



Assimilation of satellite fAPAR data within the ORCHIDEE biosphere model and its impacts on land surface carbon and energy fluxes

CAMELIA project funded by ESA



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Objectives

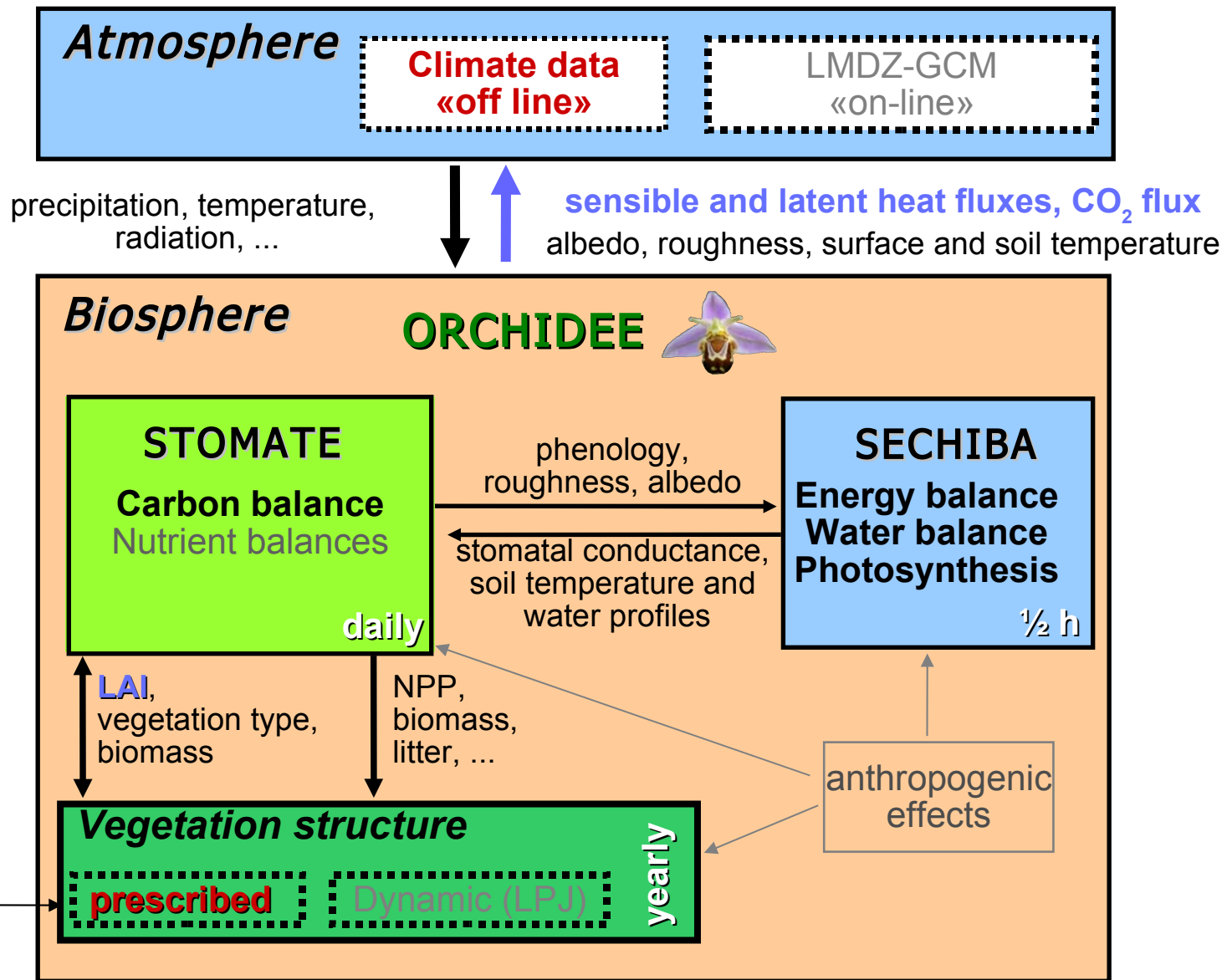
Improve ORCHIDEE vegetation model through parameter optimization

- ORCHIDEE: **global** vegetation model which parameterization may need tuning for specific sites
- Use of data at the **site level**
 - ✓ *in situ* flux measurements of Net CO₂, Sensible & Latent Heat, fluxes
 - ✓ **potentials** of satellite fAPAR time series at High and Medium spatial resolutions

Main scientific questions

- Can we combine **flux data** and **satellite fAPAR** through an assimilation process?
- What do we learn on the model strengths and weaknesses?
- What is the “optimal” spatial resolution ?

The ORCHIDEE vegetation model

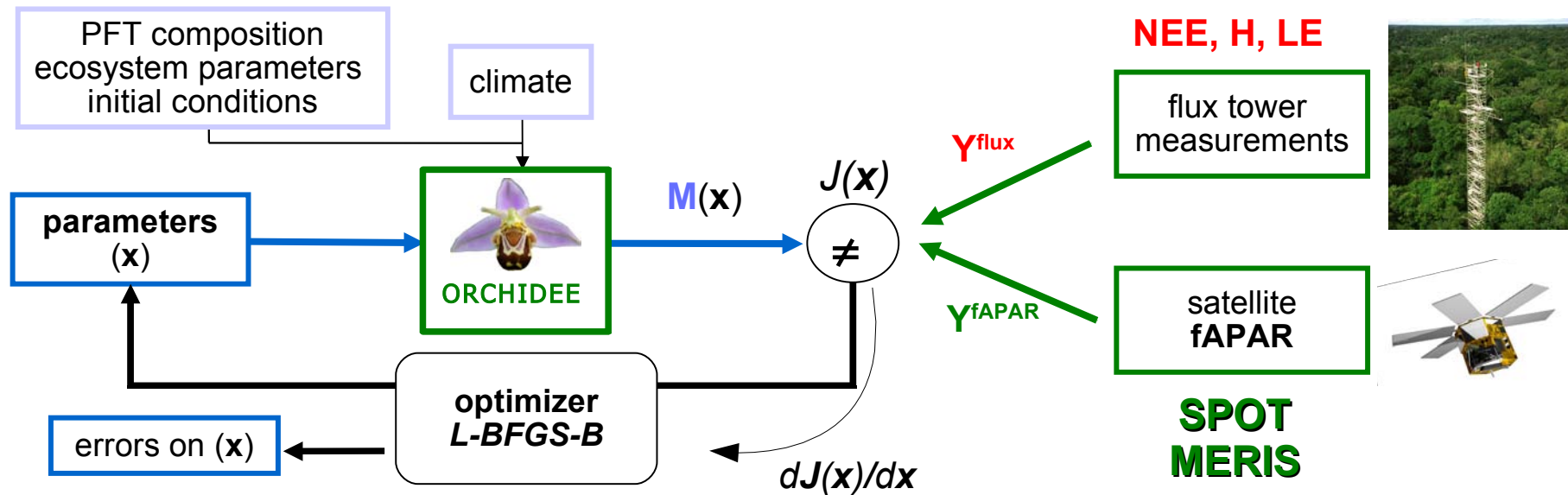


$$fAPAR(t) = \sum_i \alpha_i [1 - \exp(-K_i(t) \cdot LAI_i(t))]$$

K : extinction coefficient

α : maximum vegetation fraction

Variational assimilation system



Bayesian optimization

$$\begin{aligned}
 J(\mathbf{X}) = & (\mathbf{Y}^{\text{flux}}_{\text{daily}} - \mathbf{M}(\mathbf{x}))^T \mathbf{R}_{\text{flux}}^{-1} (\mathbf{Y}^{\text{flux}}_{\text{daily}} - \mathbf{M}(\mathbf{x})) & + & \text{daily means} \\
 & (\mathbf{Y}^{\text{fAPAR}} - \mathbf{M}(\mathbf{x}))^T \mathbf{R}_{\text{fAPAR}}^{-1} (\mathbf{Y}^{\text{fAPAR}} - \mathbf{M}(\mathbf{x})) & + & \text{weekly fAPAR} \\
 & (\mathbf{x} - \mathbf{x}_0)^T \mathbf{B}^{-1} (\mathbf{x} - \mathbf{x}_0) & & \text{prior information}
 \end{aligned}$$

