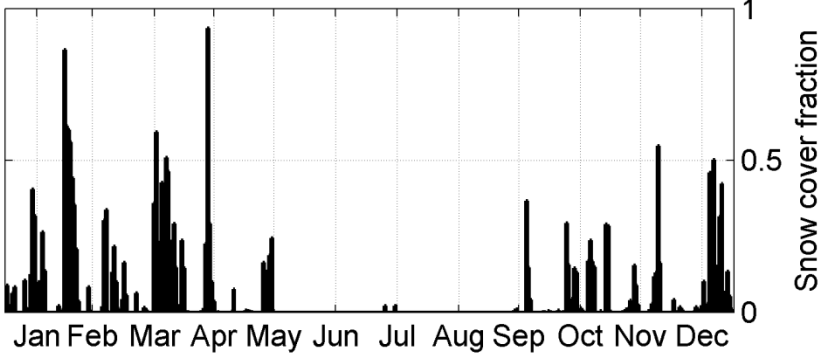
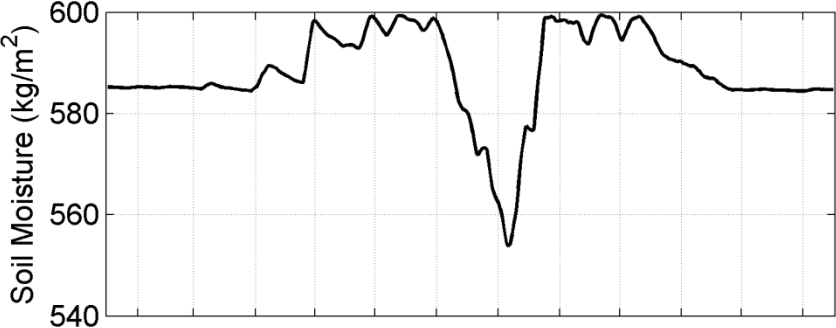
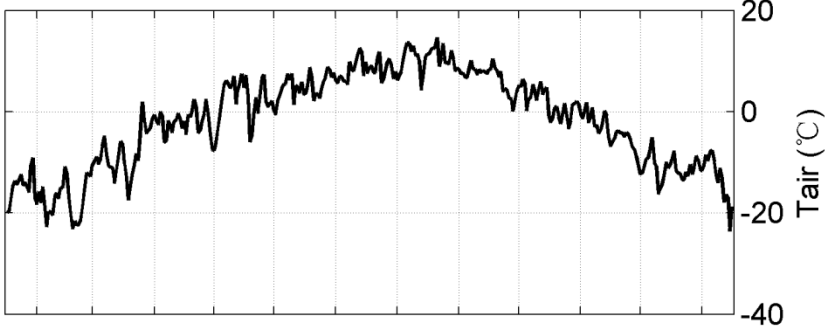


Obs
Prior
Post

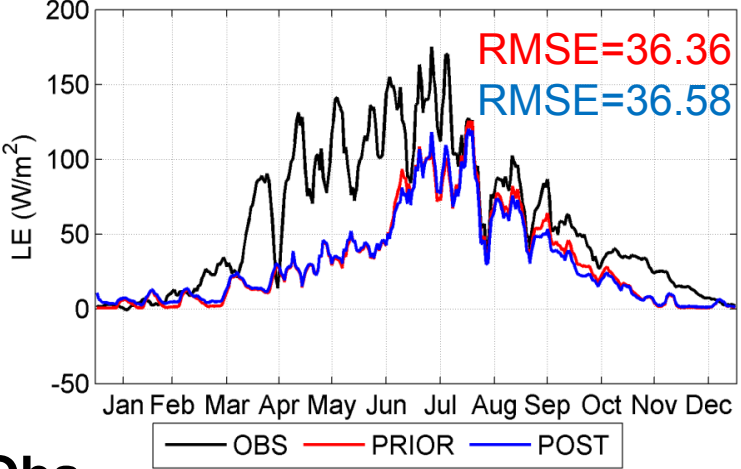
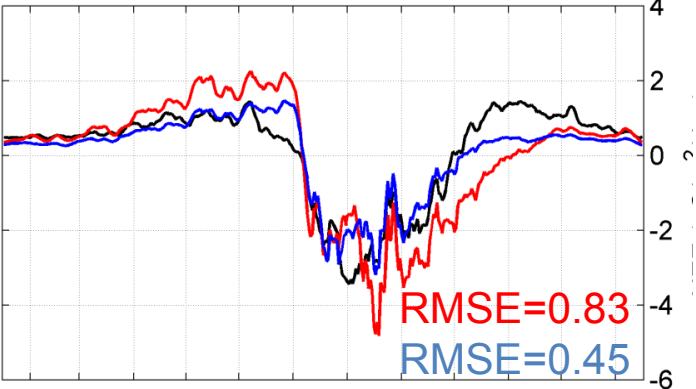
HB Swamp



