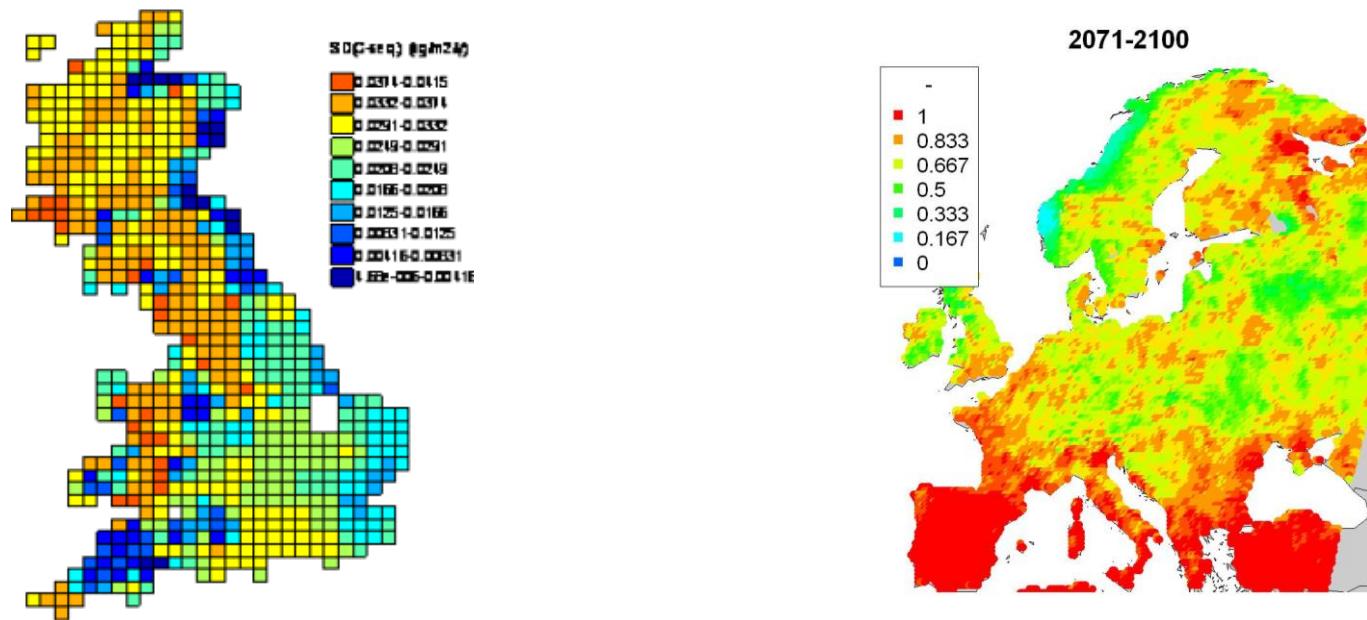


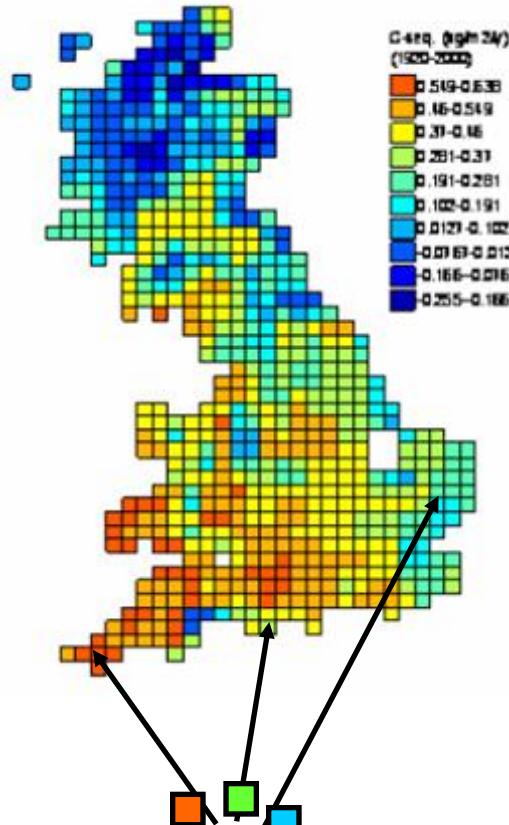
Methods for spatial upscaling and risk analysis of process-based models

Experiences from projects GREENHOUSE and Carbo-Extreme



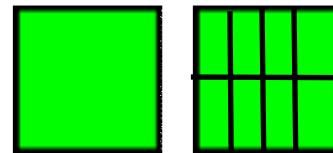
Marcel van Oijen, David Cameron, Peter Levy (CEH-Edinburgh),
and colleagues in GREENHOUSE and Carbo-Extreme

Spatial upscaling methods used for process-based models



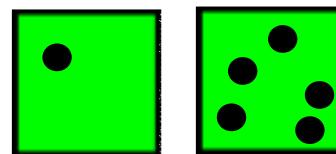
Individual grid cells (“regions”) are large → What is the model output uncertainty for each cell ?

UPSCALING METHOD



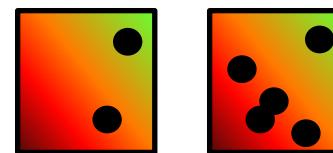
Input aggregation

$$\bar{y} = f(\bar{x})$$



Point selection

$$\bar{y} = \frac{1}{n} \sum_{i=1}^n y_i$$

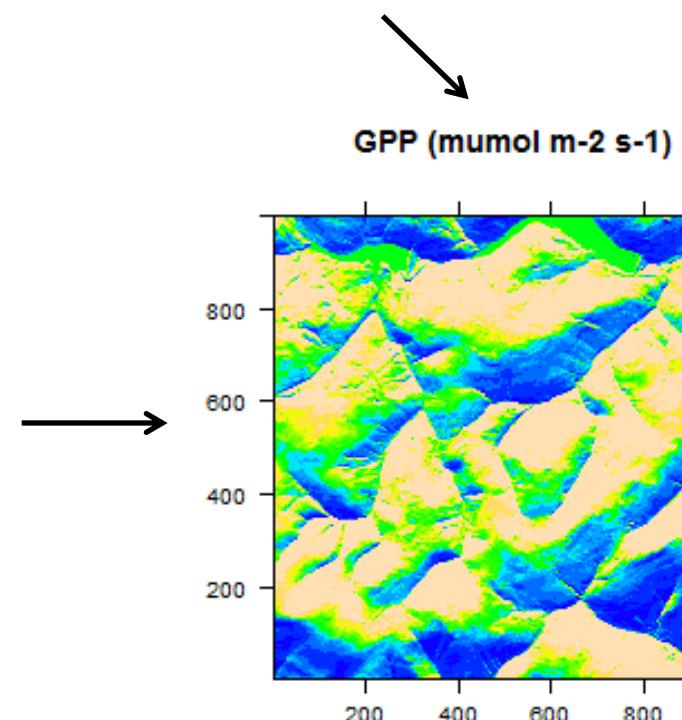
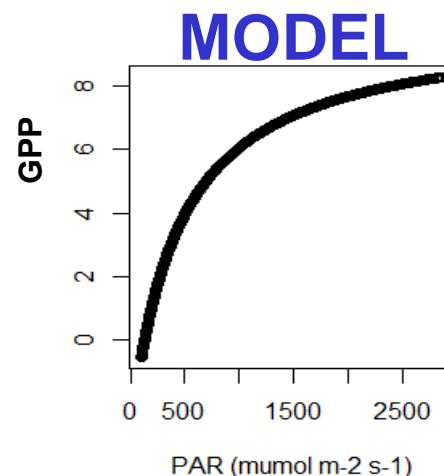
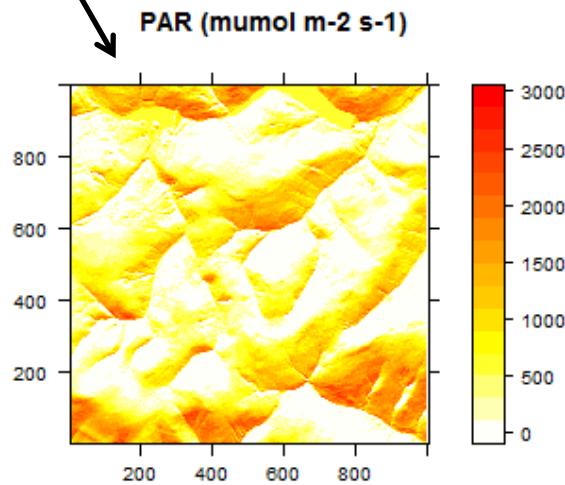


Interpolation (e.g. geostatistics)

