

# Assimilating EO Data into Terrestrial Carbon Cycle Models

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# The current challenge in C cycle research

## Objective

To produce estimates & predictions of ecosystem carbon exchange with quantifiable uncertainty.

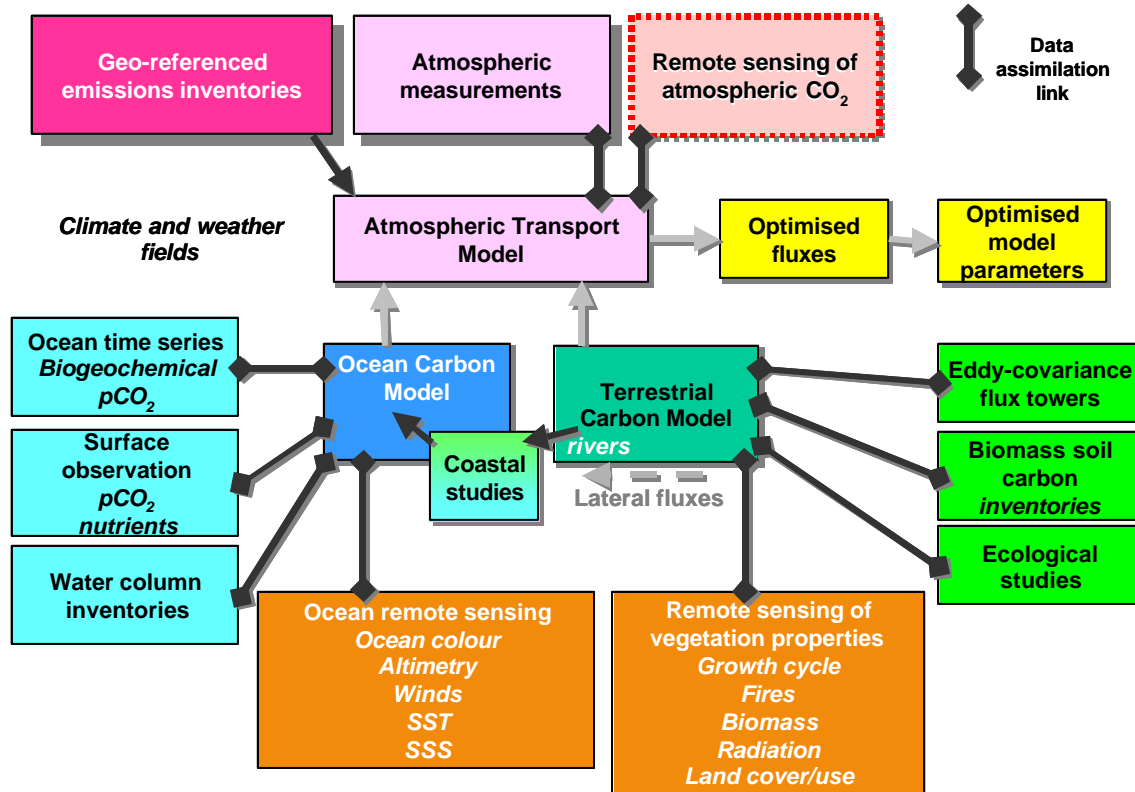
## Complications

Observations have gaps & instrumental weaknesses.  
Models tend to oversimplify and may miss key processes and linkages.

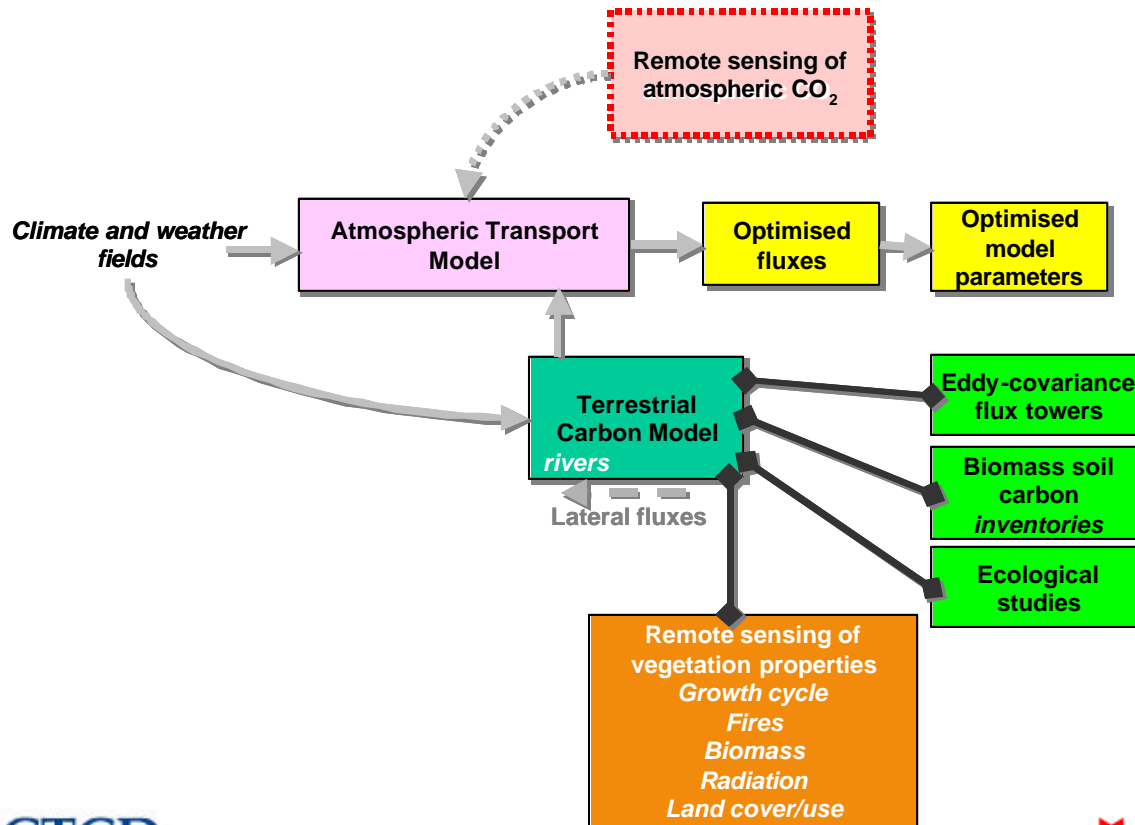
## Solution

Data assimilation provides a method to combine models and data to produce a more accurate description of ecosystem dynamics.

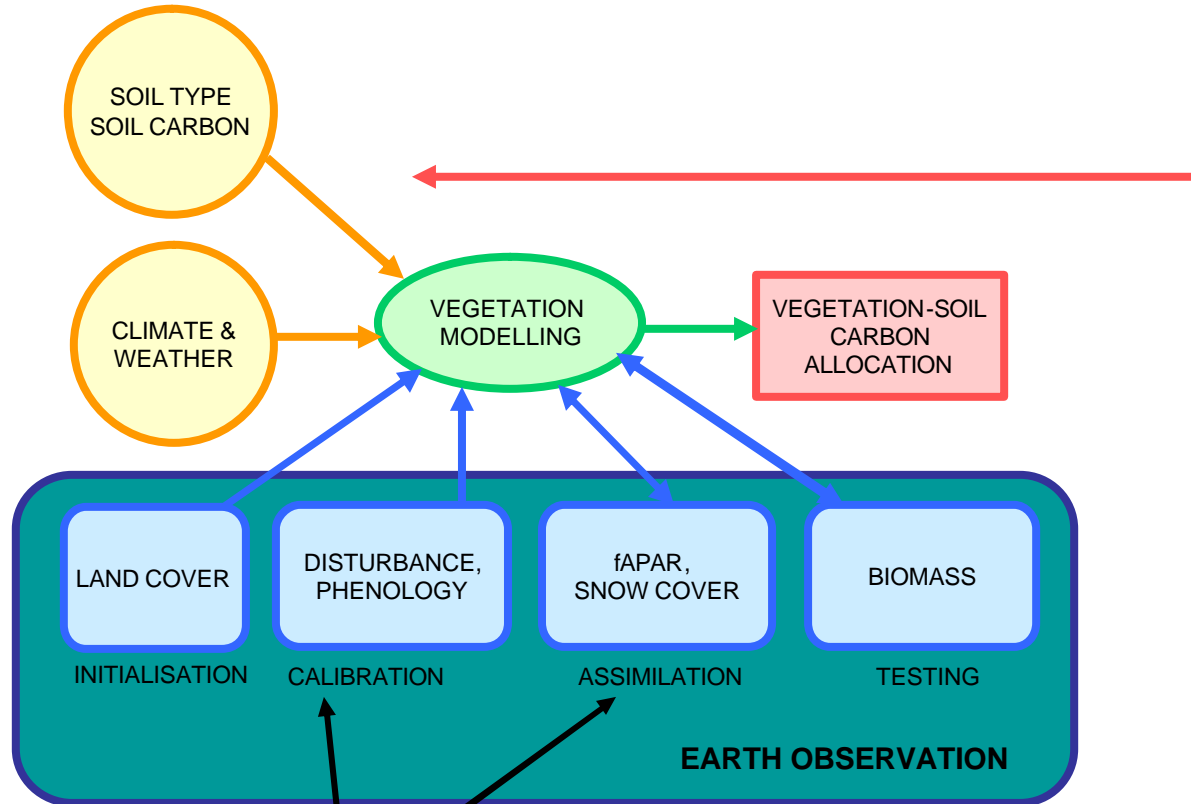
# Global Carbon Data Assimilation System