- iterative minimization of $J(\mathbf{X})$
- $dJ(\mathbf{x})/d\mathbf{x}$ computed using the Tangent Linear version of ORCHIDEE

Data

Sites considered

- **Fontainebleau - 2006**: deciduous forest (oak trees)
- **Le Bray - 2003**: pine forest
- **Puechabon - 2004**: mediterranean broadleaf evergreen forest (green oak trees)

Flux data

- Flux tower measurements of NEE, H, LE, at a half-hourly time step
- **Daily means** + temporal smoothing using a +/- 15 days moving average window

Satellite fAPAR

- **Weekly** estimates from SPOT and MERIS - Neural Network estimation algorithm
- SPOT: few observations (cloudiness) \Rightarrow temporal extrapolation by a 2-sigmoid model corrected from seasonal trends using MERIS fAPAR data at 1km
 - ✓ **40m** : mean of the pixels around the flux tower
 - ✓ **1km** : pixels having the same vegetation composition than the flux tower pixel

Compatibility between fAPAR and flux data

Methodology

- 3 assimilations: flux only / fAPAR only / flux + fAPAR
- By how much has the fit between model/observations been improved?

Governing processes and parameters to optimize

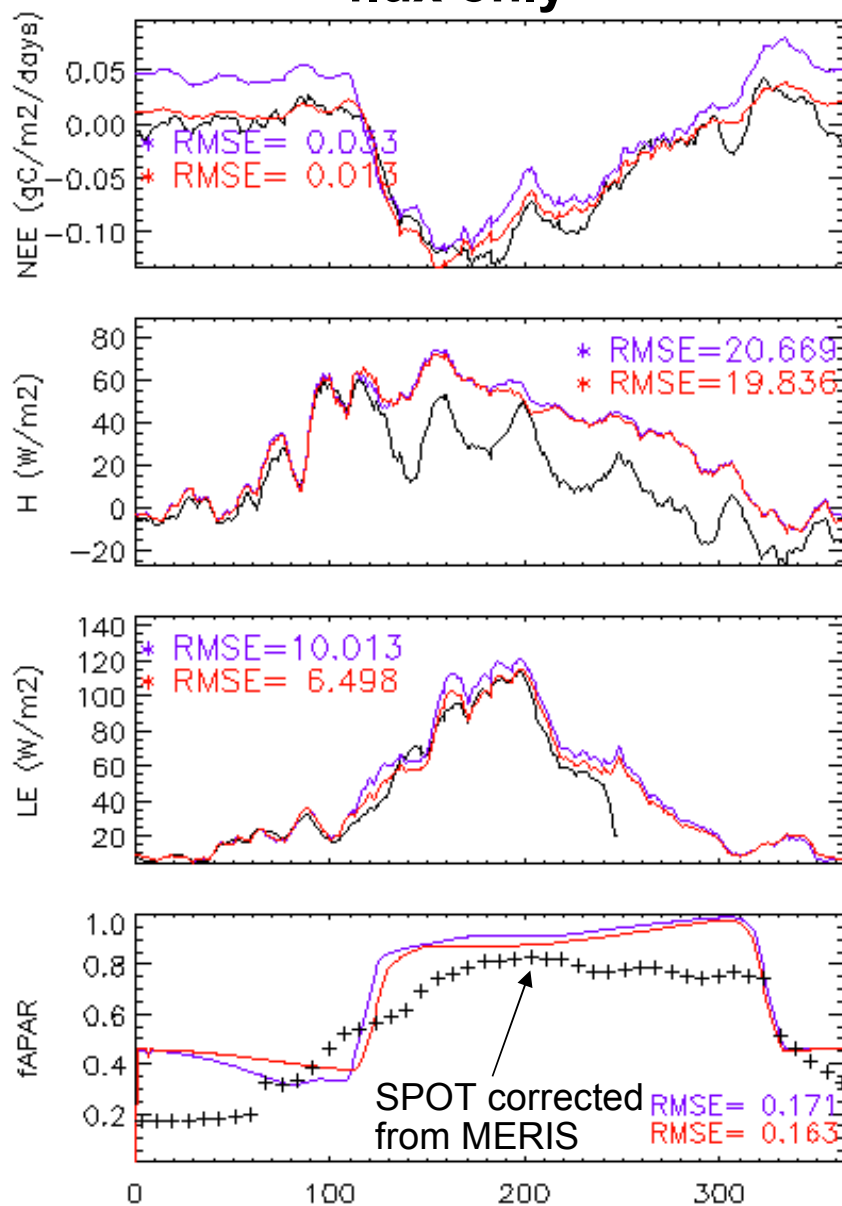
- Carbon assimilation —————▶ Vcmax_opt, Gsslope, **LAIMAX***, SLA, **Clumping***
- Autotrophic respiration —————▶ Fracgrowth_resp
- Heterotrophic respiration —————▶ Q10, KsoilC
- Plant phenology —————▶ **Kpheno_crit***, **Leafage***, **Senescence_T***, **LAI_init***
- Energy balance —————▶ Kalbedo_veg

** fAPAR assimilation only*

Fontainebleau

Fontainebleau (2006) : deciduous trees + grassland understory

flux only



observation

prior

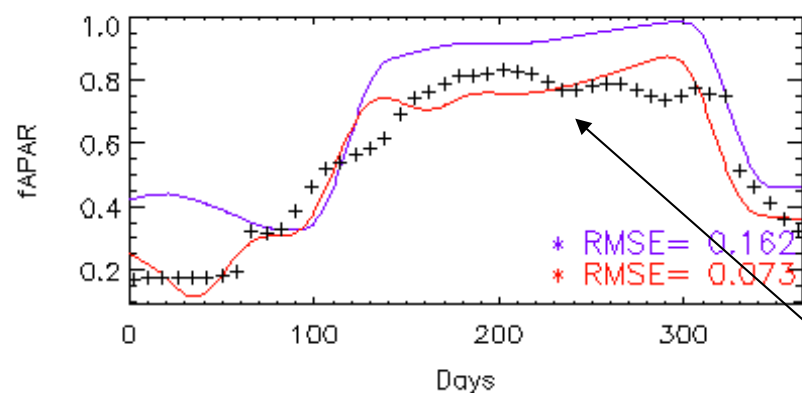
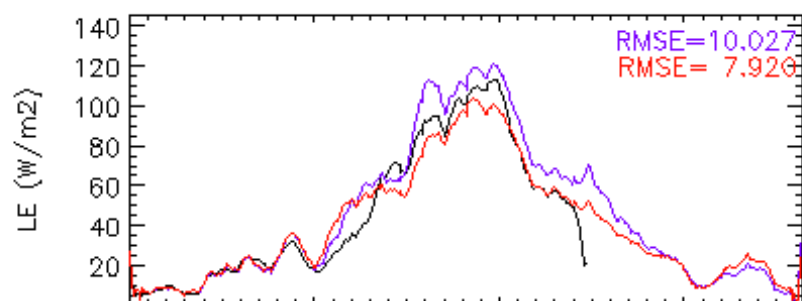
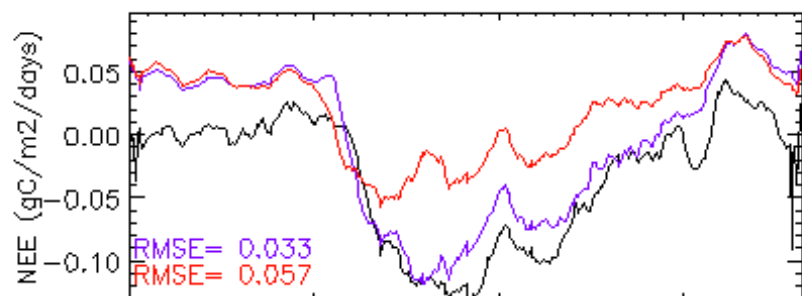
posterior

