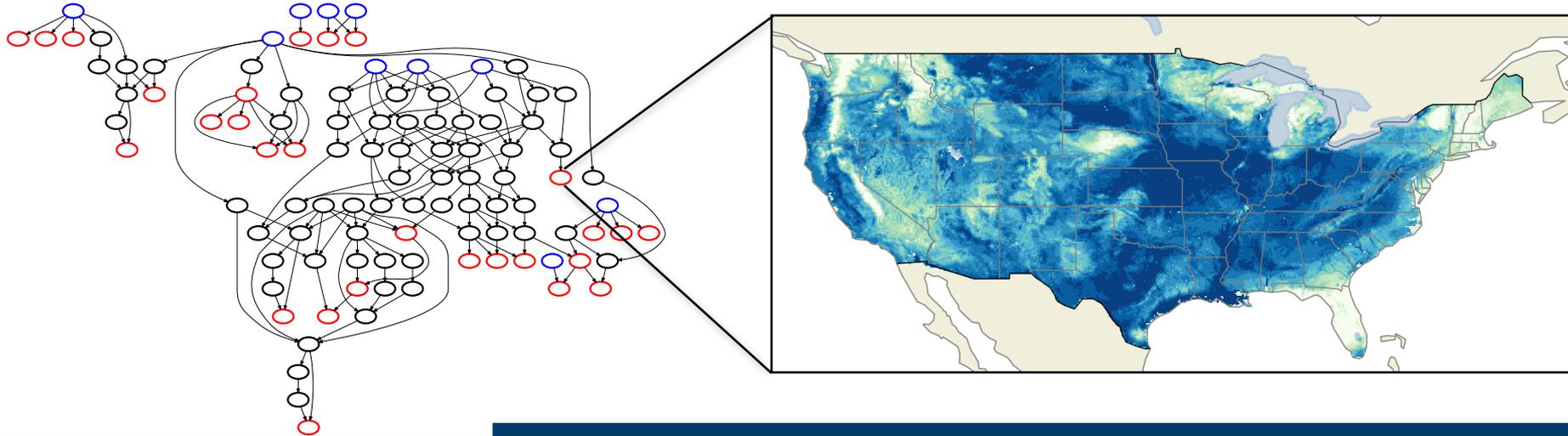


# Development of a stand-alone Multiscale Parameter Regionalization (MPR) tool for the estimation of effective model parameters for any distributed model



R. Schweppe, S. Thober, S. Attinger and L. Samaniego

EGU2019-7478 - April 12th 2019



HELMHOLTZ  
CENTRE FOR  
ENVIRONMENTAL  
RESEARCH - UFZ

# Estimating parameters for a distributed model



Van Looy, 2017 (Rev. o. Geo.)

$$O_{it} = f(I_{it}, \beta_{it})$$

$$\#\beta_i = \#i$$

$f(\cdot)$ : process representation

$I$ : process input

$\beta$ : process parameters

$O$ : process output

$i$ : cell index

$t$ : time index

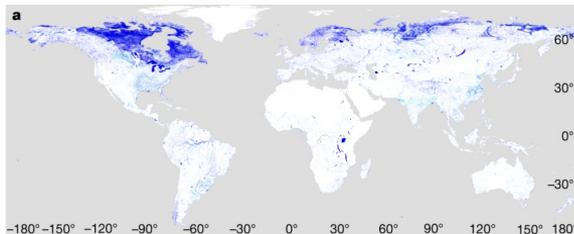
# High-resolution data



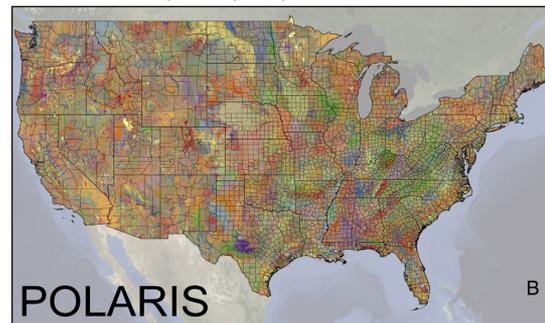
ASTER (NASA, 2001)  
elevation at 30m



MODIS (Friedl, 2019)  
land cover, LAI, VI,... at 500m

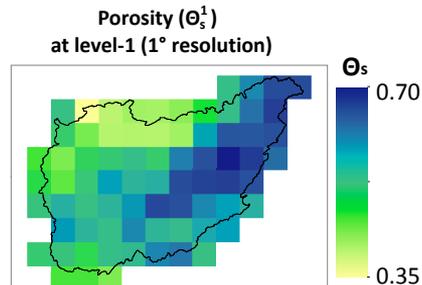
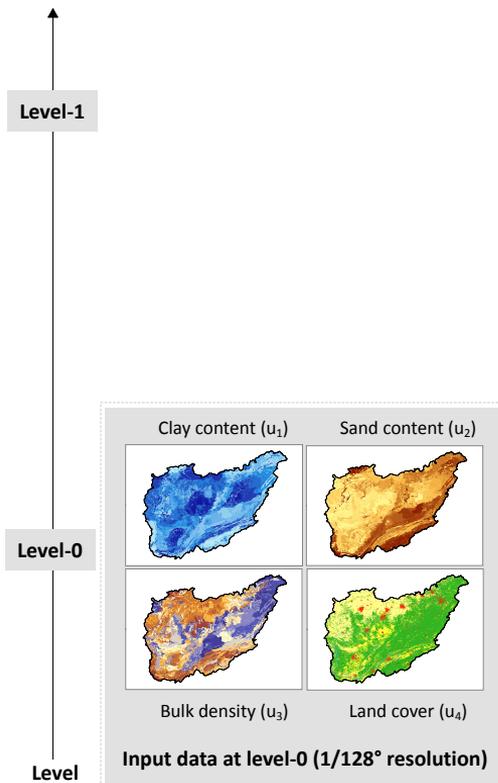


JRC GSW (Pekel, 2016 Nat)  
lake alimetry at 30m



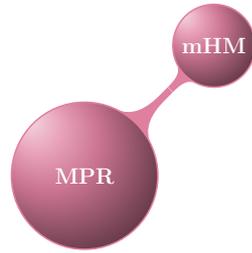
POLARIS (Chaney, 2016 Geoderma)  
soil at 30m

