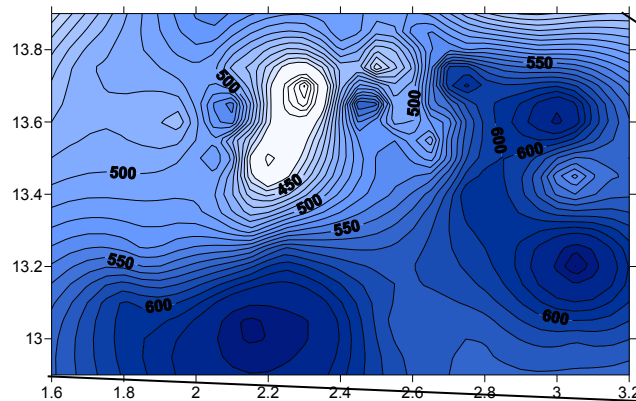
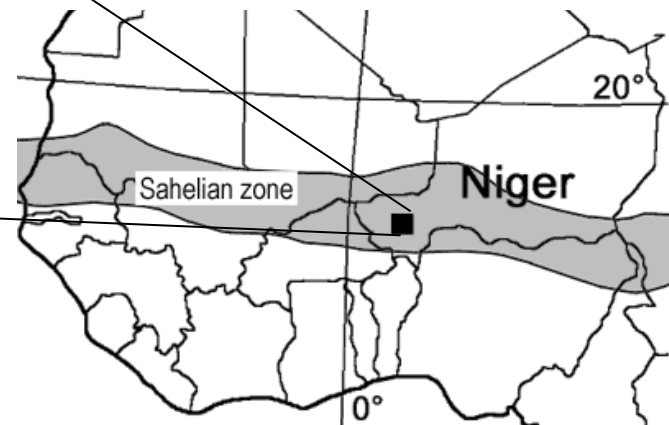


How to best use point rainfall data to drive land surface models in the Sahel ?

Feurer D., Cappelaere B., Demarty J., Vischel T., Ottlé C., Solignac P.A., Saux-Picart S., Lebel T., Ramier D., Boulain N., Charvet G., I. Mainassara, Chazarin J.P., Oi M., Quantin G.



2006 (AMMA SOP)



RBV
RÉSEAU DES BASSINS VERSANTS



LSM resolution?

- Scale of interest for **outputs**:

local/regional? daily/seasonal ?

depend of the modeling objectives (hydrology, agronomy, climatic, ...)

- Characteristic scales of **processes** ?

- Forcing processes: space-time **intermittency of convective rainfall** at small scales
- Surface processes: small-scale **endoreism**, strong **non-linearity**

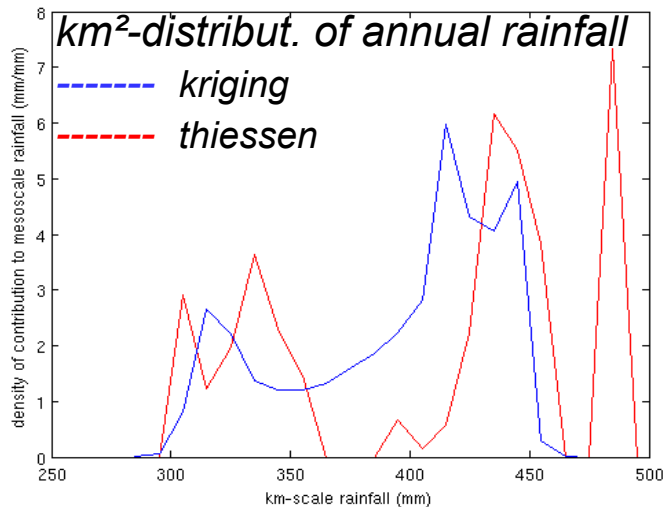
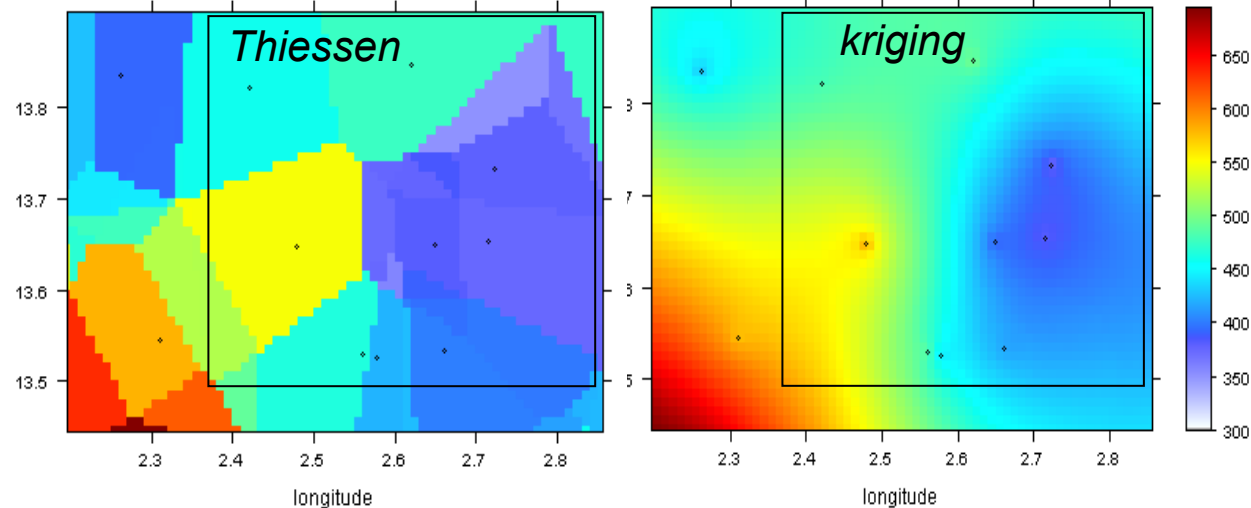
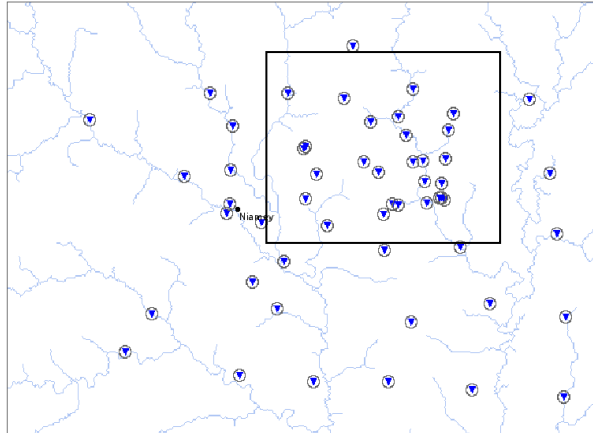
➔ **<hourly, ~km resolution** is needed to fully resolve surface processes in the Sahel (*eg., Vischel & Lebel, JH-2007*)

➔ **HR-raingauge networks** still needed despite loose space-sampling

➔➔ " *How to best use point rainfall data to drive land surface models in the Sahel ?*

Variation in annual rainfield with interpolation scheme

Amma-Catch Niger (ACN) dense



==> **Need for :**

- Assessing **performances of standard interpolation schemes** with respect to land surface applications in the Sahel context
- Evaluating **benefits from newly proposed schemes** for this region

Materials & Methods

Site : $\sim 3.10^3 \text{ km}^2$ of ACN mesosite (SW Niger) (*Cappelaere et al., JH-2009*)

Period : **2005** growing season (+ 2006 subsequent dry season)

Application model : **SETHYS** land surface model (*Saux-Picart et al., JH-2009*)
calibrated with ACN data

Rainfall interpolation schemes :

- 1 - nearest neighbor (« **Thiessen** »)
- 2 - « **standard** » (eulerian) **kriging**
- 3 - **dynamic** (lagrangian) **kriging** (*Vischel et al., JHM-2012*)
- 4 - **stochastic** (ensemble) rainfield **modeling** (*Vischel et al., JH-2009*)

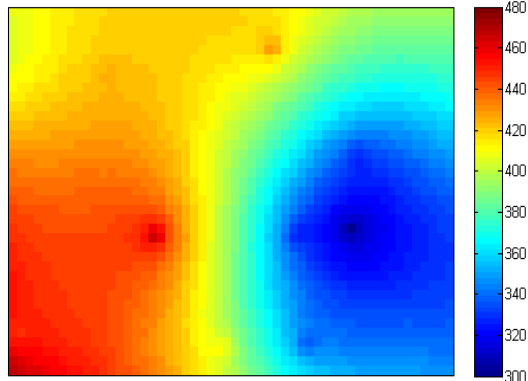
Some specific questions

- ➔ « General-purpose » deterministic method(s) ok? which ?
- ➔ Ensemble simulation necessary? To reflect uncertainty? Other?
- ➔ Or else could a single space-time rainfield do the job?

