



# The MACSUR grassland model inter-comparison with the model CARAIB

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# Framework

MACSUR task L2.4 Grassland model intercomparison :

- Phase 1 : Blind runs
- Phase 2 : Calibrated runs
- Phase 3 : Climate sensitivity analysis
- 9 models involved :
  - Grassland-specific models: AnnuGrow, PaSim, SPACSYS
  - Crop models: EPIC, STICS, ARMOSA
  - Biome models: Biome-BGC, CARAIB, LPJmL

→ Ma et al. IEMSS conference, 2014 (Phase 1)

→ Next talk today of Renáta Sándor

**The grassland model intercomparison of the MACSUR (Modelling European Agriculture with Climate Change for Food Security) European knowledge hub**

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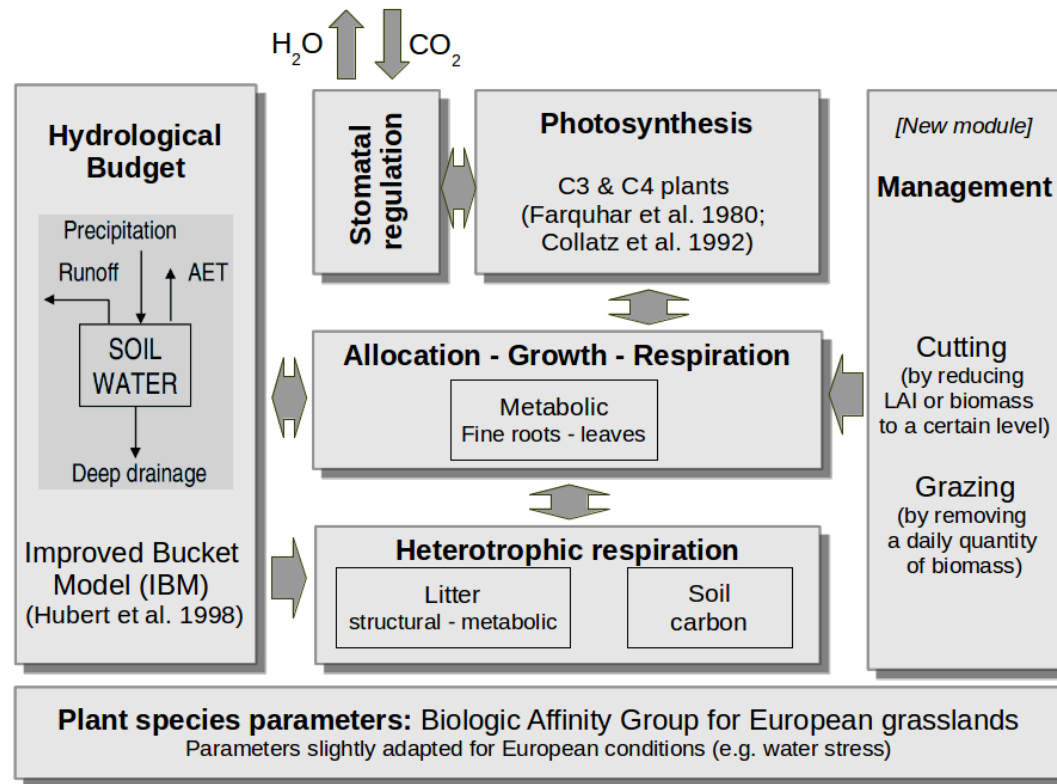
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# The grassland model : CARAIB

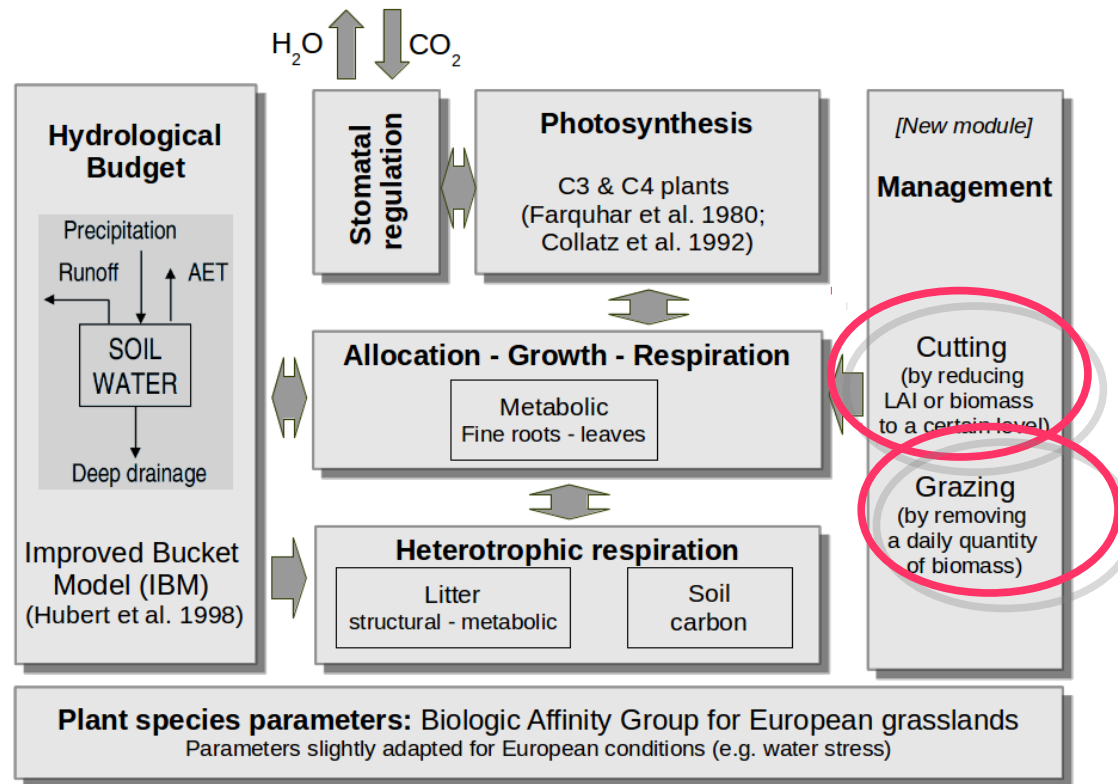
- A physically-based, mechanistic, dynamic vegetation model
- Focused on grassland
- New management functions for grassland: cut & grazing