HB Swamp



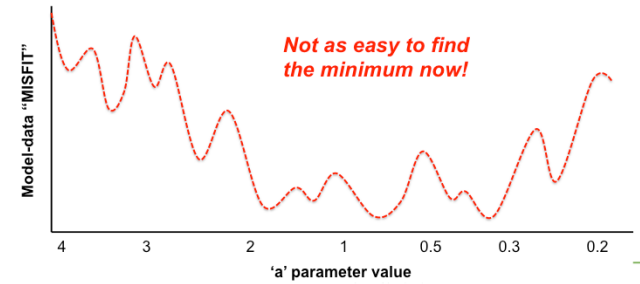
HB Swamp



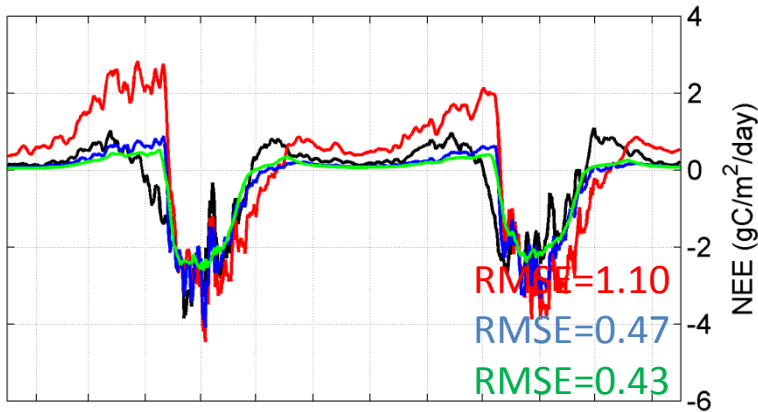
— OBS — PRIOR — POST

Obs
Prior
Post

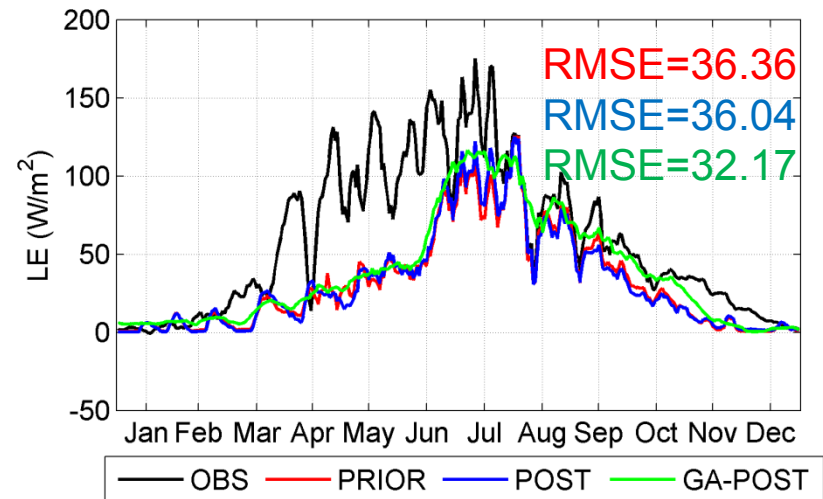
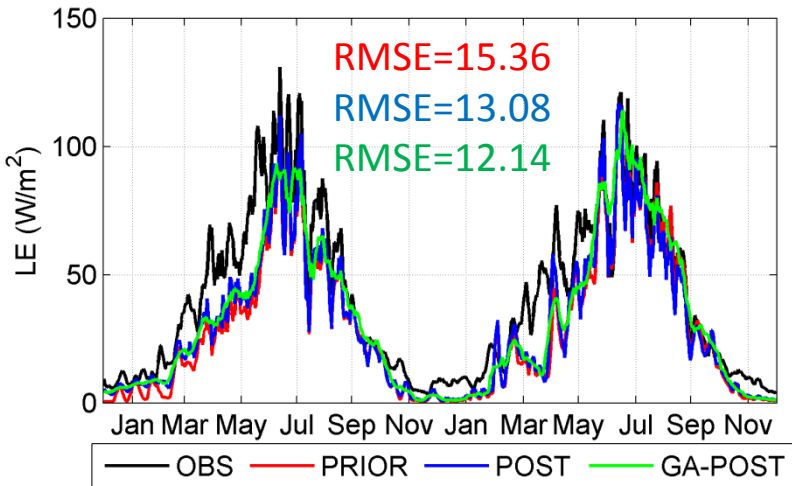
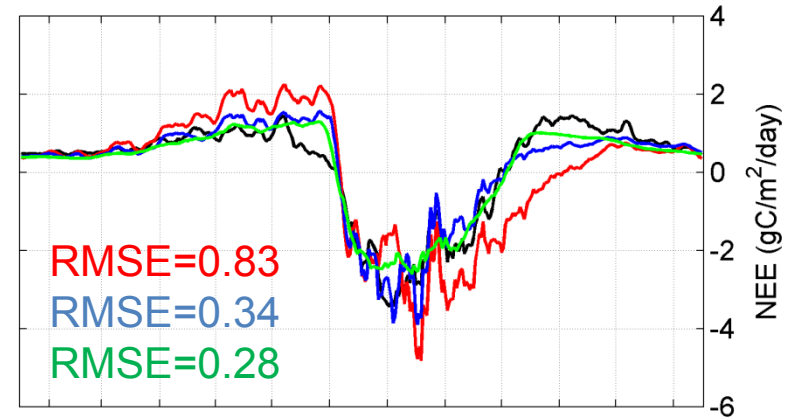
Variational versus Monte-carlo (Genetic Algo.) optimisation