# Multiscale Parameter Regionalization (MPR)

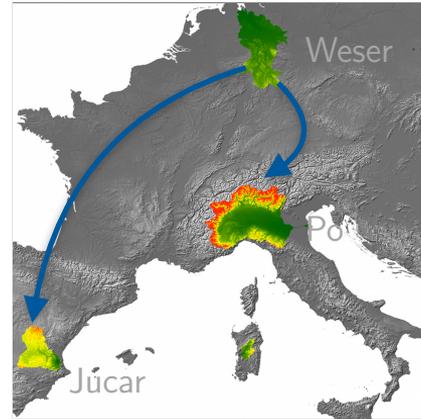


- steps:
1. transfer function
  2. upscaling

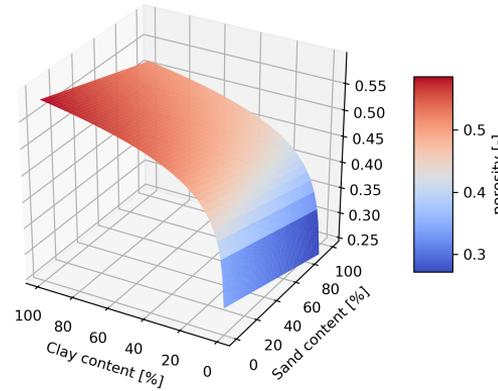
# MPR key advantages



Samaniego, 2010 (WRR)



adapted from Samaniego



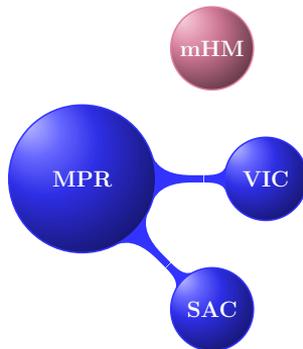
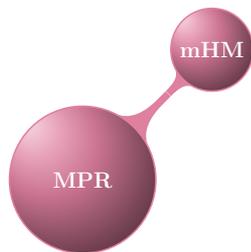
$$O_{it} = f(I_{it}, \beta_i)$$

$$\#\beta_i = \#i$$

$$\beta_i = f_u \left( f_t \left( P_{kj}, \gamma \right) \right)$$

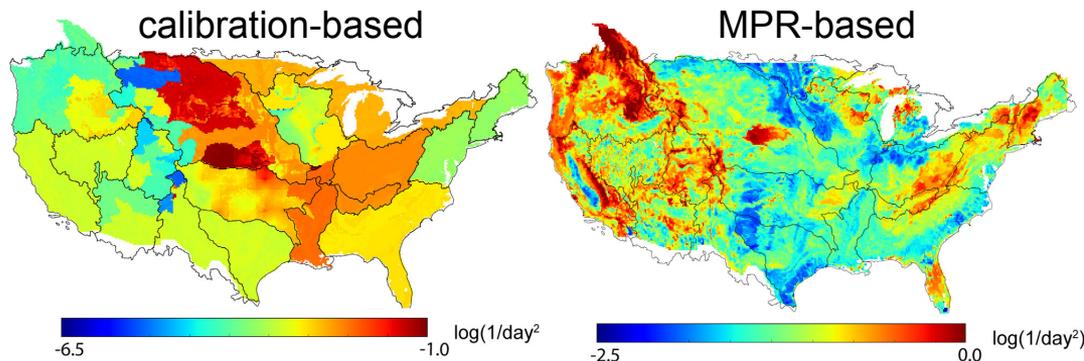
$$\#\gamma = 3$$

# MPR key advantages

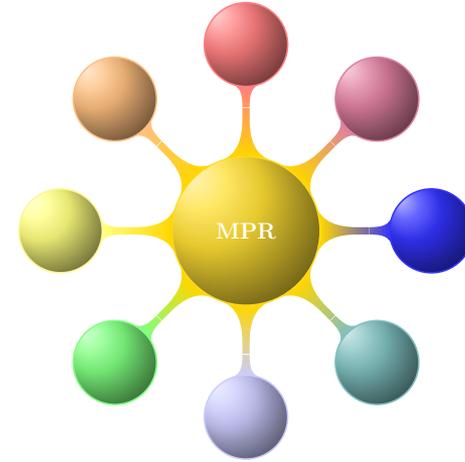
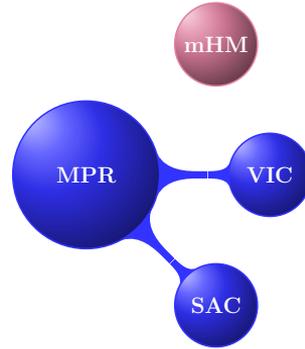
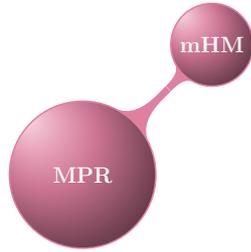


Mizukami, 2017 (WRR)

- Regularisation of parameter space at input data resolution
- Transferability across scales and location
- Seamless parameter fields



# new MPR software



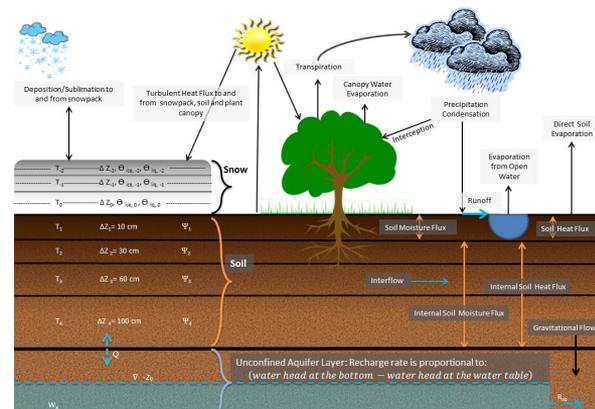
Schwepe, 2019 (in preparation)

*What is the effect of using different transfer functions on model behaviour?*

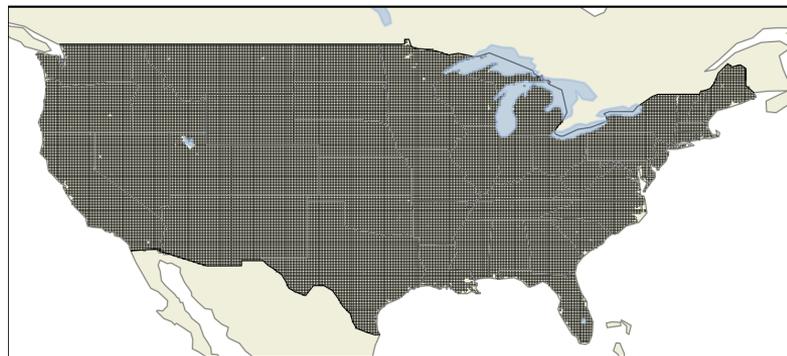
*How can MPR be implemented?*

# Land-Surface Model Noah-MP

- Part of WRF HYDRO framework
- Used in operational NOAA National Water Model
- Richards' (1931) equation & Campbell (1974) parameterization