Distributed land surface modelling

Rainfields

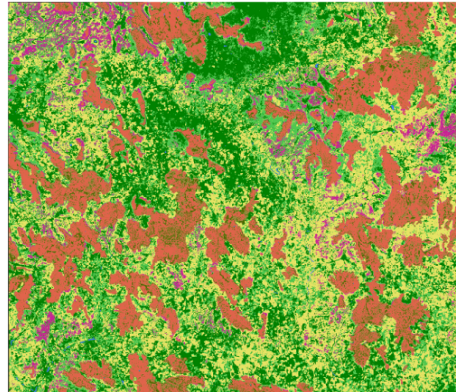
Ex : Thiessen, Std. Kriging, Dyn. kriging,
Ensemble simulations



Vischel *et al.*, 2009, 2012

Land-use

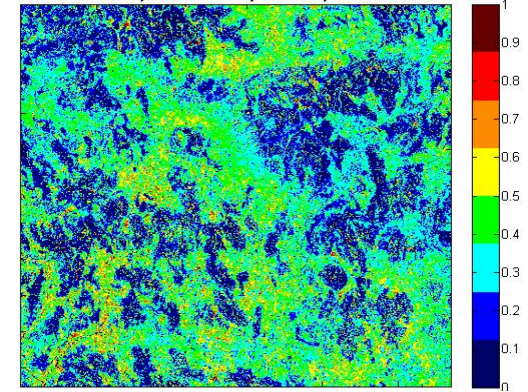
SPOT, 8 classes, 20m, 2005



Saux-Picart *et al.*, 2009

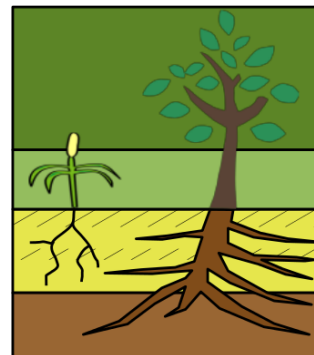
LAI

SPOT, 6 dates, 20m, 2005



Saux-Picart *et al.*, 2009

**Impacts of rainfield estimation
methods ?**



Uniform meteorological forcing

SetHyS Savannah LSM

(dX = 1 km , dT = 5 mn)

➔ Poster « Solignac *et al.* » (9B11) for model description

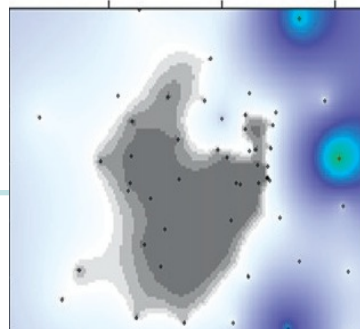
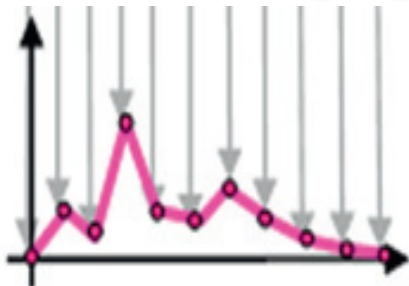
3 deterministic approaches of estimation of rainfields

Thiessen:



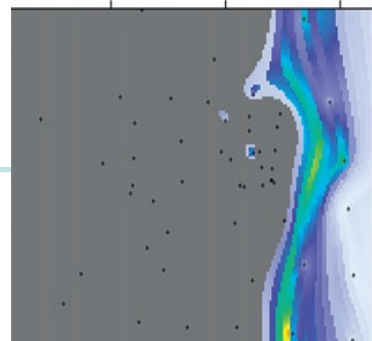
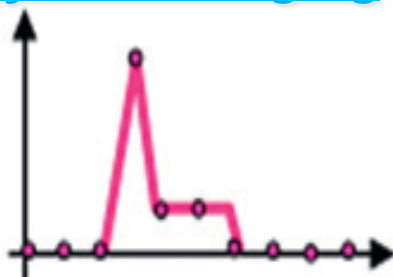
Nearest neighbor approach
Transpose point to 1km² rainfall

Standard kriging



Interpolation time step by time step (variogram)

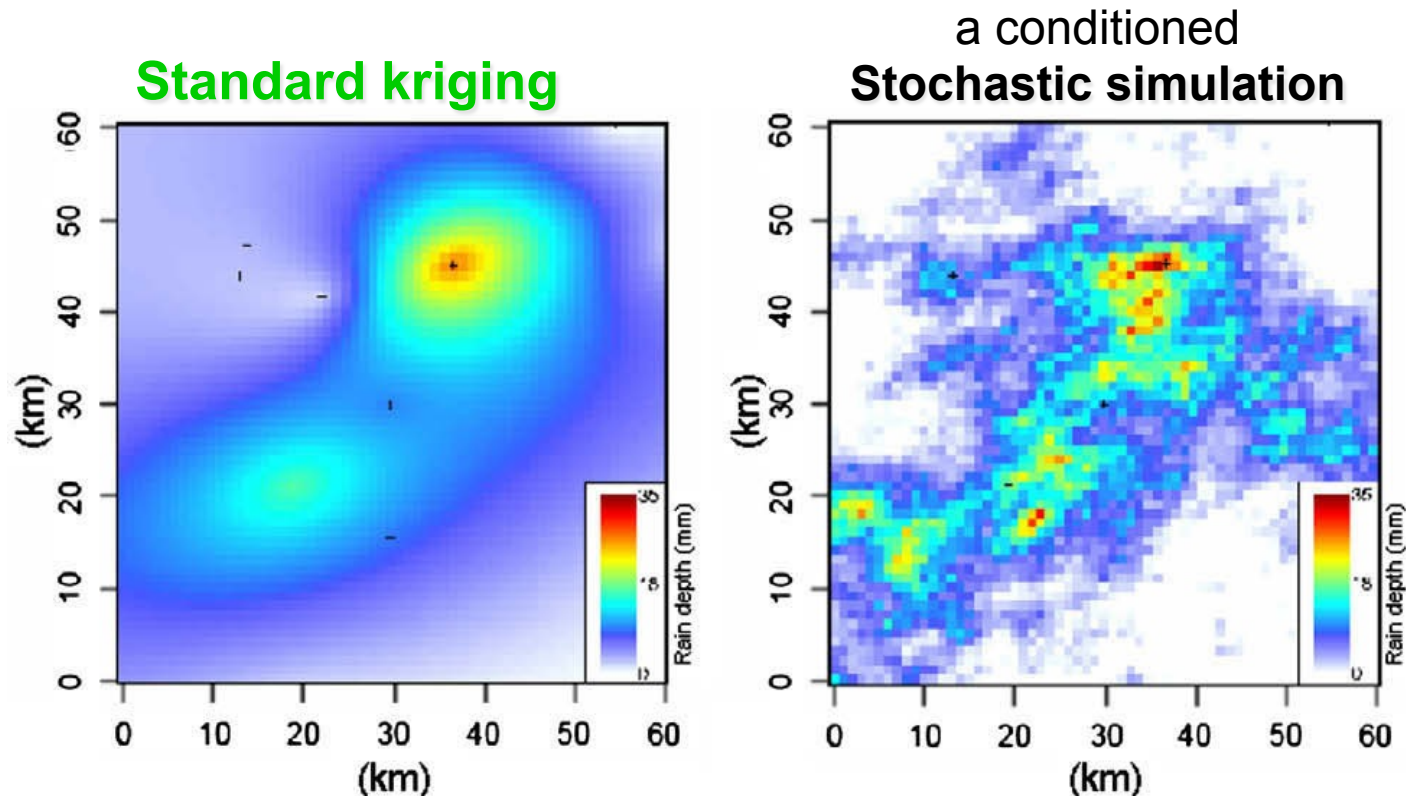
Dynamic kriging



Dynamic interpolation, superpose
hyetograph in a virtual temporal
space

Better deterministic approach

+ 1 conditioned stochastic approach of event rainfields (*Vischel et al., JH-2009*)



