

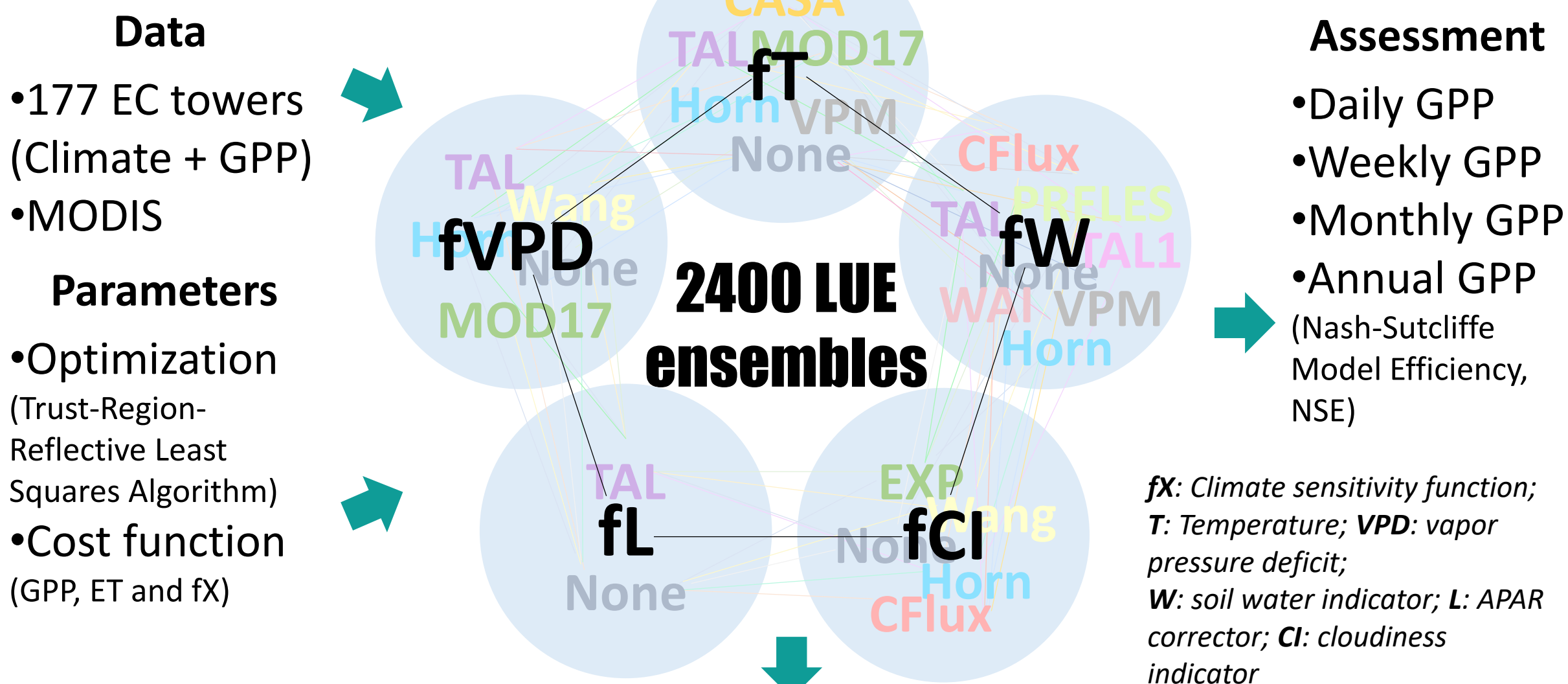
# Evaluating Light Use Efficiency (LUE) Models and Parameter-upscaling Methods

Shanning Bao ([sbao@bgc-jena.mpg.de](mailto:sbao@bgc-jena.mpg.de)), Fabian Gans, Simon Besnard, Sujan Koirala, Alvaro Moreno, Sophia Walther, Ulrich Weber, Martin Jung, Miguel Mahecha, and Nuno Carvalhais

LUE model structure

$$GPP = \varepsilon_{max} \cdot APAR \cdot fT \cdot fVPD \cdot fW \cdot fL \cdot fCI$$

## 1. Which is the best model ?



Best LUE model (NSEmedian,d/w/m/a = 0.73/0.79/0.84/0.54)

$$GPP = \varepsilon_{max} \cdot APAR \cdot fT_{CASA} \cdot fVPD_{TAL} \cdot fW_{Horn} \cdot fL_{TAL/None} \cdot fCI_{EXP}$$