# Terrestrial component



# Carbon Cycle – Earth Observation Interfaces



# Issues

## ◆ Monitoring

Consistency of models and data:

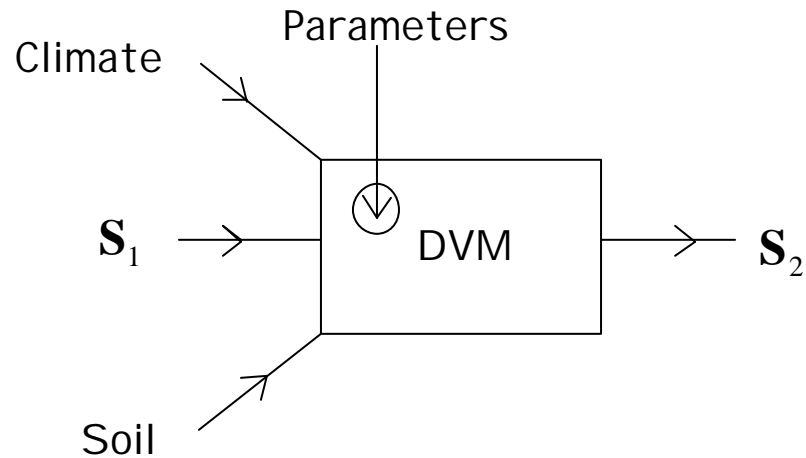
- are model and measurement quantities compatible?
- comparability of values

Timescales (re-analysis)

## ◆ Prediction

Are model internal processes and parameters testable and credible?

# The Functioning of a DVM



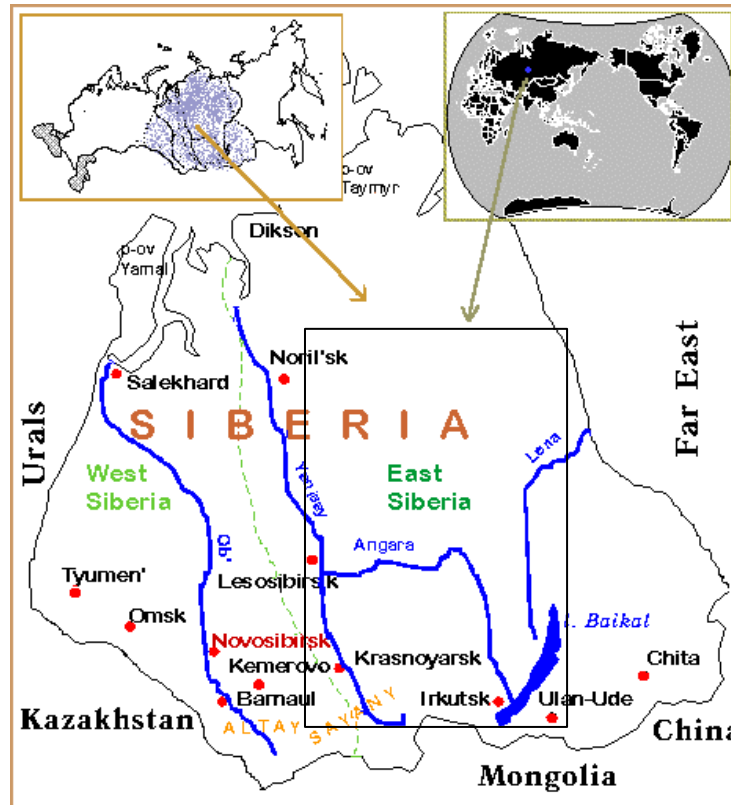
$S$  is the state vector describing the vegetation-soil system.

# Calibrating the SDGVM phenology module with EO data

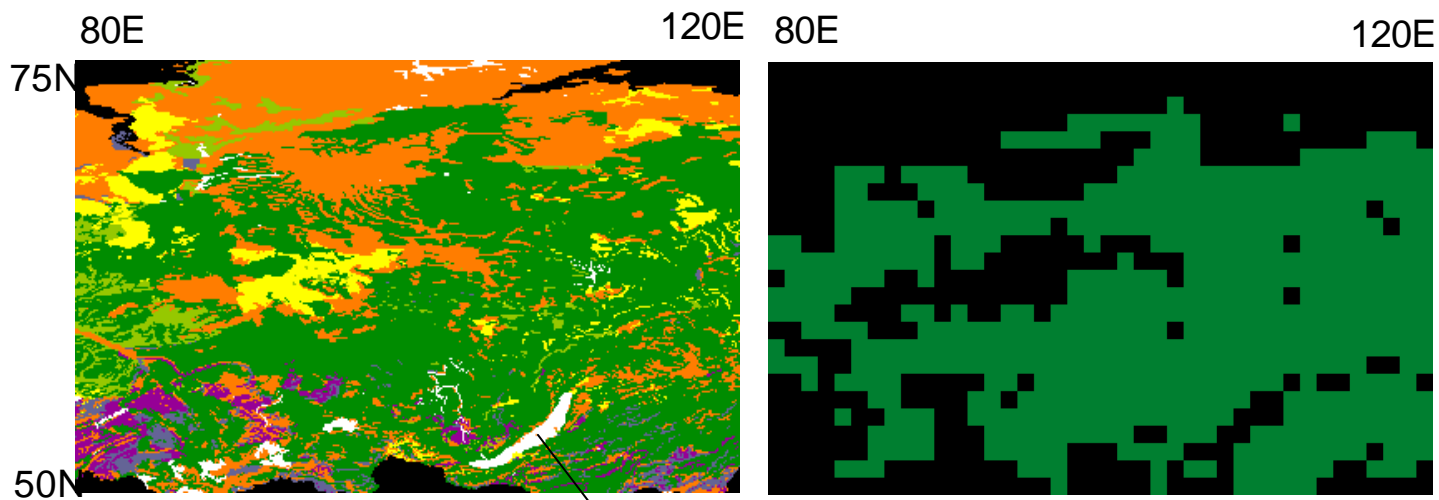


# SIBERIA-II: Multi-Sensor Concepts for Greenhouse Gas Accounting of Northern Eurasia

5th Framework Project , 2002-2005



# The Central Siberia dataset: ~ 2 M km<sup>2</sup>



Land cover (IIASA)

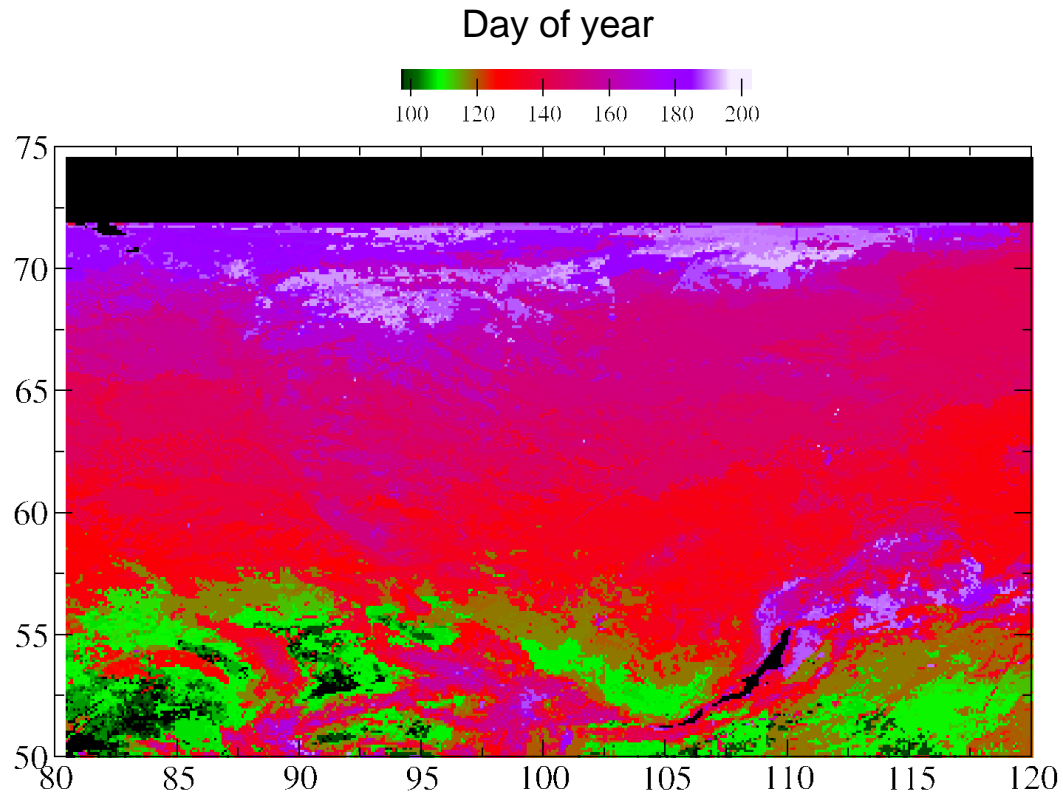
1° x 1° forest map

Lake Baikal

# The CESBIO budburst algorithm

- ◆ Data set: SPOT-VEG 1999-2001
- ◆ Based on minimum in time-series of NDWI data
- ◆ Uncertainties in recovered budburst date ~ 7 days