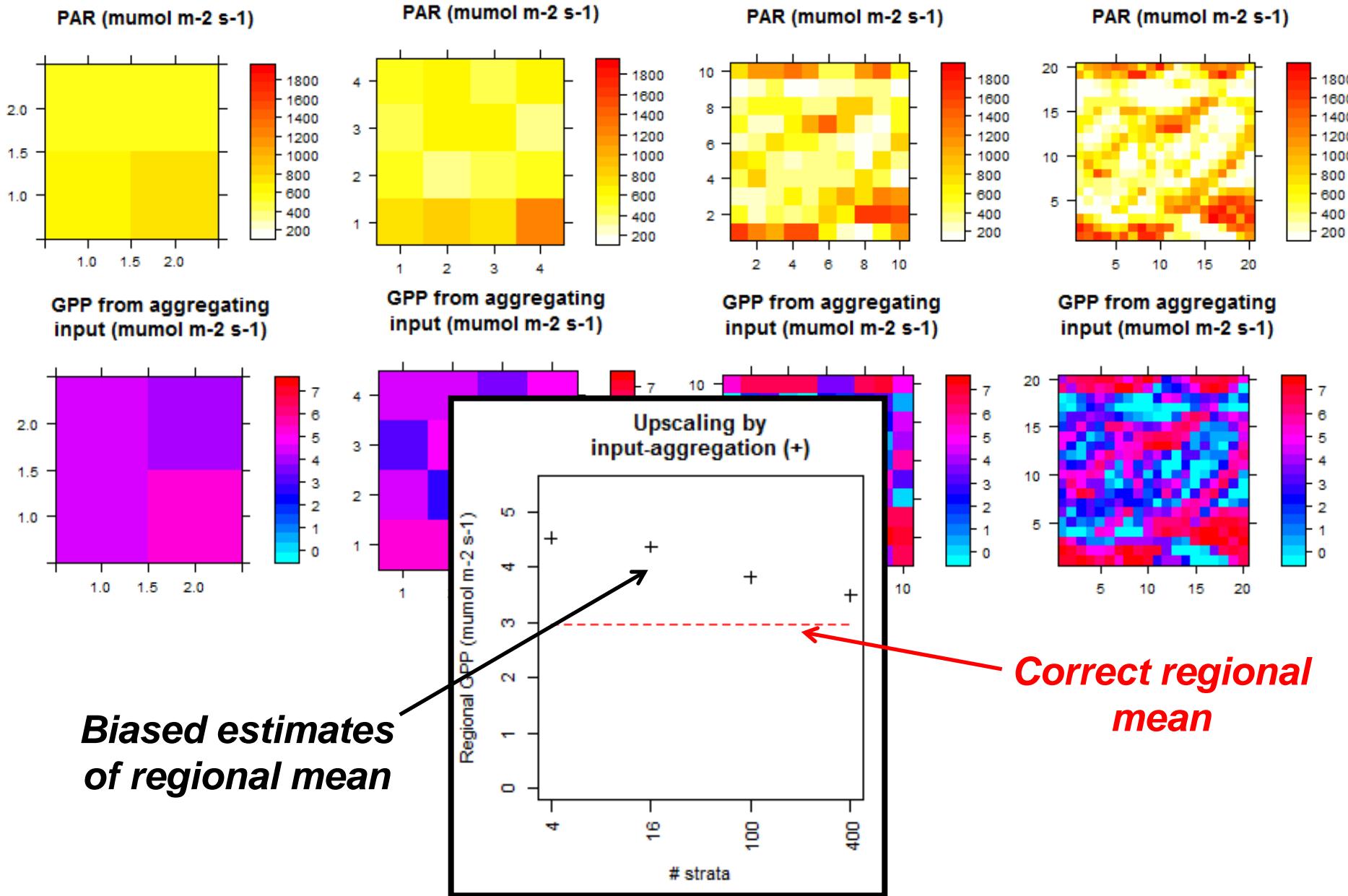
$$p(\bar{y} | \{y_i\}) = \int p(\bar{y} | \theta, \{y_i\}) p(\theta | \{y_i\}) d\theta$$

↑
Θ = Spatial hyperparameters
(variance, correlation length)

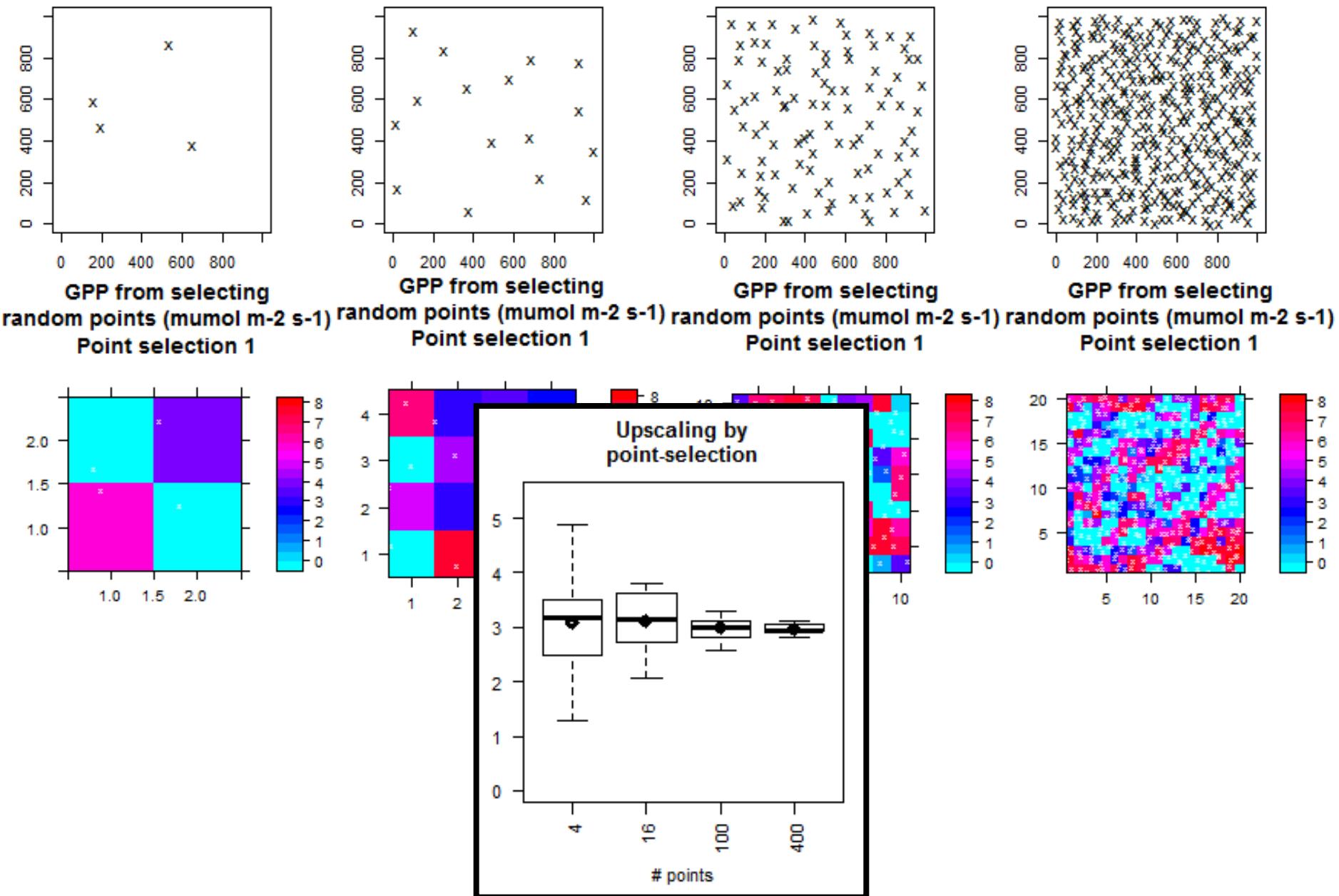
Example: region 5 x 5 km



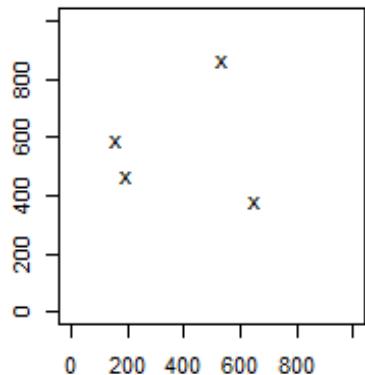
UPSCALING METHOD 1: Input Aggregation



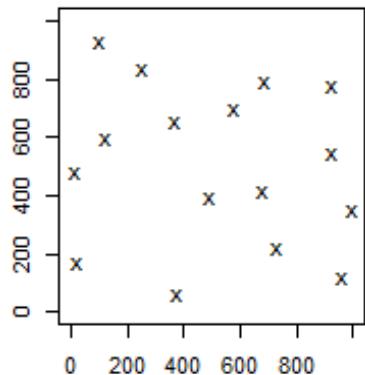
UPSCALING METHOD 2: Point Selection



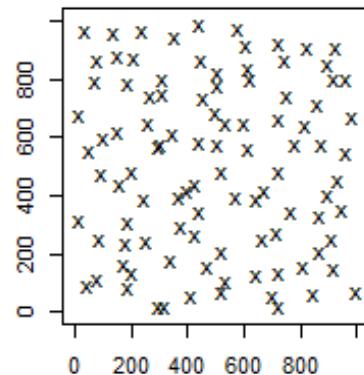
UPSCALING METHOD 3: Bayesian Kriging



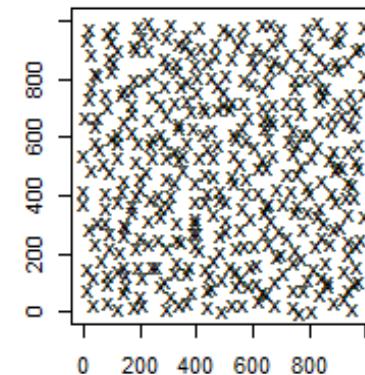
GPP from geostats
(mumol m⁻² s⁻¹)
Point selection 1