## 2. How to upscale parameters ?

## Take home message

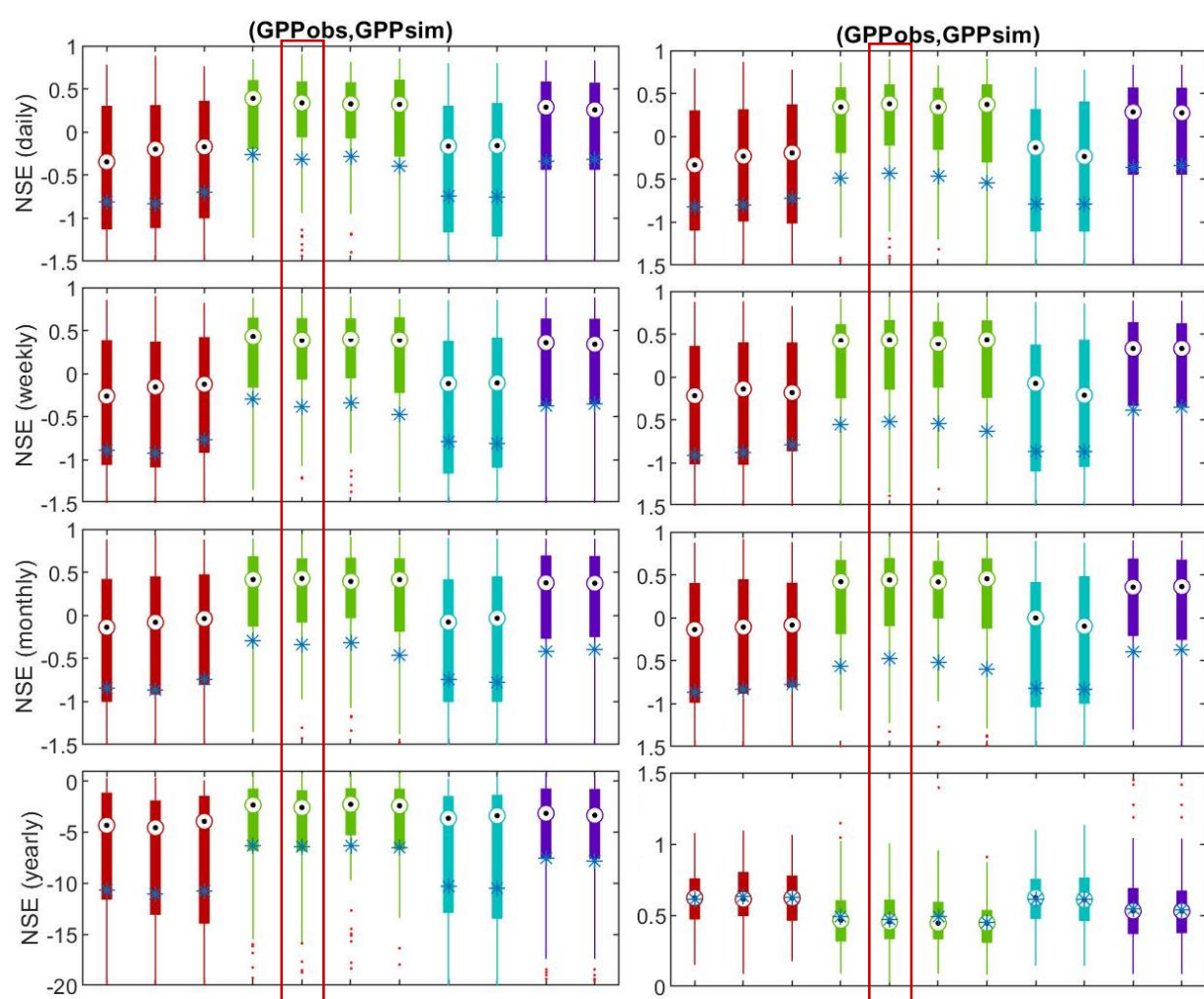


Fig.1 NSE of GPP using upscaled parameters in cross validation for model I (left) and model II (right)