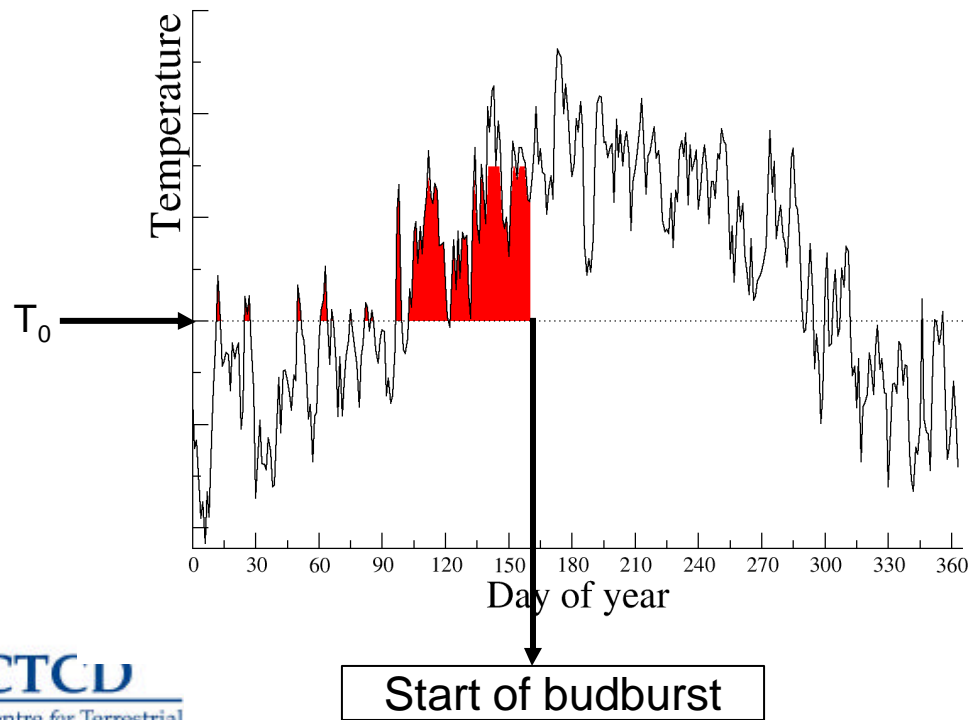
# The Date of budburst derived from minimum NDWI (VGT sensor, 2000) N. Delbart, CESBIO



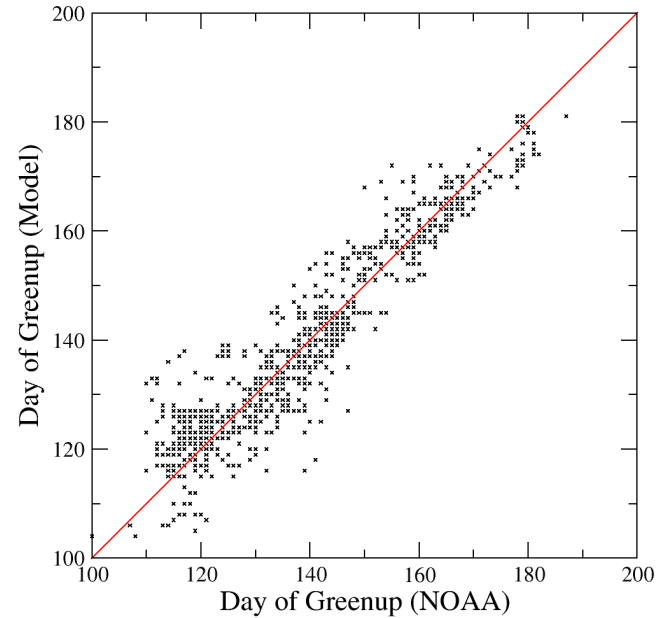
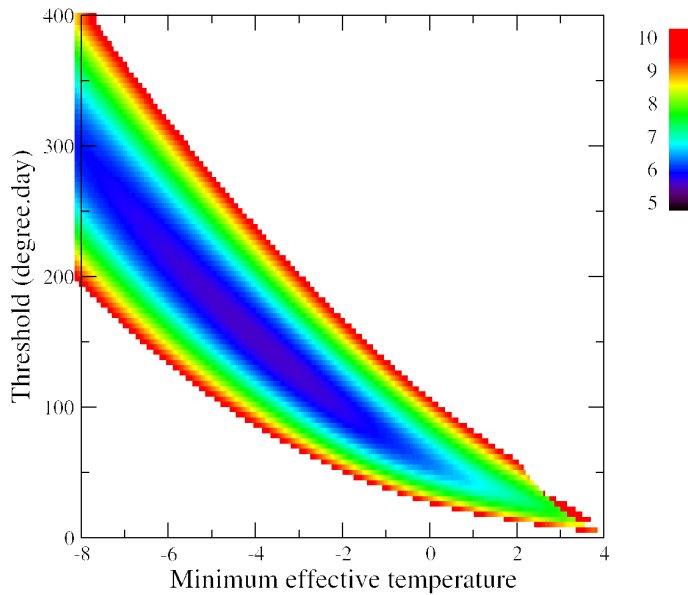
# The SDGVM budburst algorithm

When  $\sum_{\text{days}} \min(0, T - T_0) > \text{Threshold}$ , budburst occurs.

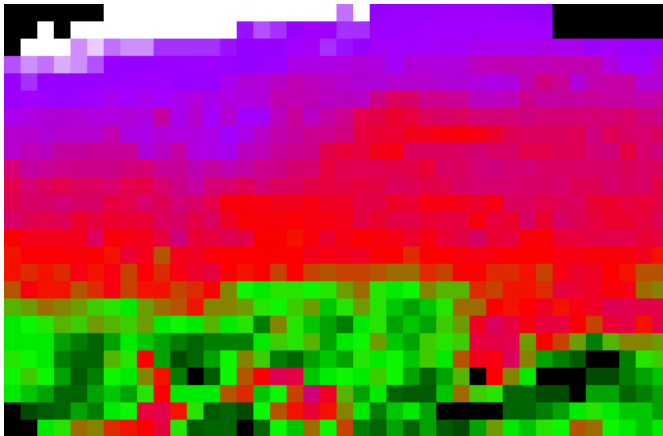
The sum is the **red** area. Optimise over the 2 parameters, Threshold and  $T_0$  (minimum effective temperature).



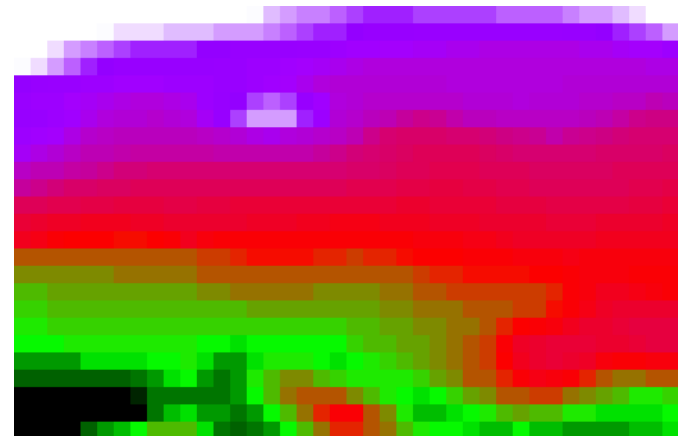
# The calibration procedure



# Data - model comparison 1999



Budburst from NDWI data



Model budburst:  
optimised parameters

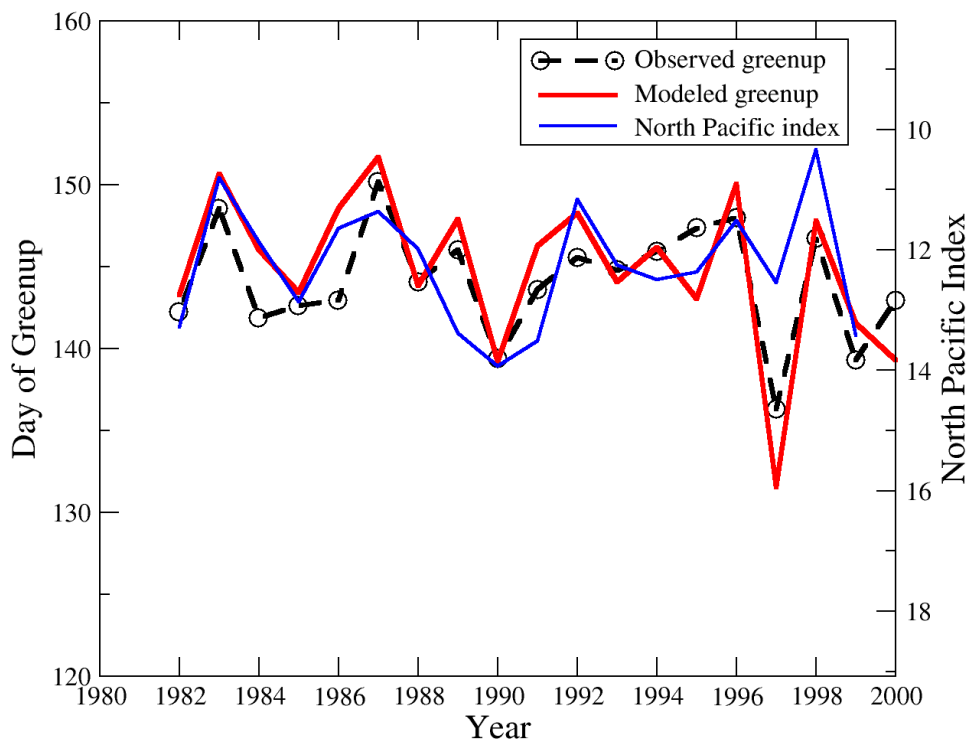
## Calibration parameters (forest only)