- Very good agreement between ORCHIDEE and NEE & LE observations
- Misfit between model / measurements for H after leaf onset
- Low impact on posterior fAPAR simulations
- Smooth temporal variation of fAPAR in the satellite products / abrupts changes in the model (leaf onset, senescence)
 - processing of fAPAR satellite data
 - land-cover variability seen by satellites / mean behaviour simulated by ORCHIDEE
- No vertical mixing of vegetation in ORCHIDEE

Fontainebleau

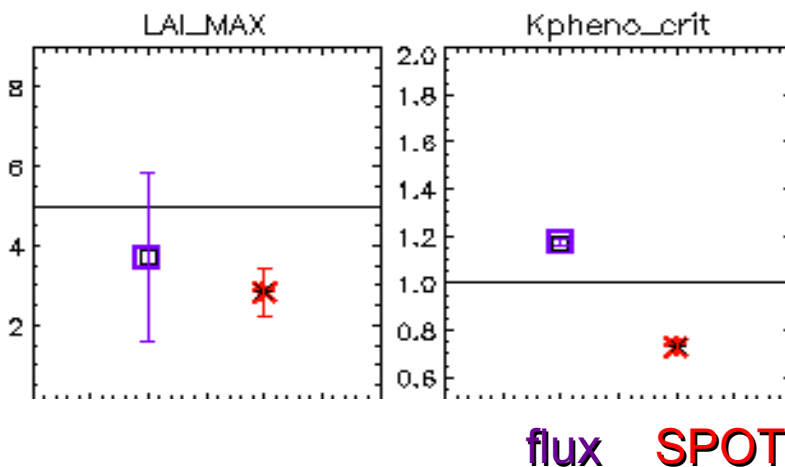
Fontainebleau (2006) : deciduous trees + grassland understory

SPOT fAPAR only



- Strong reducing of the carbon uptake in summer (NEE)
- Advance of the start of the growing season

Optimized parameters



observation

prior

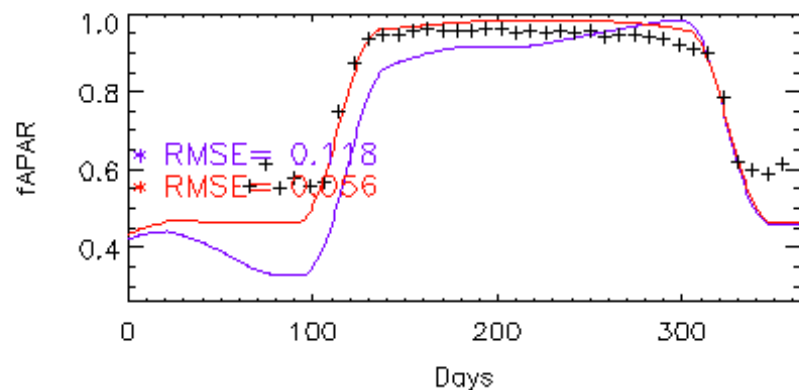
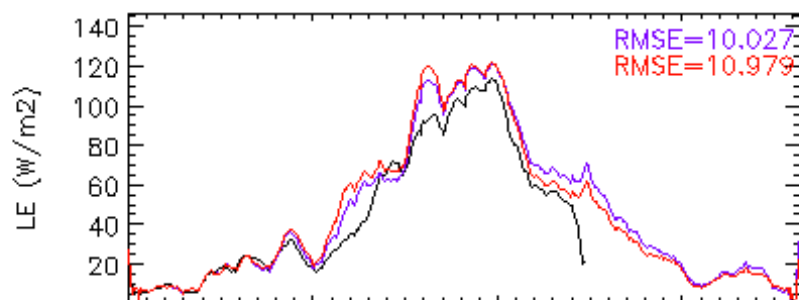
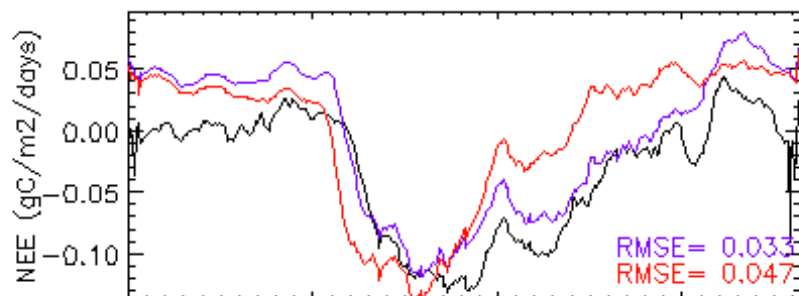
posterior

SPOT corrected
from MERIS

Fontainebleau

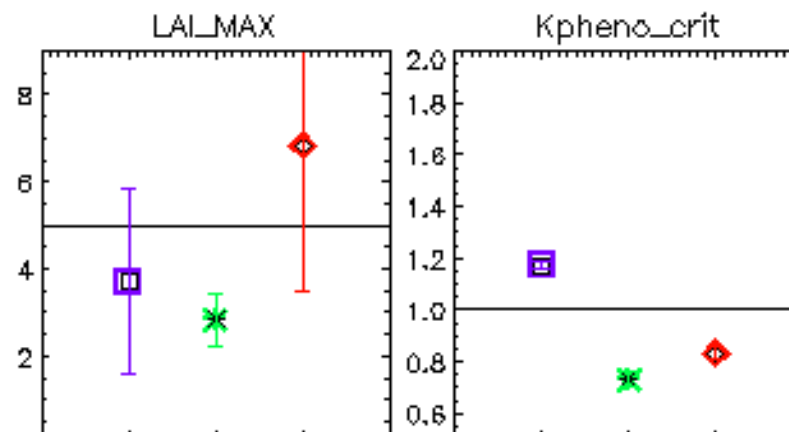
Fontainebleau (2006) : deciduous trees + grassland understory

fAPAR *in situ* only



- Model / fAPAR data agreement: length of the growing season, slope of fAPAR increase
- Slight advance of the date of leaf onset:
 - ✓ assimilation problem of tuning initial LAI
 - ✓ model deficiency: different phasing between NEE and fAPAR?