- **Mean** of per climate type (Koeppen-Geiger, K-G)
- **Mean** of per Plant Functional Type (PFT)
- **Mean** of per PFT and K-G (first 2 characters)
- **Median** of per K-G
- **Median** of per PFT
- **Median** of per PFT and K-G (first 2 characters)
- **Median** of per plant type
- **Random Forest (RF) Regression** using bioclimatic variables and corresponding vegetation indexes (VI)
- **RF Regression** using bioclimatic variables
- **Site similarity** using PFT, VI and mean seasonal cycle (MSC) climate variables
- **Site similarity** using PFT, VI, MSC climate and ET

- On daily, weekly, monthly and yearly scale, 36 models were significantly better than the others.
- The best two models as above had the best global NSE (NSE for all sites) over other models for the four time scales.
- Using the median parameters per PFT had the best performance to upscale parameters from site-level to global-level.
- We further explore the relationship between parameters/climate sensitivity functions and environmental drivers as well as biophysical plant traits using global retrieval of SIF.

# Functional Responses of Primary Productivity to Climate

**SHANNING BAO**, FABIAN GANS, SIMON BESNARD, SUJAN KOIRALA, ALVARO MORENO, SOPHIA WALTHER, ULRICH WEBER, MARTIN JUNG, MIGUEL MAHECHA, AND NUNO CARVALHAIS

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# Light Use Efficiency (LUE) models

$$GPP = \varepsilon_{max} \cdot APAR \cdot fT \cdot fVPD \cdot fW \cdot fL \cdot fCI$$

|  |  |
|--|--|
| <b>GPP:</b>                            | Gross Primary Productivity                   |
| <b><math>\varepsilon_{max}</math>:</b> | maximum light use efficiency                 |
| <b>APAR:</b>                           | Active Photosynthetically Absorbed Radiation |
| <b>fT:</b>                             | Temperature sensitivity function             |
| <b>fVPD:</b>                           | Vapor Pressure Deficit sensitivity function  |
| <b>fW:</b>                             | soil Water indicator sensitivity function    |
| <b>fL:</b>                             | Light (APAR) sensitivity function            |
| <b>fCI:</b>                            | Cloudiness Index sensitivity function        |

|       |       |        |      |       |
|-------|-------|--------|------|-------|
| MOD17 | MOD17 | TAL    | TAL  | CFlux |
| TAL   | TAL   | Horn   | none | Wang  |
| CASA  | CASA  | VPM    |      | EXP   |
| Horn  | Wang  | CFlux  |      | Horn  |
| VPM   | none  | PRELES |      | none  |
| none  |       | none   |      |       |



# Questions

- Which is the best LUE model?
- Which are the best climate sensitivity functions of GPP?
- Does the climate sensitivity change with environmental condition and biophysical traits of vegetation?

# Assumptions

- The LUE model which has the best model efficiency on different time scales and less parameters is the best model.
- The climate sensitivity functions (fXs) of the best LUE model can best represent the response of vegetation photosynthesis rate to climate change.
- The model parameters which controls the fXs trends change with environmental condition and biophysical traits of vegetation.