Reference website: [http://www.umccb.ulg.ac.be/Sci/m\\_car\\_e.html](http://www.umccb.ulg.ac.be/Sci/m_car_e.html)

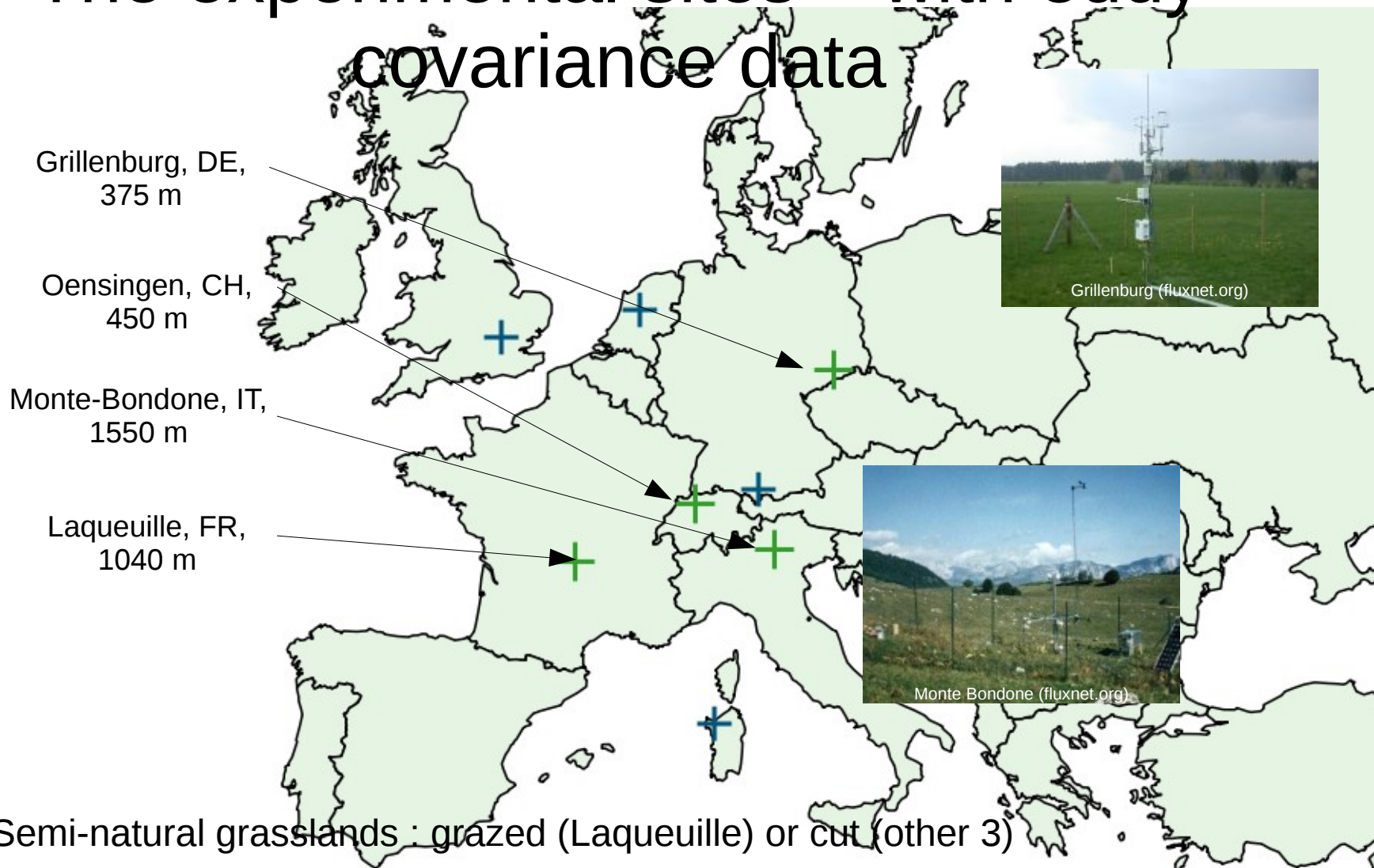