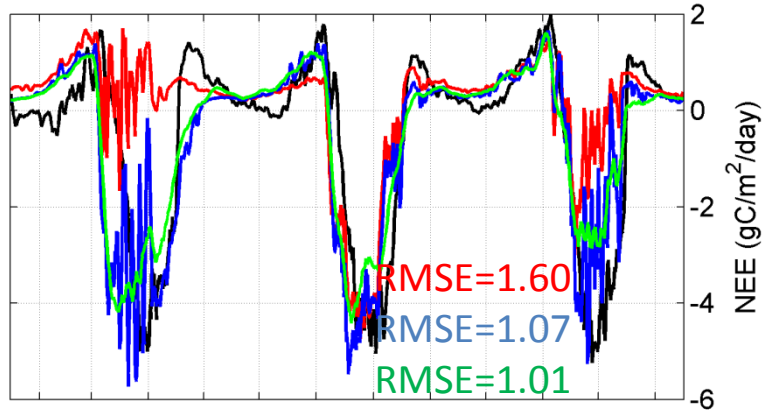
HB Shrub



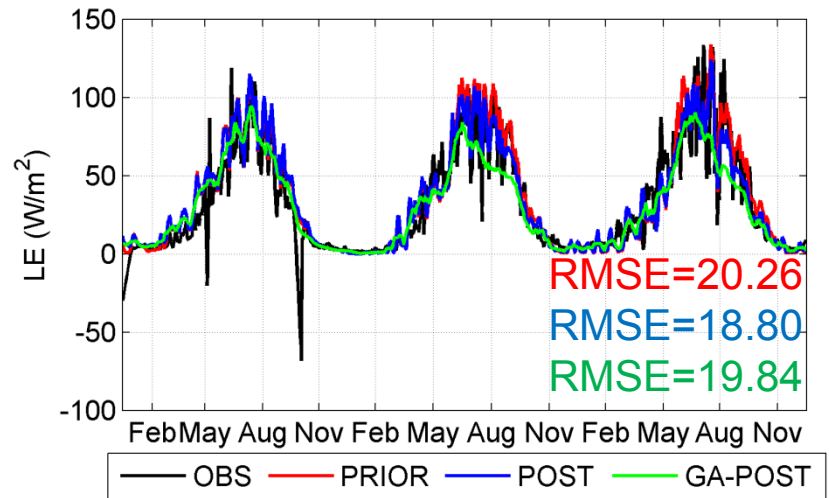
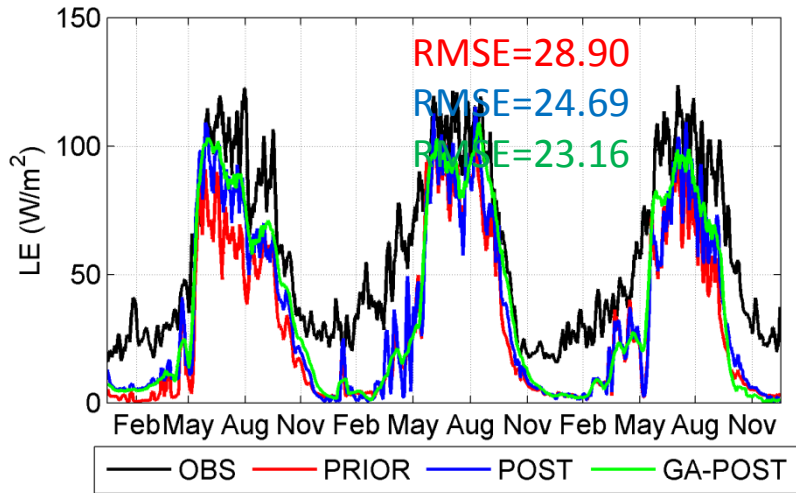
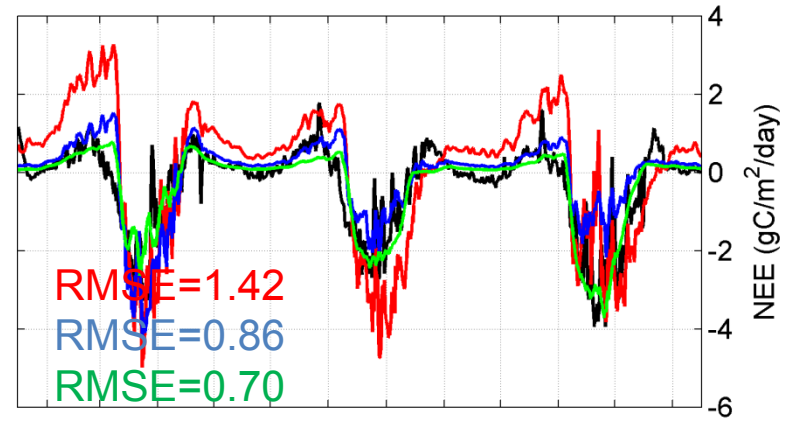
HB Swamp



