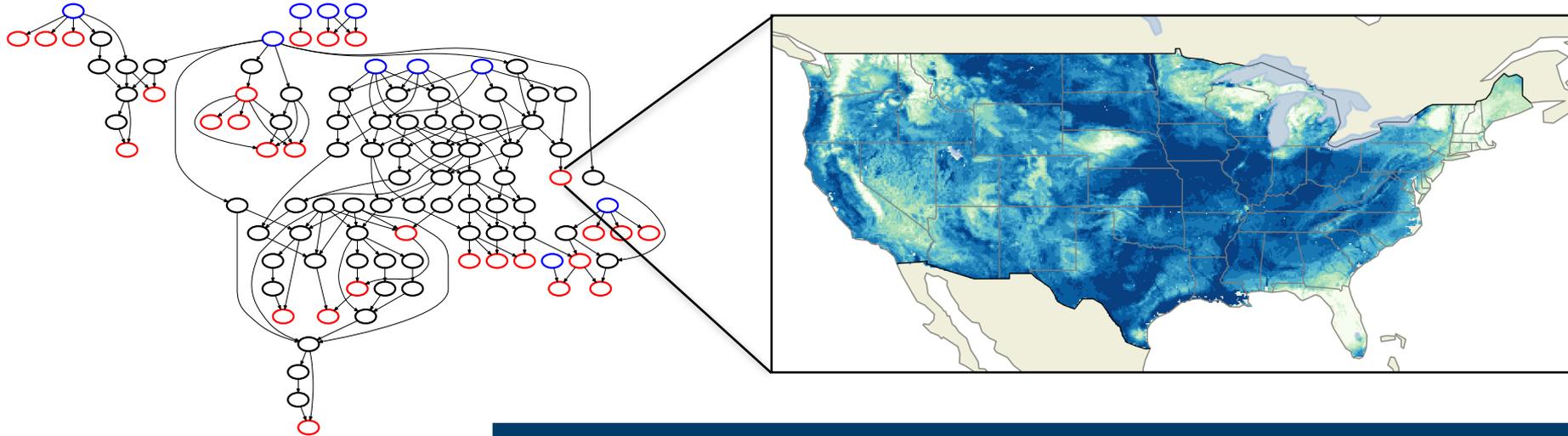


Development of a stand-alone Multiscale Parameter Regionalization (MPR) tool for the estimation of effective model parameters for any distributed model



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

EGU2019-7478 - April 12th 2019



HELMHOLTZ
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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