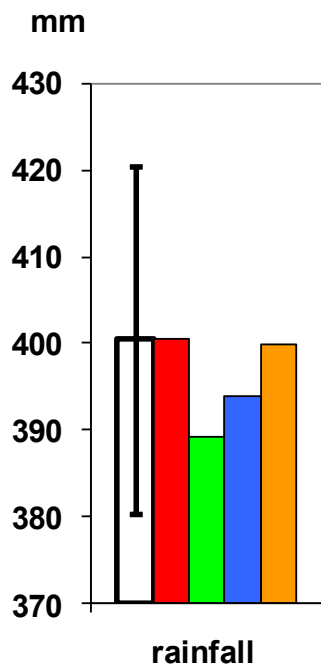
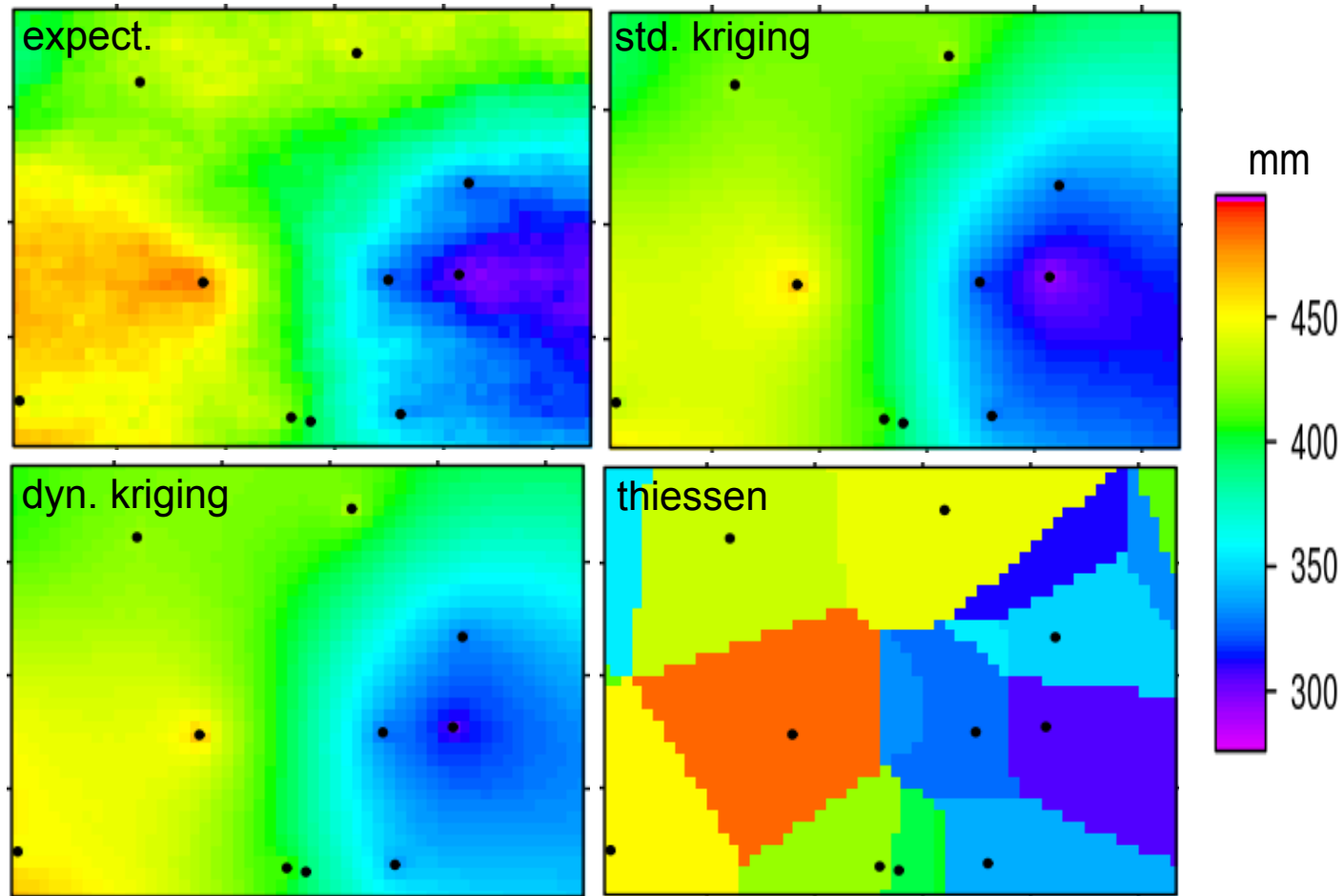
→ Poster « *Vischel et al.* » (9B9)

Application

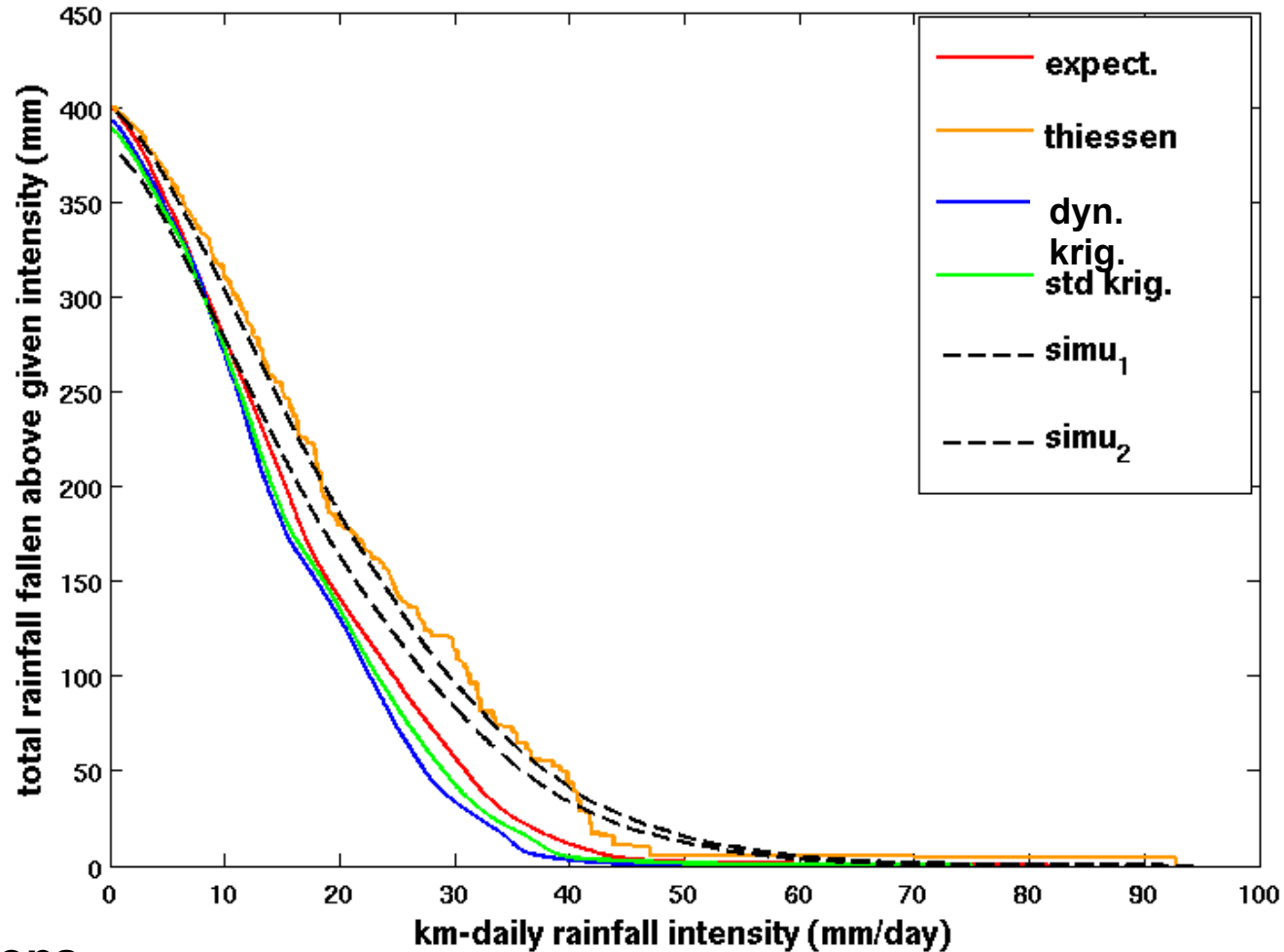
- 50-member ensemble simulation (event resolution)
- Dynamical kriging-based time disaggregation (event to 5-min)
- simulation averages used as reference for comparisons here