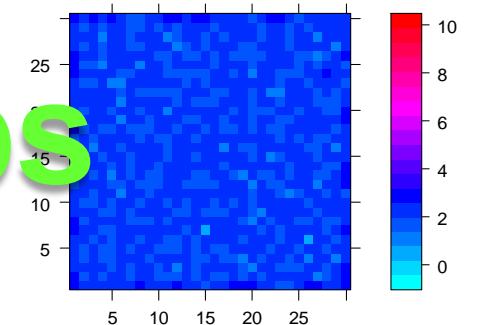
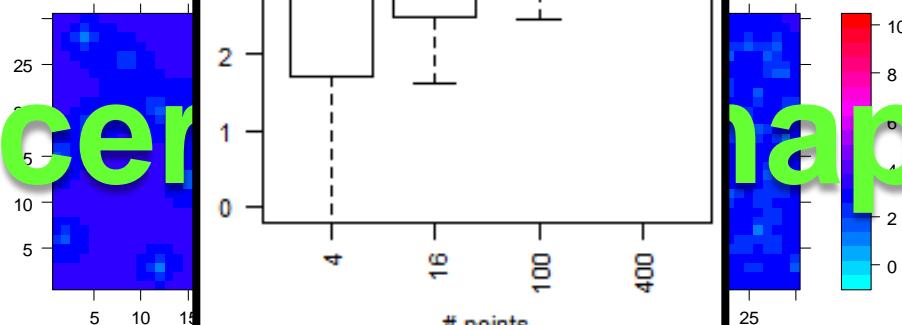
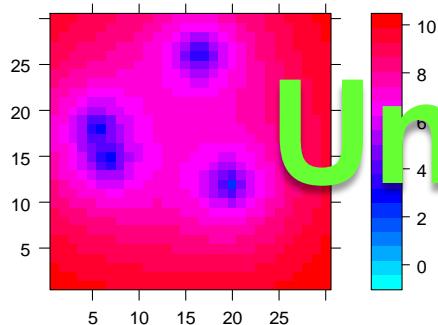
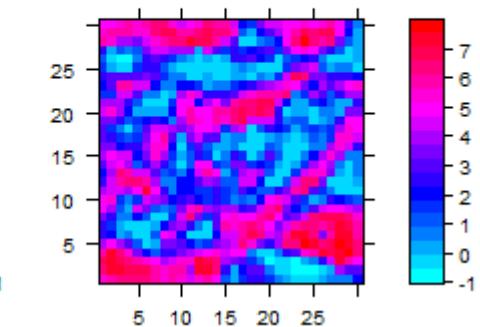
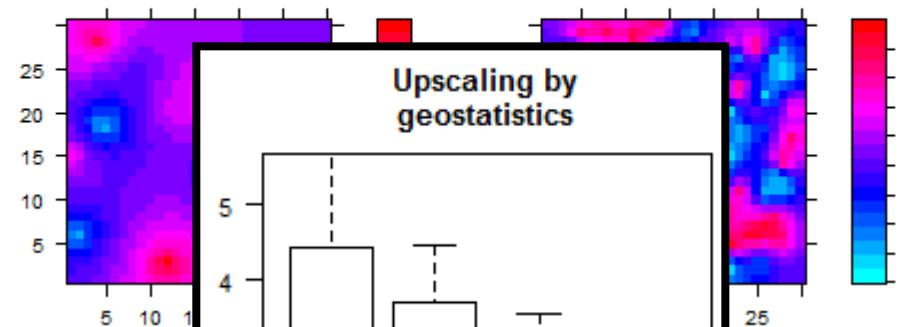
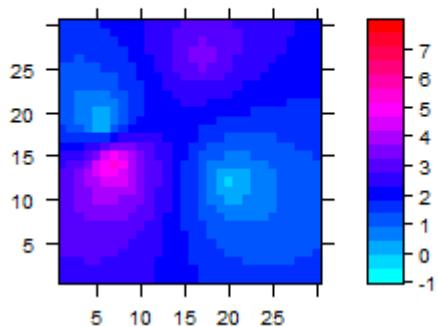
GPP from geostats
(mumol m⁻² s⁻¹)
Point selection 1