Optimized parameters



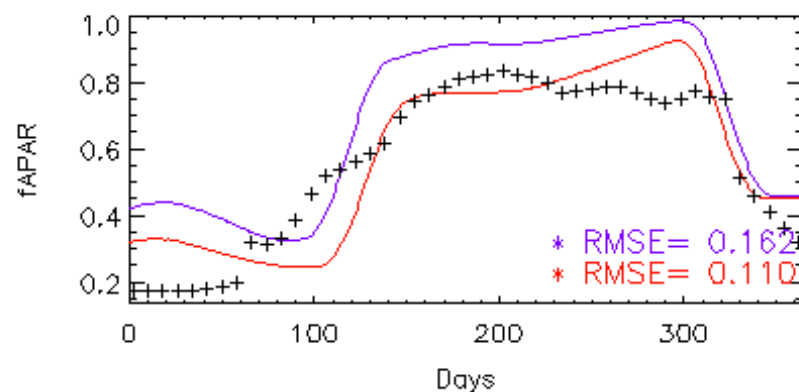
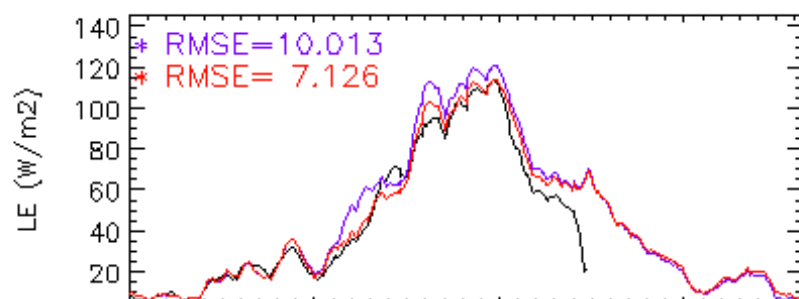
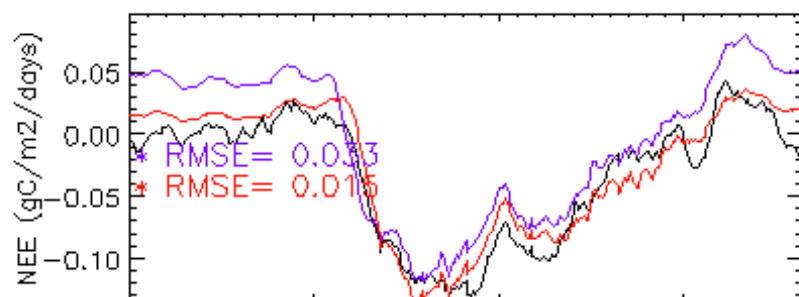
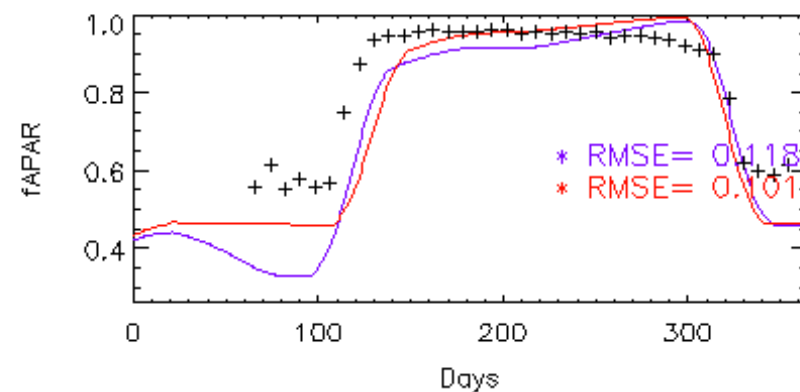
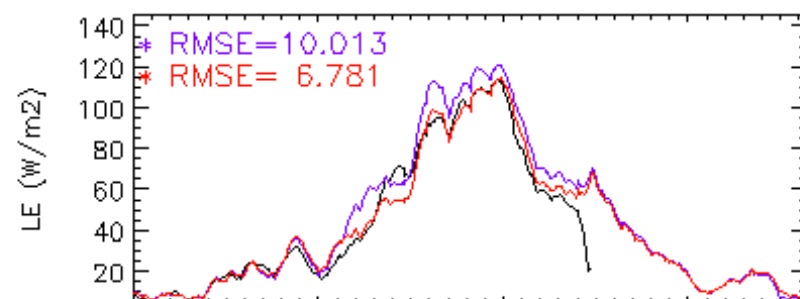
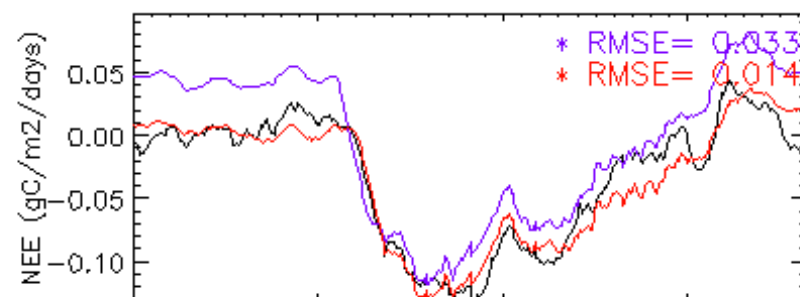
observation prior posterior

flux SPOT in situ fAPAR

Fontainebleau

Fontainebleau (2006) : deciduous trees + grassland understory

flux + SPOT fAPAR

flux + *in situ* fAPAR

observation
prior
posterior

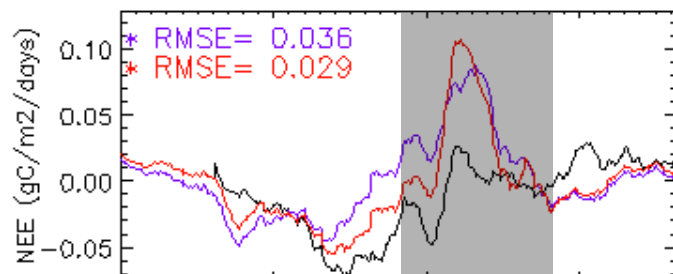
- flux data drive the assimilation results
- better agreement with *in situ* fAPAR data

Le Bray

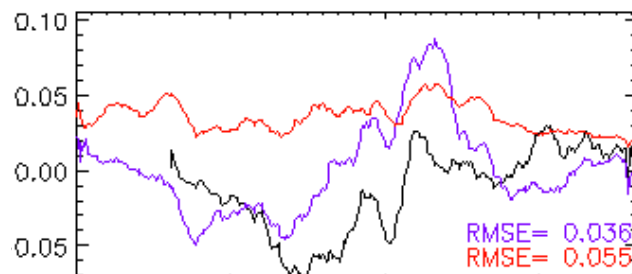
Le Bray (2003) : pine trees + grassland understory