2005 –aggregated rainfields (start. June 15)

- ensemble mean
- expected rainfall
- std. kriging
- dyn. kriging
- Thiessen



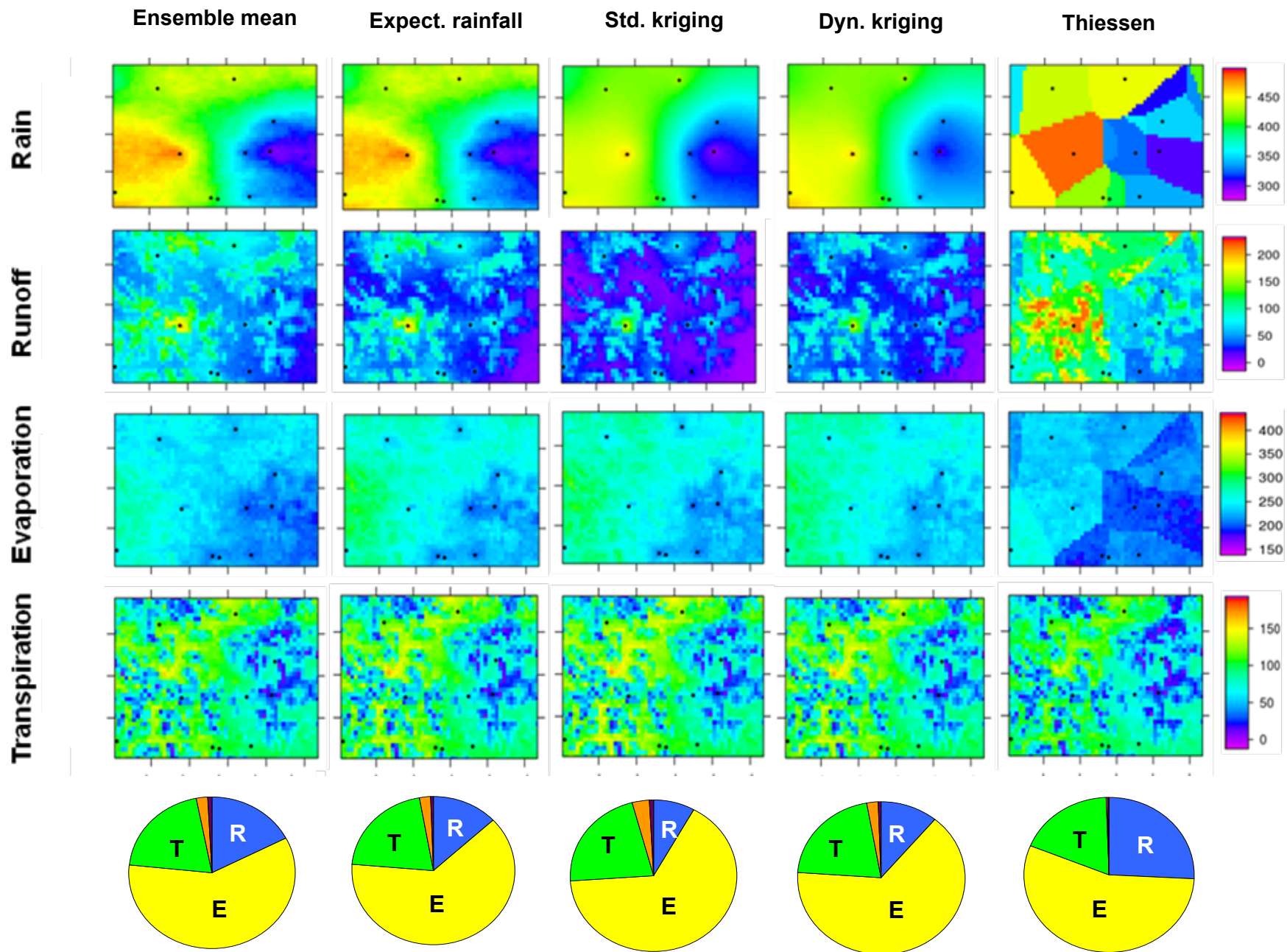
km-day intensity distribution of total rainfall



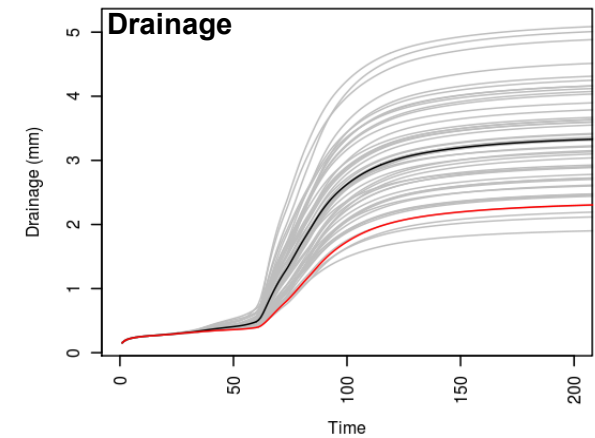
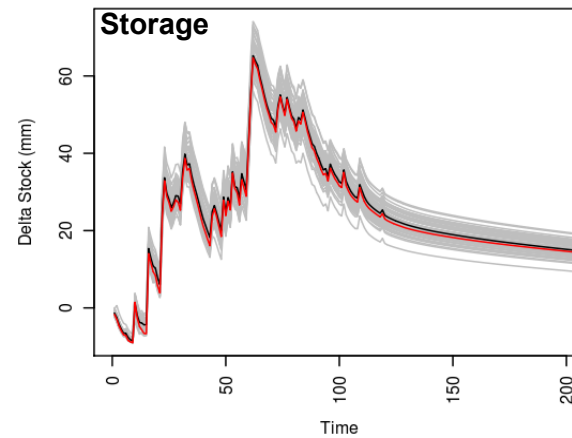
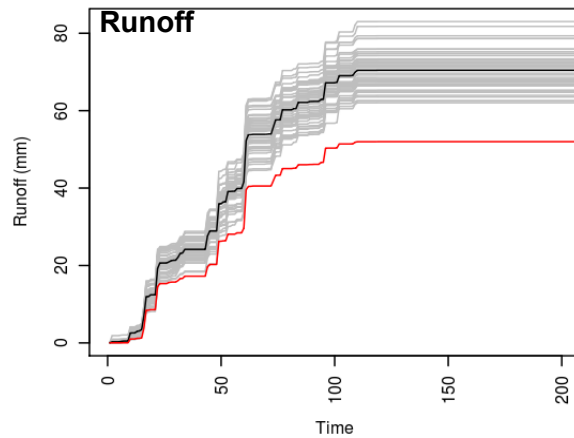
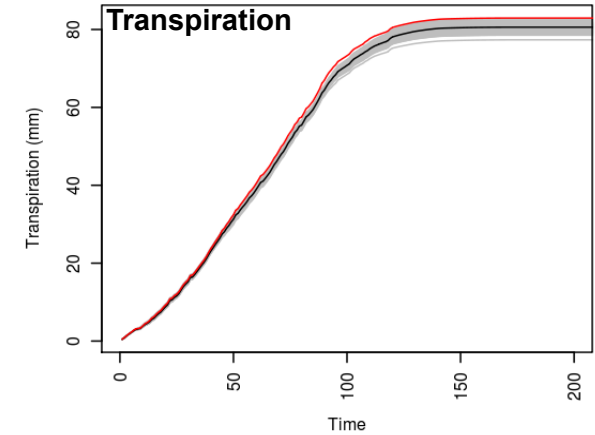
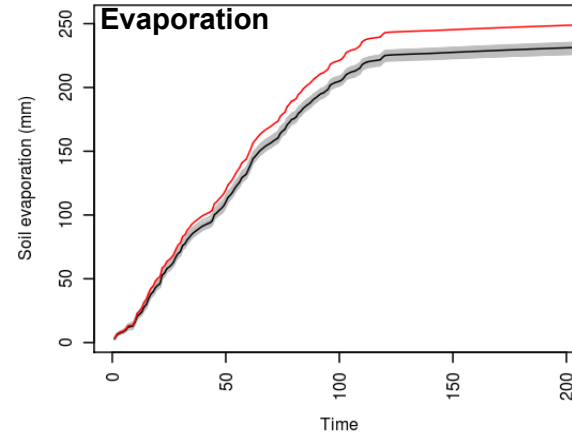
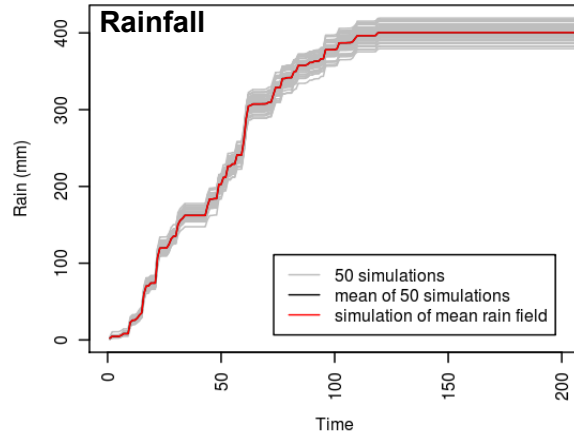
Conclusions