DX swamp



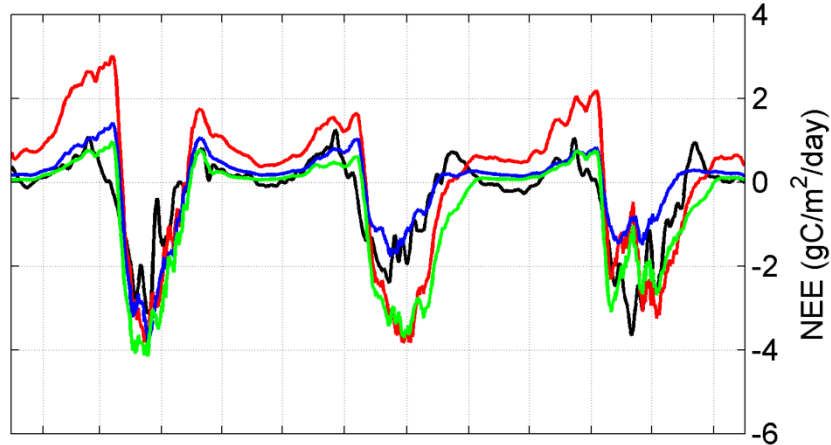
HB meadow



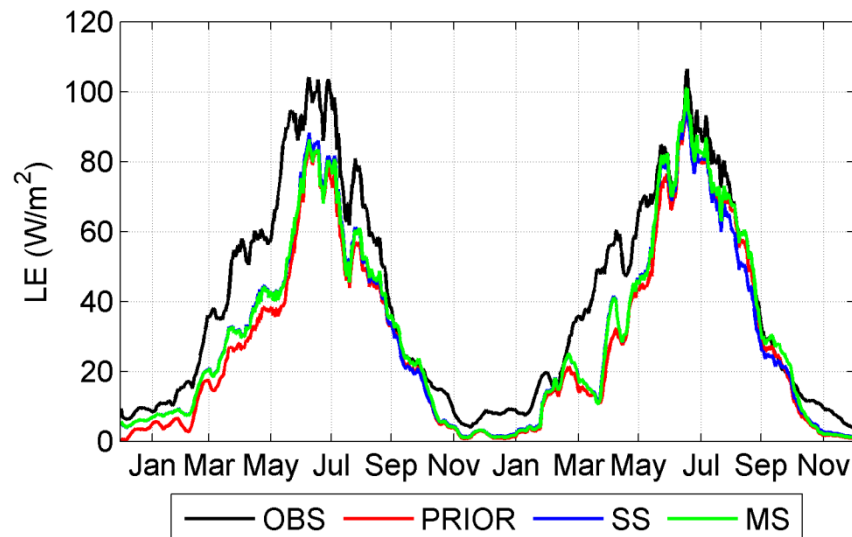
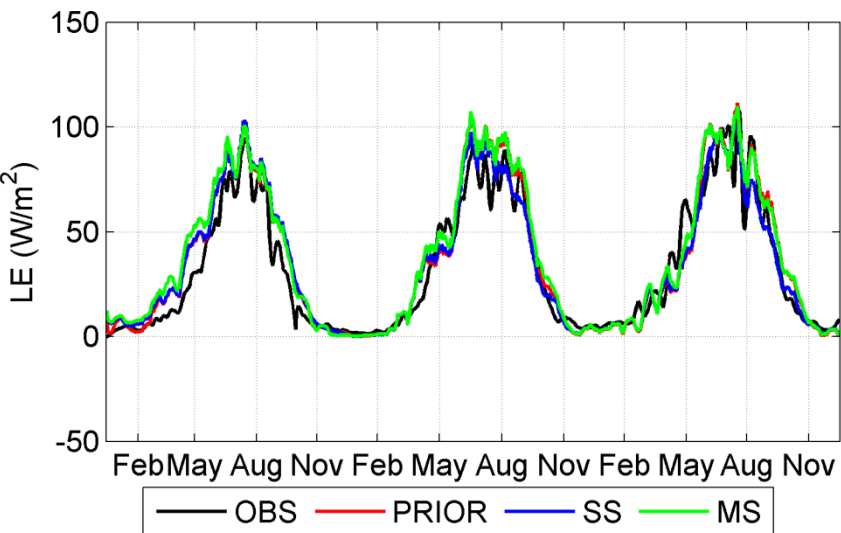
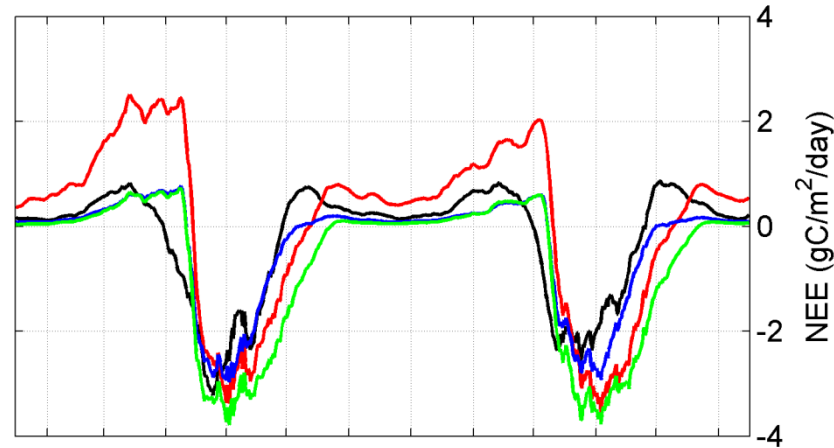
The optimization using Genetic Algorithm is generally better!