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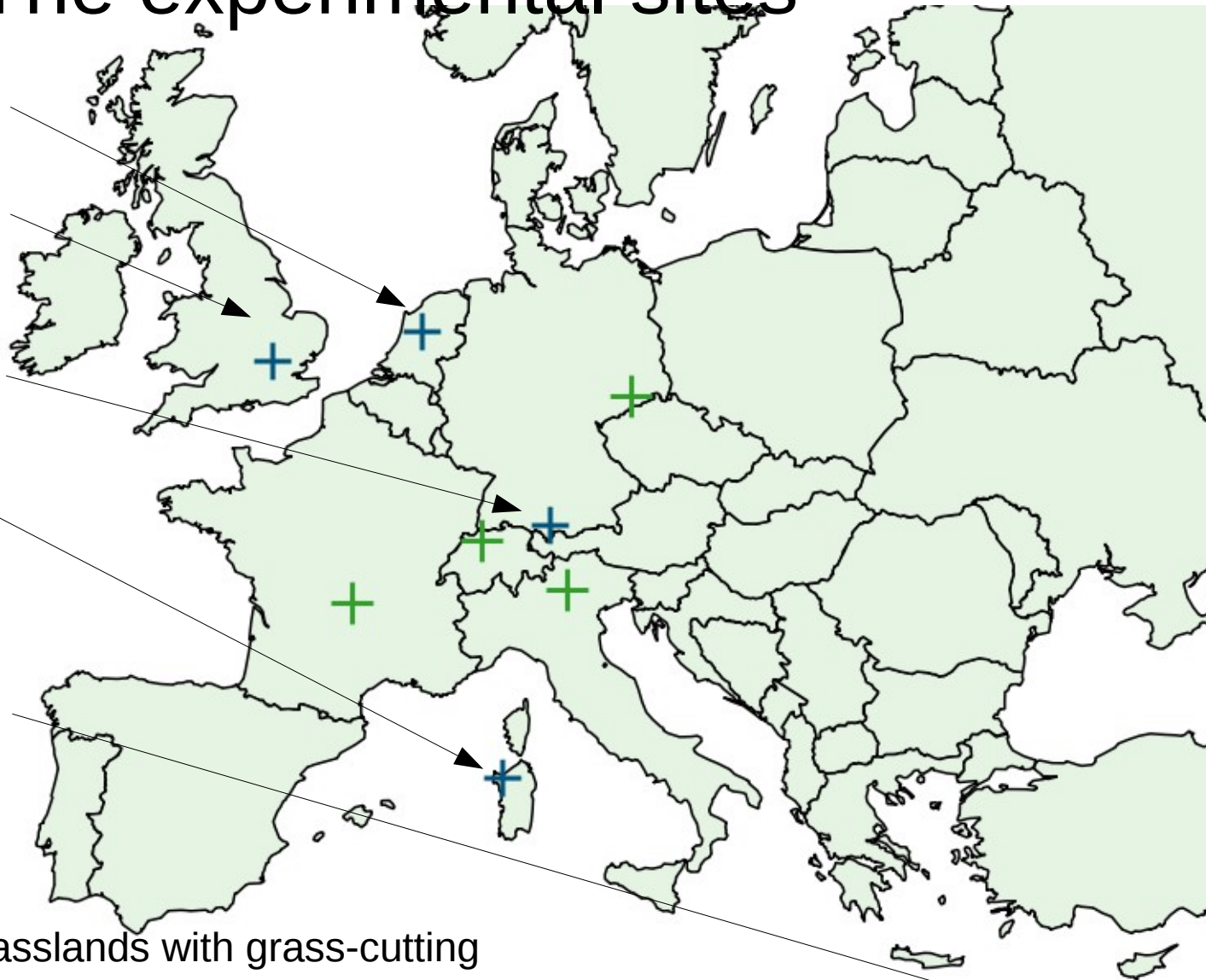
# The experimental sites – with eddy covariance data