- Thiessen, stochastic : Maximise high/low intensities
- Kriging, expected rainfall : Minimize high/low intensities
- Great impacts on hydrological processes

Simulated hydrological variables



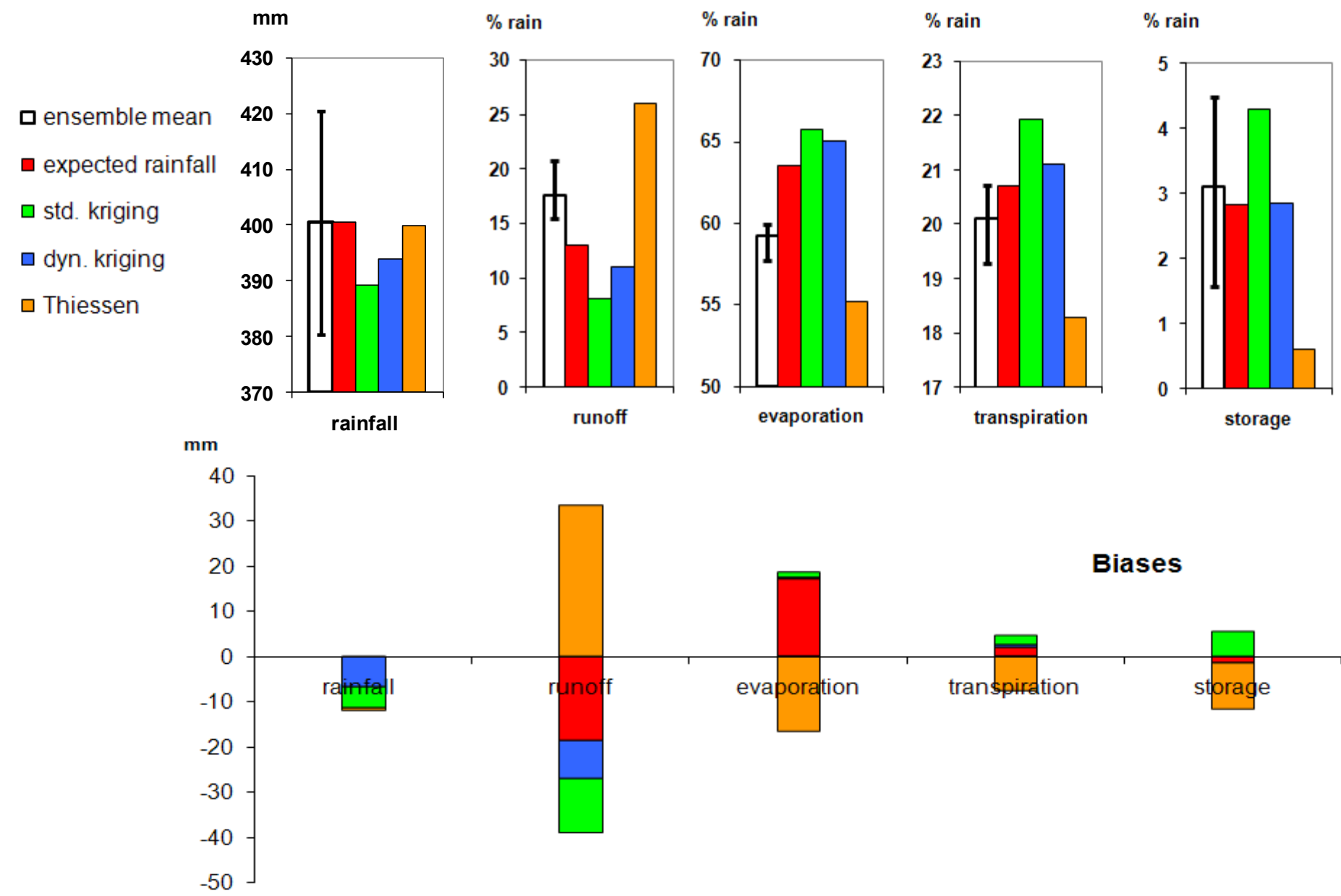
Cumulative mesoscale hydro. variables for ensemble & expected rainfall



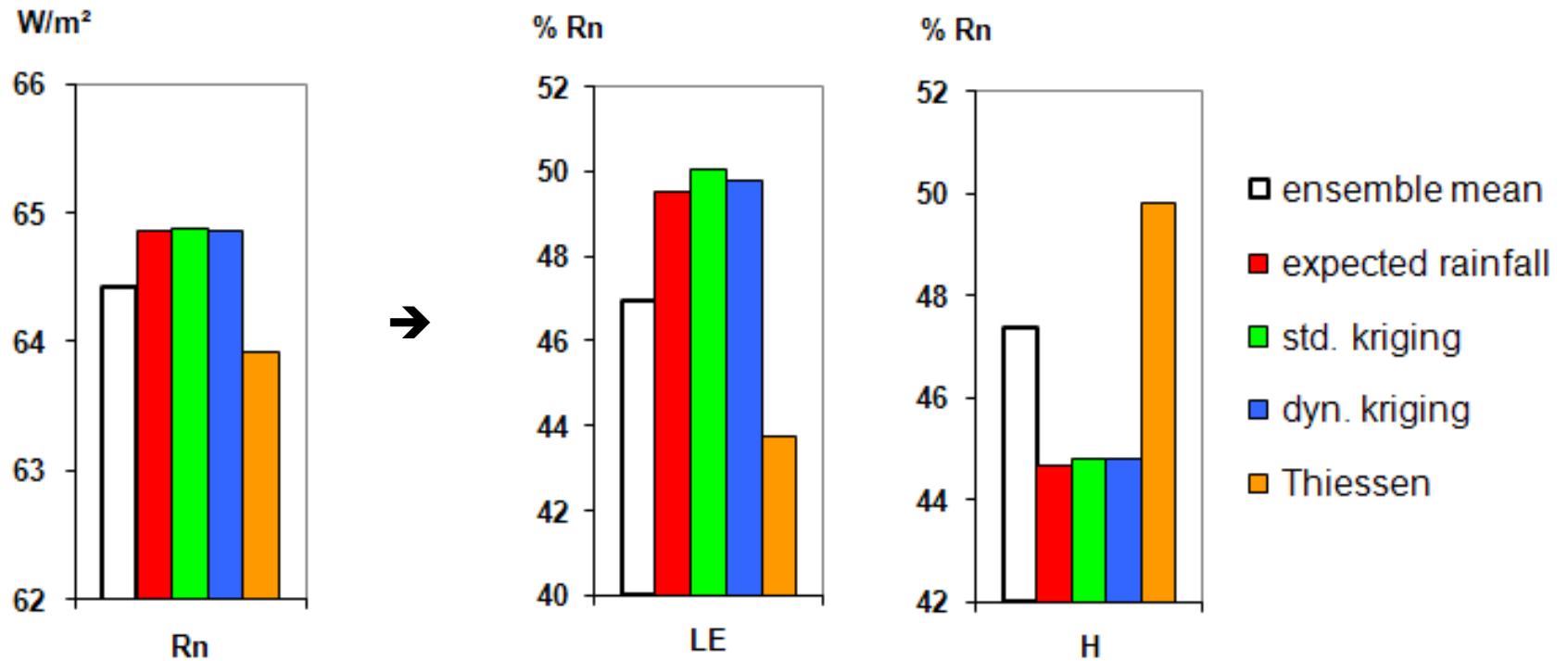
➔ **Uncertainty propagation** : % amplified for runoff & storage, dampened for evaporation