β : process parameters

O : process output

i : cell index

t : time index

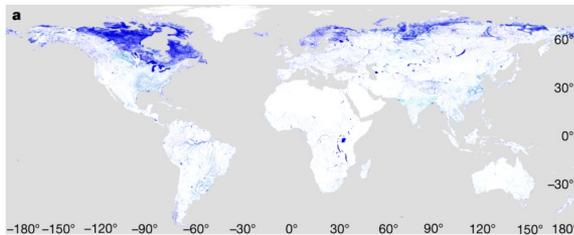
High-resolution data



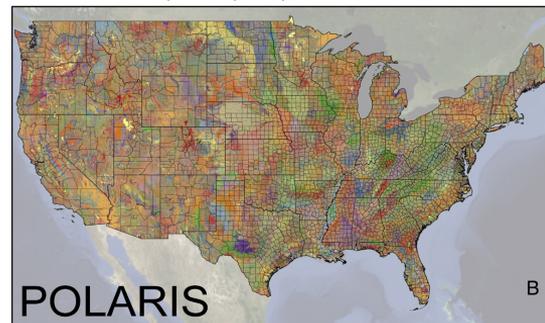
ASTER (NASA, 2001)
elevation at 30m



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

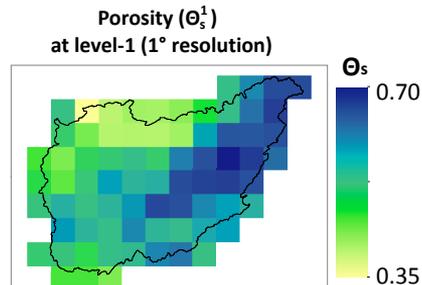
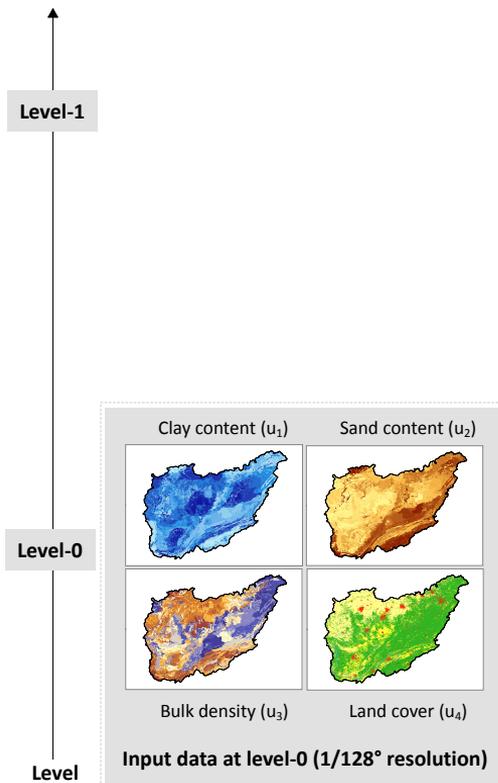


JRC GSW (Pekel, 2016 Nat)
lake alimetry at 30m



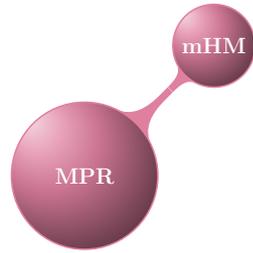
POLARIS (Chaney, 2016 Geoderma)
soil at 30m

Multiscale Parameter Regionalization (MPR)

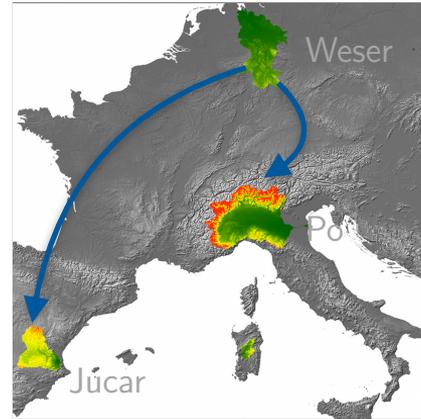


- steps:
1. transfer function
 2. upscaling

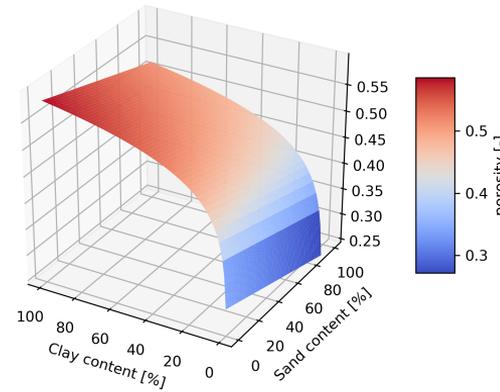
MPR key advantages



Samaniego, 2010 (WRR)



adapted from Samaniego



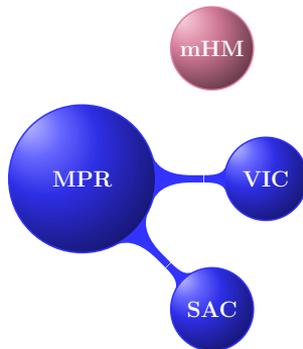
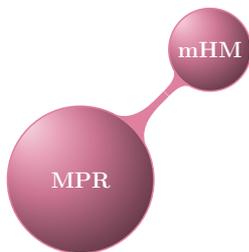
$$O_{it} = f(I_{it}, \beta_i)$$