- Semi-natural grasslands : grazed (Laqueuille) or cut (other 3)
- 4 Eddy-covariance sites : flux measurements available : GPP, RECO, ET, ...

# The experimental sites

Lelystad, NL  
Rothamsted, UK  
Kempton, DE  
Sassari, IT  
Matta, IS

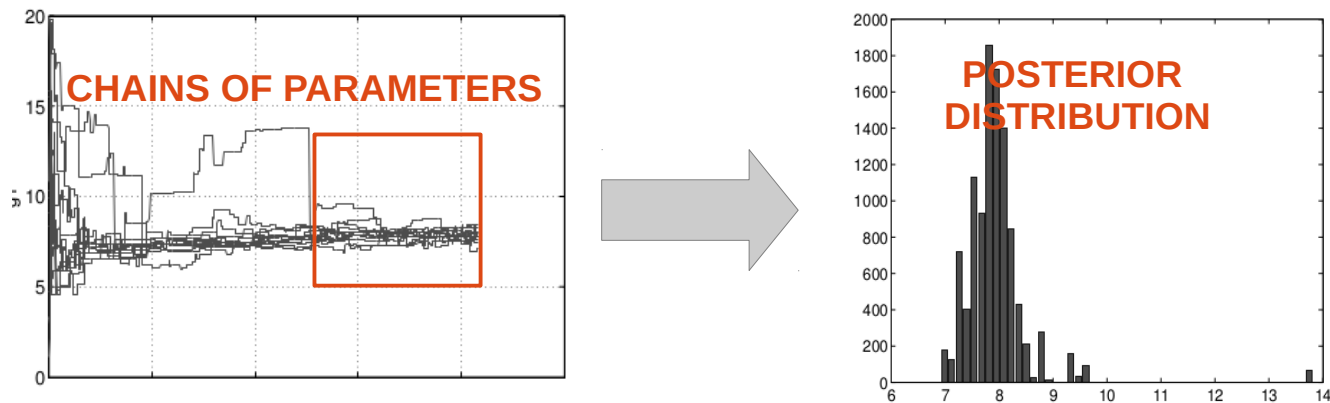


- Semi-natural grasslands with grass-cutting



# 2nd phase : Model calibration

- Inverse problem:  $Optimal\ parameters = argmin(Observations - Modeled(parameters))$
- DREAM\_ZS: a Markov-Chain Monte-Carlo sampler (Bayesian method)
- 10 parameters were sampled using 3 measurements variables from Eddy covariance: GPP, RECO, ET + their uncertainties ( $U_{GPP}$ ,  $U_{RECO}$ ,  $U_{ET}$ )
- A multi-objective cost function (CF) was used :  $CF = f(GPP, U_{GPP}, RECO, U_{RECO}, ET, U_{ET})$



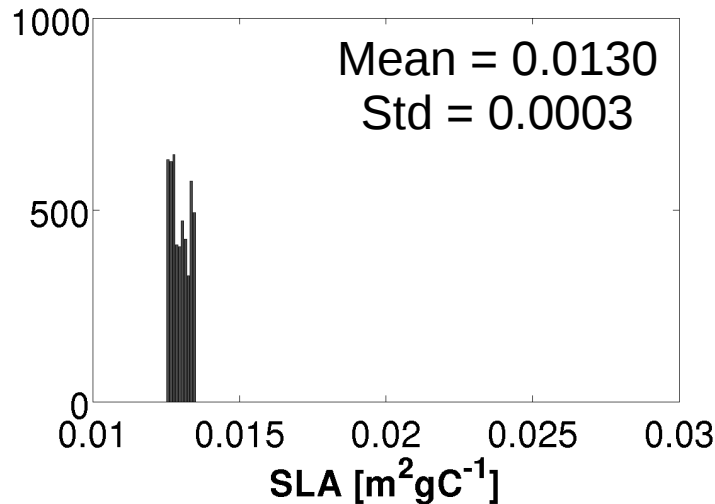
Laloy, E., and J.A. Vrugt, High-dimensional posterior exploration of hydrologic models using multiple-try DREAM\_(ZS) and high-performance computing, *Water Resources Research*, 48, W01526, 2012

Vrugt, J.A., C.J.F. ter Braak, C.G.H. Diks, D. Higdon, B.A. Robinson, and J.M. Hyman, Accelerating Markov chain Monte Carlo simulation by differential evolution with self-adaptive randomized subspace sampling, *International Journal of Nonlinear Sciences and Numerical Simulation*, 10(3), 273-290, 2009.

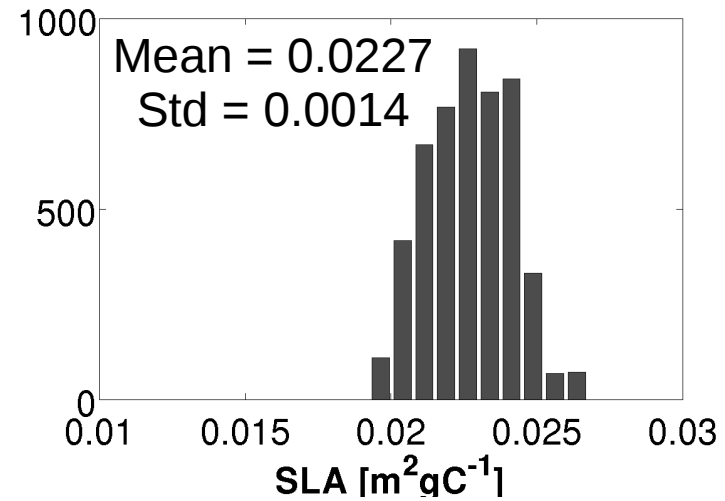
# Results 2nd phase: parameter samplings

Specific leaf area (SLA) [ $\text{m}^2/\text{gC}$ ]