Year	$T_0$ (degrees)	Threshold (degree-days)	MMSE (days)
1999	-2.9	117	6.0
2000	4.4	29	7.0

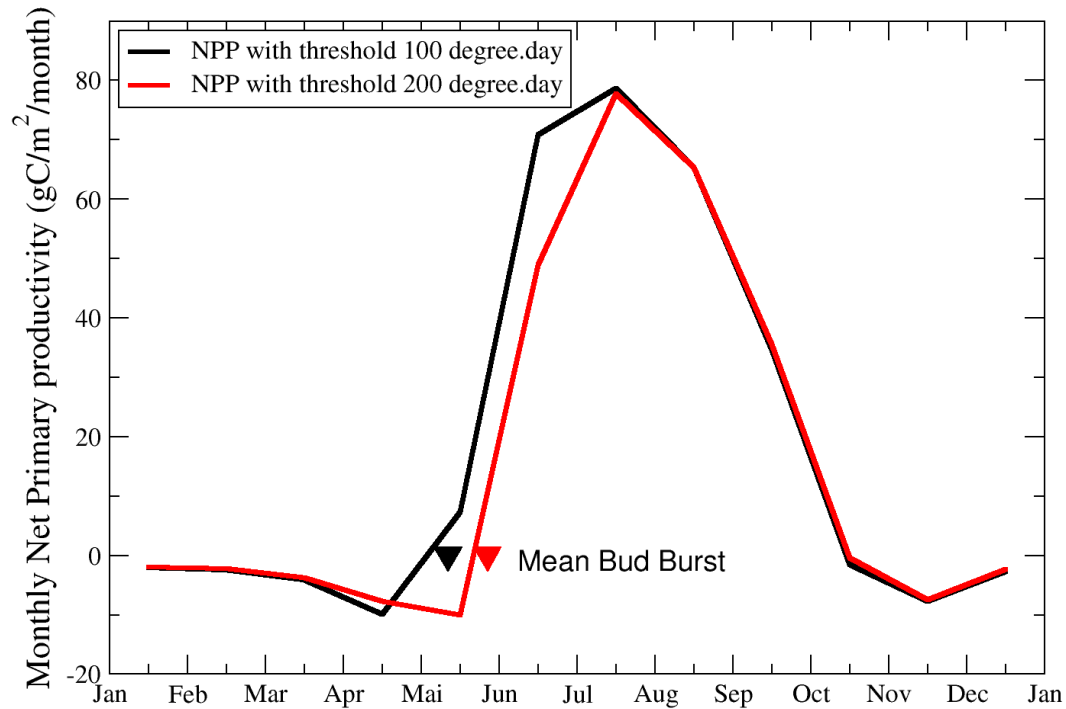


# Green-up relation to N Pacific Index

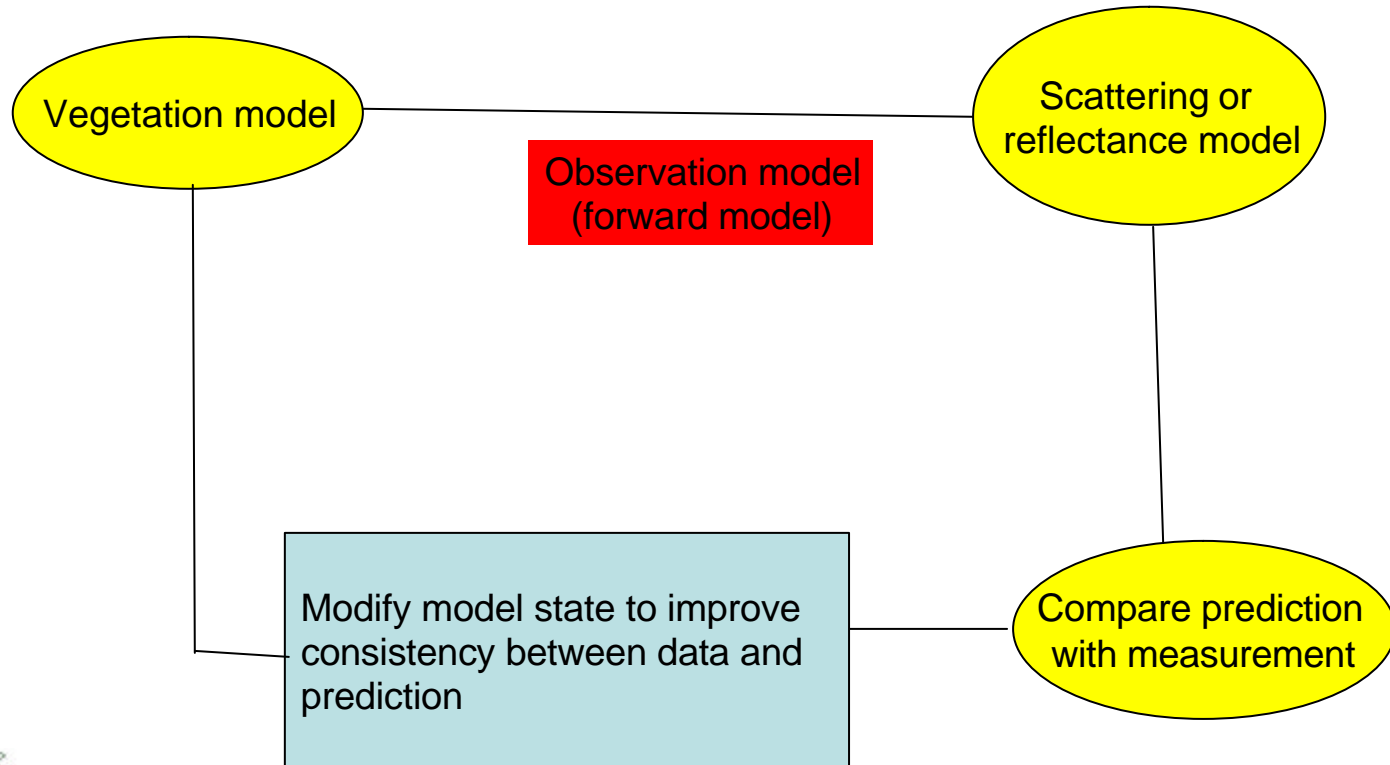
The NP Index is averaged from April the previous year to March the present year



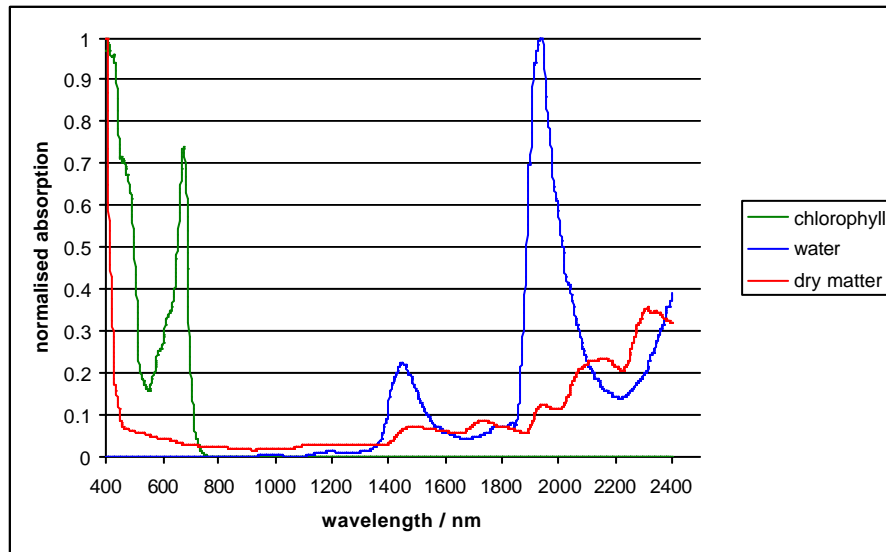
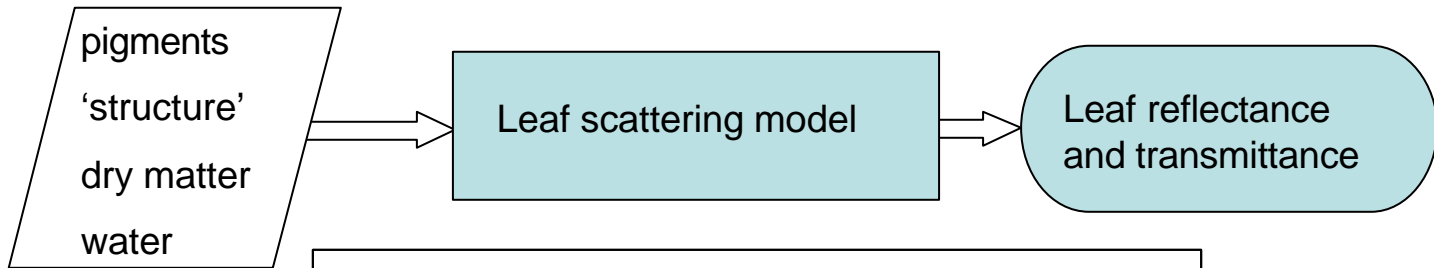
# Effect of uncertainty in green-up day