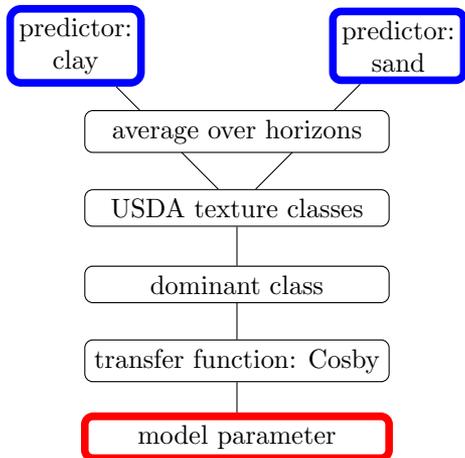
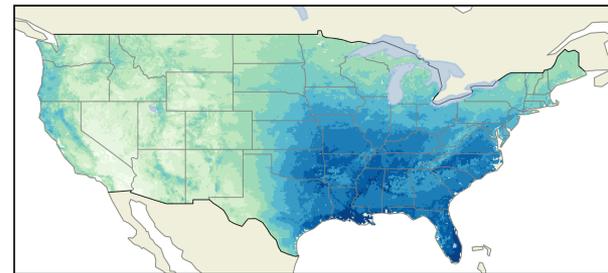
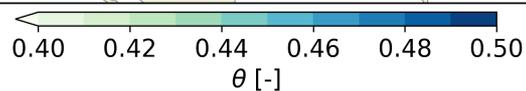
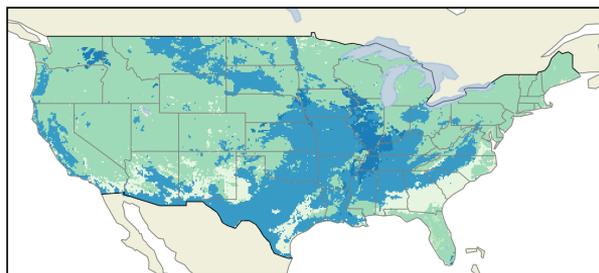
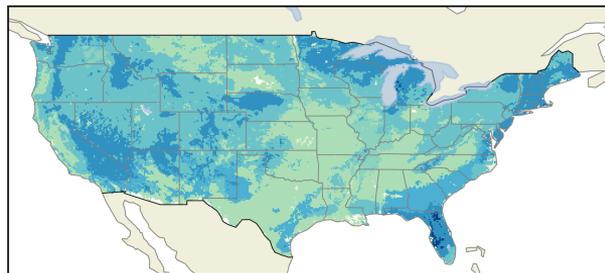


University of Texas at Austin

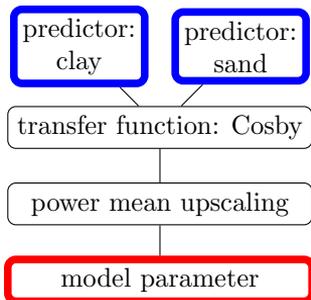
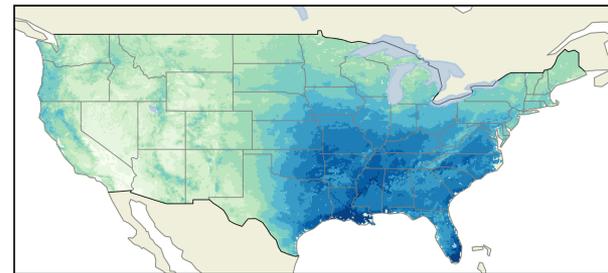
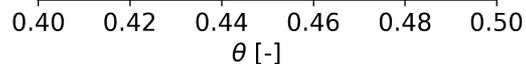
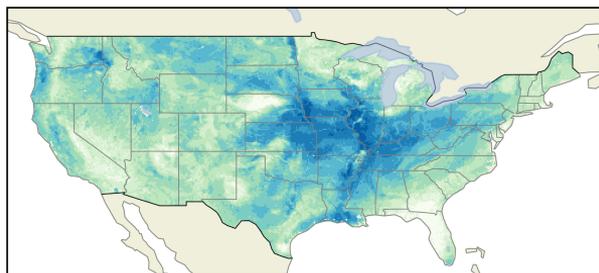
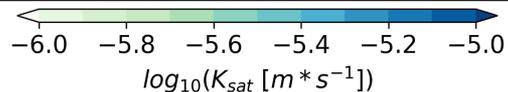
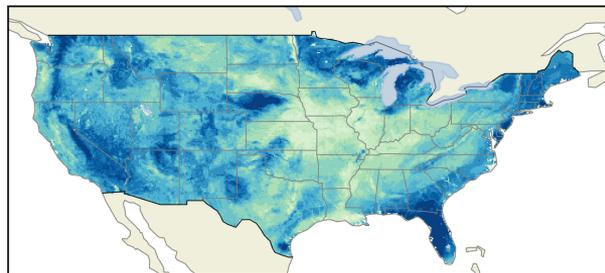


- NLDAS2 forcing
- Hourly time step,  $1/8^\circ$  resolution

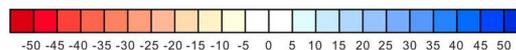
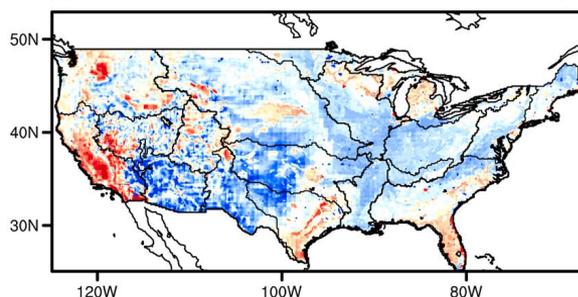
# Application with Noah-MP



# Application with Noah-MP

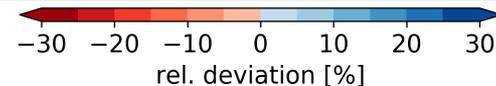
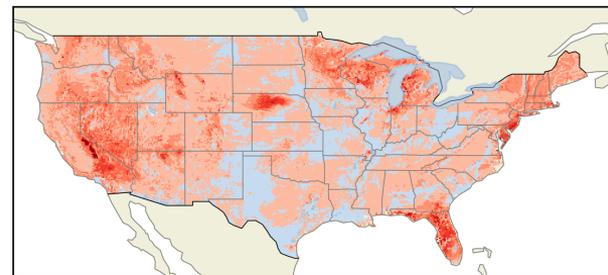


(b) Relative bias of ET (%)