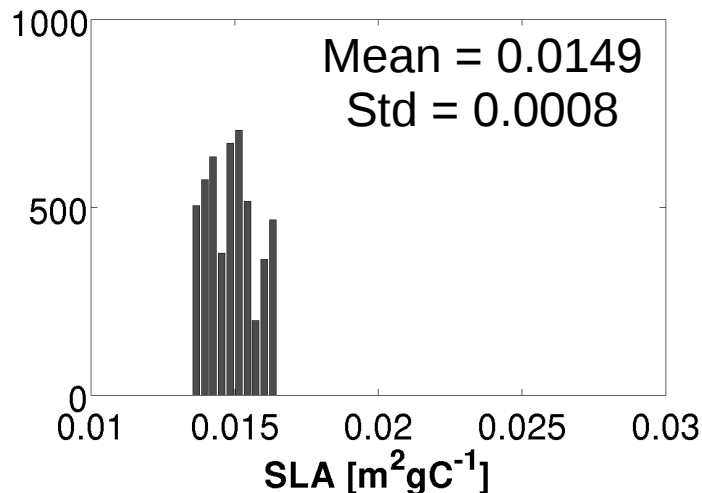
Grillenburg



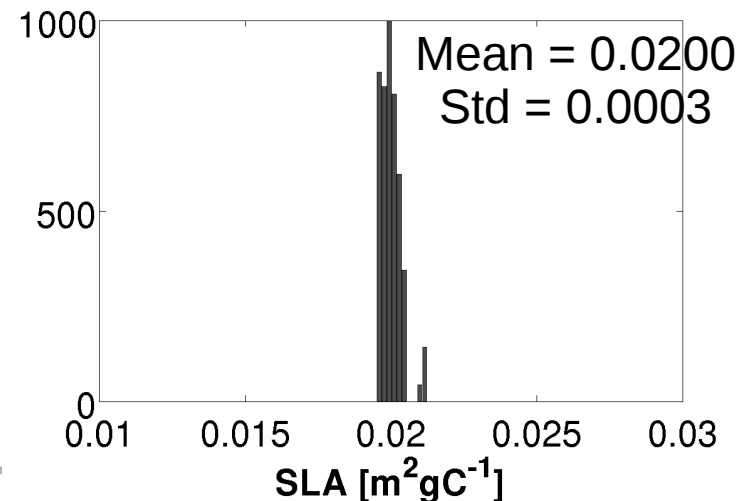
Oensingen



Monte-Bondone



Laqueuille

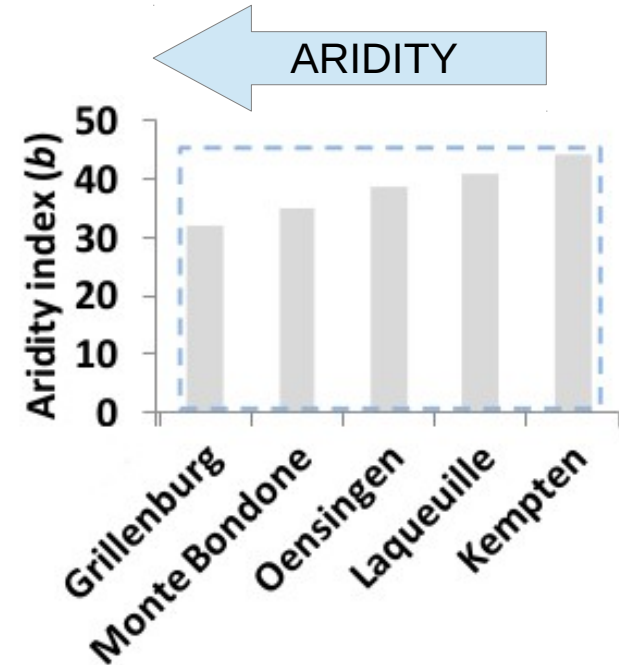
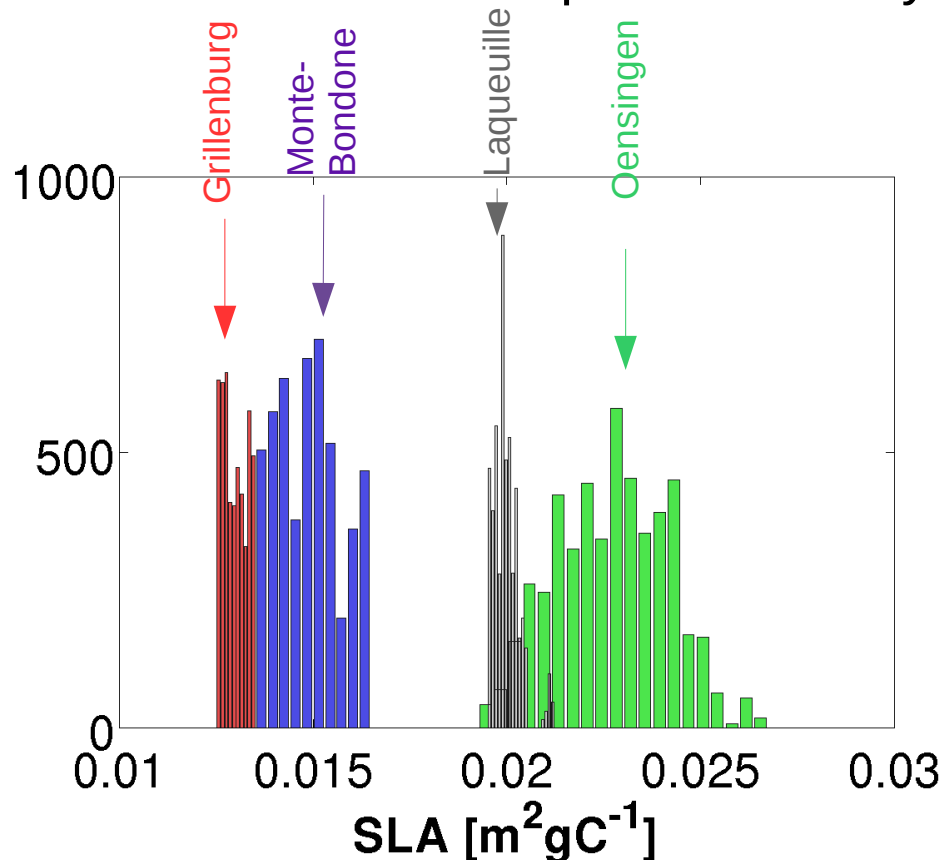