observation **prior** **posterior**

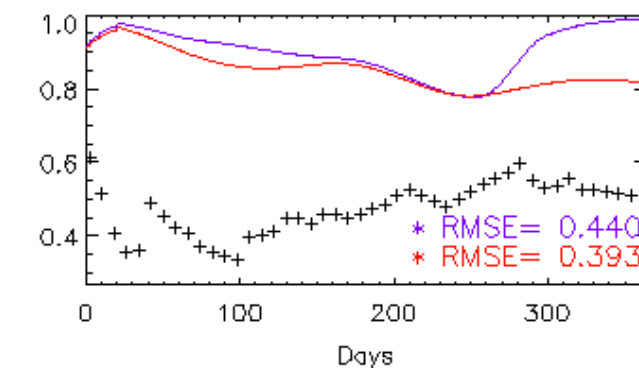
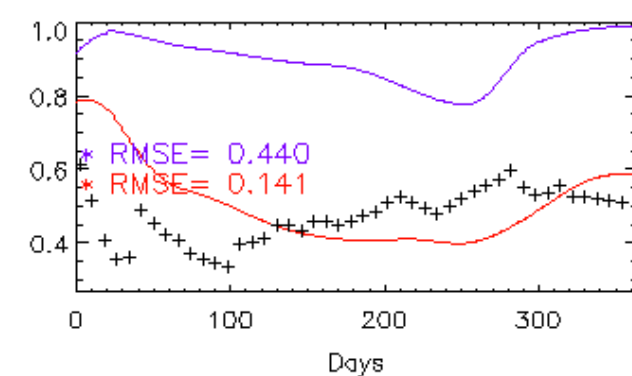
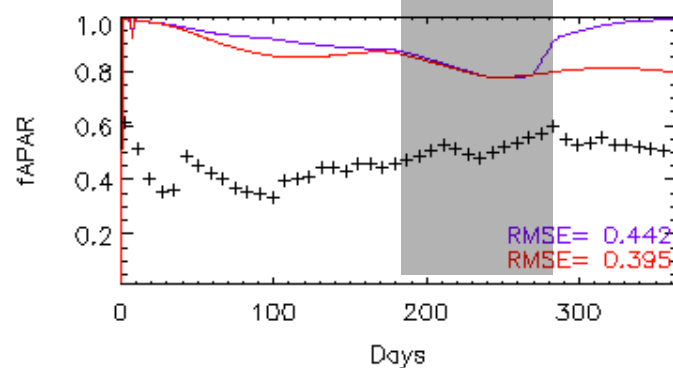
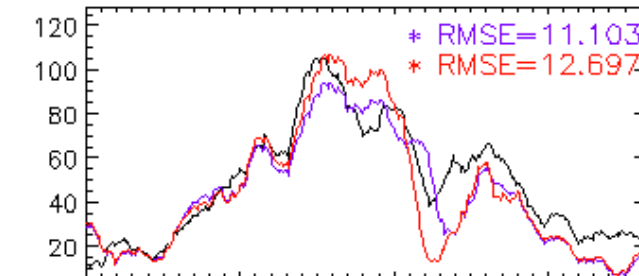
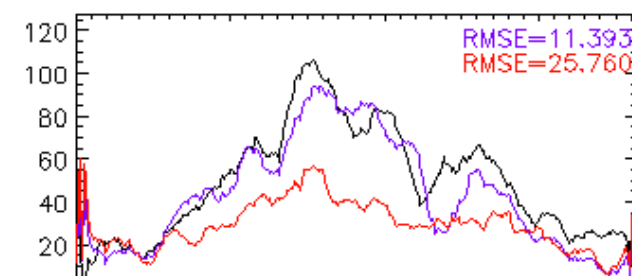
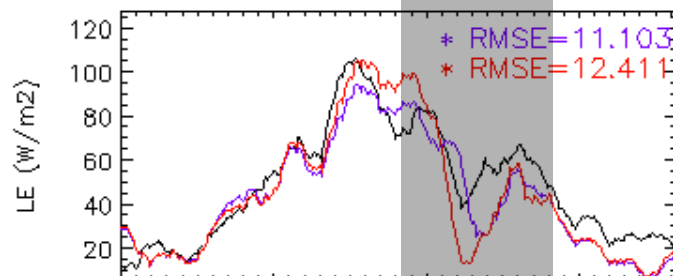
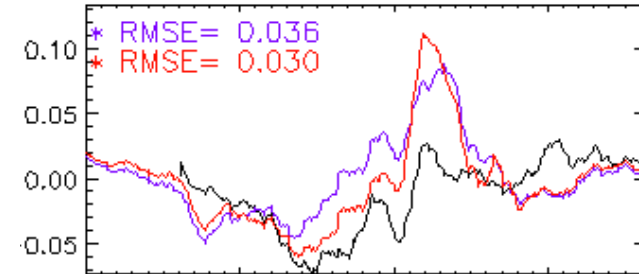
flux only



SPOT fAPAR only



flux + fAPAR



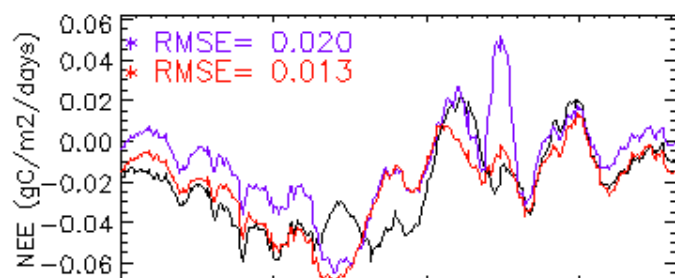
- 2003 drought summer event not well captured by the model
- Incompatibility of fAPAR levels between model/satellite product

Puechabon

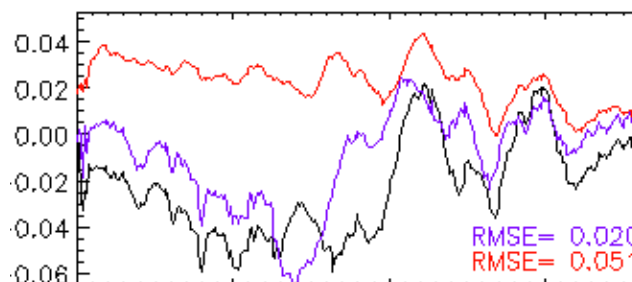
Puechabon (2004) : mediterranean broadleaf evergreen trees