➔ **Bias generation** with expect. rainfall: larger than uncertainty for runoff and evaporation

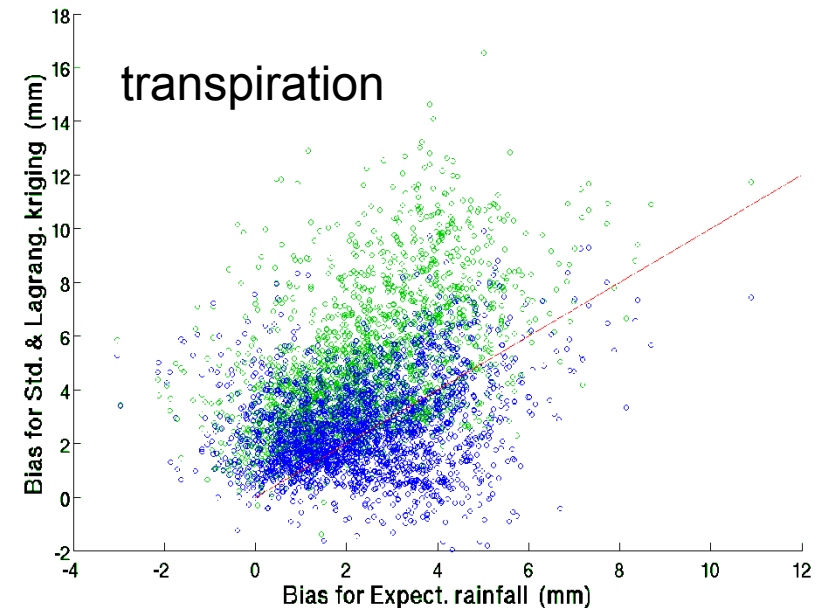
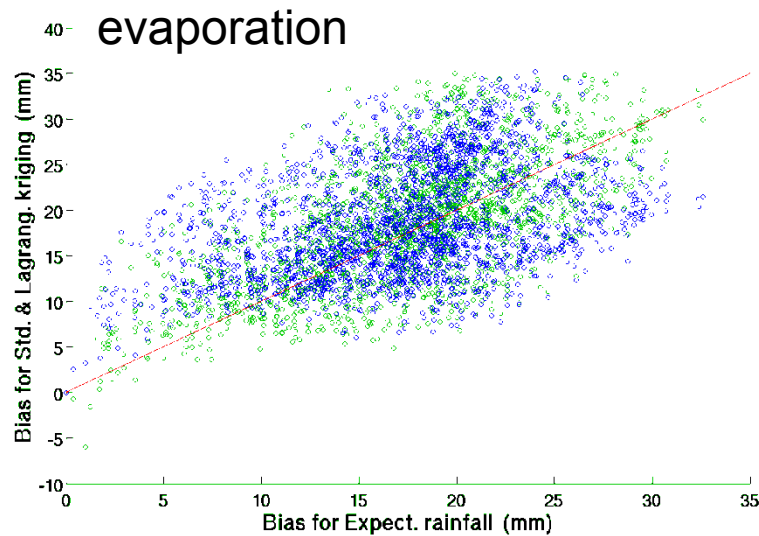
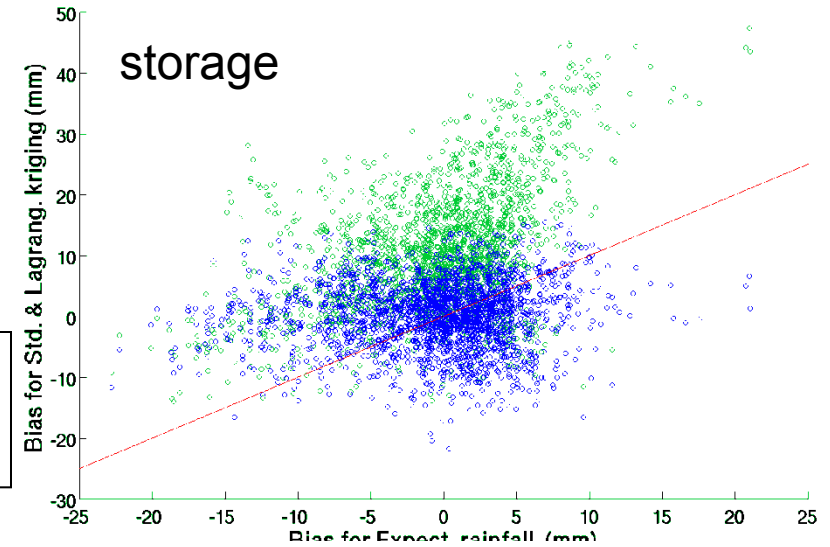
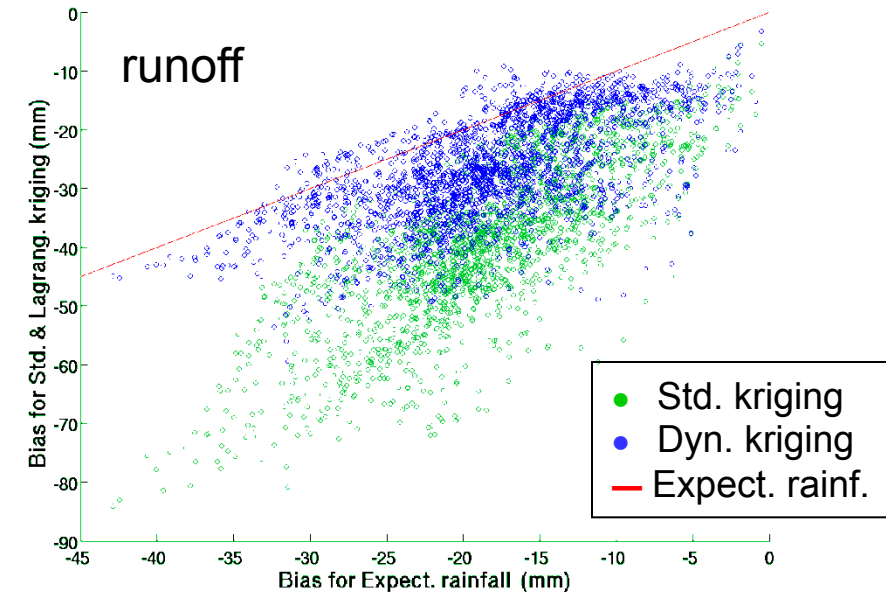
Season / mesoscale hydro. variables for 4 methods