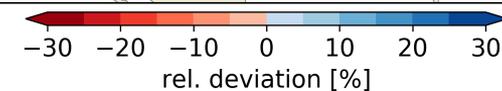
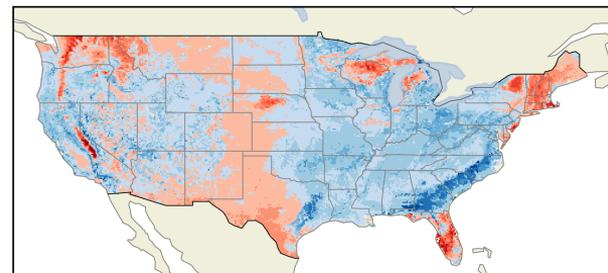
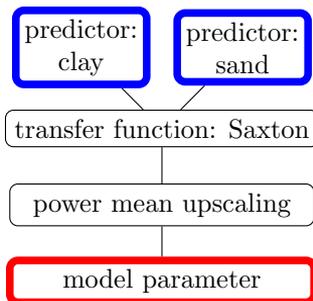
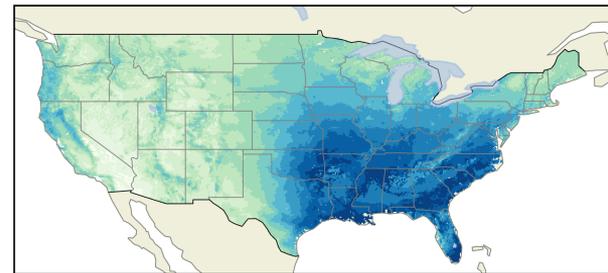
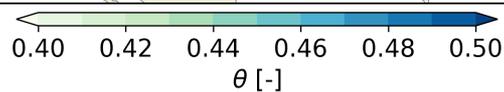
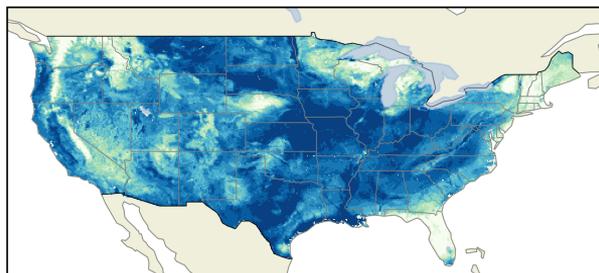
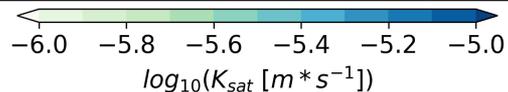
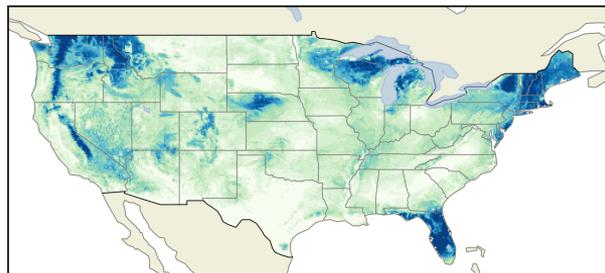


Ma, 2017 (JGRA)

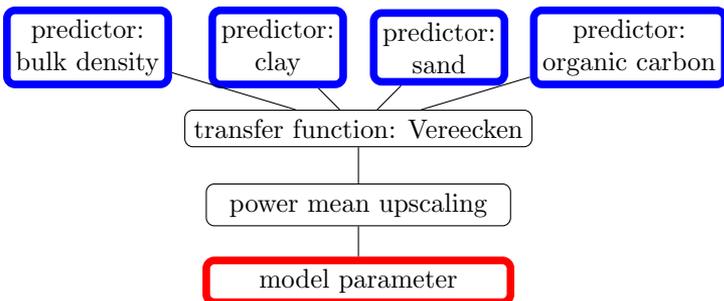
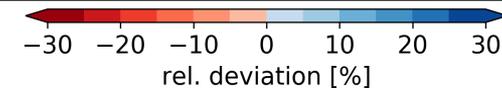
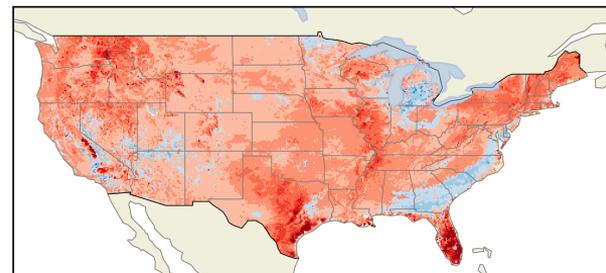
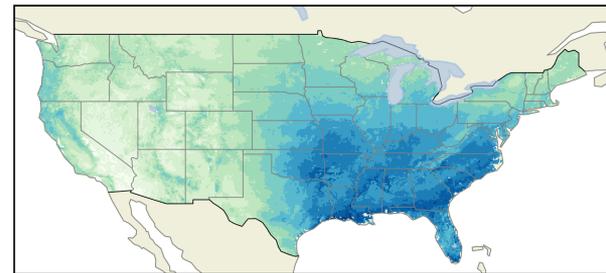
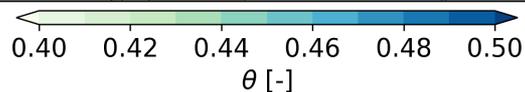
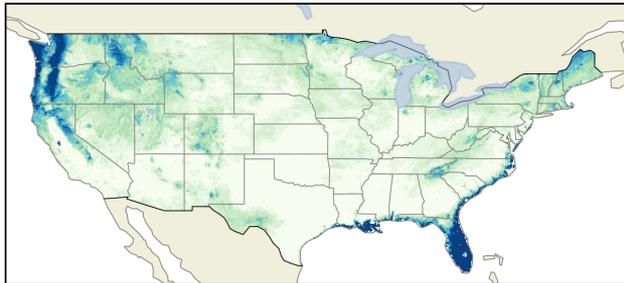
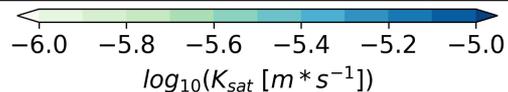
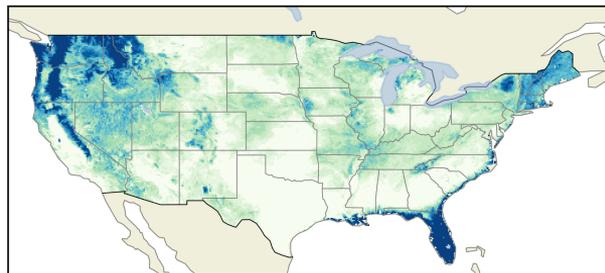
Noah-MP ET ( $mm yr^{-1}$ )