# Experiment design

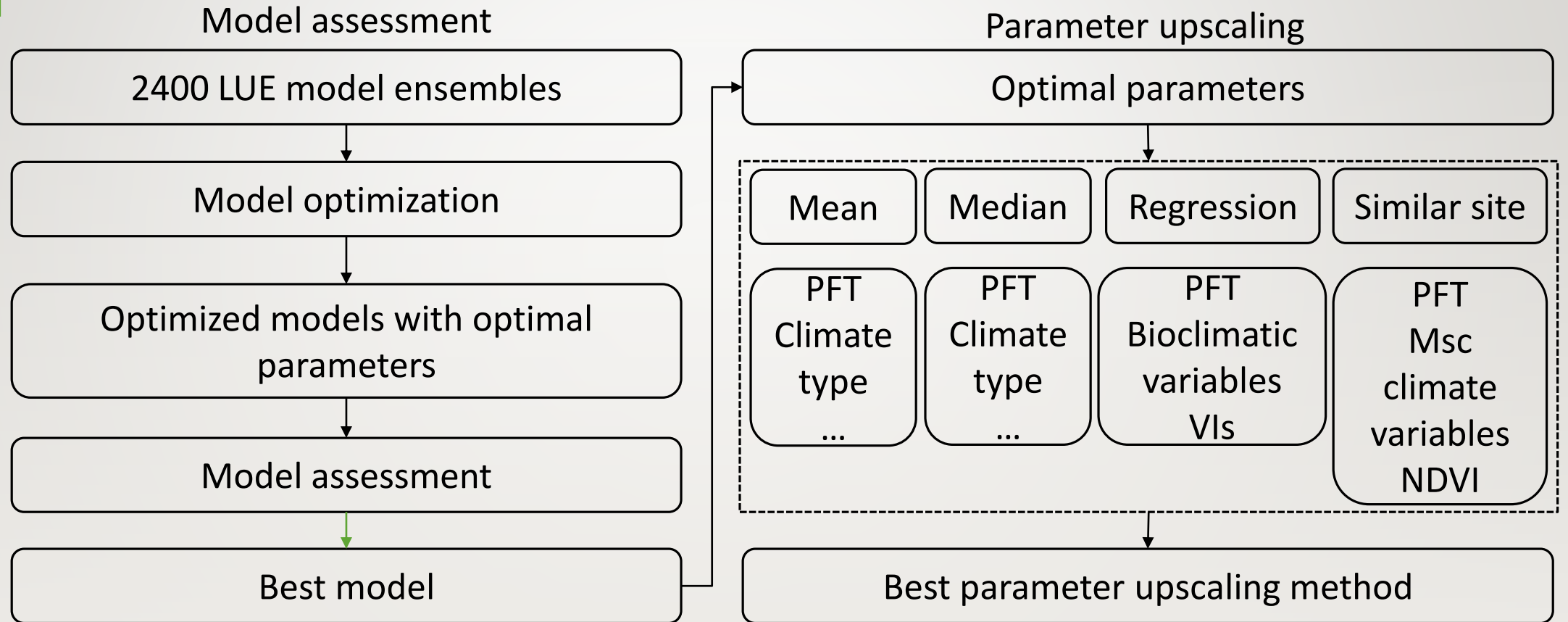
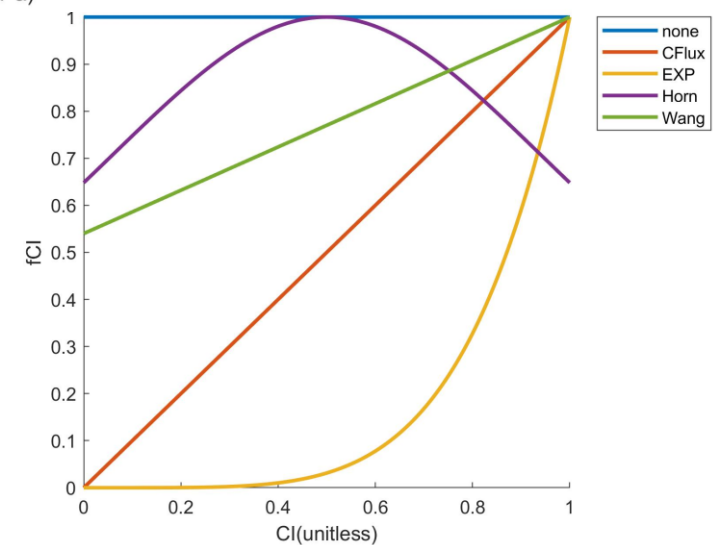
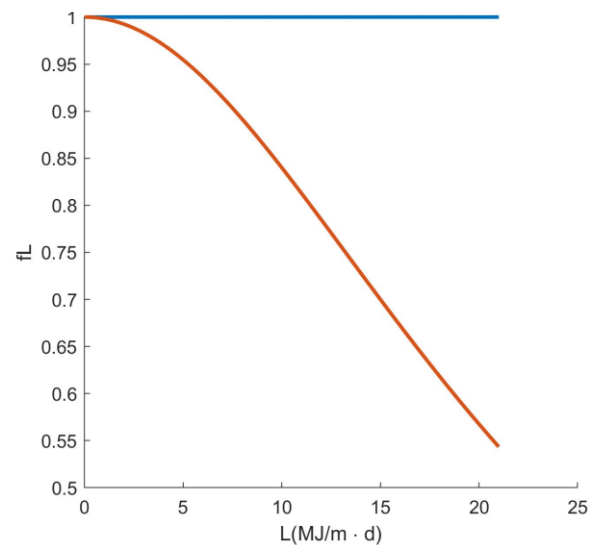