Same for turbulent fluxes in energy budget



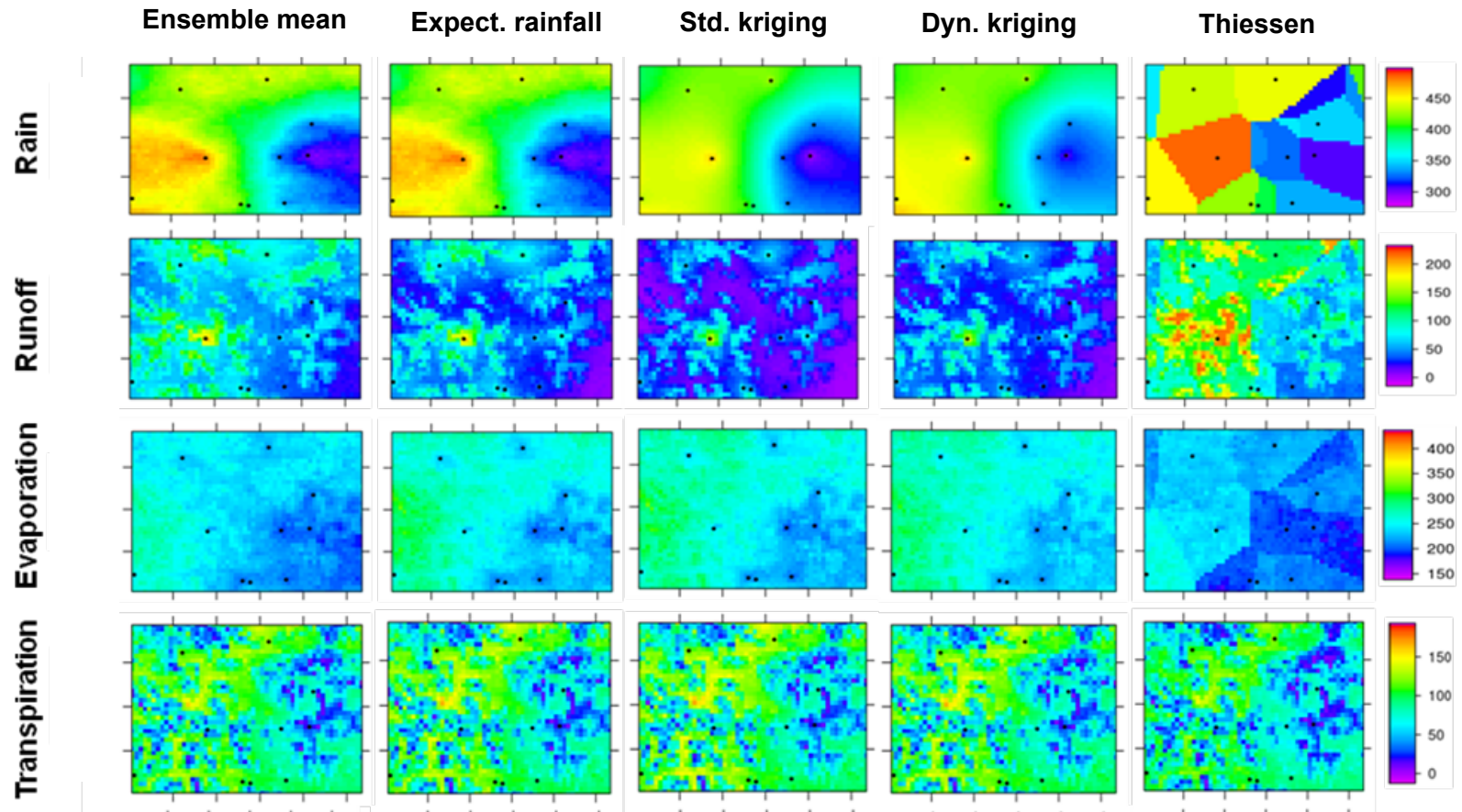
Bias from kriging methods vs. from expect. rainfall (year x km² scale)

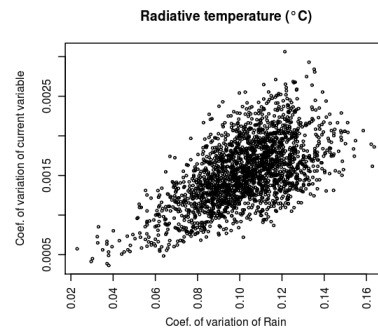
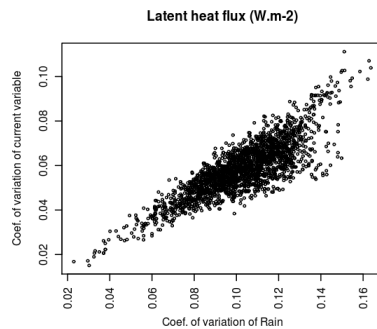
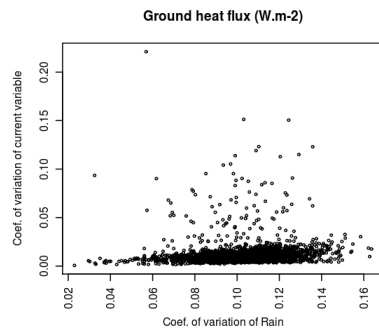
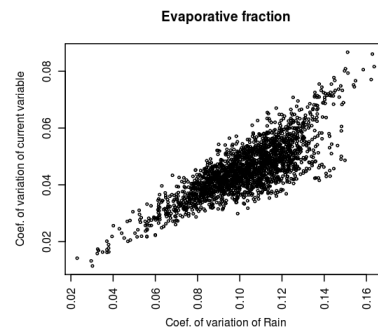
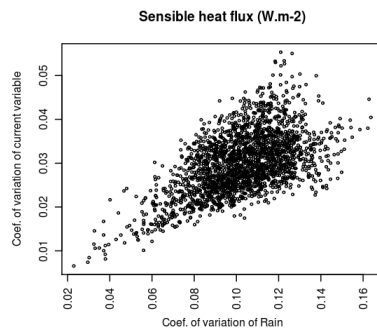
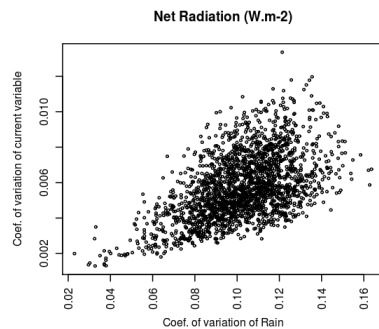
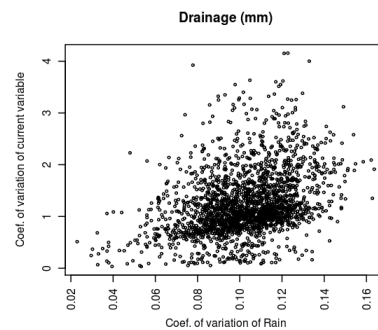
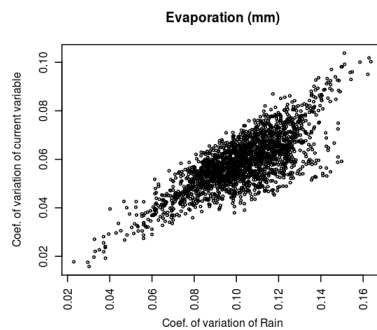
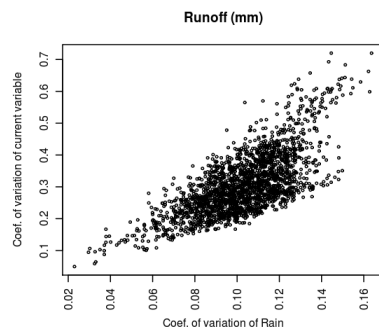
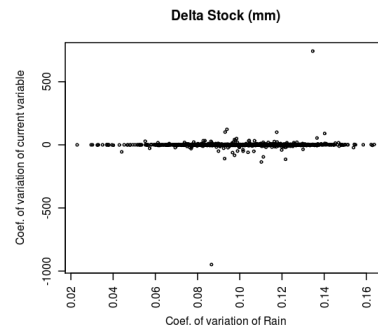
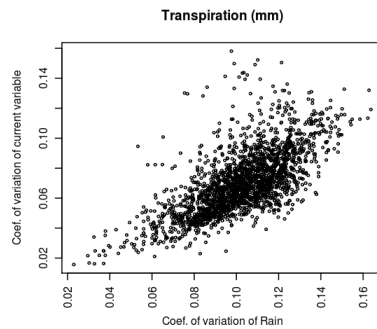
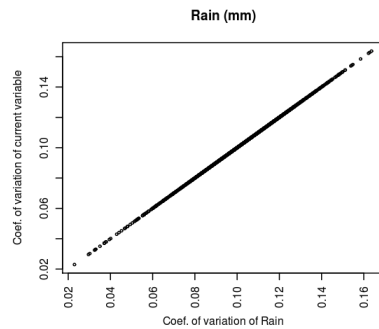