But this optimization increase computing time, and only improves a little.

Single Site optimization vs. Multi Site optimization

HB Meadow



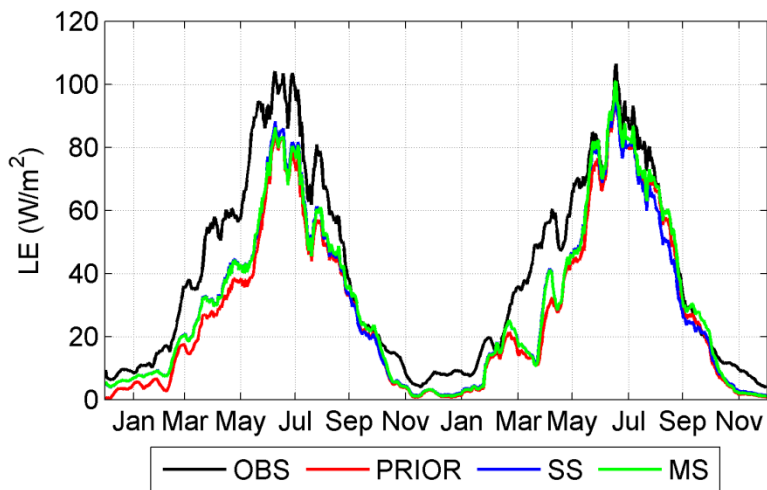
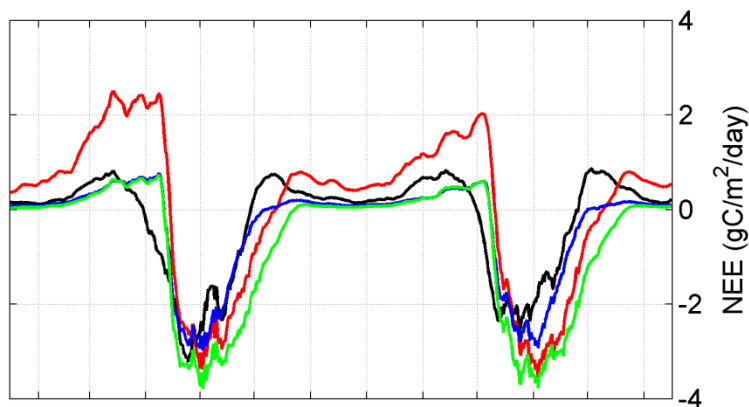
HB Shrub