GPP from geostats
(mumol m⁻² s⁻¹)
Point selection 1

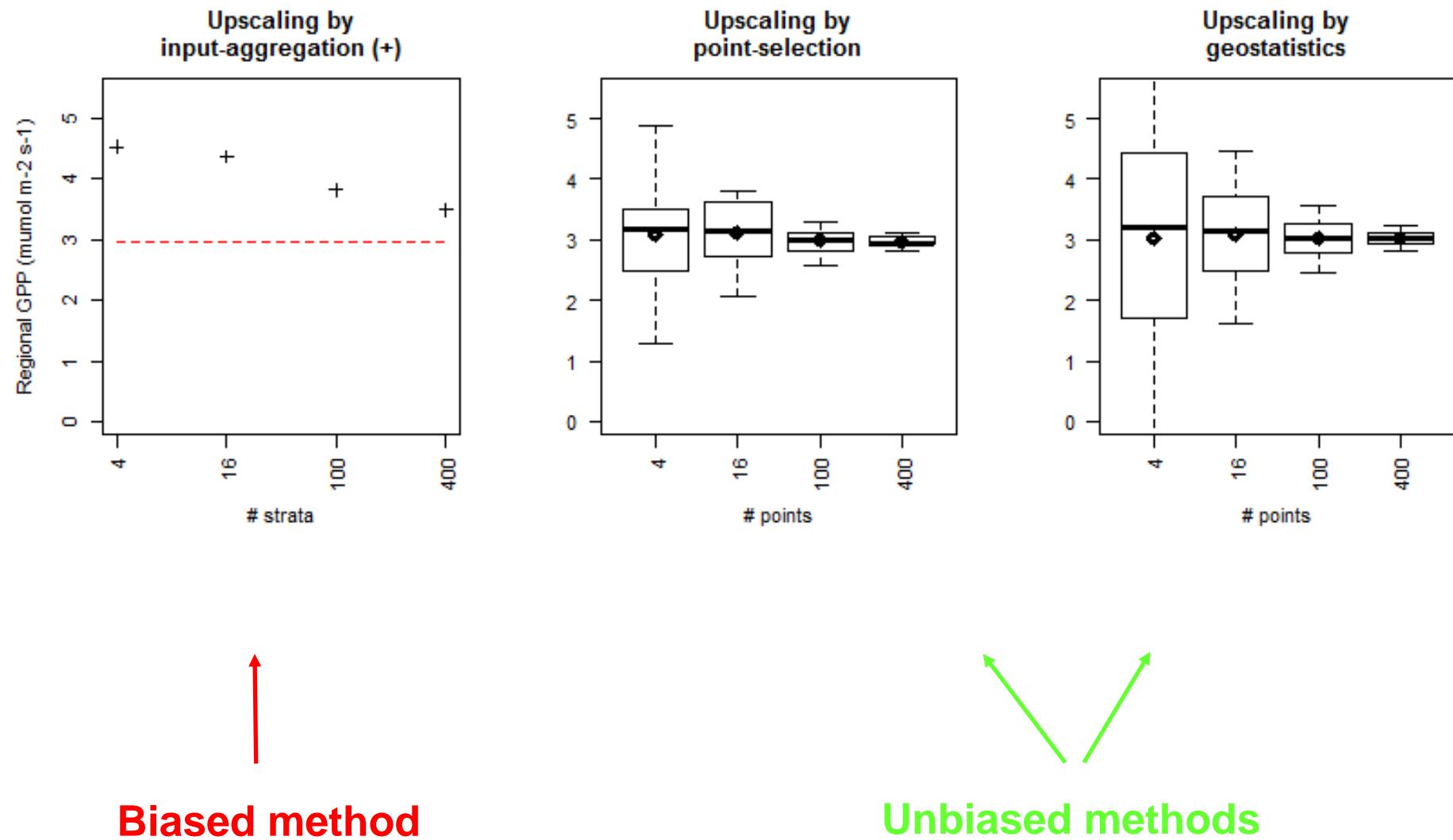


GPP from geostats
(mumol m⁻² s⁻¹)
Point selection 1

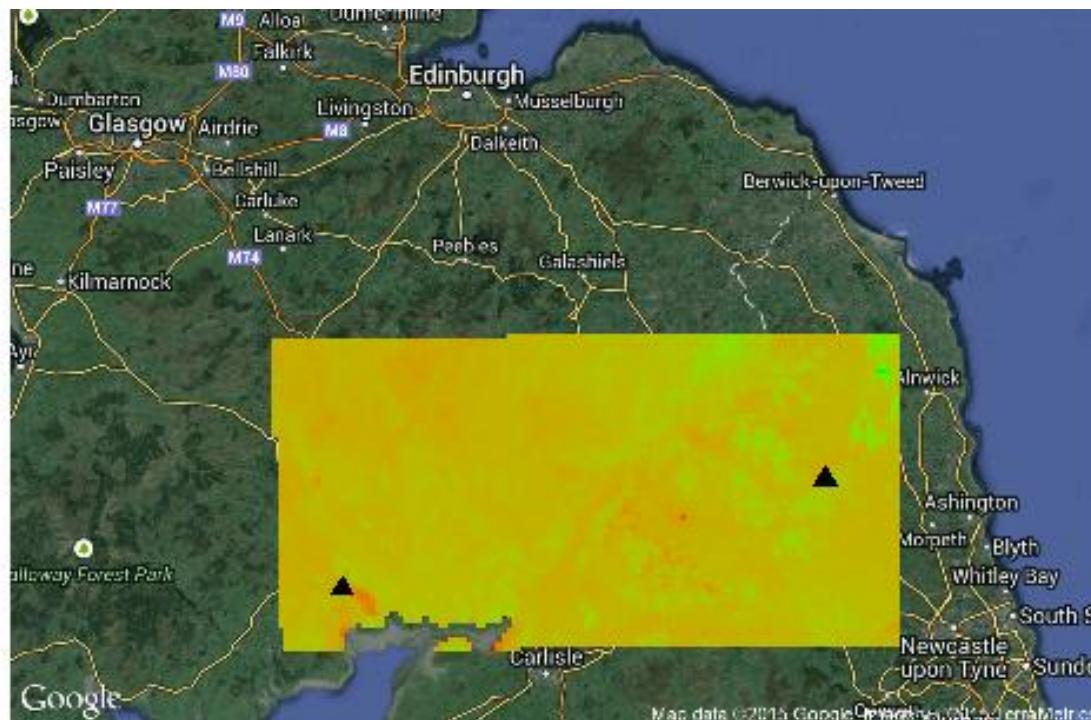


Uncertainty maps

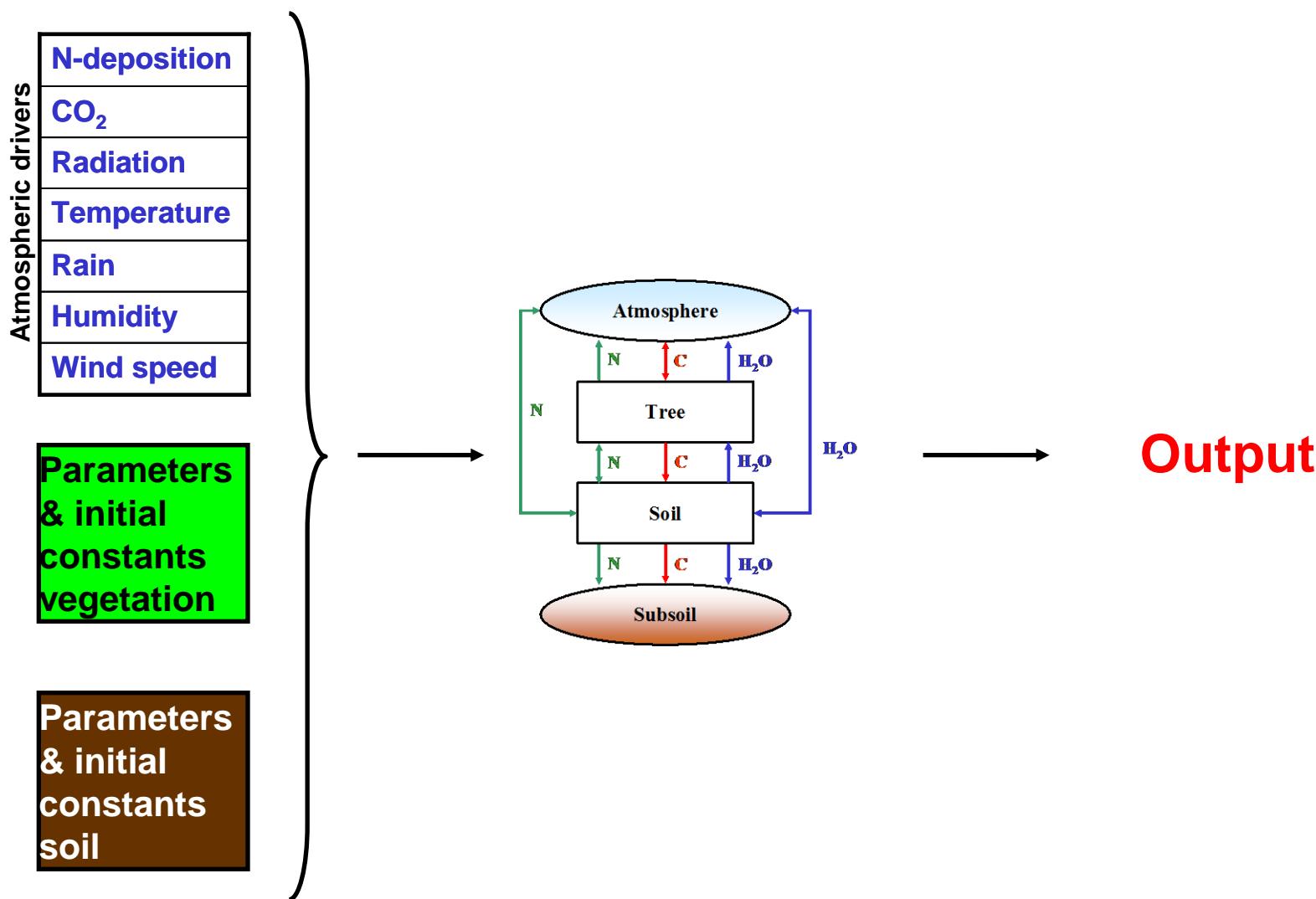
EXAMPLE 1: Comparison of methods



Example region: 128 x 64 km



Model BASECO

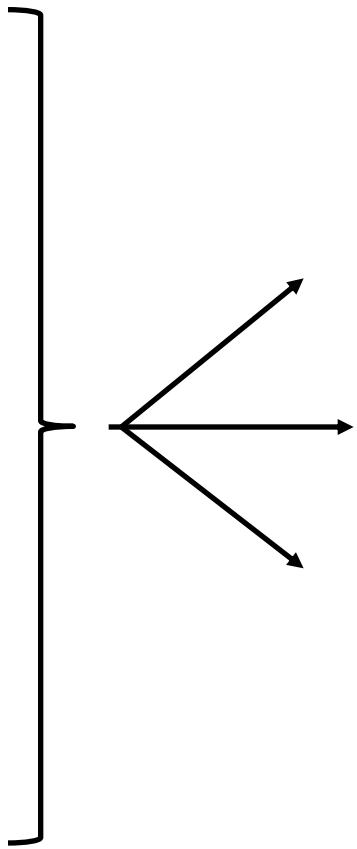
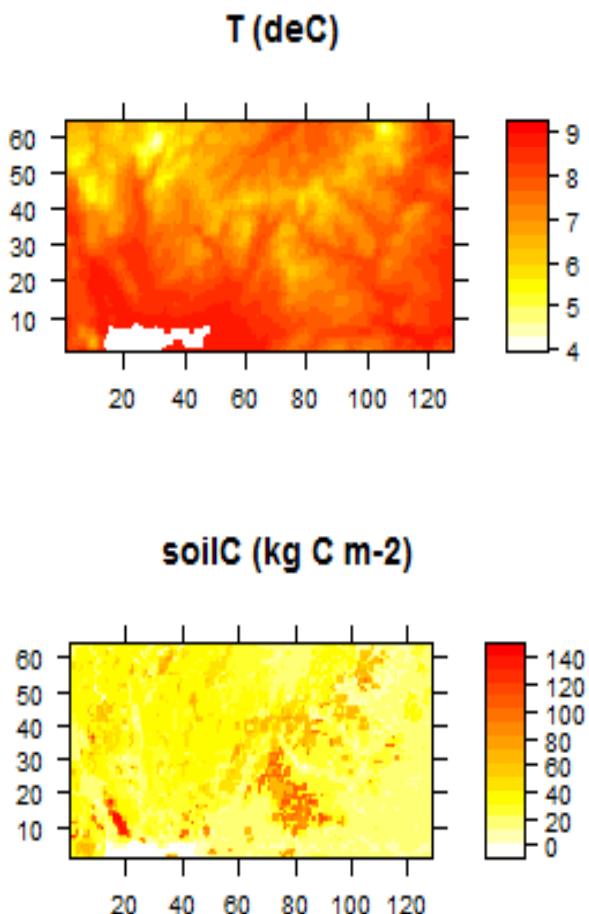


Input

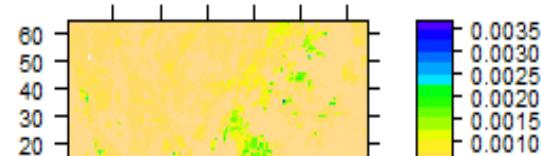
Model

Output

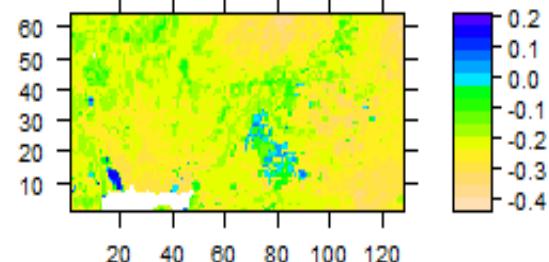
Some inputs and outputs of BASECO



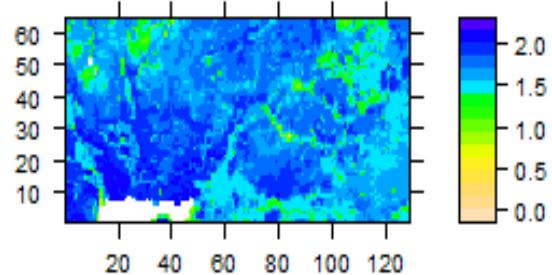
N₂O ($\text{kg N m}^{-2} \text{y}^{-1}$)



NEE ($\text{kg C m}^{-2} \text{y}^{-1}$)



GPP ($\text{kg C m}^{-2} \text{y}^{-1}$)

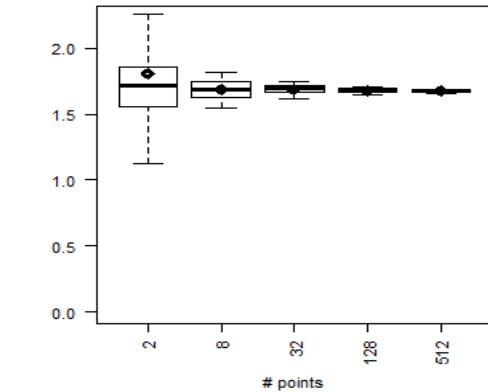
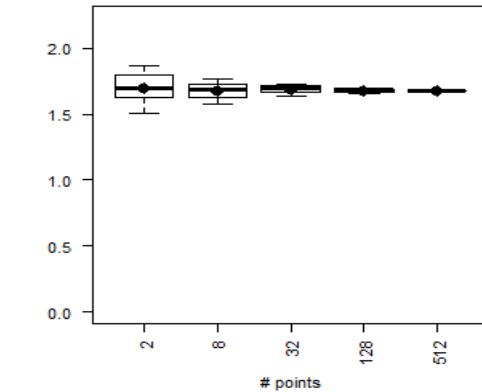
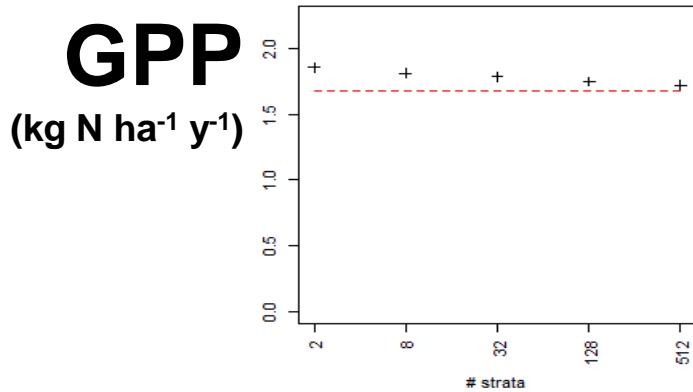
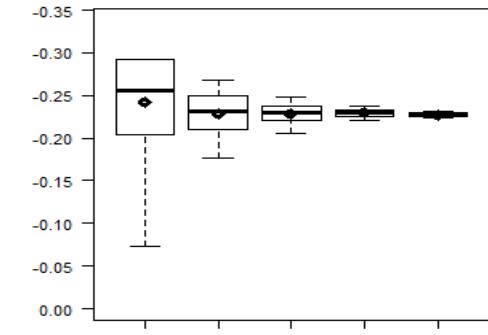
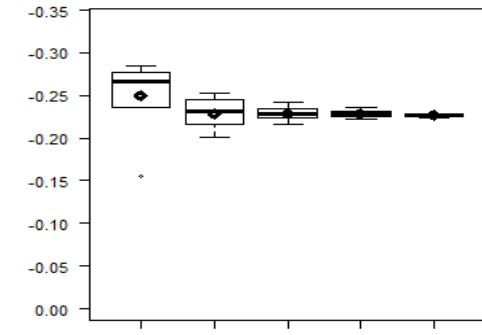
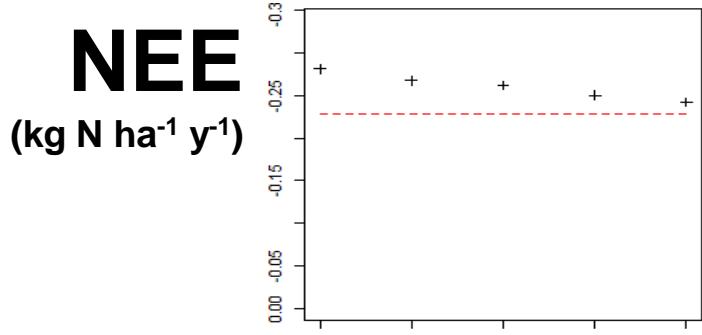
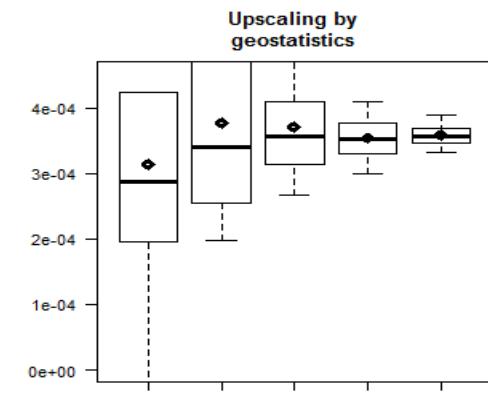
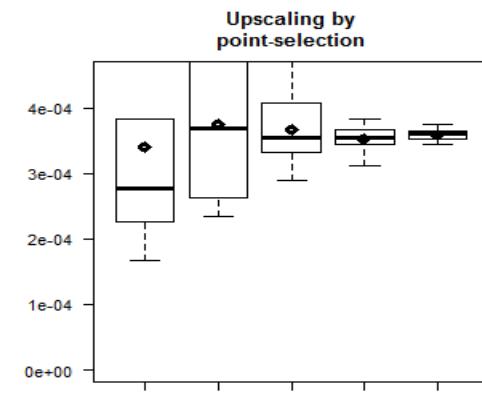
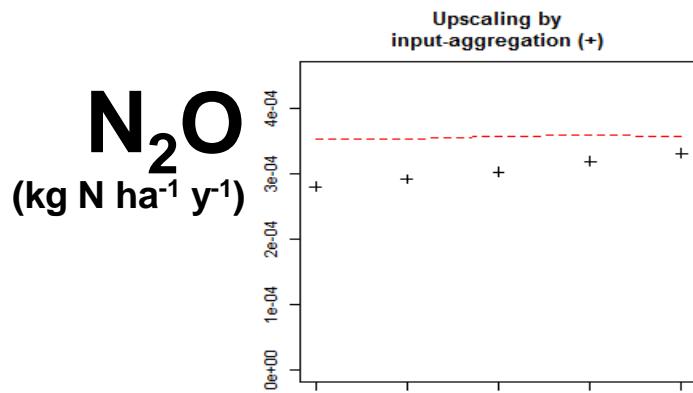