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

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

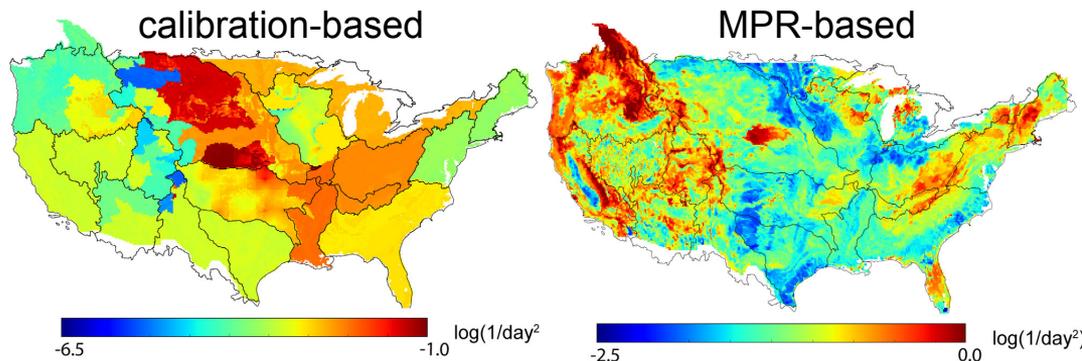
$$\#\gamma = 3$$

MPR key advantages

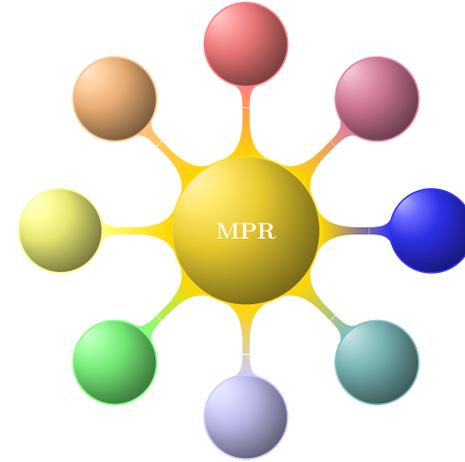
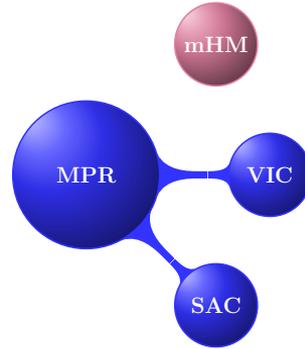
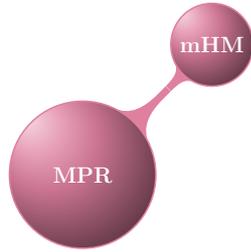


Mizukami, 2017 (WRR)

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



new MPR software



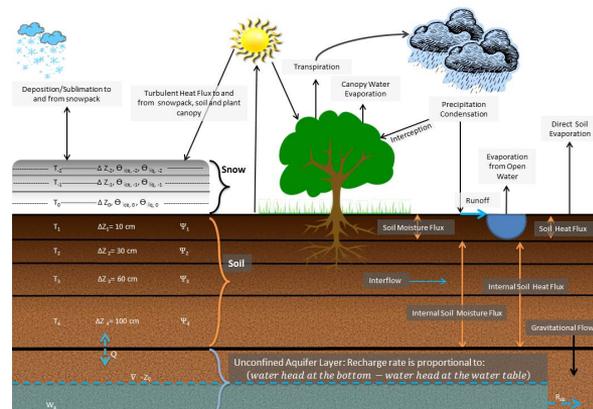
Schwepe, 2019 (in preparation)

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

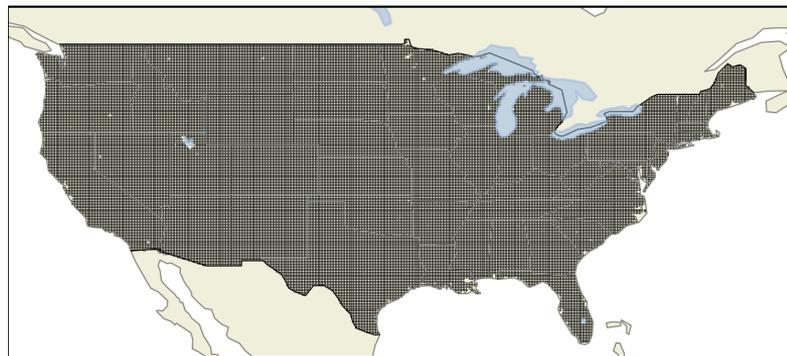
How can MPR be implemented?