# Application with Noah-MP

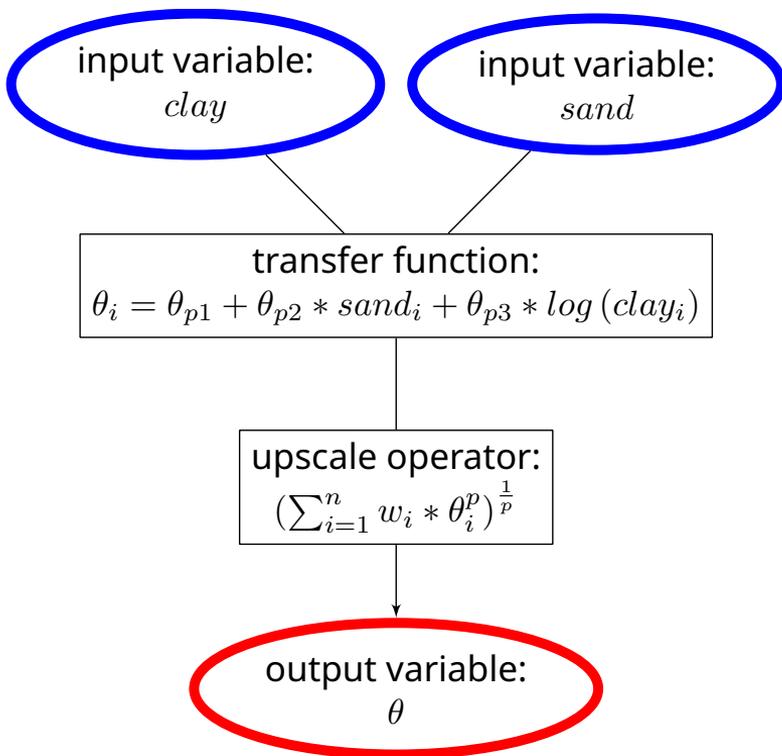


# Application with Noah-MP



Variable	2000-50 sand	50-2 silt	Carbon content, %	Bulk density, $g\ cm^{-2}$
<b>Maximum value</b>	97.80	80.70	6.60	1.230
<b>Minimum value</b>	5.60	0.00	0.01	1.040

# MPR configuration is simple



what you want...

```

&mainconfig
out_filename = "MyParams.nc"
dim_name_alias(:,1) = "x_in", "x_out"
dim_name_alias(:,2) = "y_in", "y_out"
dim_name_alias(:,3) = "z_in", "z_out"
/

&Data_Arrays
names(1) = "clay"
from_file(1) = "PathTo/MyNetcdfFile.nc"
names(2) = "sand"
from_file(2) = "PathTo/MyNetcdfFile.nc"
names(3) = "theta"
transfer_funcs(3) = "theta_p1 + theta_p2 * sand + theta_p3 *
log(clay)"
from_data_arrays(1:2,3) = "sand", "clay"
target_dim_names(1:3,3) = "z_out", "y_out", "x_out"
upscale_ops(1:3,4) = "1.0", "1.0", "1.0"
/

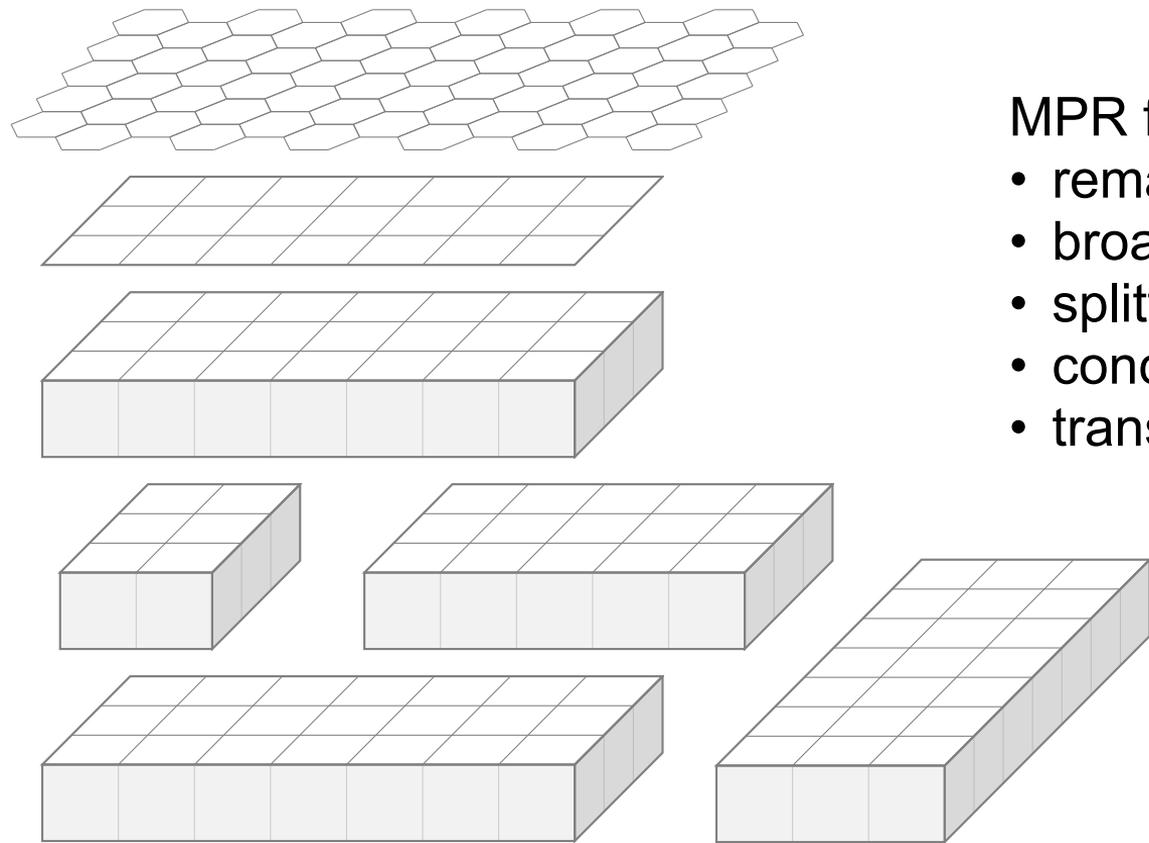
&Dimensions
dim_names(1:3) = "x_out", "y_out", "z_out"
dim_reference(1:3) = "center", "center", "end"
dim_step(1:2) = 0.125, 0.125
dim_bound(3) = 0.0
dim_vector(:,3) = 0.1, 0.4, 1.0, 2.0
/

&Parameters
parameter_names(1:3) = "theta_p1", "theta_p2", "theta_p3"
parameter_values(1:3) = 0.505, -0.00142, -0.00037
/
    
```

what you type...