Input

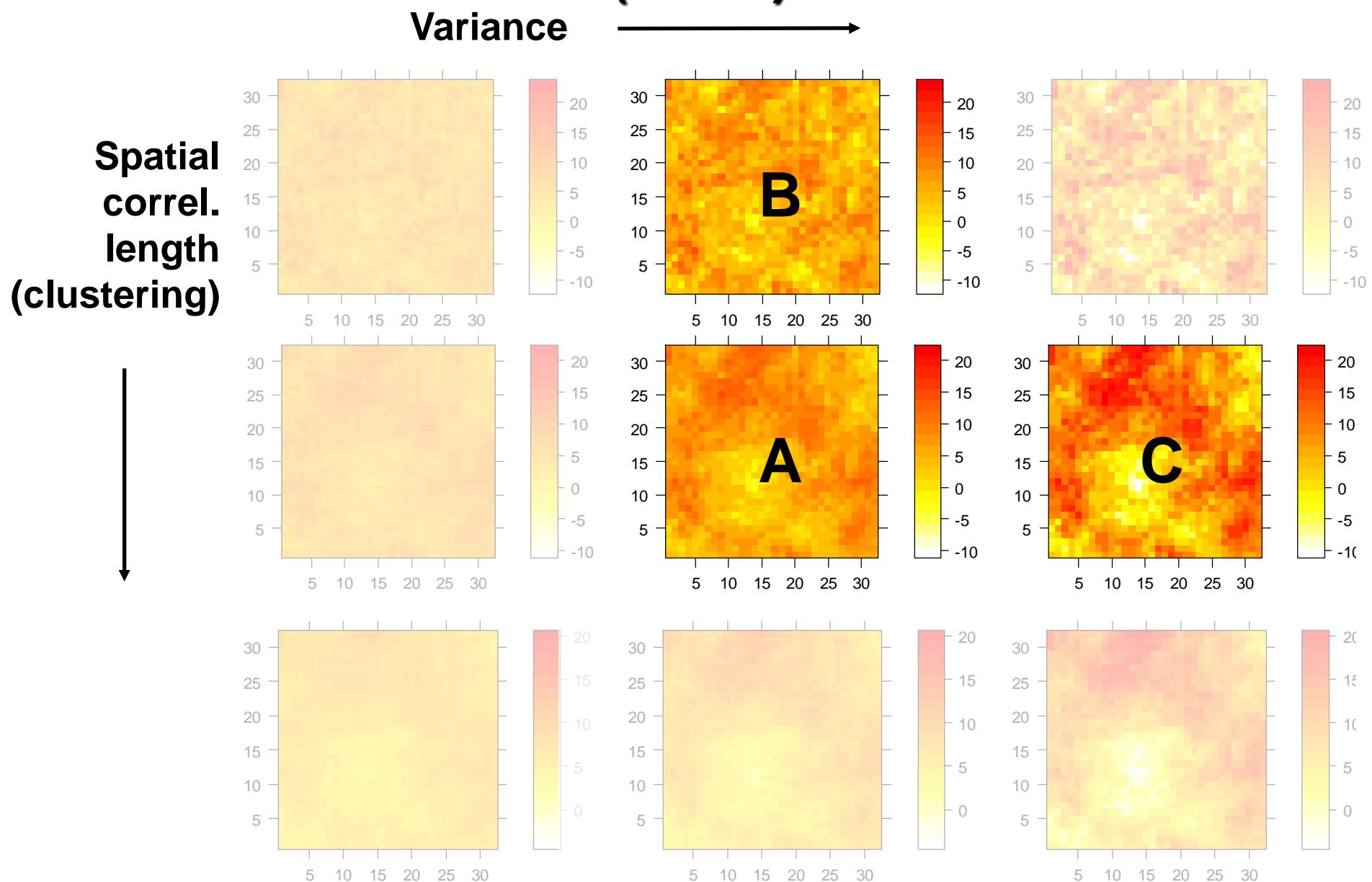
Model

Output

EXAMPLE 2: Comparison of methods

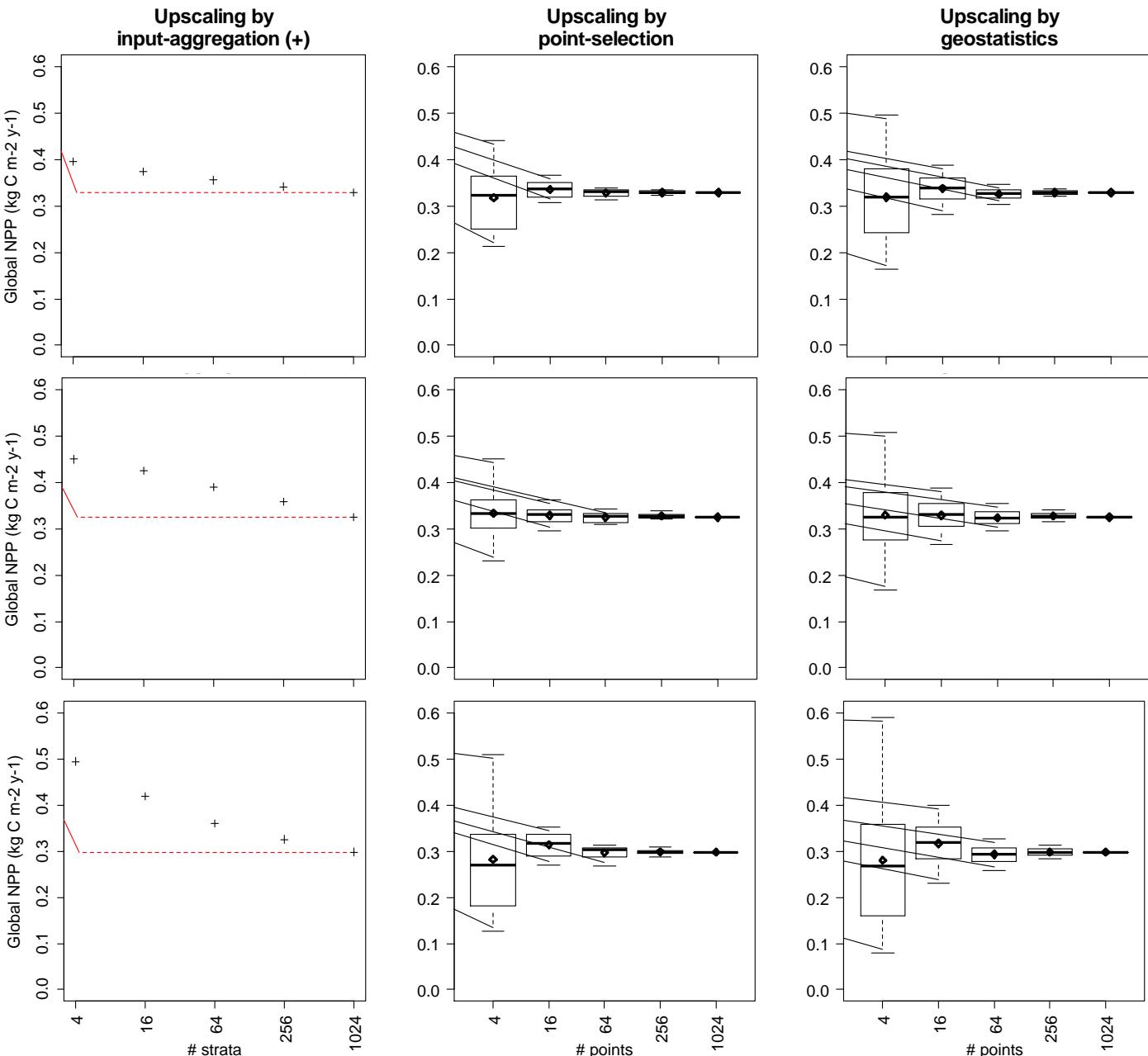


Spatial heterogeneity: Gaussian Random Fields (GRFs)



Impact of spatial correlation & variance on upscaling

A.