Land-Surface Model Noah-MP

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

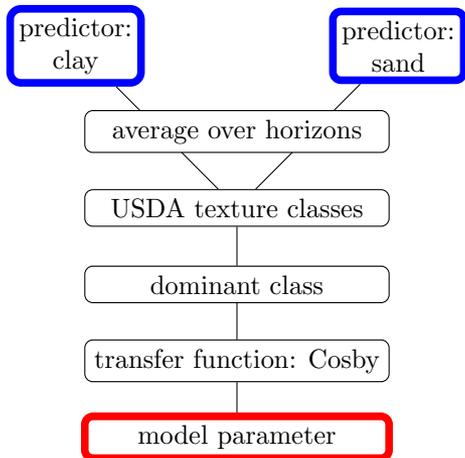
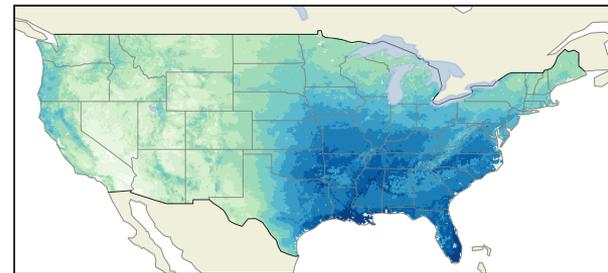
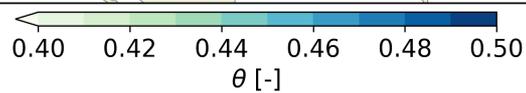
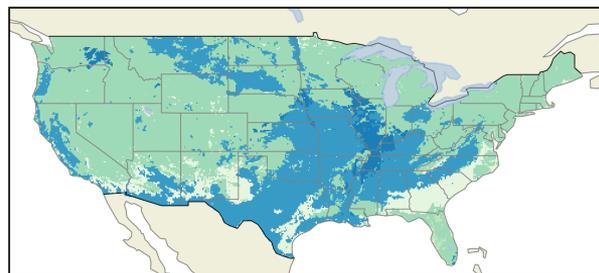
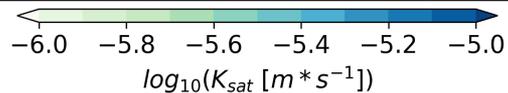
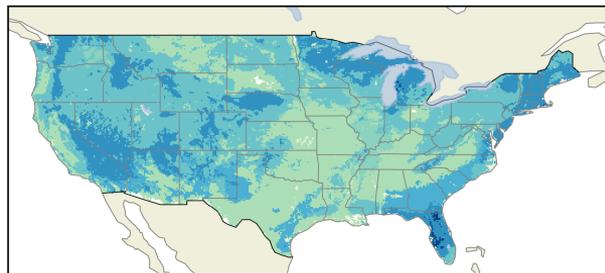