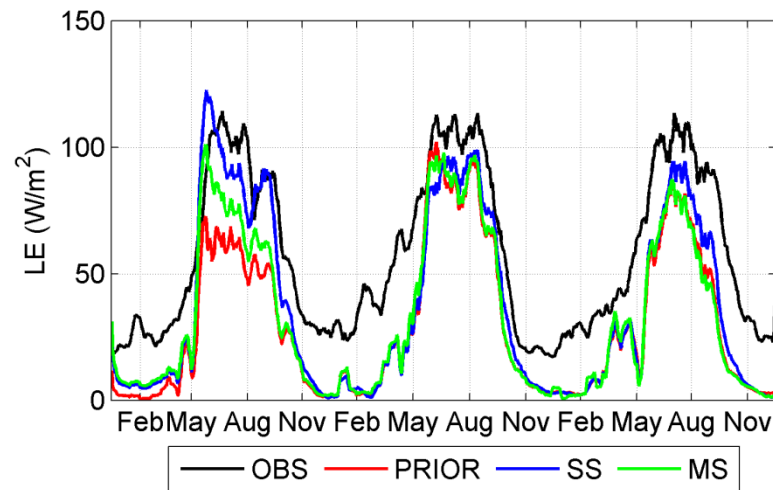
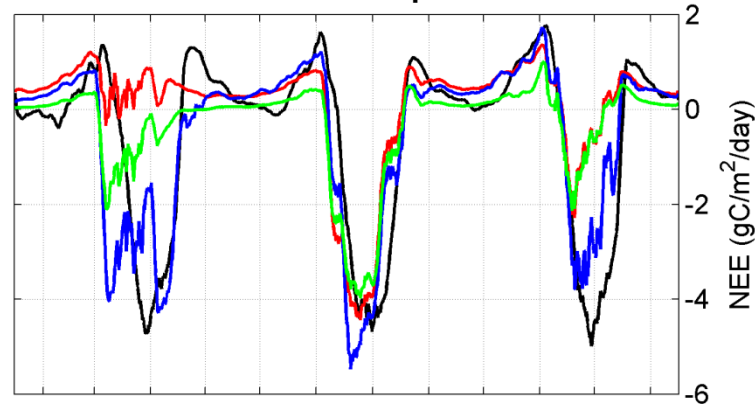
1. the MS parameter set significantly improves the model-data fit at most sites
2. The results of the MS optimization are often comparable to those of the SS optimizations