# Results 2nd phase: parameter samplings

Specific leaf area (SLA) [ $\text{m}^2/\text{gC}$ ]

- SLA in CARAIB : *effective* SLA for a plant functional type !
- Actually, SLA is variable between leaves and along the season
- SLA is known to depend on aridity (-) and intensification (+)

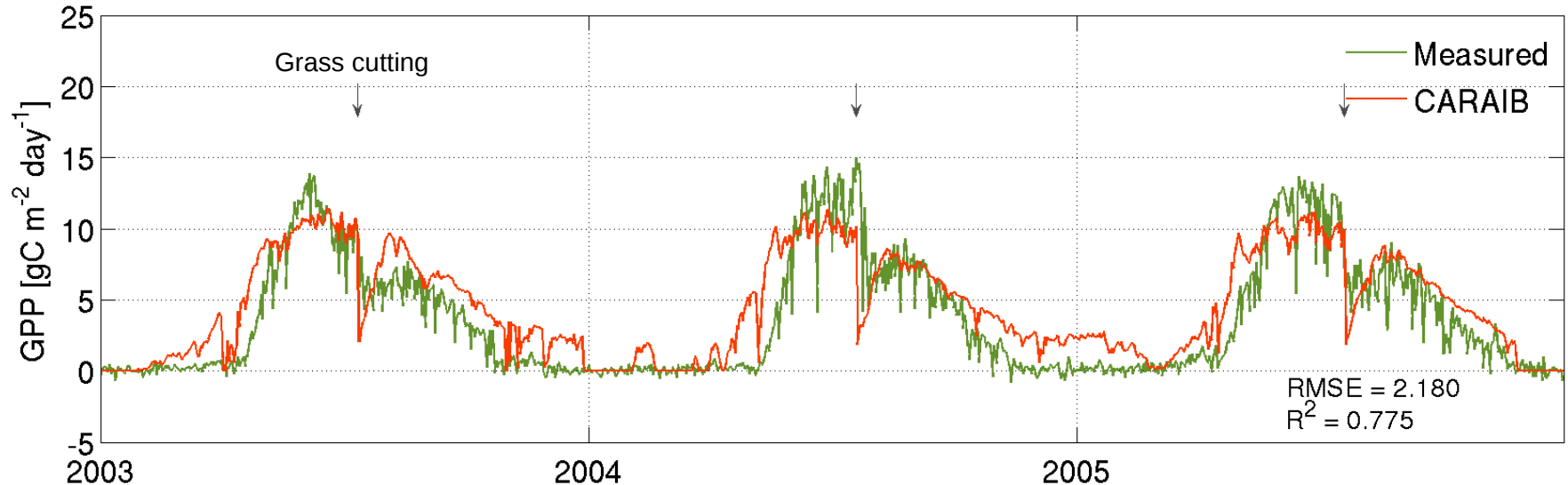


De Martonne-Gottmann aridity index  
from Ma et al. iEMSS, 2014

# Results 1st phase – blind runs

## Measured vs modelled Gross Primary Productivity (GPP)

Measured VS modelled gross primary productivity (GPP), blind run, Monte-Bondone

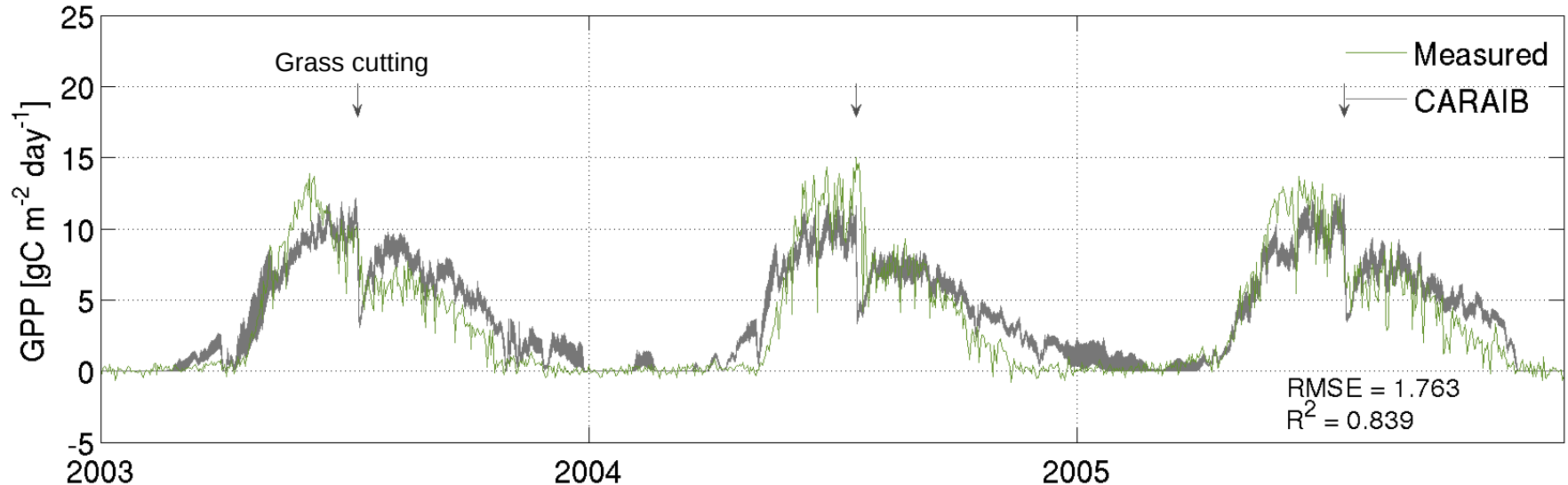


- Blind run without site-specific model calibration.
- Monte-Bondone : 1 grass cutting /year.