# 'True' Assimilation



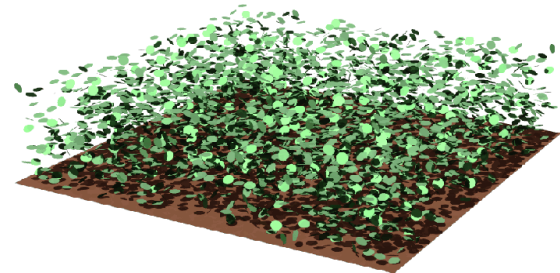
# Basis of radiation models (optical)



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## ◆ Model of canopy scattering:

- Leaf properties
- Scattering object density (LAI), orientation, and spatial distribution
- Soil / understorey properties for low density canopies



## ◆ solutions by analytical or numerical methods

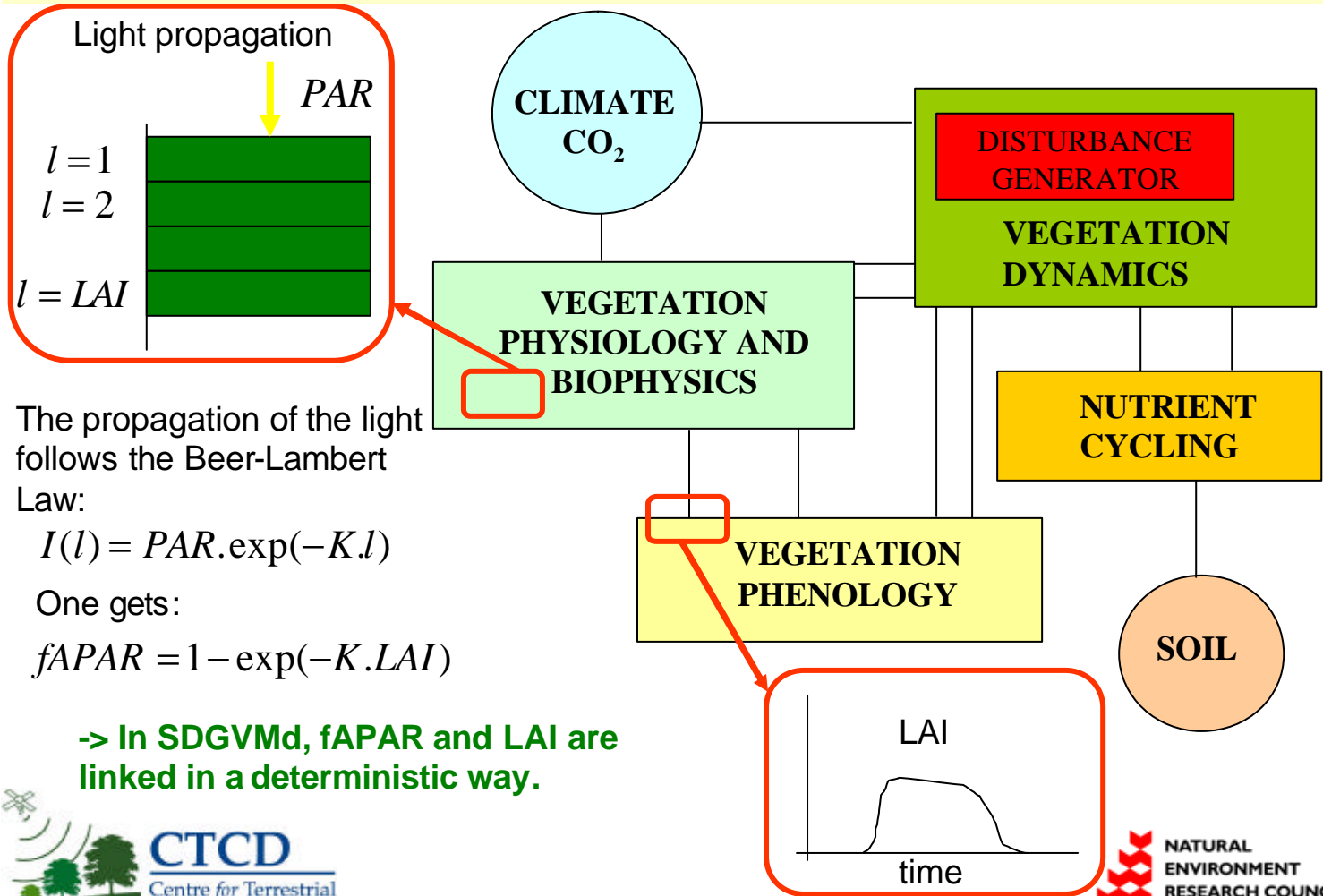
# Link to the C models

- ◆ **C models include concept of radiation model**
  - For calculation of intercepted radiation
- ◆ **Observation model**
  - Provides link from *subset* of C-model state variables to EO observation
  - Main linkages:
    - ◆ LAI, Density (for limited conditions)
    - ◆ leaf properties (hyperspectral data)
      - leaf dry matter, chlorophyll (nitrogen), water
      - xanthophyll cycle (light use efficiency)

# Exploiting quantities derived from radiance



# Sheffield Dynamic Vegetation Global Model



The propagation of the light follows the Beer-Lambert Law:

$$I(l) = PAR \cdot \exp(-K \cdot l)$$

One gets:

$$fAPAR = 1 - \exp(-K \cdot LAI)$$

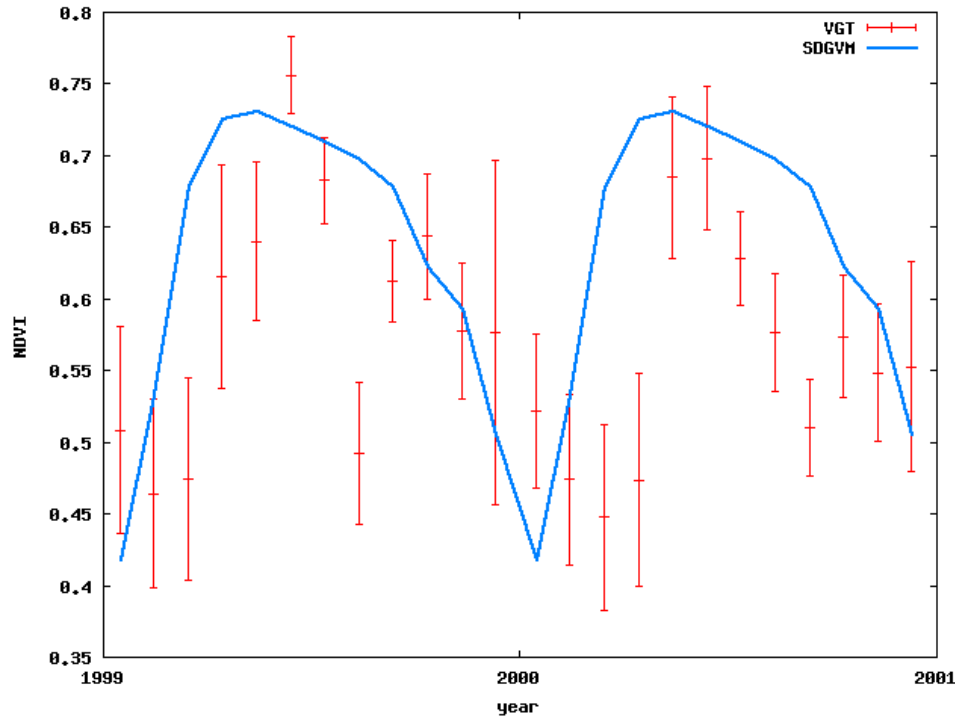
-> In SDGVMd, fAPAR and LAI are linked in a deterministic way.