observation **prior** **posterior**

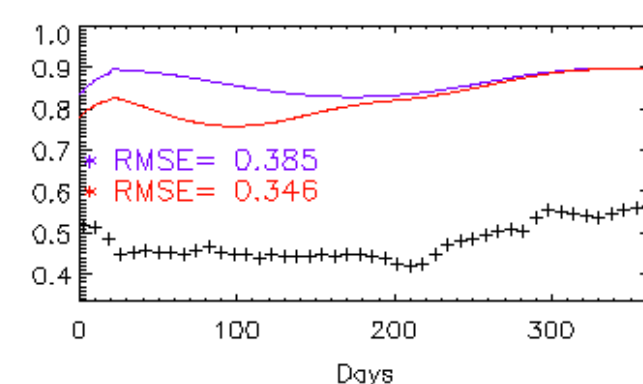
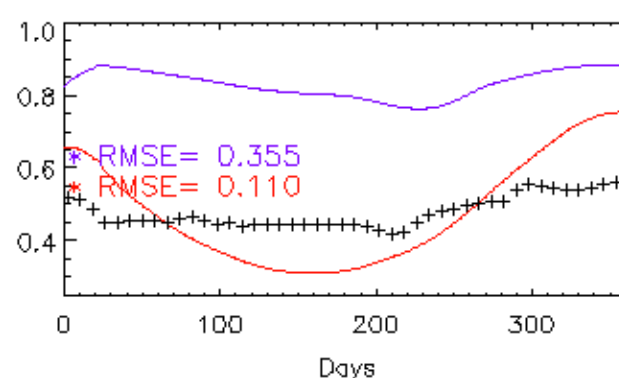
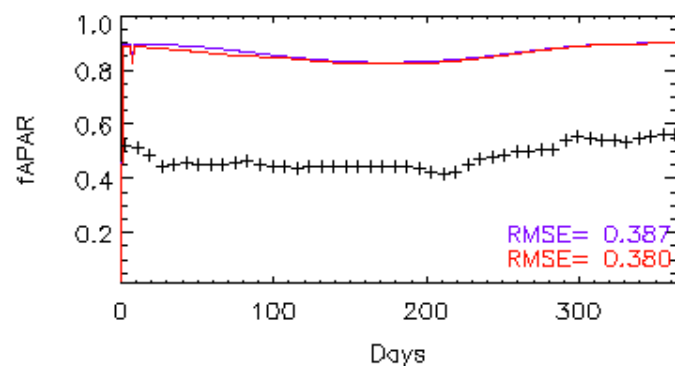
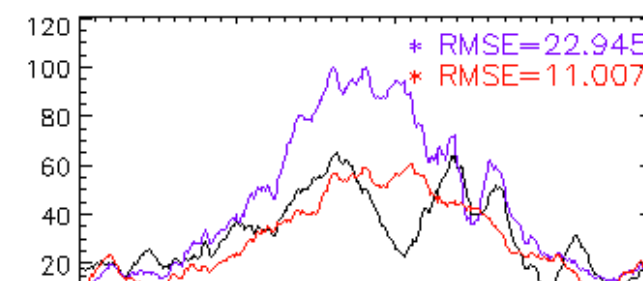
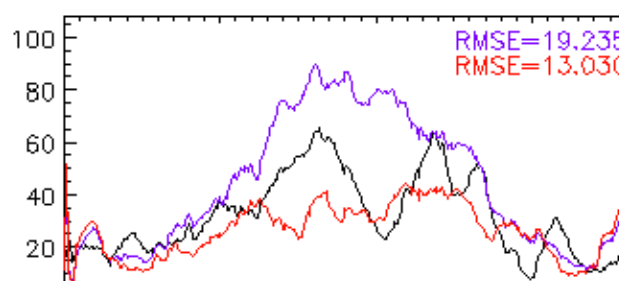
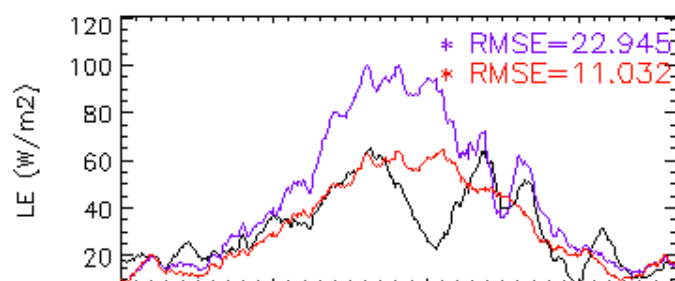
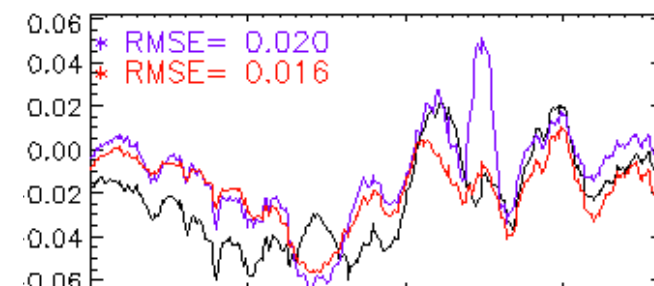
flux only



MERIS fAPAR only



flux + fAPAR



- simulated fAPAR reproduce the bowl-shape variation as in in site measurements, as well as the mean fAPAR level / low level for the satellite product
- similar findings than for Le Bray

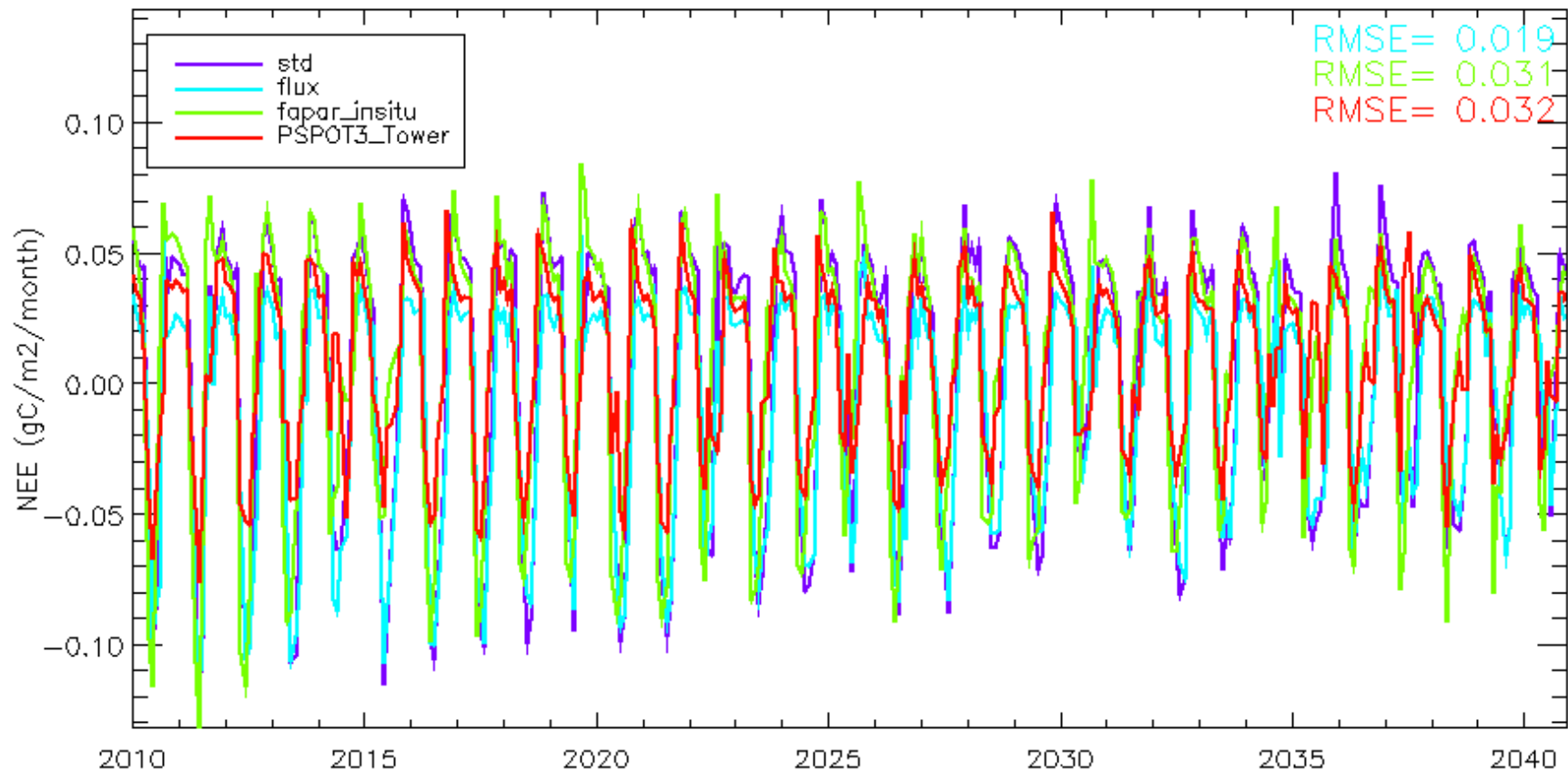
Conclusions

- **ORCHIDEE** simulates quite well the flux seasonal variations (NEE in particular)
- Assimilation of fAPAR data creates some inconsistencies between NEE observations / model due to differences in:
 - ✓ fAPAR levels
 - ✓ seasonality (timing, smoothness)
 - need for **high temporal resolution / high spatial resolution** fAPAR data
 - need in site validation datasets
- Use of HR fAPAR data requires model improvements:
 - ✓ vegetation vertical mixing
 - ✓ phenology for some specific PFT
- Joint assimilation of flux & fAPAR data seems possible
- Need careful examination of retrieved parameters and uncertainties