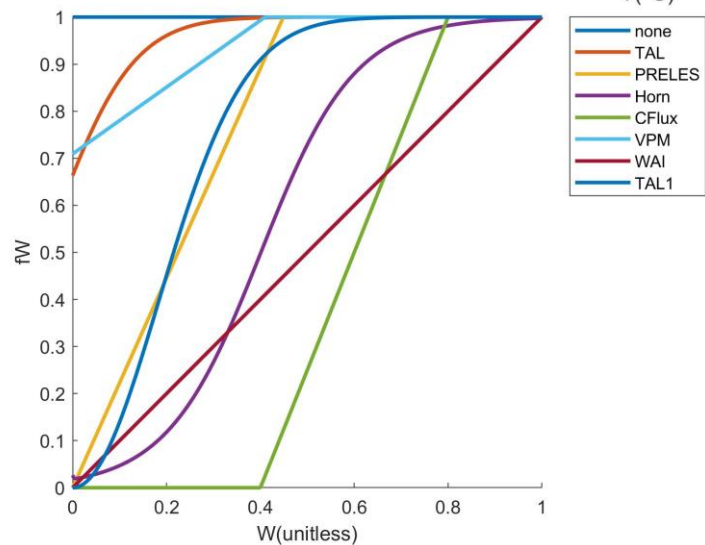
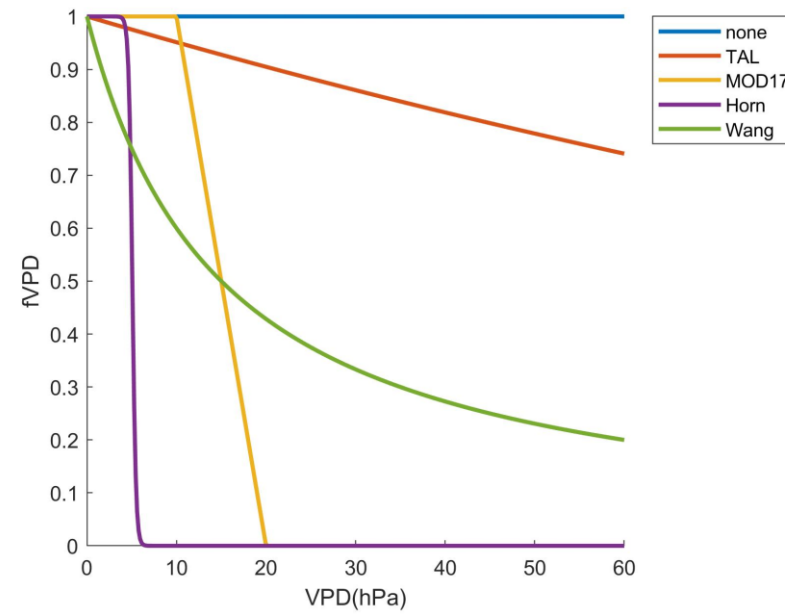
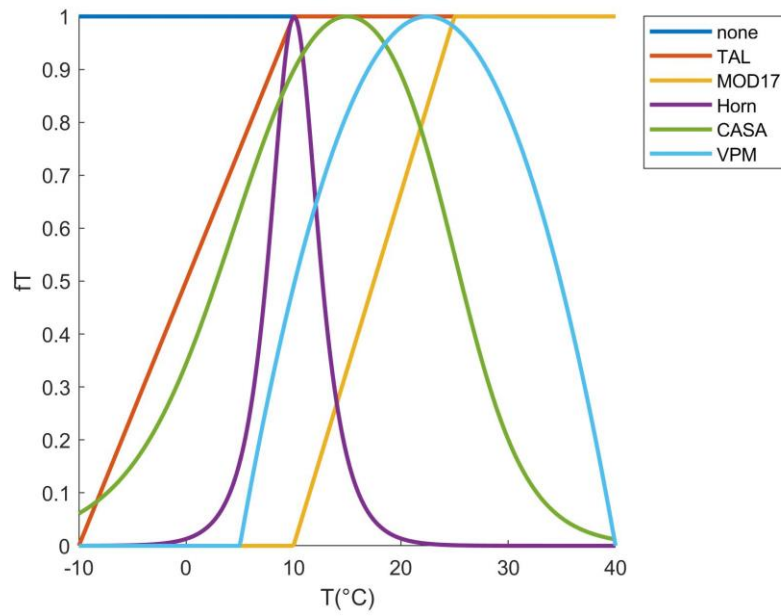