# MPR configuration is flexible

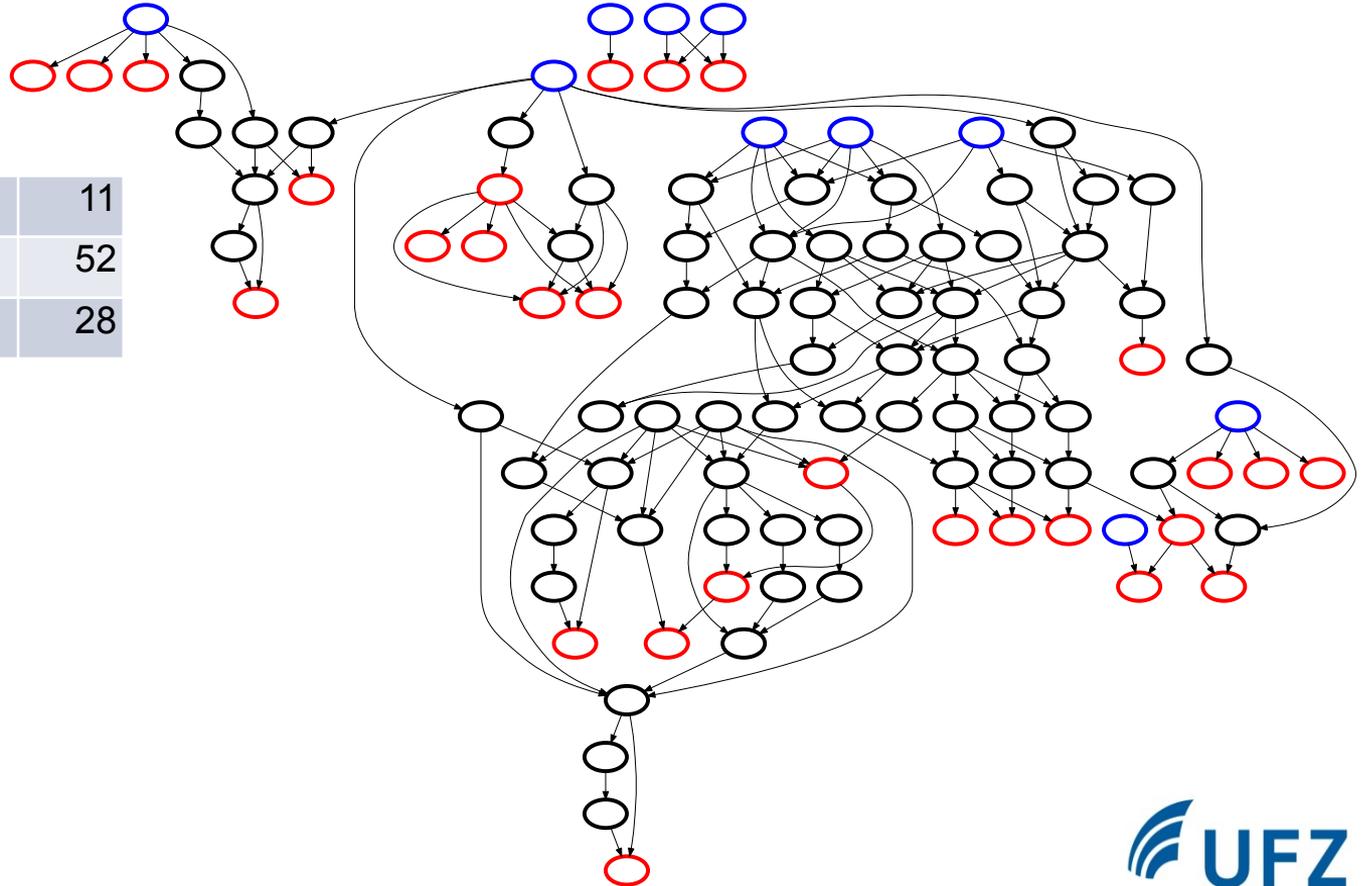


MPR features dimension:

- remapping of irregular shapes
- broadcasting
- splitting
- concatenation
- transposing

# MPR configuration is modular

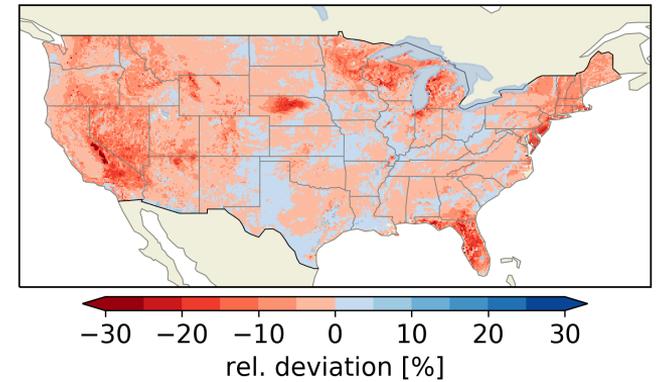
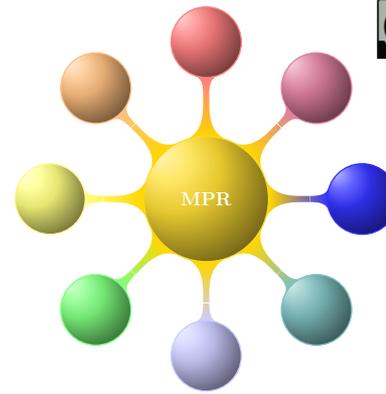
predictor variables	11
global parameters	52
model parameters	28



# Summary

- MPR uses transfer functions and upscaling operators to estimate model parameters from high-resolution data
- Simple, flexible, modular setup, can be coupled to any model
- MPR reveals uncertainty in transfer functions and aggregation methods
- Code development on [git.ufz.de/CHS/MPR](https://git.ufz.de/CHS/MPR)