Impact of retrieved parameters on prognostic NEE simulations

- Meteorological forcing derived from the coupled IPSL model **2010-2040**
- A1B scenario of CO₂ emission from IPCC
- Various sets of ORCHIDEE parameters derived from assimilation of various classes of data

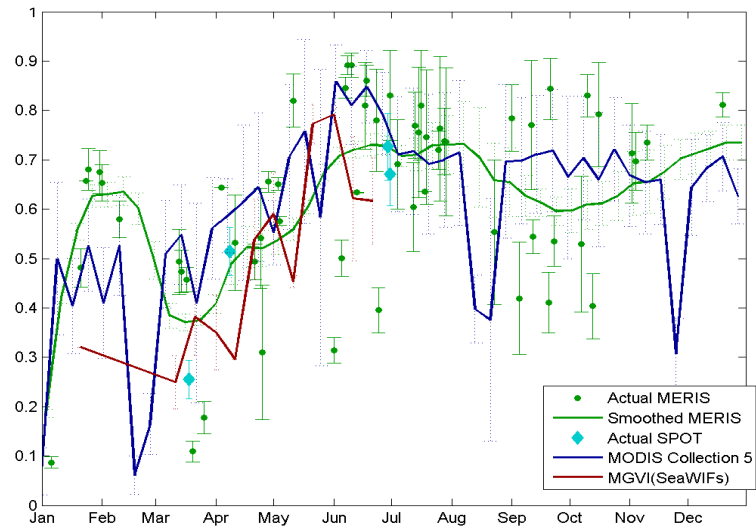
Fontainebleau (monthly means)



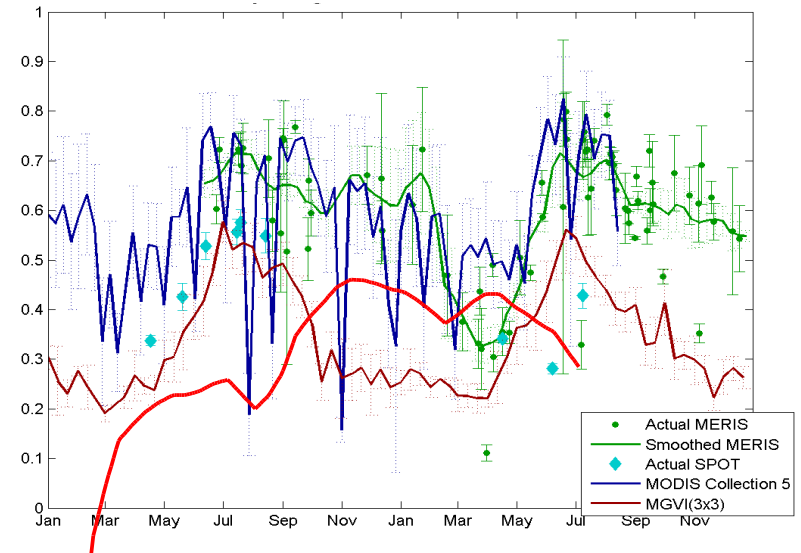


Satellite fAPAR data comparison

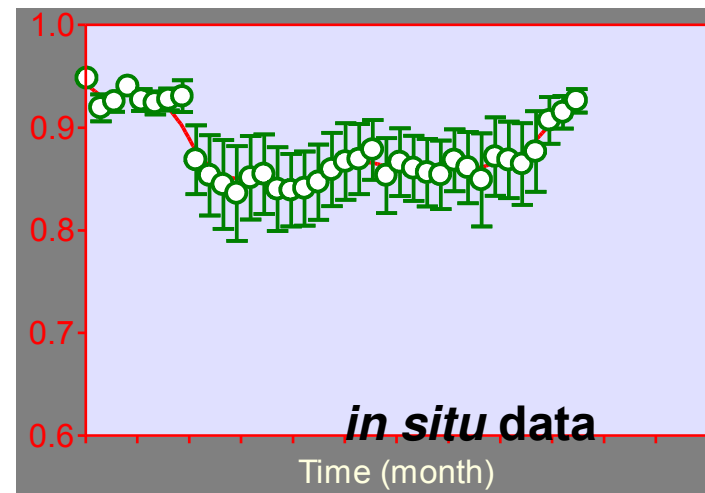
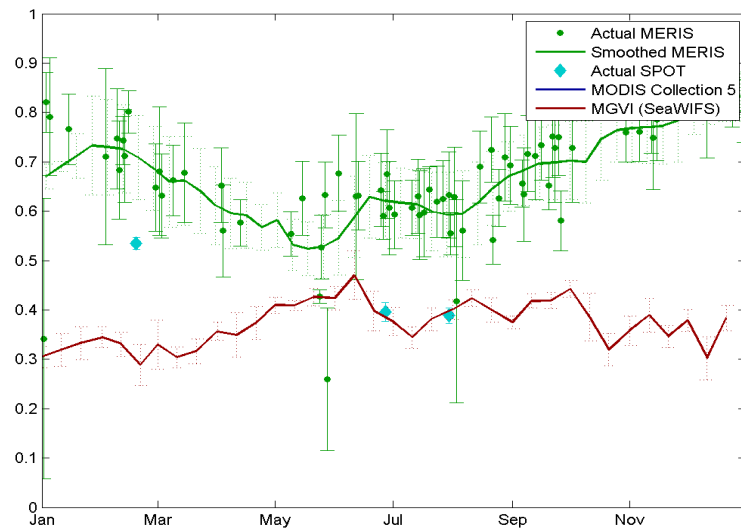
Fontainebleau (3x3 km²)



Le Bray (3x3 km²)



Puechabon (3x3 km²)



Compatibility model / satellite fAPAR with the spatial scale

Principles

- Assimilation of fAPAR data at 40m / 1km / 10km
- Assess the model/measurements misfit