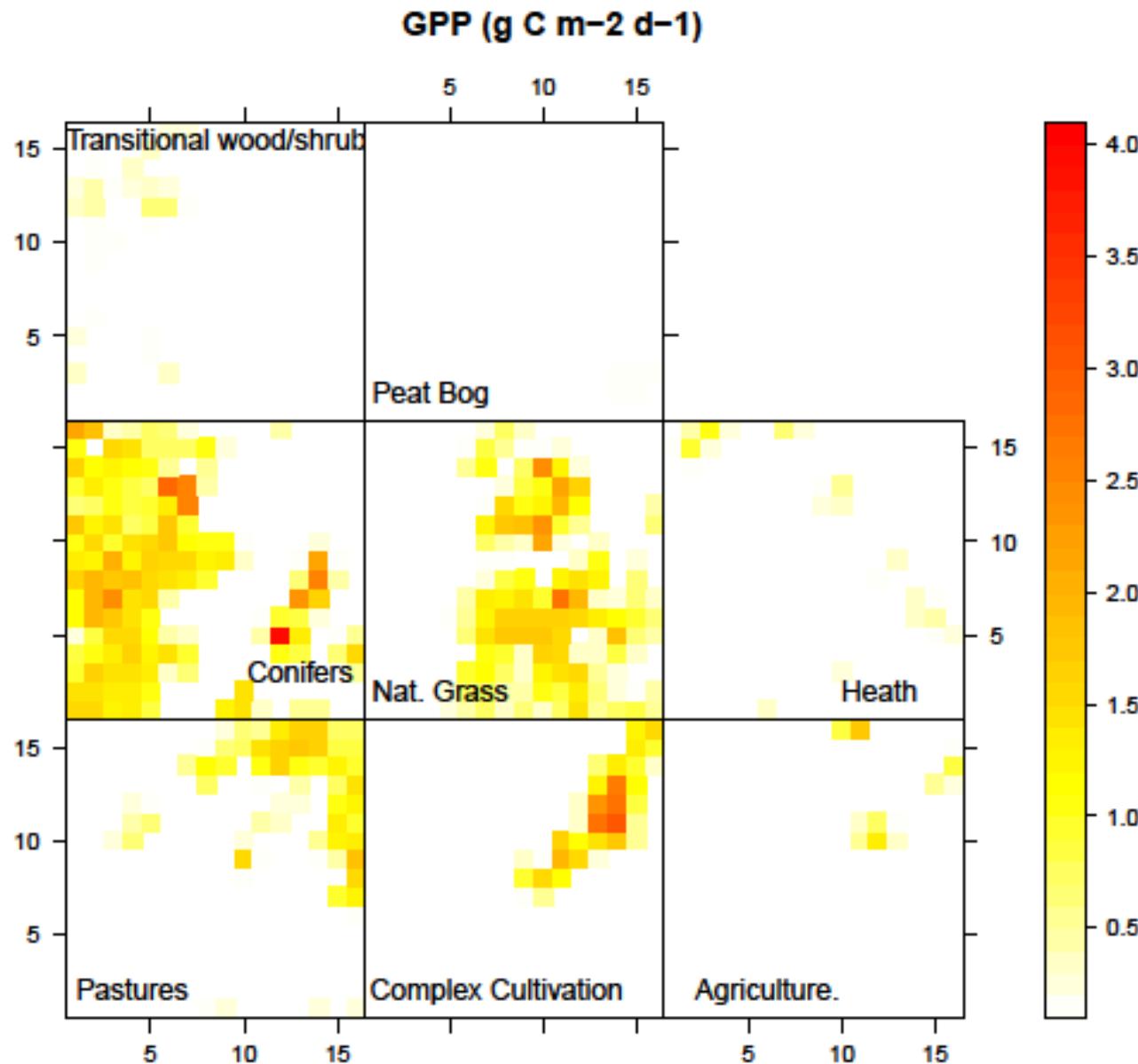
B.

Smaller spatial
correlation
length (less
clustering)

C.

Greater variance

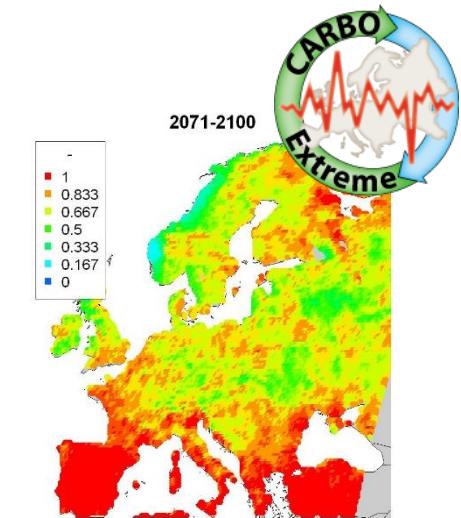
Current work: upscaling for multiple land-use types



Upscaling methods: Discussion

- Upscaling errors depend on:
 - sampling intensity and spatial distribution
 - output variable
 - variance & autocorrelation length
- Input Aggregation gives biased estimates of the regional mean; the other methods are unbiased
- Uncertainty estimation is part of (geostatistical) Interpolation; the other methods require data from representative regions

Risk analysis in project Carbo-Extreme



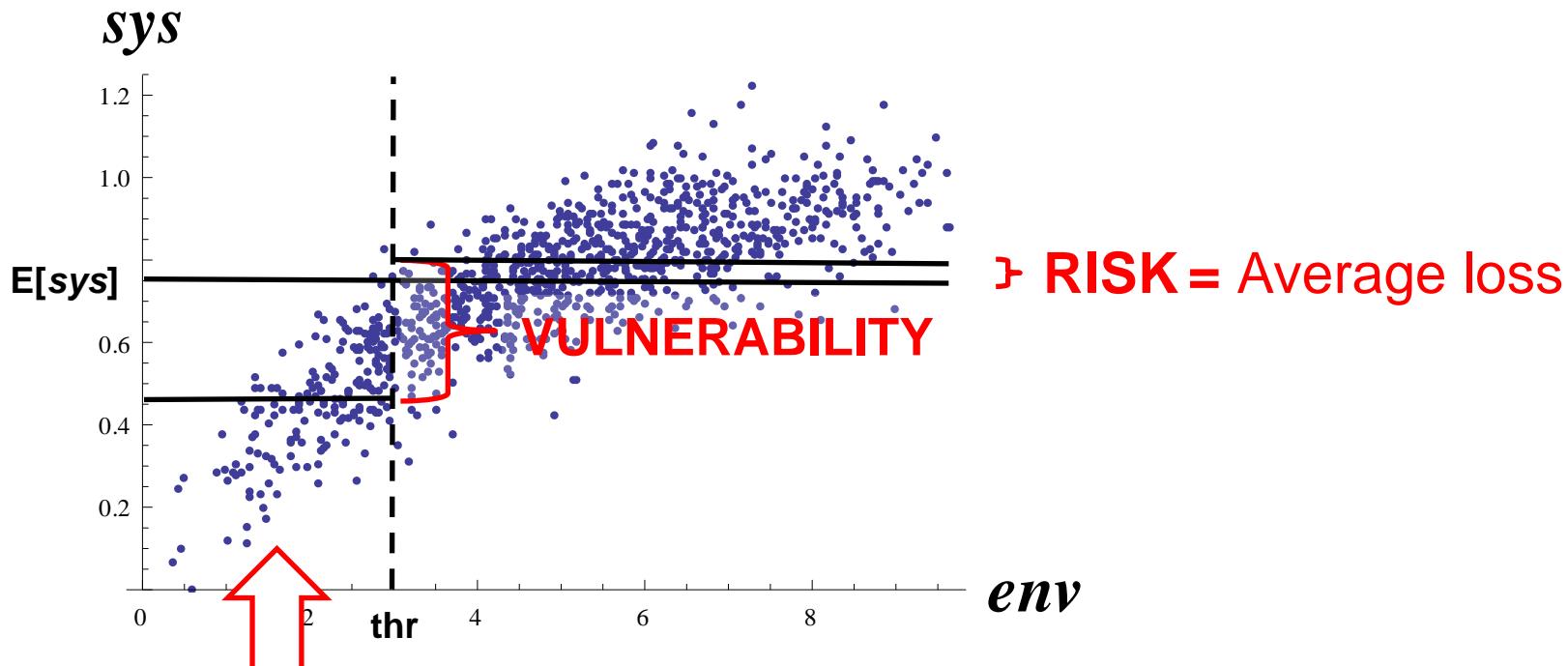
Theory:

Van Oijen, M., Beer, C., Cramer, W., Rammig, A., Reichstein, M., Rolinski, S., Seneviratne, S. & Soussana, J.-F. (2013). A novel probabilistic risk analysis to determine the vulnerability of ecosystems to extreme climatic events. *Environmental Research Letters* 8: 015032. <http://iopscience.iop.org/1748-9326/8/1/015032>

Application:

Van Oijen, M., Balkovič, J., Beer, C., Cameron, D., Ciais, P., Cramer, W., Kato, T., Kuhnert, M., Martin, R., Myneni, R., Rammig, A., Rolinski, S., Soussana, J.-F., Thonicke, K., Van der Velde, M. & Xu, L. (2014). Impact of droughts on the C-cycle in European vegetation: A probabilistic risk analysis using six vegetation models. *Biogeosciences* 11: 6357-6375. <http://www.biogeosciences.net/11/6357/2014/bg-11-6357-2014.html>

How does the analysis work?