# Results 2nd phase – calibrated runs

## Measured vs modelled GPP

Measured VS modeled gross primary productivity (GPP), after calibration, Monte-Bondone



- Model runs after calibration using a Bayesian algorithm : 1000's of modelled signals → model sensitivity.
- Better agreement...

# 3rd phase - Climate sensitivity

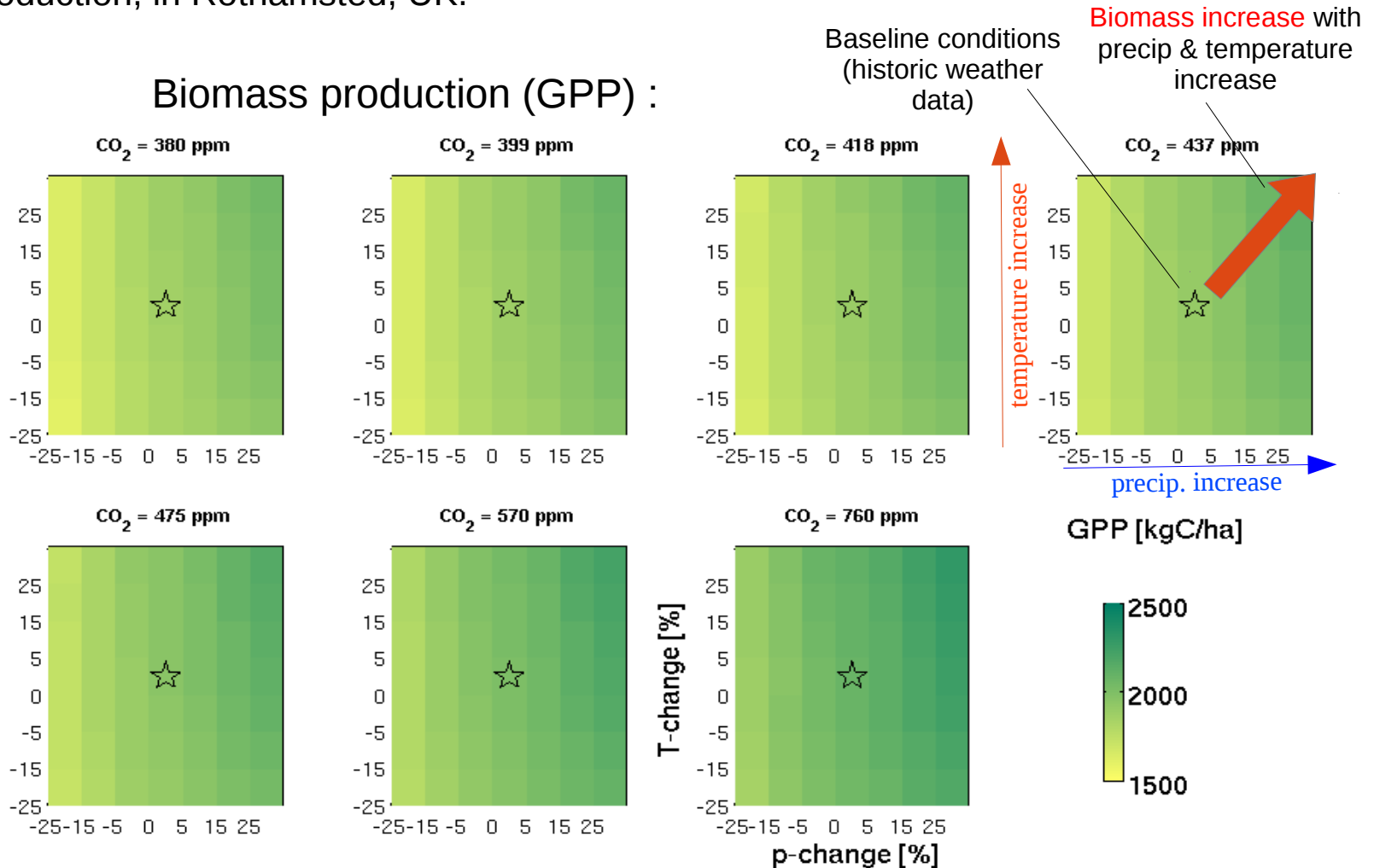
## Methodology

- Analyse sensitivity of model to temperature, precipitation and [CO<sub>2</sub>] change
- For precipitation and temperature, altered values are  $y = y \pm x\% \cdot SD$  where percent changes are -25%, -15%, -5%, +5%, +15% & +25% and SD the standard deviation of temperature and precipitation.
- For [CO<sub>2</sub>], altered values are the baseline (380 ppm) + 5%, 10%, 15%, 25%, 50% & 100%.
- 9 sites, 4 to 30 years,  $6 \times 6 \times 6 = 216$  scenarios