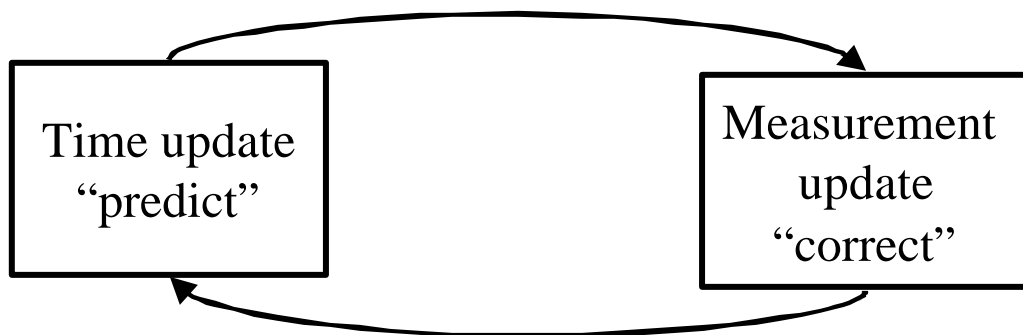
# Application of the models

## ◆ Testing C models (SDGVM)

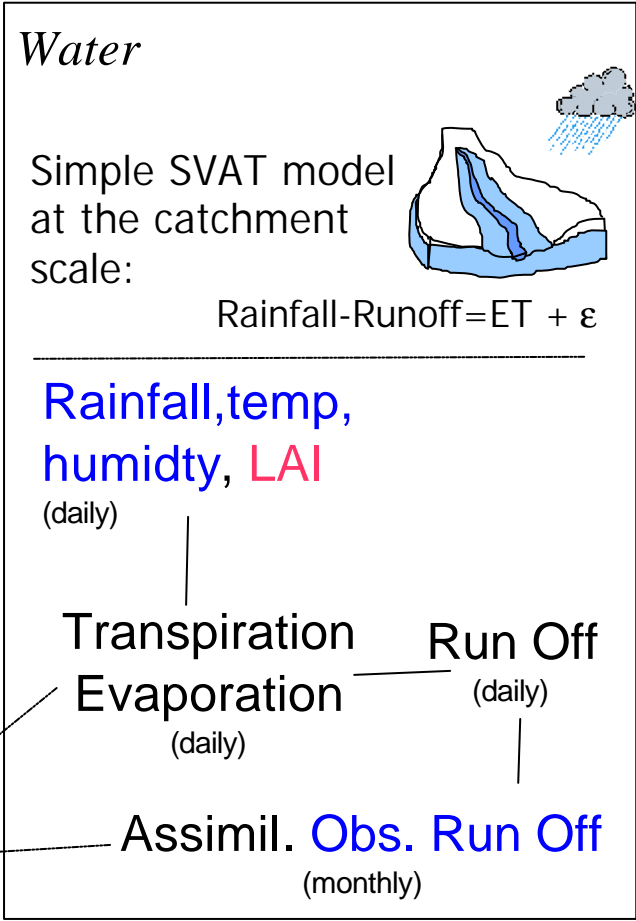
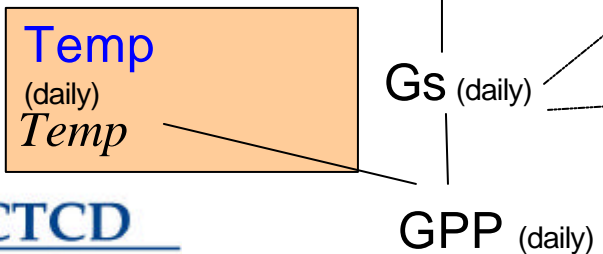
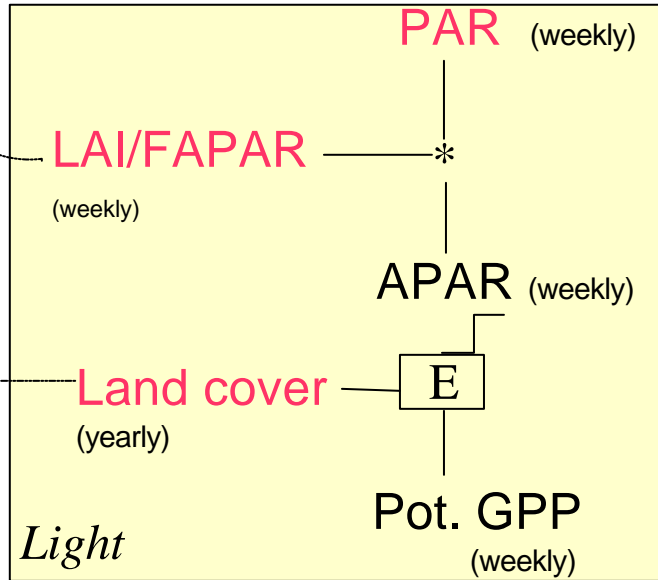
- Confront predictions with observations



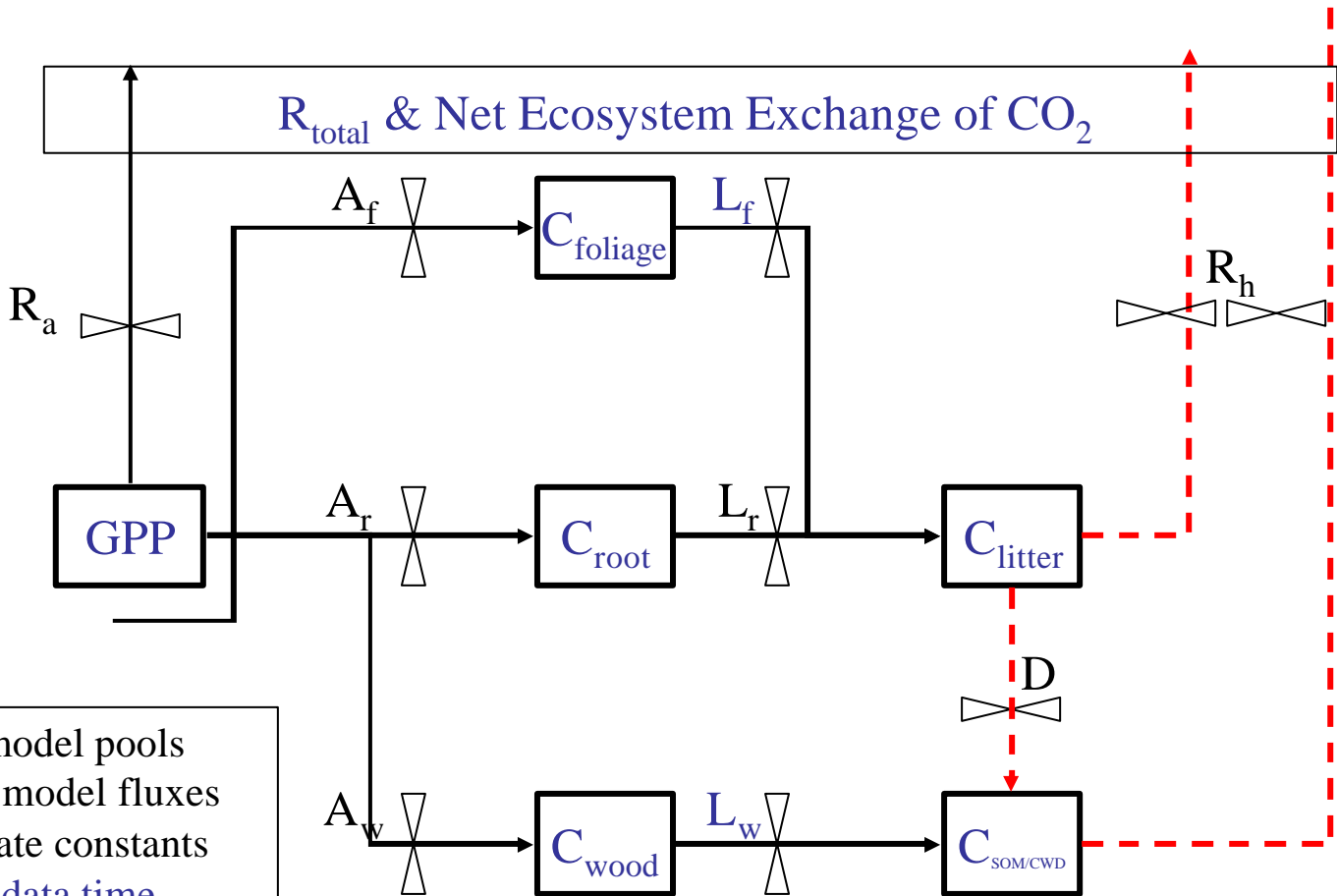
# A prediction-correction system



Earth Observation

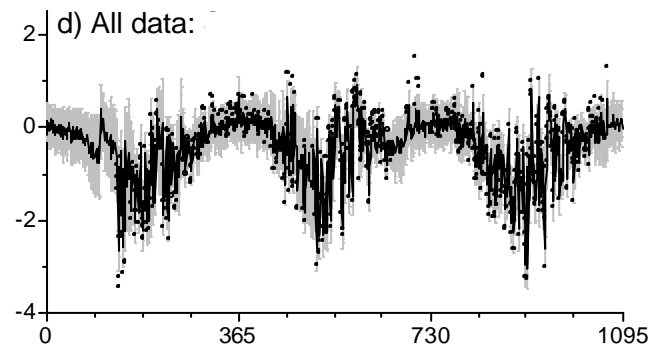
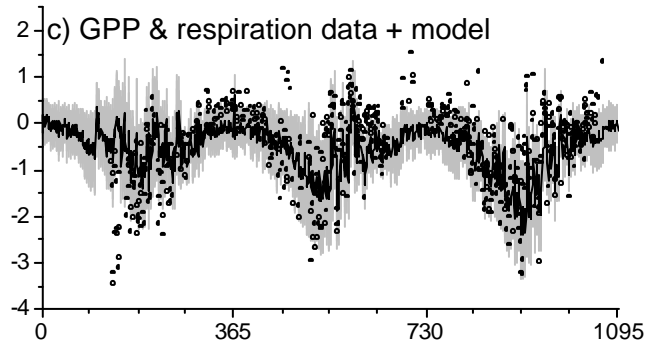
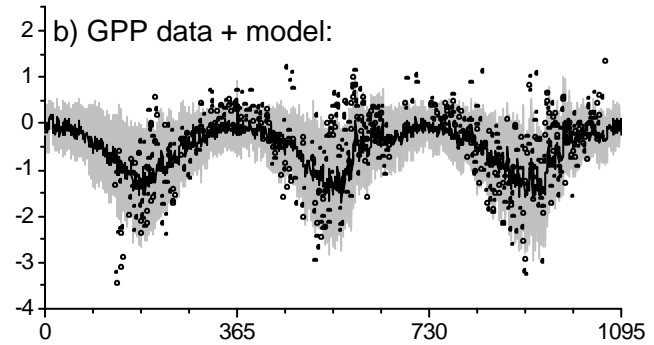
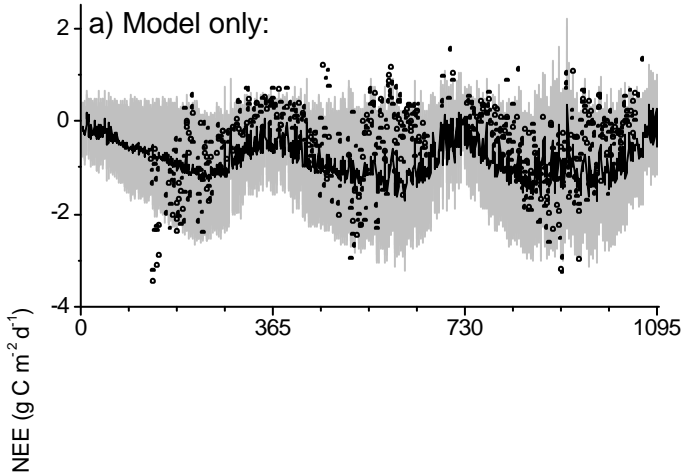