$P(\text{Hazardous}) =$
frequency $env < thr$

RISK =
VULNERABILITY * $P(\text{Hazardous})$

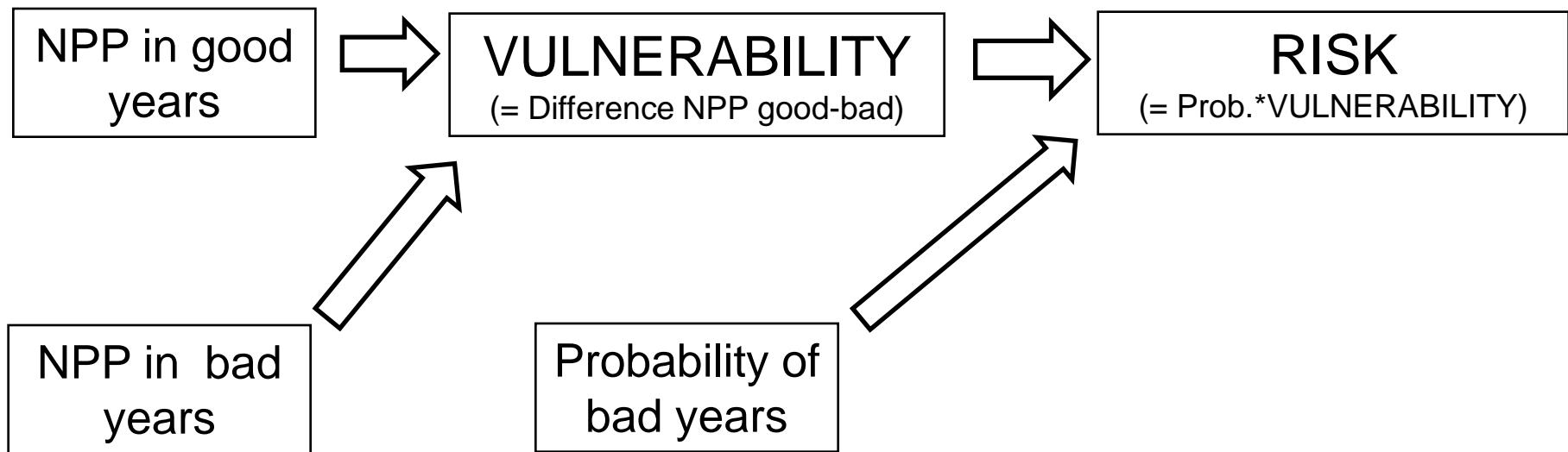
Risk analysis carried out in Carbo-Extreme

- Time periods: 1971-2000, 2071-2100
- sys: NPP, NEP, R_h , SWC, ET
- env: Drought (*)
- Models:

Group	Model	Ecosystem
PIK	LPJ	generic
LCSE	ORCHIDEE	generic
MPI	JSBACH	generic
CEH	BASFOR	forest
IIASA	EPIC	crops
INRA	PASIM	grassland

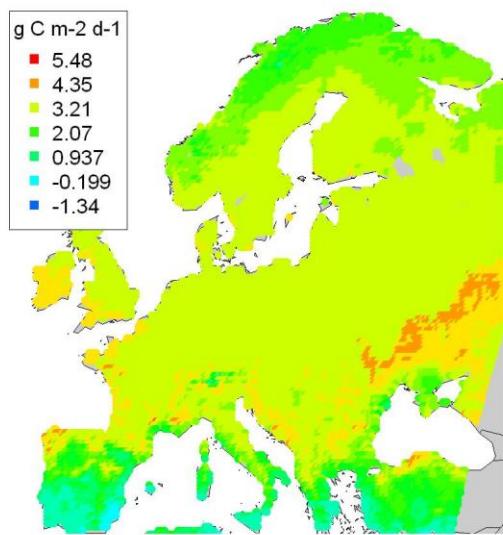
(*) Threshold: Standardized Precipitation Evapotranspiration Index (SPEI) < -1

Example: Risk analysis for NPP

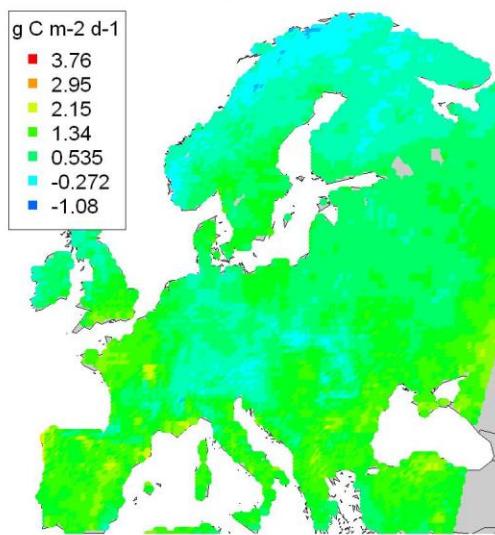


Risk analysis NPP (model PIK) 1971-2000

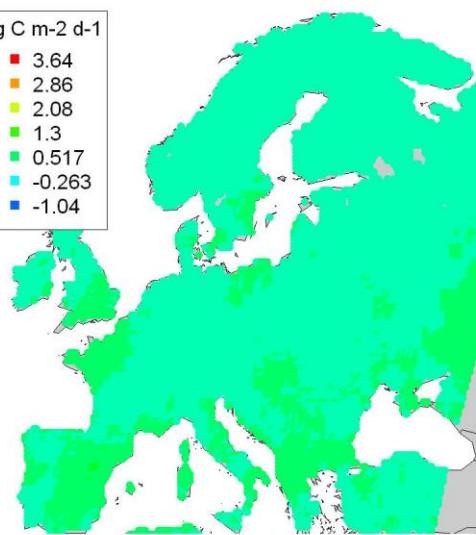
A: $E[NPP | SPEI > -1]$
1971-2000



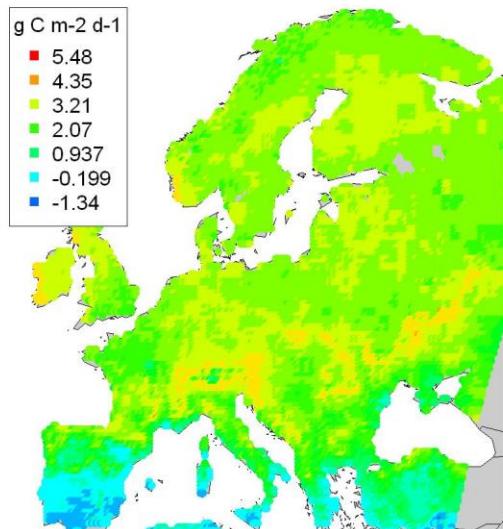
C: Vulnerability[NPP | SPEI < -1]
1971-2000



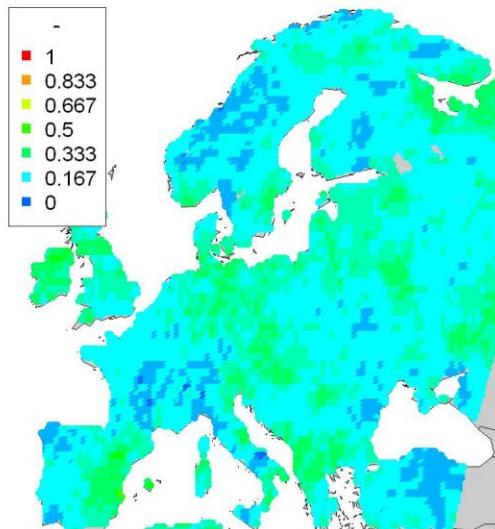
E: Risk[NPP | SPEI < -1]
1971-2000



B: $E[NPP | SPEI < -1]$
1971-2000

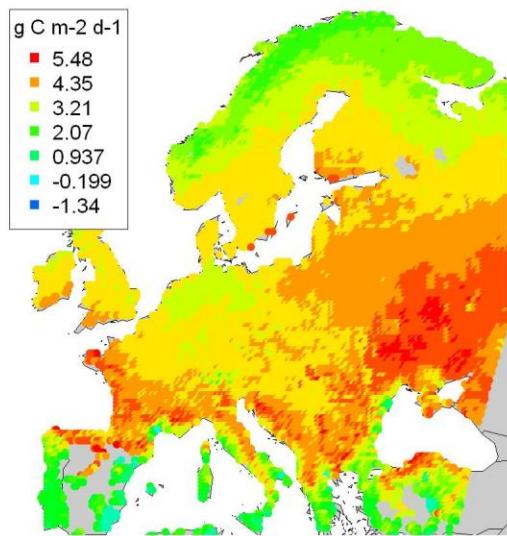


D: $P[SPEI < -1]$
1971-2000

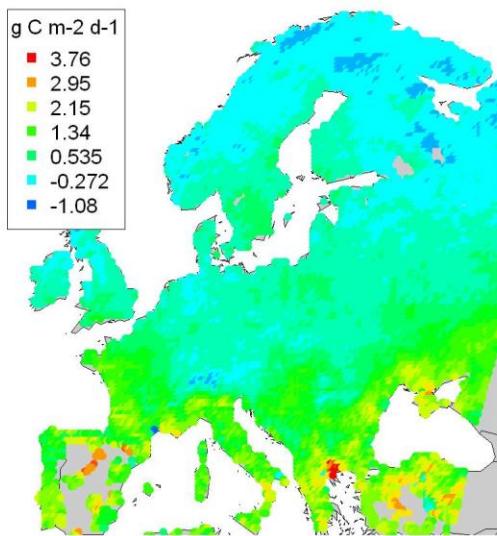


Risk analysis NPP (model PIK) 2071-2100

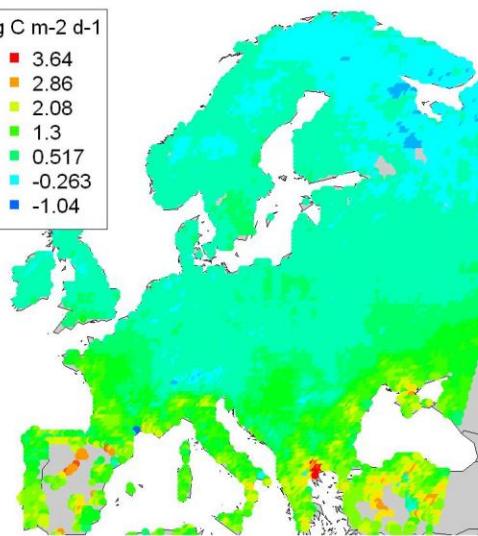
A: $E[NPP | SPEI > -1]$
2071-2100



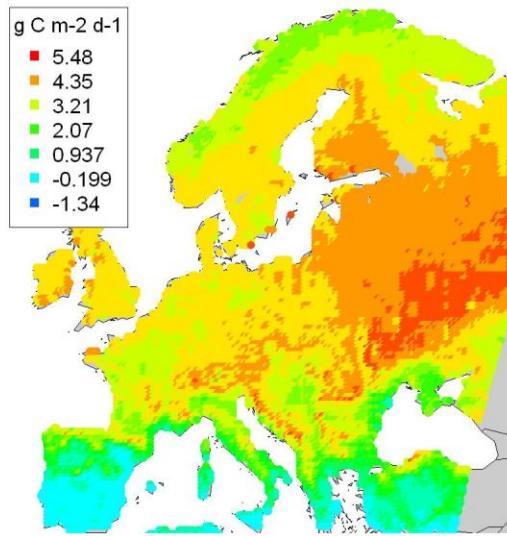
C: Vulnerability[NPP | SPEI < -1]
2071-2100