Fig.1 Workflow of this study

# Climate sensitivity functions in LUE models



# Results

## 1. Best model selection

$$GPP = \varepsilon_{max} \cdot APAR \cdot fT_{CASA} \cdot fVPD_{TAL} \cdot fW_{Horn} \cdot fL_{TAL} \cdot fCI_{EXP} \quad (I)$$

$$GPP = \varepsilon_{max} \cdot APAR \cdot fT_{CASA} \cdot fVPD_{TAL} \cdot fW_{Horn} \cdot fL_{None} \cdot fCI_{EXP} \quad (II)$$

| fX                  | Equation   | Reference                       |
|---------------------|--|---------------------------------|
| fT <sub>CASA</sub>  | $\frac{2 \times \cosh(5 \times T_{ab})^2}{\left( \cosh\left(T_{ab} \times (T_{opt} - T)\right) + \cosh(10 \times T_{ab}) \right)}$ $, T_{ab} = (T < T_{opt}) \times T_a + (T \geq T_{opt}) \times T_b$ | (Potter, Randerson et al. 1993) |
| fVPD <sub>TAL</sub> | $e^{\kappa \times VPD}$  | (MÄKelä, Pulkkinen et al. 2007) |
| fW <sub>Horn</sub>  | $1 / \left( 1 + e^{k_W \times (WAI_f - W_l)} \right)$ $WAI_{f_k} = (1 - \alpha) \times WAI_k + \alpha \times WAI_{f_{k-1}}, \text{ k is time}$   | (Horn and Schulz 2011)          |
| fL <sub>TAL</sub>   | $1 / (\gamma \times APAR + 1)$   | (MÄKelä, Pulkkinen et al. 2007) |
| fL <sub>None</sub>  | 1  | -                               |
| fCI <sub>EXP</sub>  | $CI^\mu$   | This study                      |



# Results

## 1. Best model selection

- The Nash-Sutcliffe Model Efficiency(NSE) of the two models:

| NSE                | Model | Daily | Weekly | Monthly | Annual |
|--------------------|-------|-------|--------|---------|--------|
| Median of site NSE | I     | 0.726 | 0.788  | 0.836   | 0.544  |
|                    | II    | 0.724 | 0.782  | 0.834   | 0.510  |
| Global NSE         | I     | 0.755 | -      | -       | -      |
|                    | II    | 0.753 | -      | -       | -      |

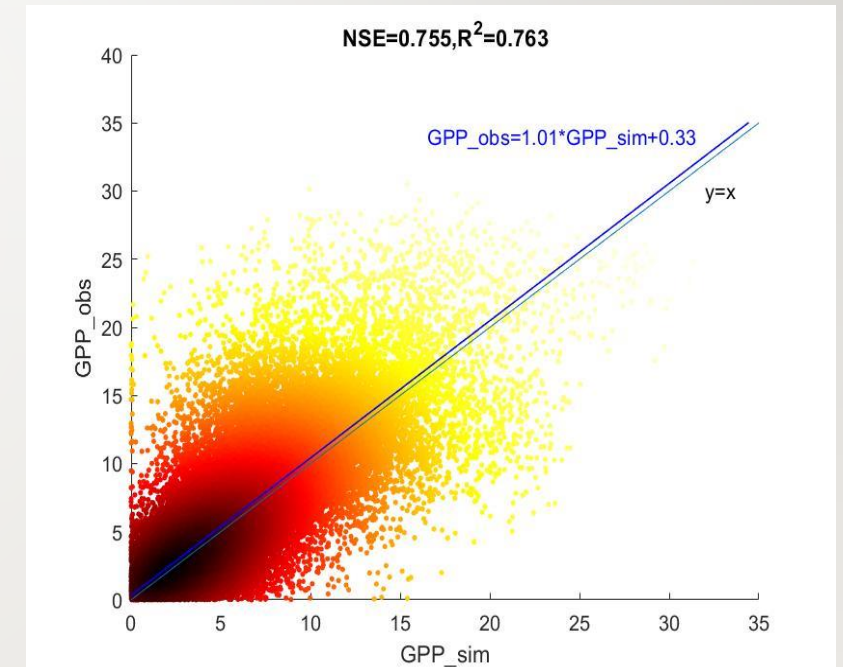


Fig.2 Model I simulated GPP against observed GPP (color represent the density)