Single Site optimization vs. Multi Site optimization

HB Shrub



DXSwamp



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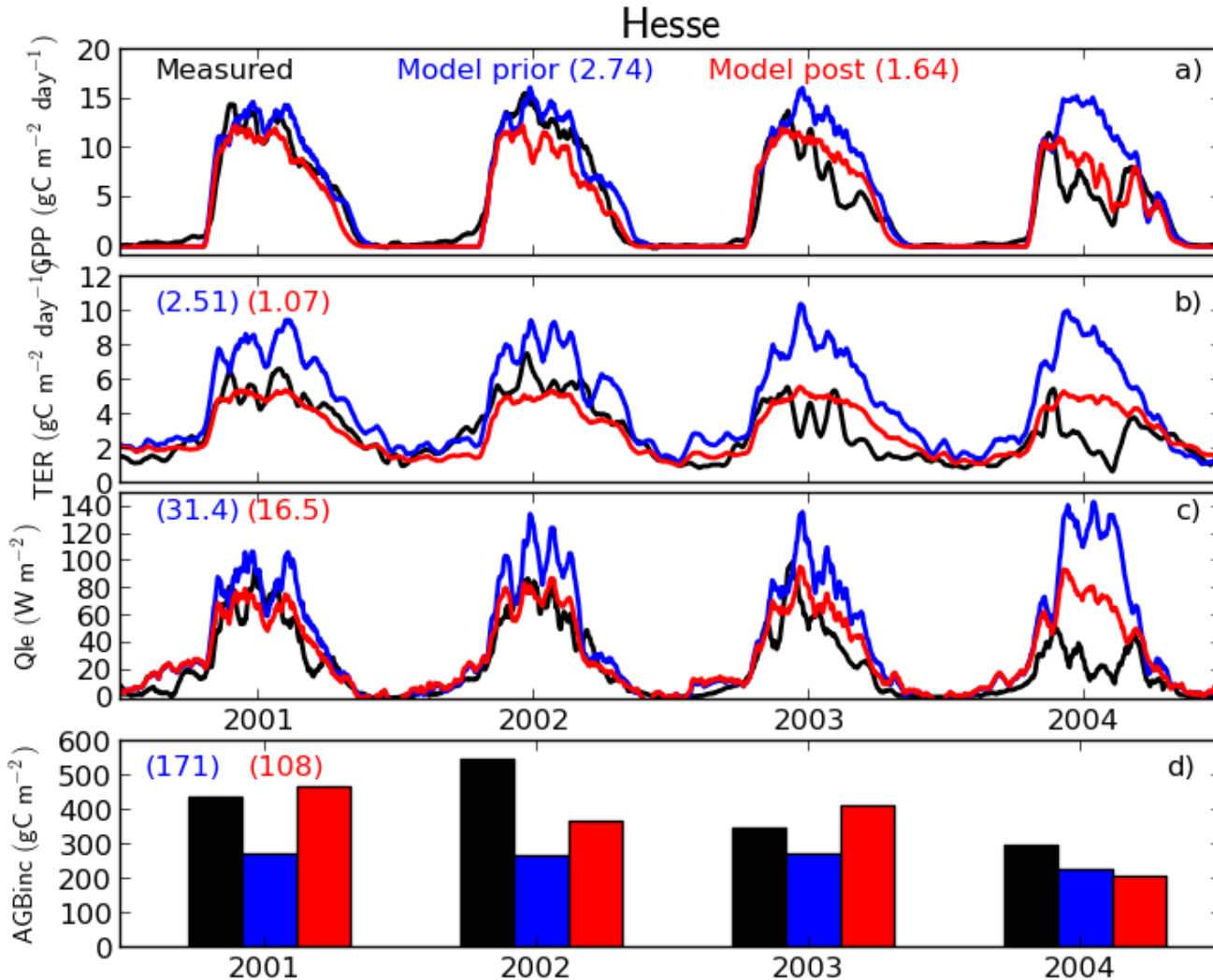
Summary and on-going work

- Preliminary results with Tibetan grassland sites
 - Optimization → significantly improve NEE simulation
 - Promising approach to improve ORCHIDEE for « high altitude » grassland ecosystems
- Using multiple data streams constrains different aspects of the model
- Simultaneous optimisation results in better fit overall
- BUT there are difficulties! → May be inconsistencies in model or between model and obs





Joint assimilation of biomass and in situ fluxes



➔ Assimilate obs:

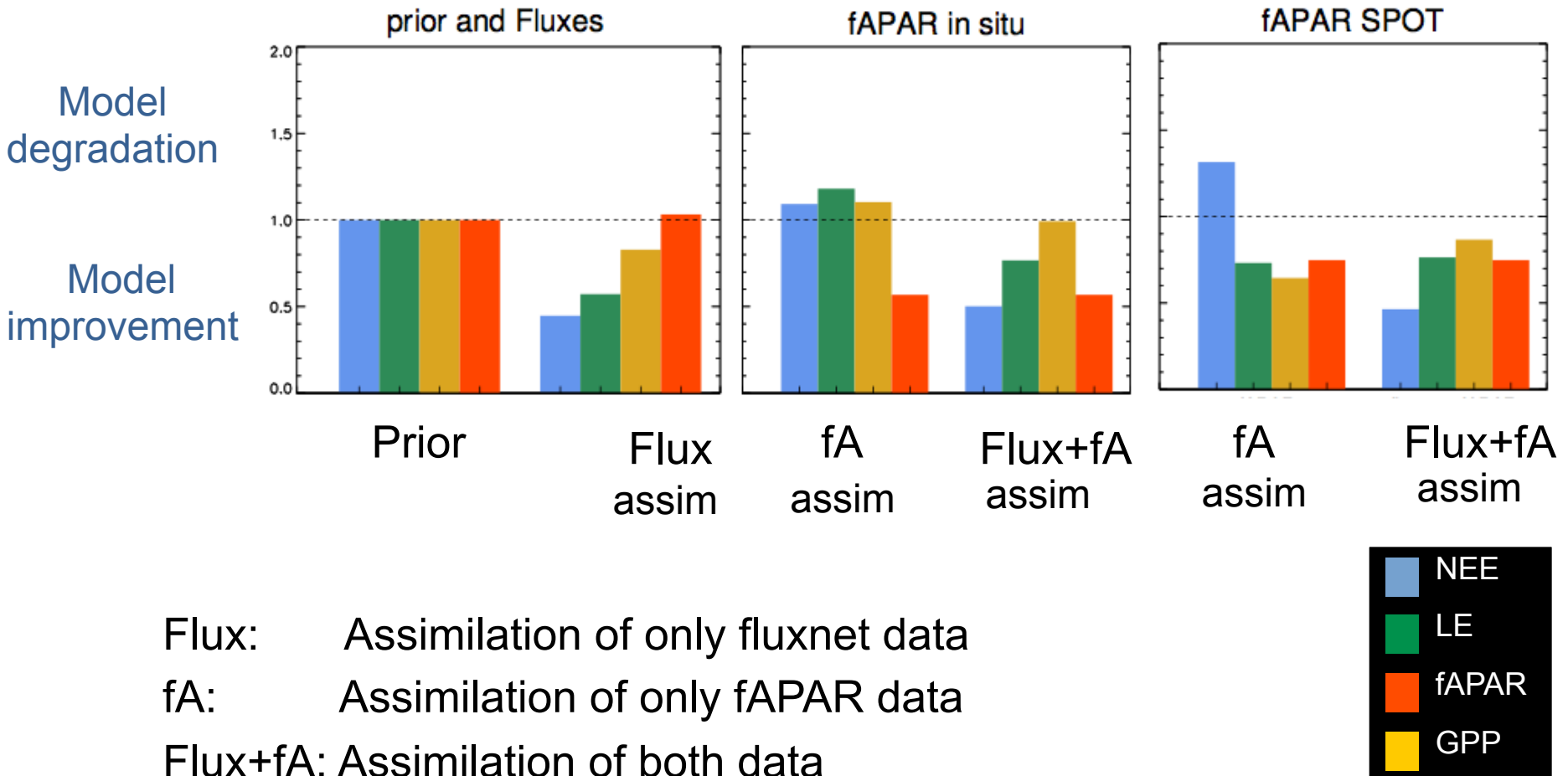
- GPP
- TER
- LE
- AGB increment

➔ Optimise params:

- photosyn
- resp
- energy balance
- soil water avail.
- phenology
- allocation

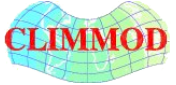
Joint assimilation of FluxNet and MODIS-fAPAR

→ Fontainebleau (Oak forest) : RMSE_poste / RMSE_prior





Thank you!



orchidas.lsce.ipsl.fr



Work undertaken as part of several projects:



Name	Description	Value	Name	Description	Value
- Carbon					
Vcmax_opt	Maximum carboxylation rate		Senescence_temp_c	Temperature threshold for senescence	
Tphoto_min_c	Minimal photosynthesis temperature		Leafagecrit	Average critical age for leaves	
Tphoto_max_c	Maximum photosynthesis temperature		Qsintcst	interception reservoir coefficient	
Senescence_temp_c	Temperature threshold for senescence		Maint_resp_c	Maintenance respiration coefficient	
LAI_MAX	Maximum LAI per PFT		Maint_resp_slope_c	slope of maintenance respiration coefficient	
SLA	Specific leaf area		Rsol_cste	Evaporation resistance	
Dpu_cste	Total depth of soil water pool		Residence_time	residence time of trees	
S0	sapwood allocation		So_capa_dry	Dry Soil Capacity	
Z0_over_height	Characteristic rugosity length		So_capa_wet	Wet Soil Capacity	



Name	Description	Value
-Energy and Hydrology		
Klaihappy	LAI threshold to stop carbohydrate use	
Tau_leafinit	time required to develop a minimal LAI	
Frac_growthresp	fraction of GPP which is lost as growth respiration	
Humcste	Root profile	
Defc_mult	Constant in the computation of surface resistance	
Q10	Temperature dependence of heterotrophic respiration	
Z_decomp	Maximum depth for soil decomposer's activity	
Moistcont_a	Soil moisture coefficient	
Moistcont_b	Soil moisture coefficient	
Moistcont_c	Soil moisture coefficient	