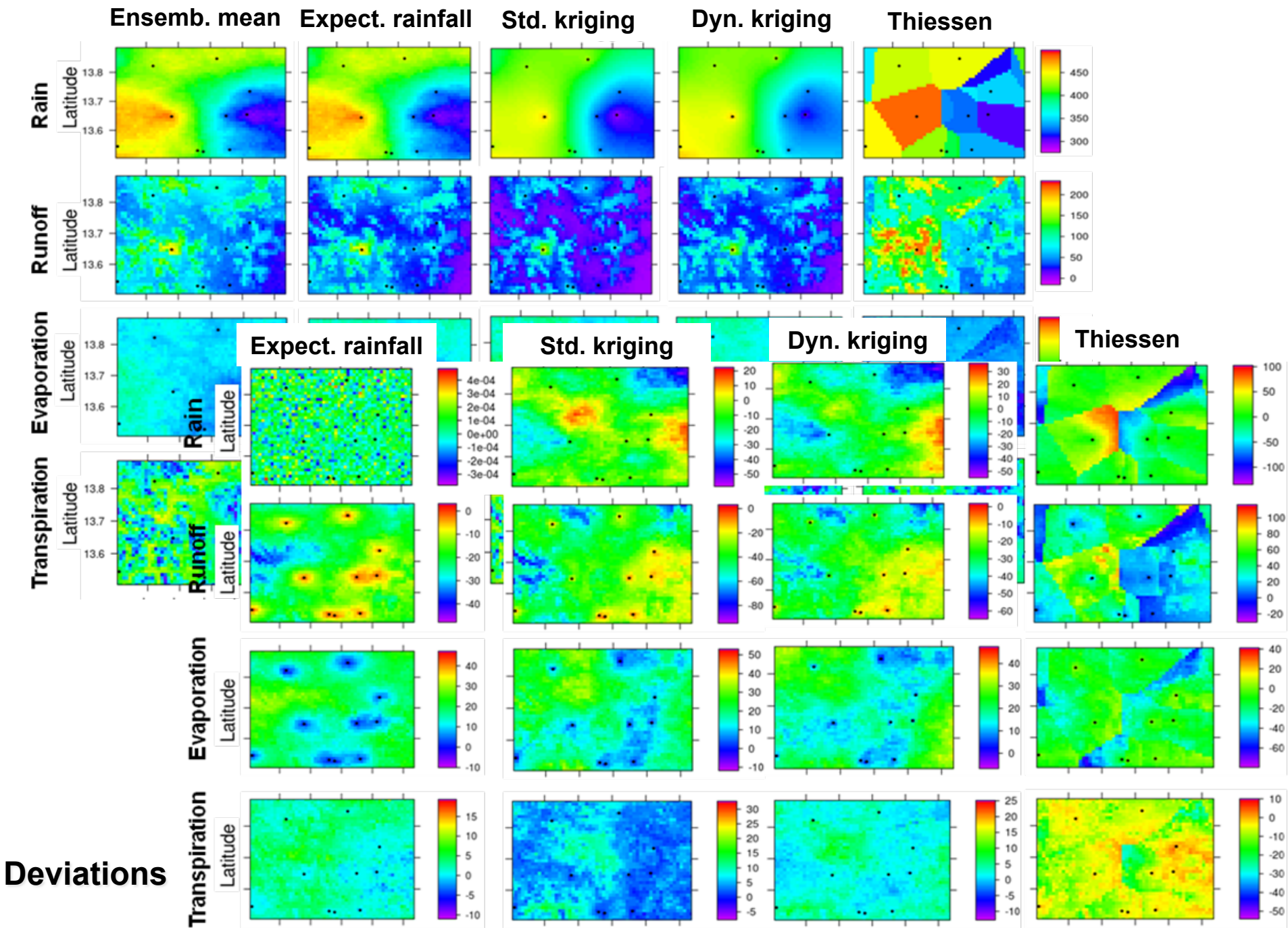
Tentative conclusions

- Runoff most sensitive to rainfall uncertainty,
 - 5-25% rain in 2005 over the mesosite
 - compensates ~evenly between evaporation, transpiration, deep storage
- Rainfall uncertainty generates biases on all hydrological variables with single rainfield methods (incl. expected rainfield)
- Improvement by dynamic kriging over standard kriging, coming close to expected rainfield
- Single stochastic rainfield better than any deterministic method (for mesoscale outputs) ?

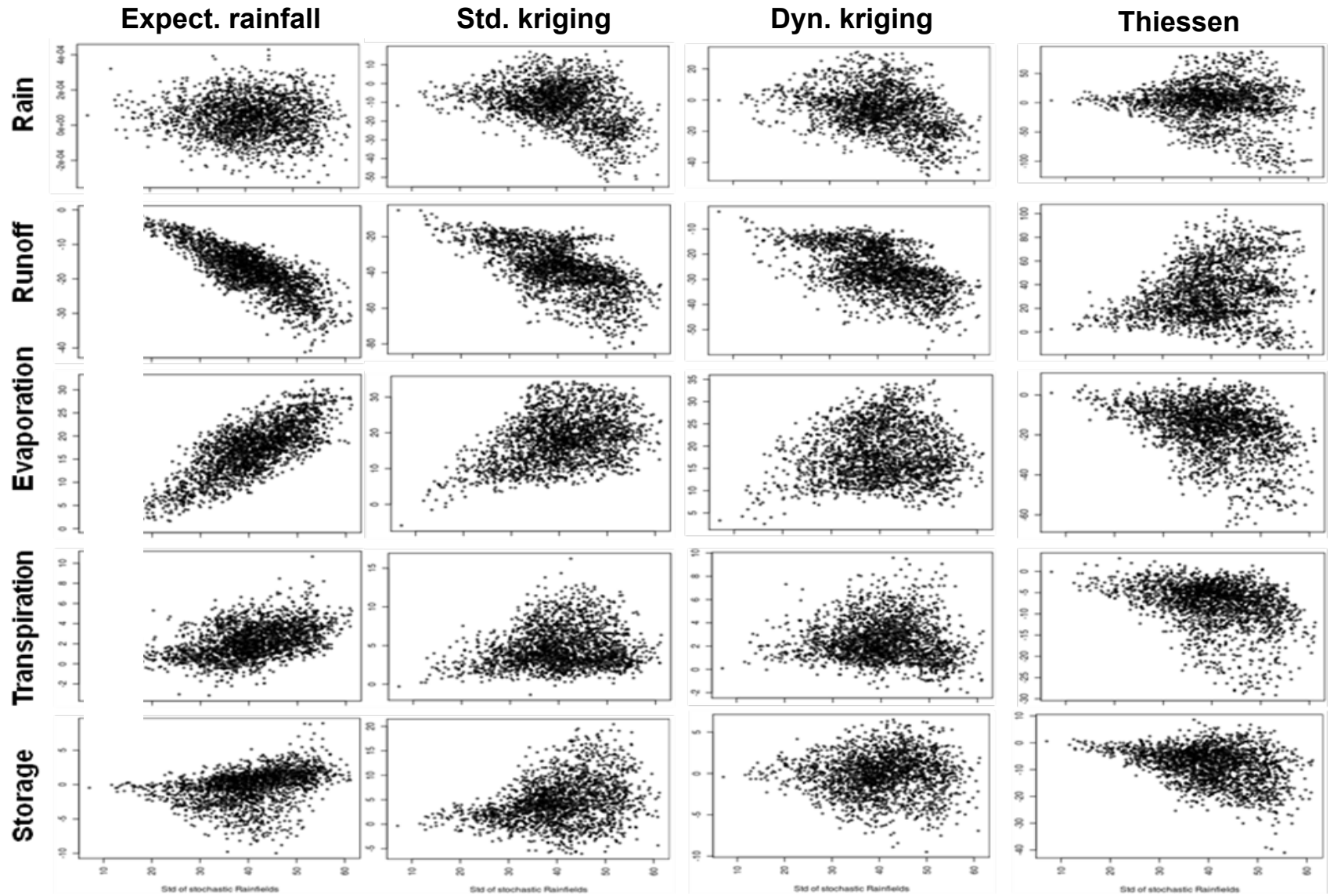
Simulated hydro variables







Bias vs. Uncertainty (season x km² scale)



The End