# Results

## 2. Parameter upscaling

- Mean of per climate type (Koeppen-Geiger, K-G)
- Mean of per Plant Functional Type (PFT)
- Mean of per PFT and K-G (first 2 characters)
- Median of per K-G
- Median of per PFT
- Median of per PFT and K-G (first 2 characters)
- Median of per plant type
- Random Forest (RF) Regression using bioclimatic variables and corresponding vegetation indexes (VI)
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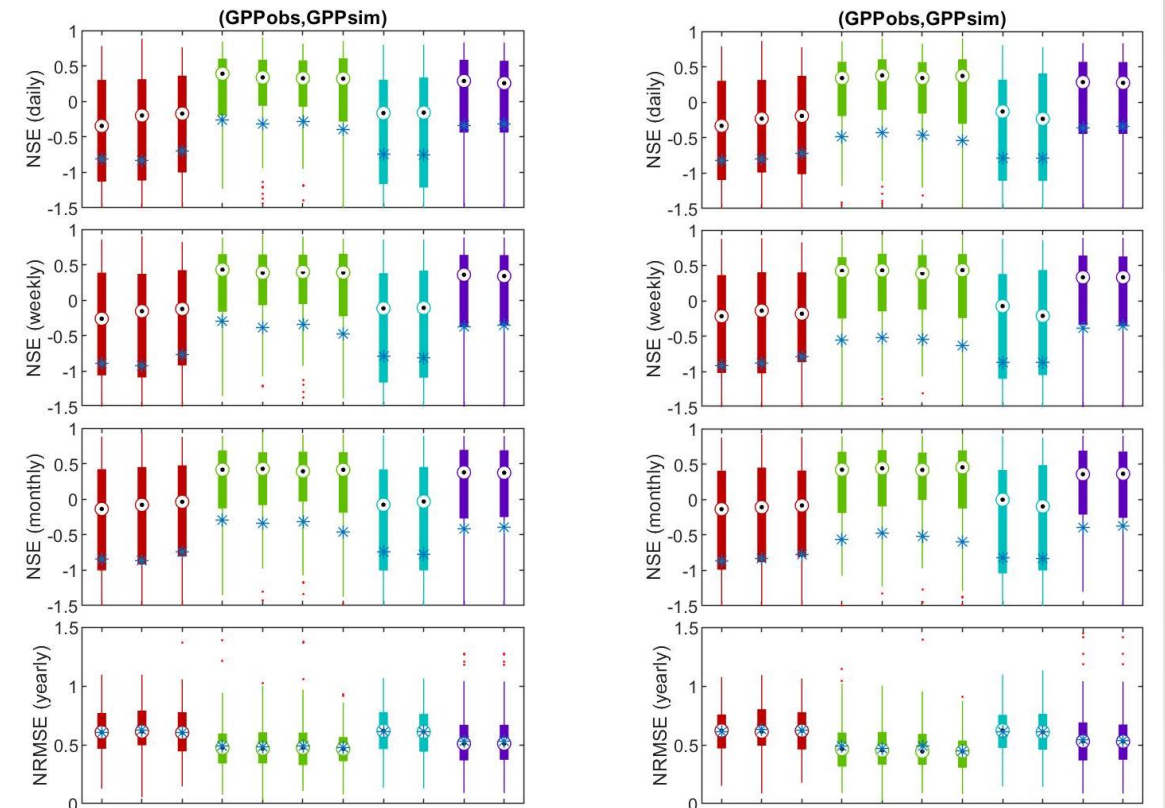


Fig. 3 NSE of GPP simulated by model I(left) and model II (right) and upscaled parameters

# Conclusions

- On daily, weekly, monthly and yearly scale, 36 models were significantly better than the others.
- The best two models as above had the best global NSE (NSE for all sites) over other models for the four time scales.
- Using the median parameters per PFT had the best performance to upscale parameters from site-level to global-level.
- Since the limitation of sparse EC towers, we further explore the relationship between parameters/climate sensitivity functions and environmental drivers as well as biophysical plant traits using global retrieval of SIF.



Thanks for your attention!