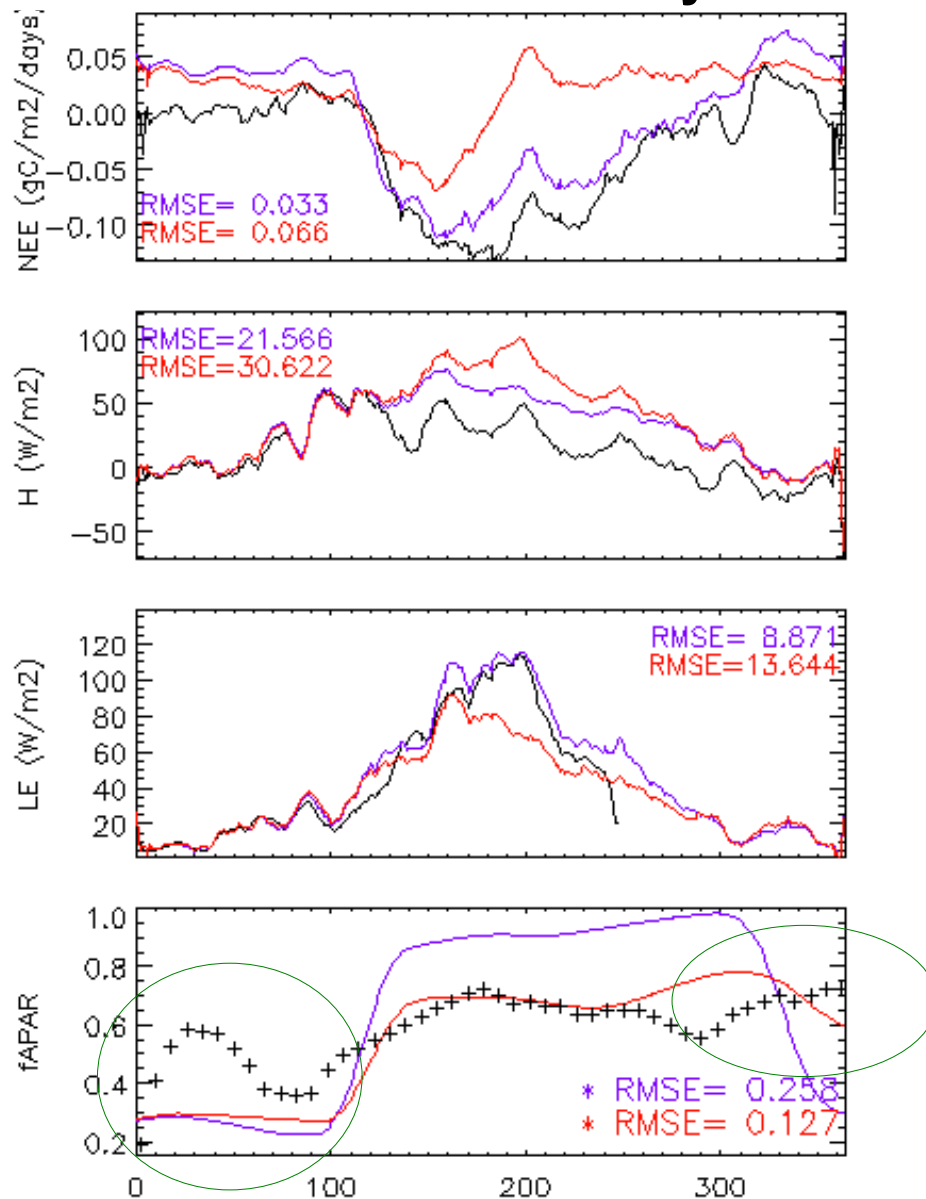
Increased fAPAR agreement at larger spatial resolution

- **HR:** local heterogeneity seen by the satellite / mean behaviour simulated by ORCHIDEE
- **MR:** spatial aggregation smoothes out the errors in satellite data and model (**mis-partitioning the scene into PFTs smoothes**)
-
- **Impact on the retrieved parameters:**
 - ✓ differences in the parameter estimates comprised within their error bars (except for Kpheno_crit)
 - ✓ increased uncertainty at medium & coarse spatial resolutions (lesser visibility of a parameter, higher degrees of freedom)

Compatibility model / satellite fAPAR with the spatial scale

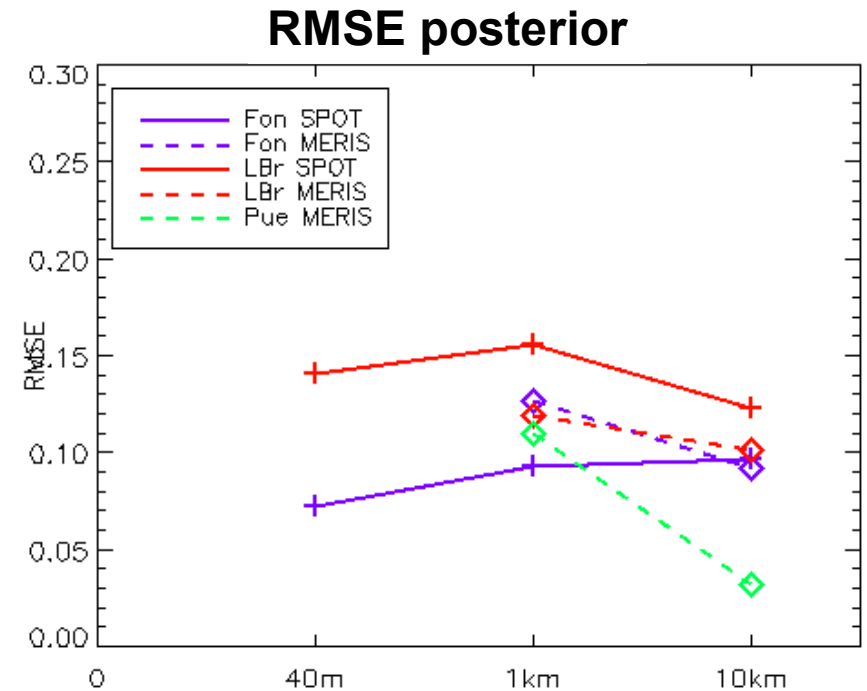
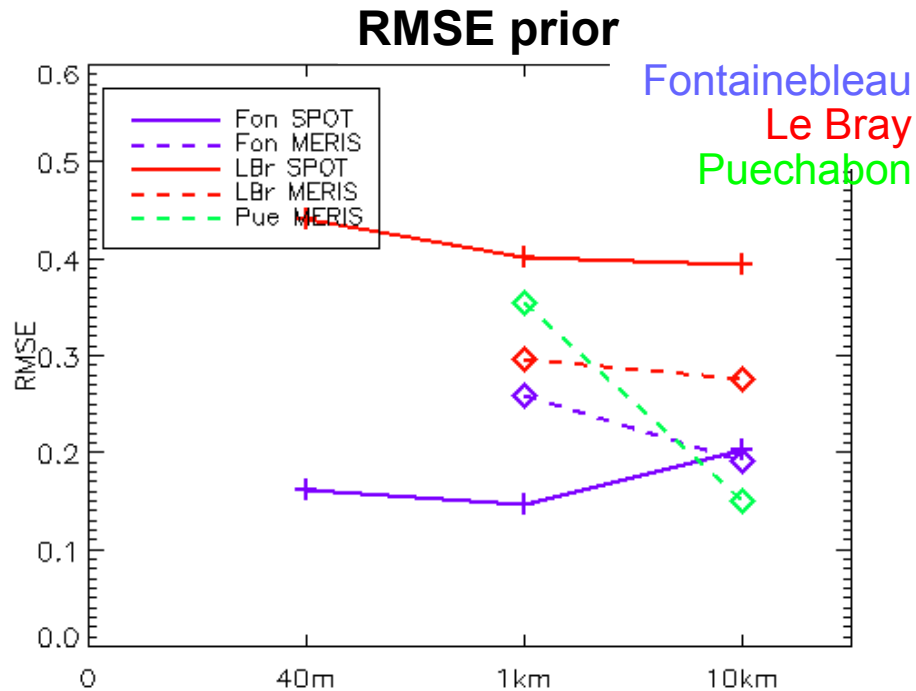
Fontainebleau

MERIS fAPAR only



- Problems of fAPAR levels in winter
 - ➔ residual atmospheric effects
 - ➔ high illumination angles not properly accounted for in the estimation algorithm
- Again, strong decrease of the net carbon assimilation in summer (NEE)

Compatibility model / satellite fAPAR with the spatial scale



- Better agreement at 10km than at 40m
 - ✓ **model:** each individual Vegetation Type has its individual “different” seasonal response and spatial mean is smoother than each individual
 - ✓ **satellite:** locally strong spatial heterogeneities in the satellite products are smoothed out at larger spatial resolutions
- **PFT dependant / spatial heterogeneity**