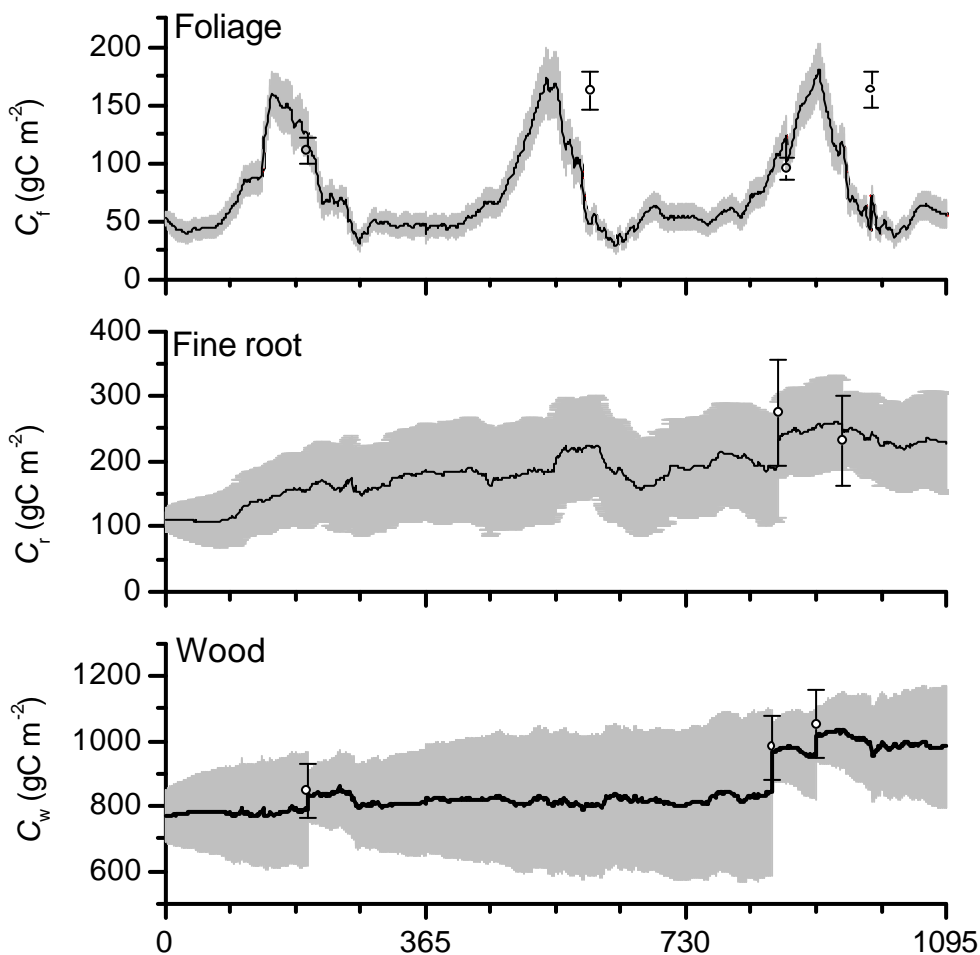


6 model pools  
 10 model fluxes  
 9 rate constants  
 10 data time series

C = carbon pools  
 A = allocation  
 L = litter fall  
 R = respiration (auto- & heterotrophic)



Time (days, 1= 1 Jan 2000)

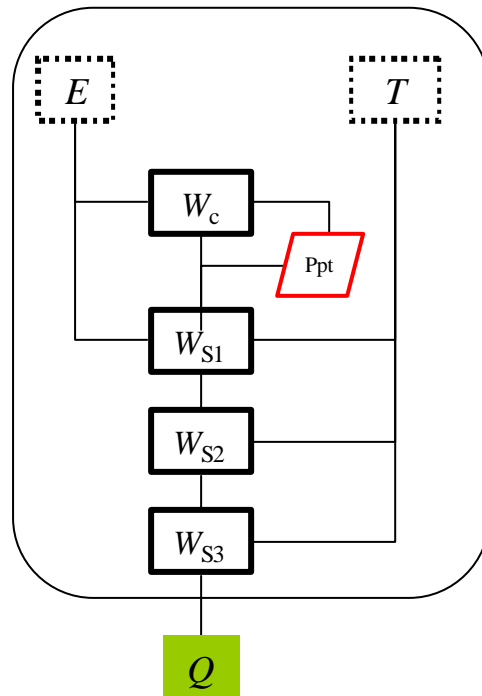
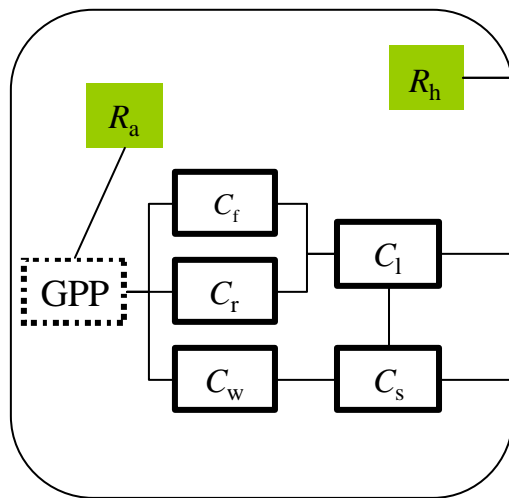