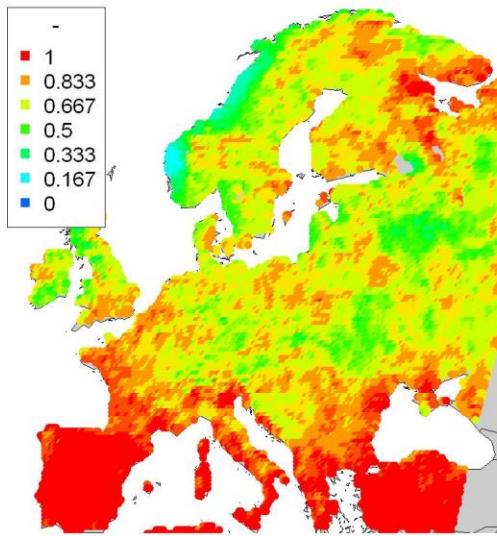
E: Risk[NPP | SPEI < -1]
2071-2100



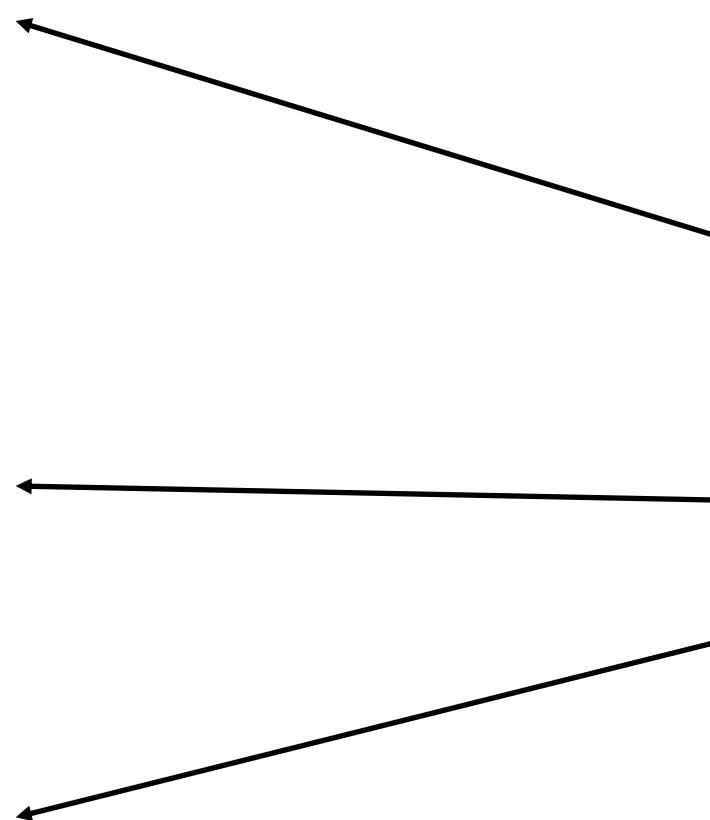
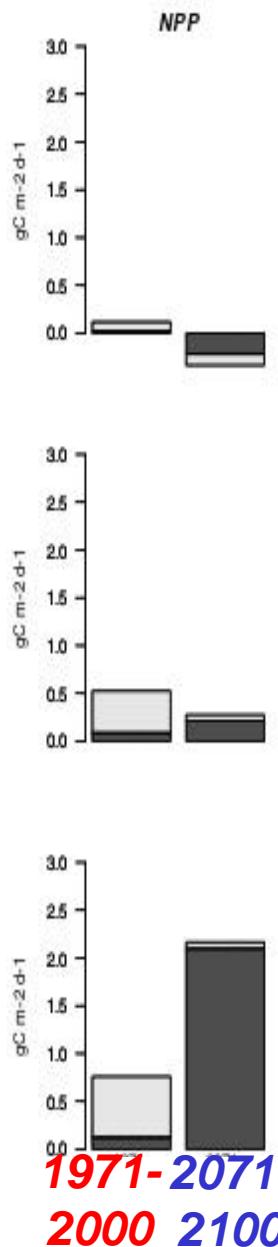
B: $E[NPP | SPEI < -1]$
2071-2100



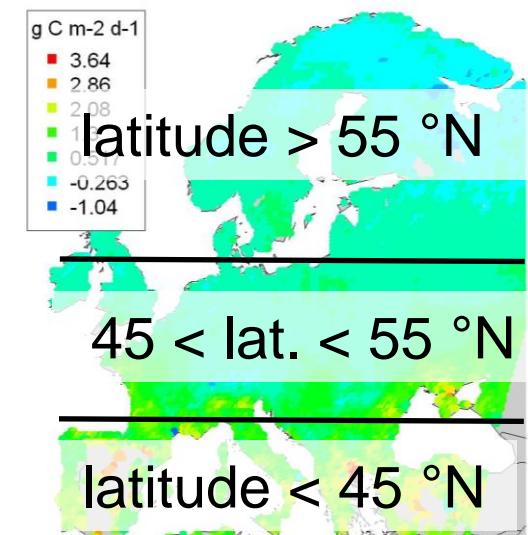
D: $P[SPEI < -1]$
2071-2100



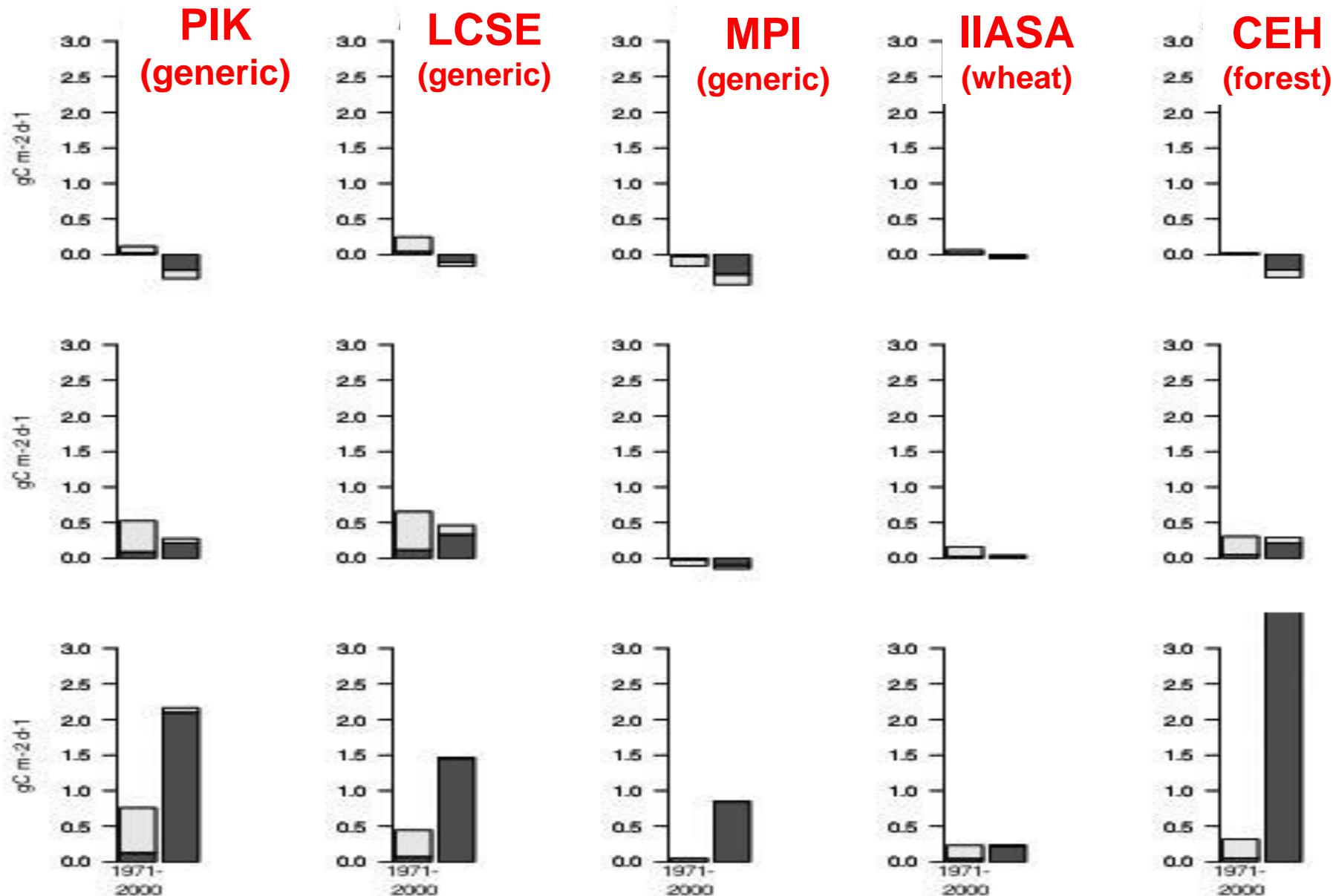
Risk analysis NPP (model PIK) per latitudinal band



E: Risk[NPP | SPEI < -1]
2071-2100



Risk analysis NPP: model comparison



Risk analysis: Discussion

- Definitions of risk (R), vulnerability (V) and probability of hazard ($P(H)$) allow decomposition:
 $R = V * P(H)$
- Analysis method is applicable to data as well as to model outputs
- For drought, both V and R expected to increase in the south, but large model (vegetation) differences