1st MPR workshop

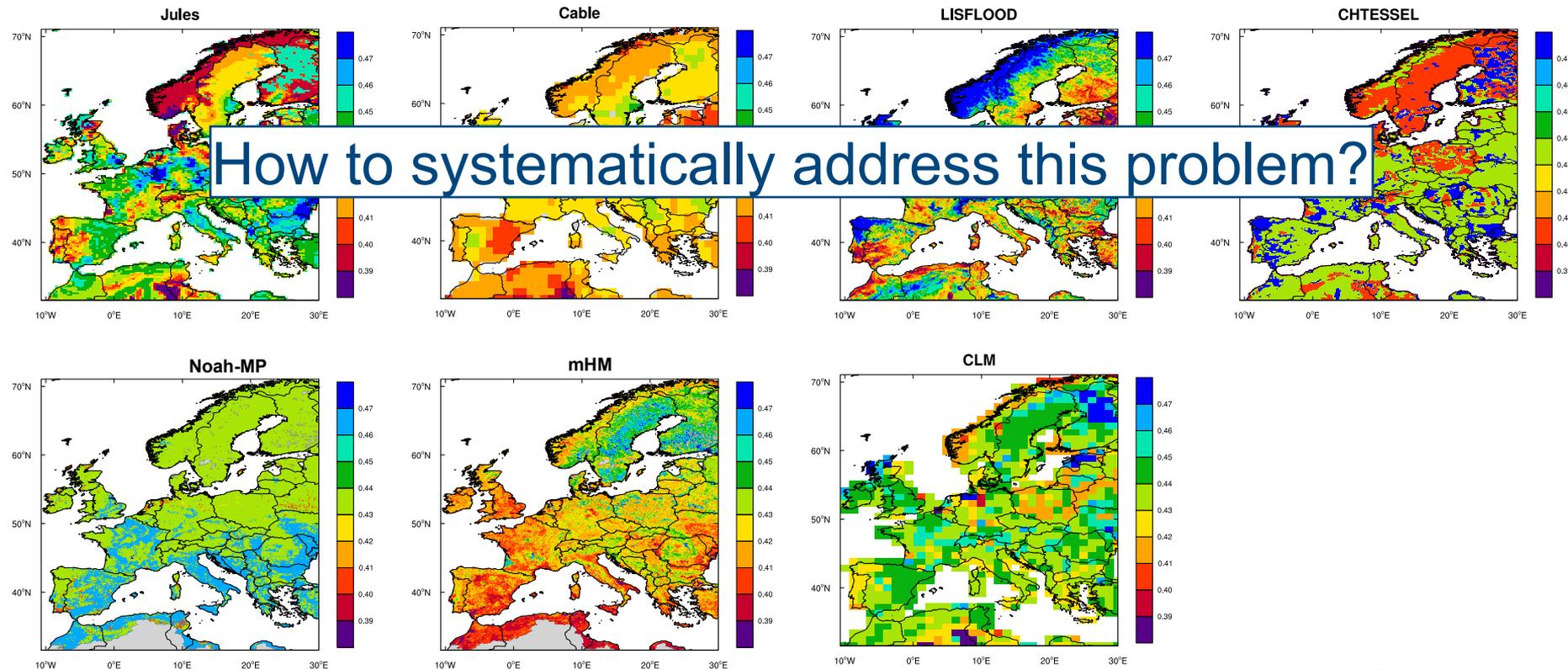


Thank you!