———— Flux

# Pools and fluxes

## Hydrology

## Carbon model

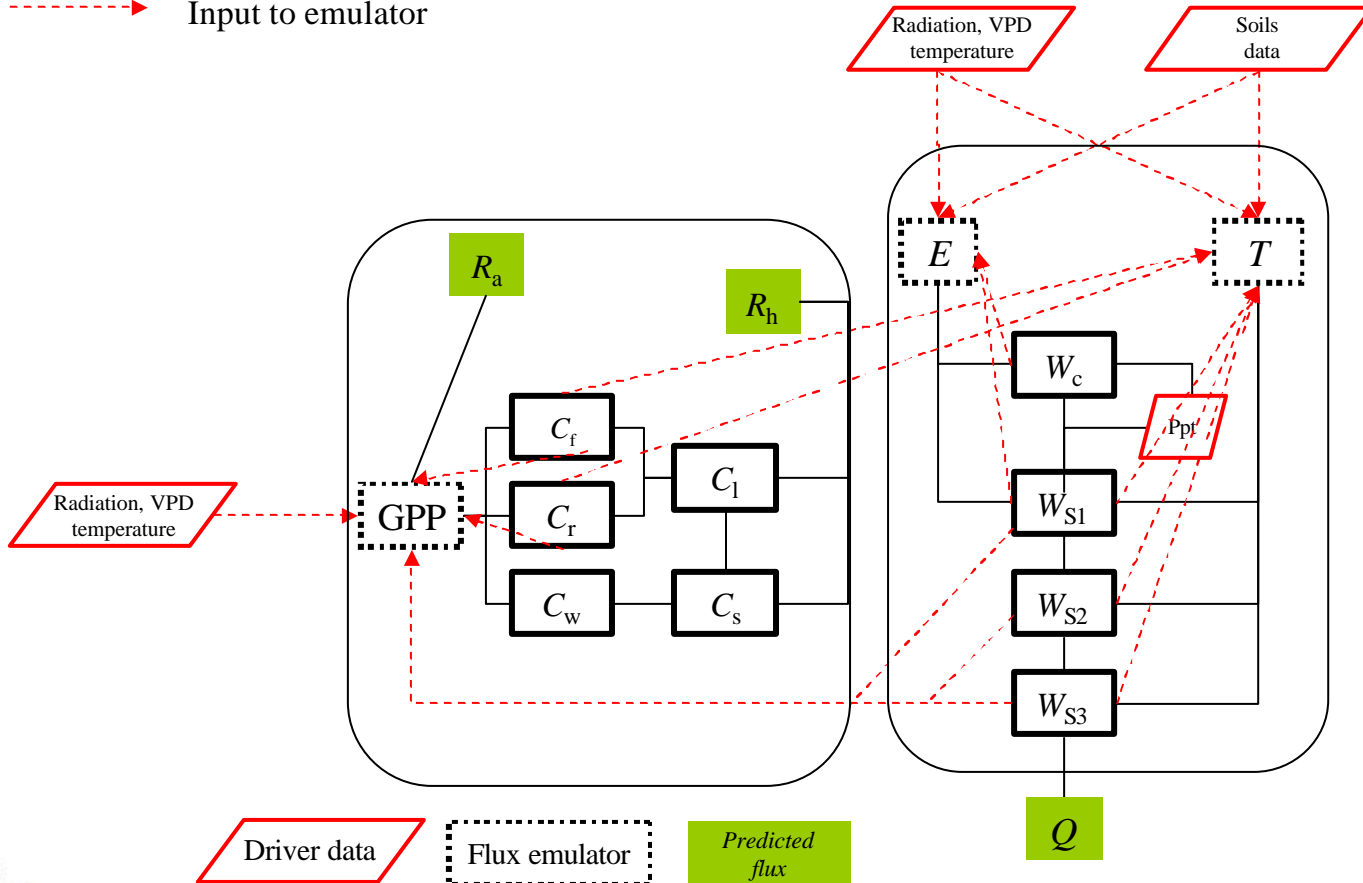




# Linking C and water fluxes

— Flux

- - - -> Input to emulator



Driver data

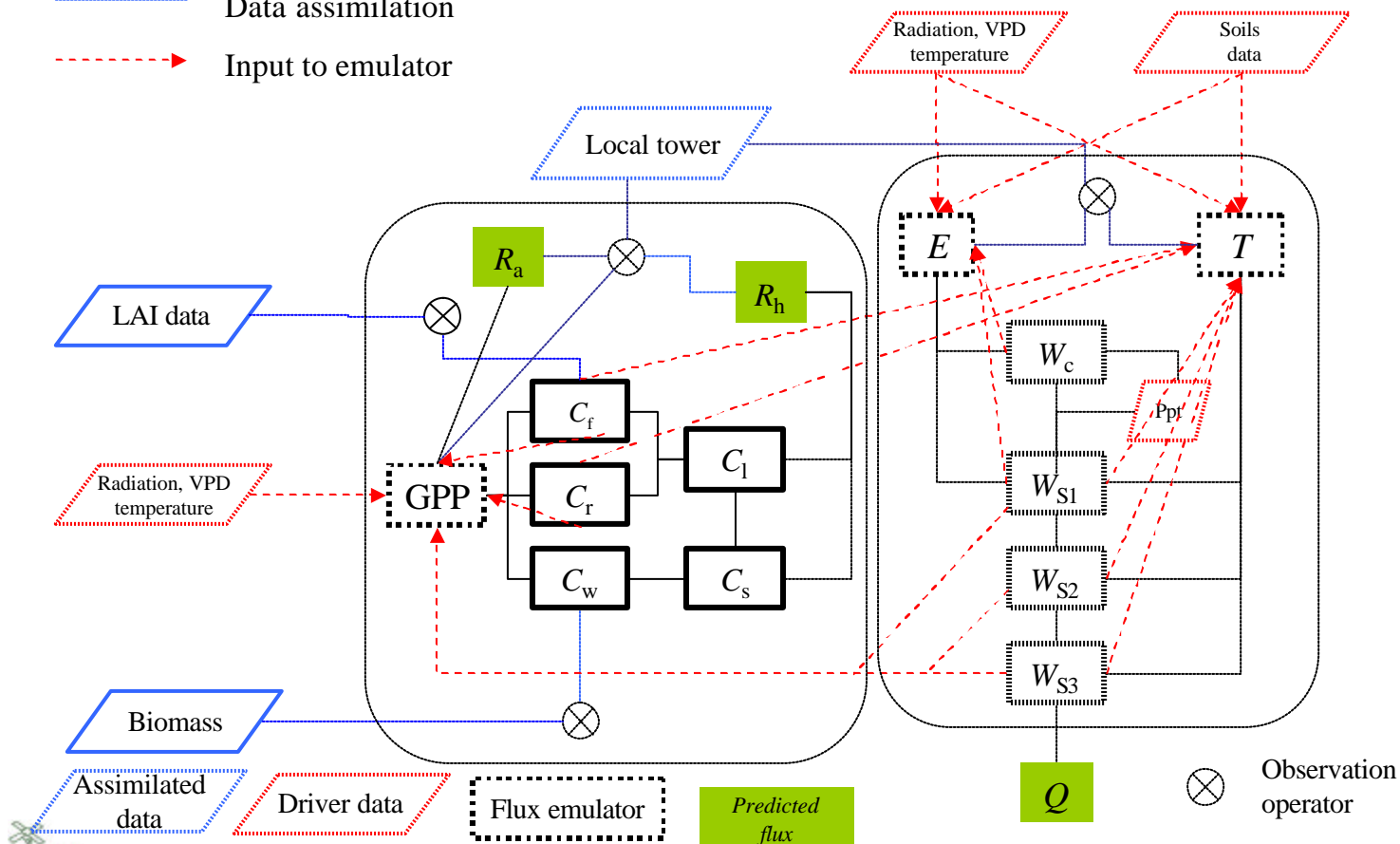
Flux emulator

Predicted flux

State variable

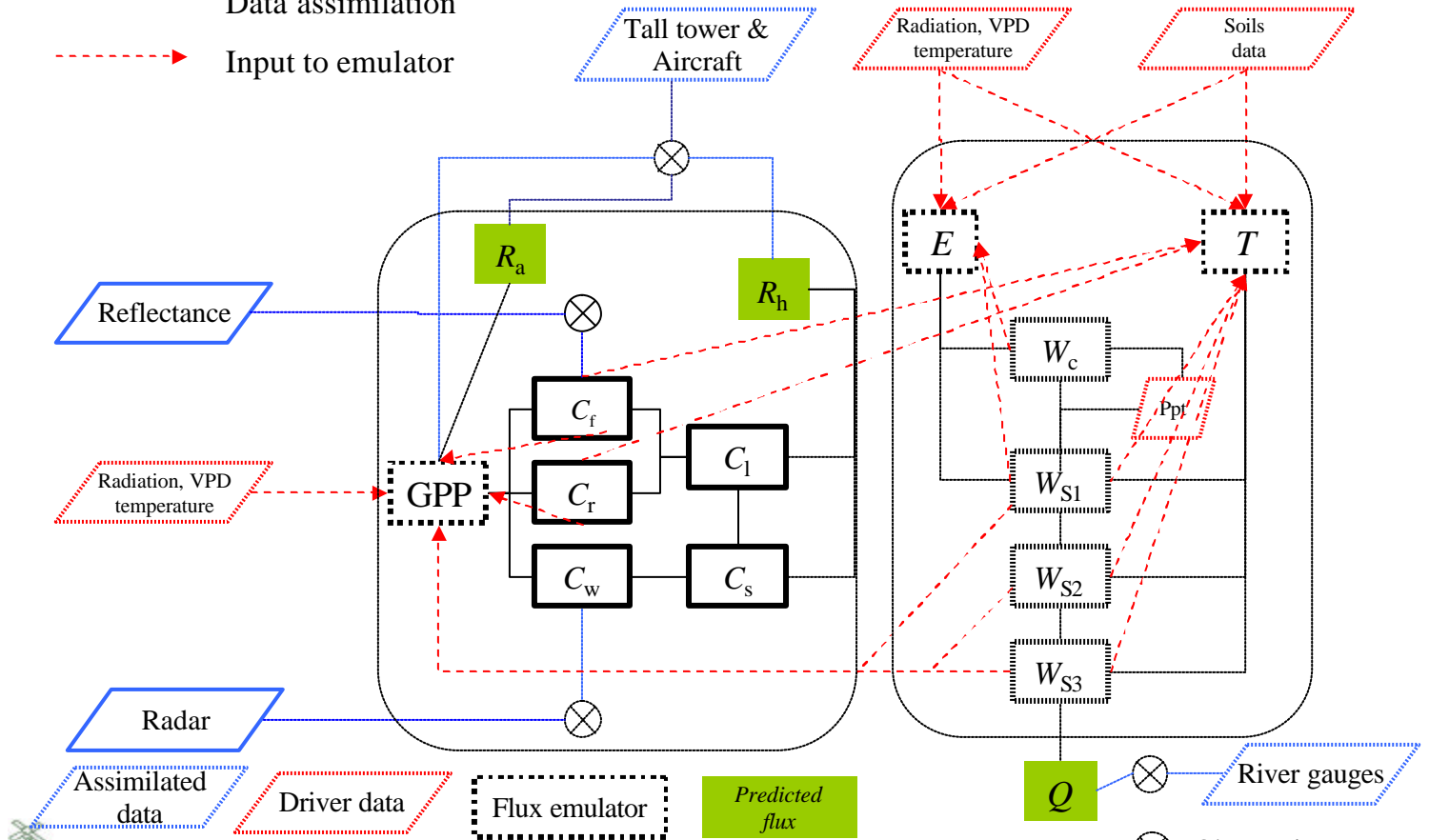
# Local assimilation

- Flux
- Data assimilation
- - - Input to emulator



# Regional assimilation

- Flux
- Data assimilation
- - -> Input to emulator



# Conclusions

- ◆ Calibration of DVM parameters with EO data provides a means to improve the **predictive** power of the models, e.g., phenology, fire.
- ◆ Well-developed forward models for scattering and reflectance exist; a current challenge is to interface them to biospheric models for **monitoring and assimilation**.
- ◆ Because of possible problems in derived products, such as fAPAR, assimilation of radiances seems preferable. However, this is dependent on how radiation absorption is represented in the biospheric model.
- ◆ Successful assimilation schemes have just been developed for biospheric models. By using existing forward models, these provide a framework for assimilating EO data.
- ◆ Watch this space!