# Results 3rd phase : Climate sensitivity

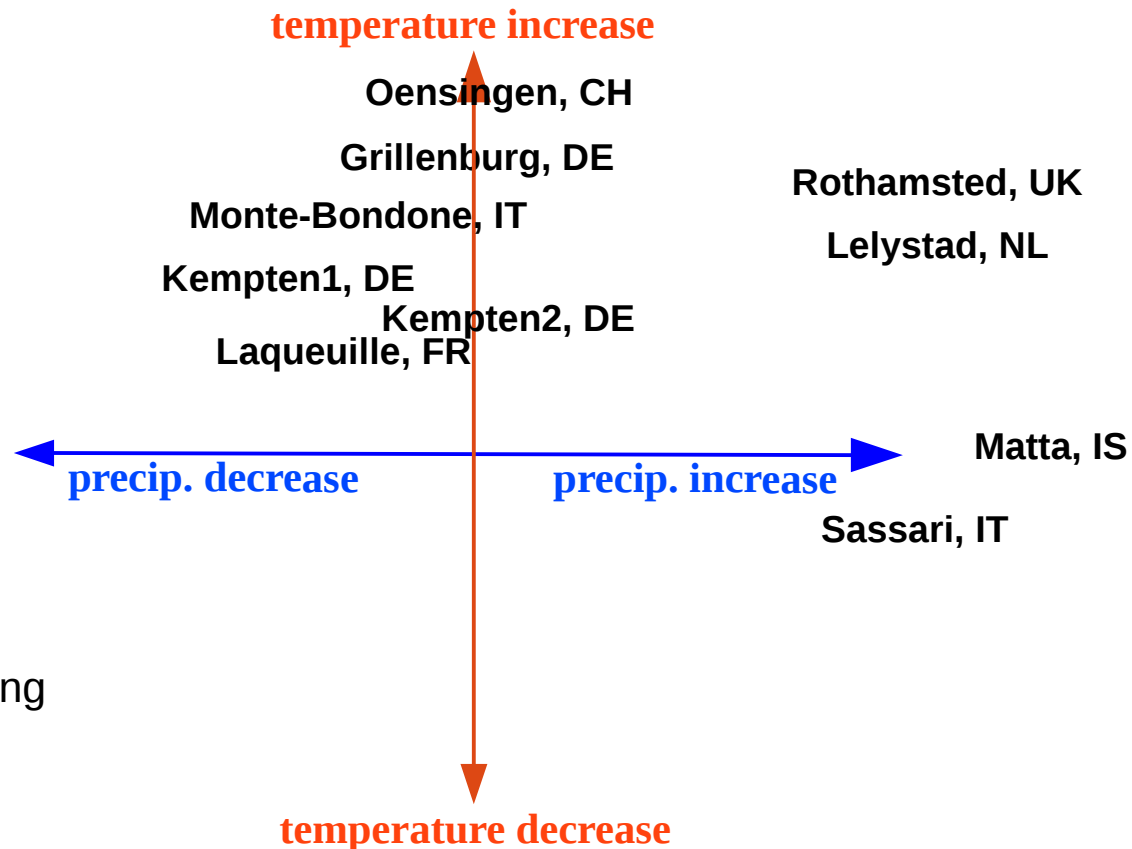
Mean (over years) impact of temperature, precipitation and [CO<sub>2</sub>] change on grassland production, in Rothamsted, UK.

## Biomass production (GPP) :



# Results 3rd phase : Climate sensitivity

Sites can be classified according to their climatic sensitivity for biomass increase (after Pirttioja et al. 2014, MACSUR CropM WP4) :



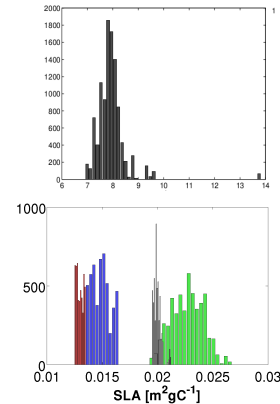
Note : All sites show an increase in biomass production with increasing CO<sub>2</sub>.

Not only sites but also grasslands models can be classified...

# Conclusion

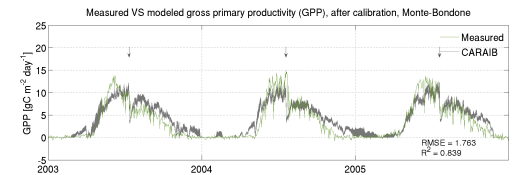
2nd phase : Bayesian sampling with DREAM\_ZS :

- Obtain a uncertainty assessment on model parameters
- Compare distributions of parameters between the sites
- Obtain an interval on model output due to parameters uncertainties



3rd phase : Climate sensitivity

- Huge amount of results
- Grasslands sites can be compared and classified
- Still on-going work...





# Thanks for your attention