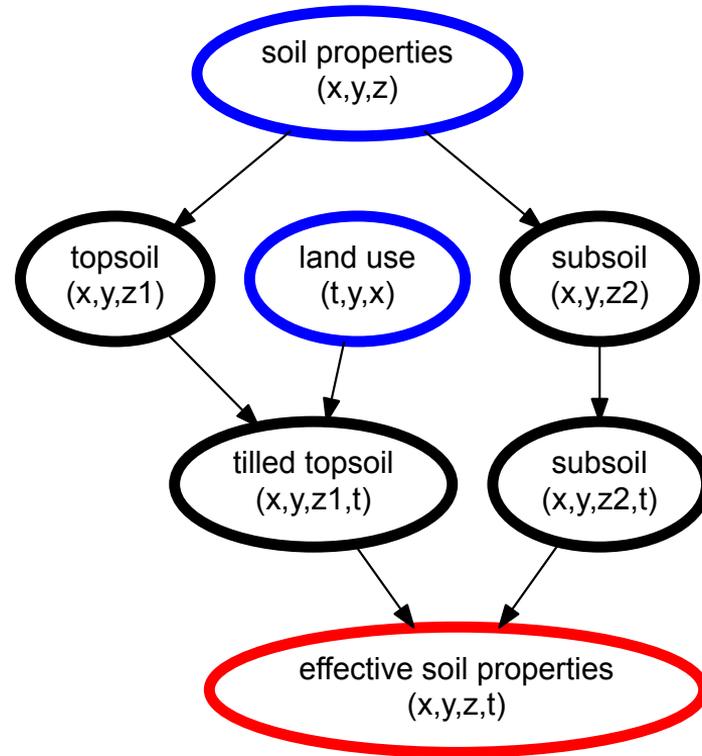


# Appendix

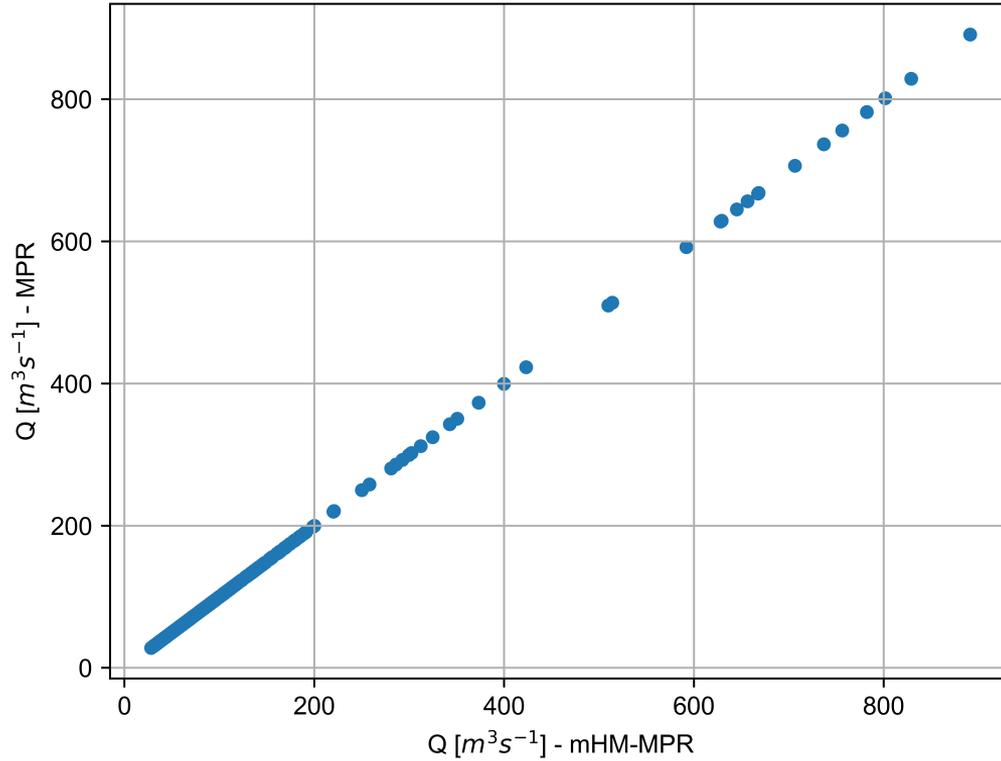
# Porosity in different models



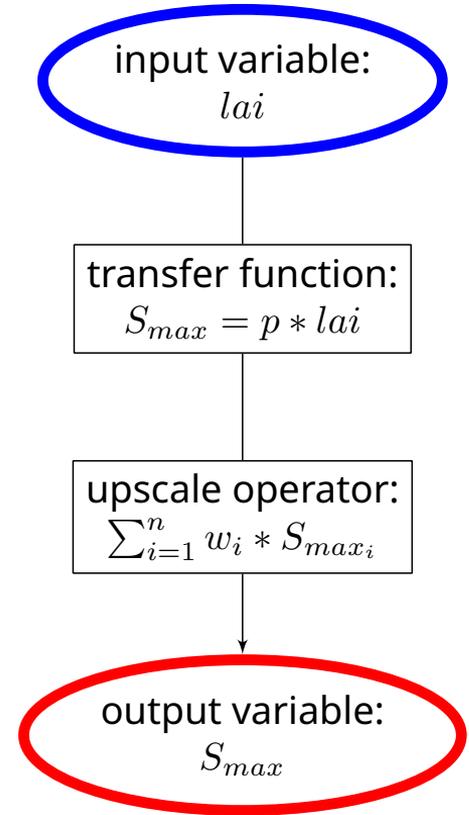
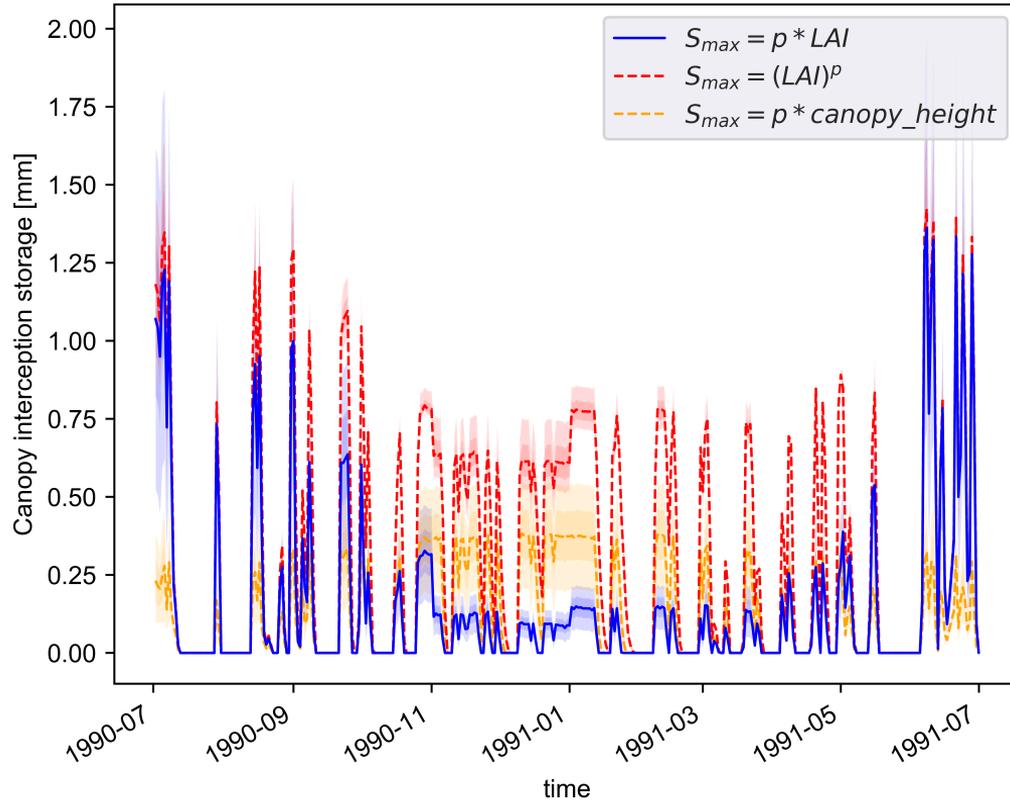
# MPR configuration is flexible



# MPR verification

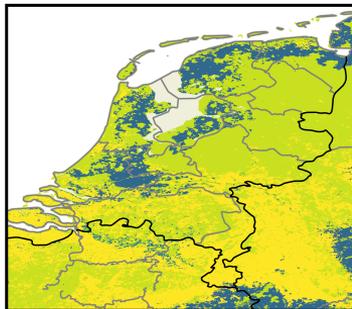


# MPR coupling to mHM

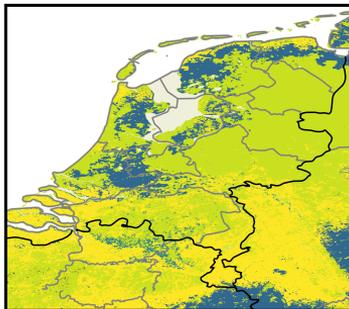


# How MPR validation with EU-SoilHydroGrids

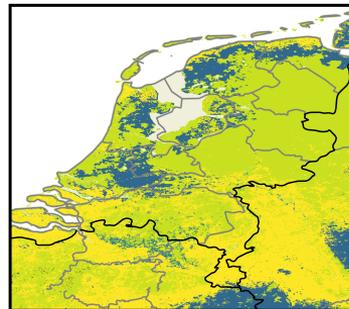
a) EU-SoilHydroGrids<sup>1</sup>



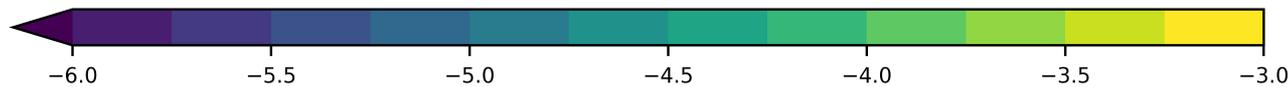
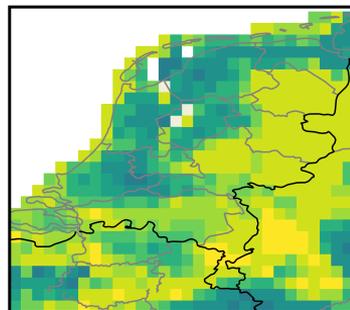
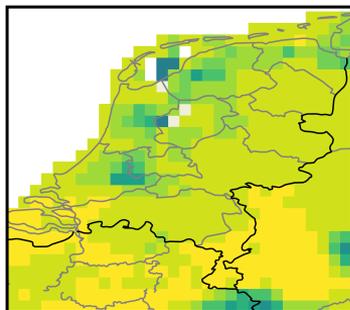
b) MPR applied on SoilGrids<sup>2</sup>



c) R eu-ptf<sup>3</sup> applied on SoilGrids<sup>2</sup>



d) MPR upscaled (arithm. mean) e) MPR upscaled (harmon. mean)



$\log_{10}(Ks [(m * s^{-1})])$

<sup>1</sup> Tóth et al., 2017 (HP) <sup>2</sup> Hengl et al., 2017 (PLOS) <sup>3</sup> Weynants & Tóth, 2014

# Questions

- Regionalization approach classes are defined as (Beck, 2016 (WRR)):
  - (i), catchment-by-catchment calibration followed by regression;
  - (ii), simultaneous calibration and regression;
  - (iii), geographic proximity;
  - (iv), physiographic and/or climatic similarity;
  - (v), regional calibration; and
  - (vi), Q signatures.