University of Texas at Austin

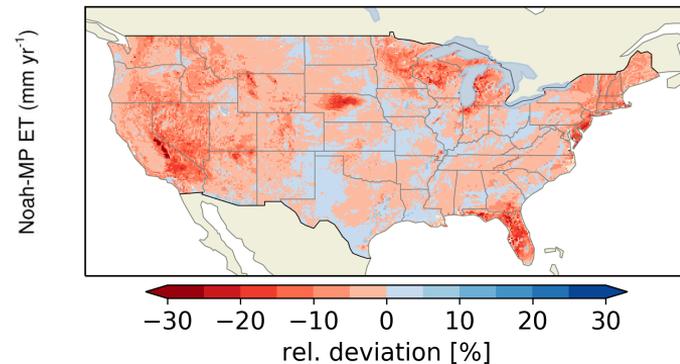
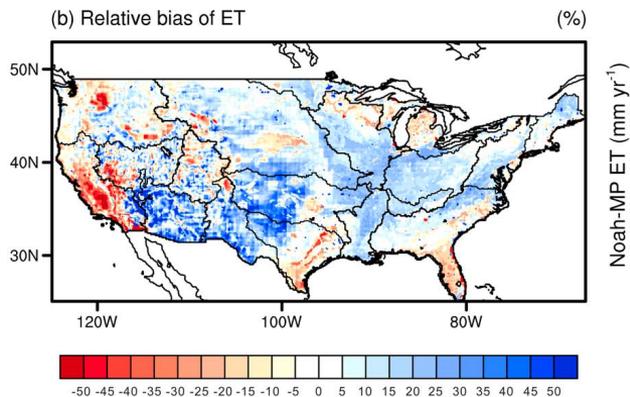
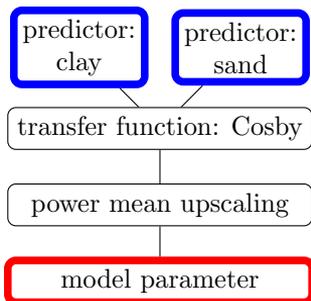
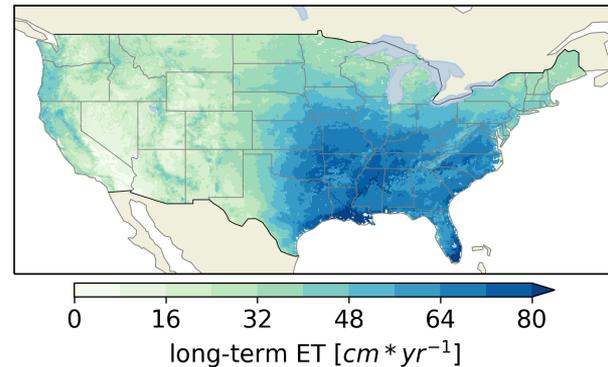
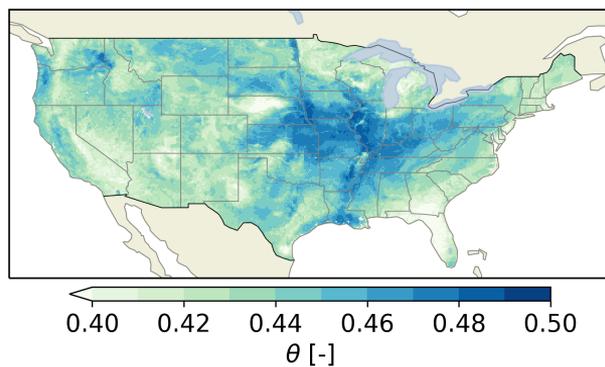
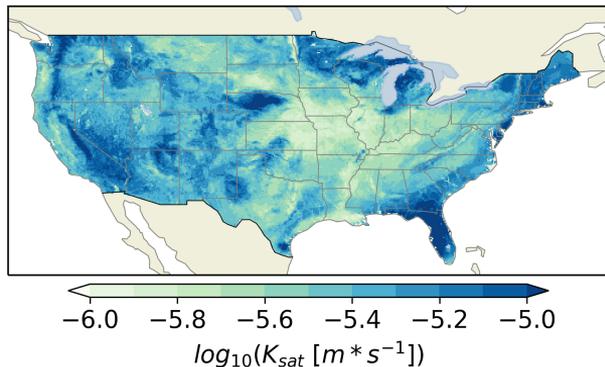


- NLDAS2 forcing
- Hourly time step, $1/8^\circ$ resolution

Application with Noah-MP

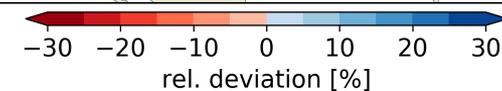
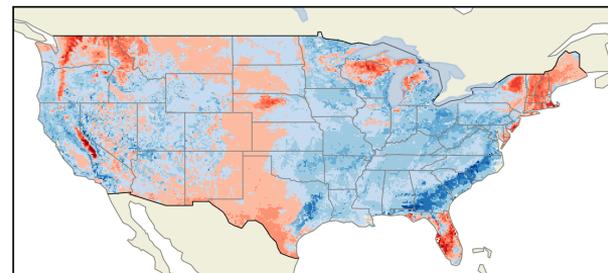
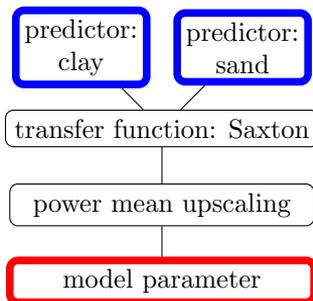
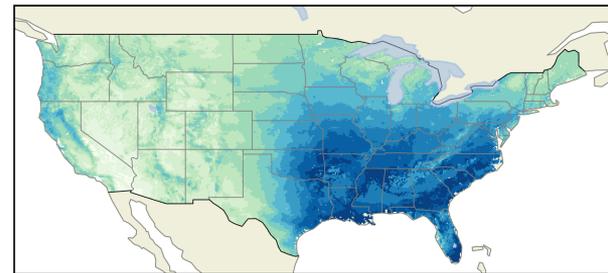
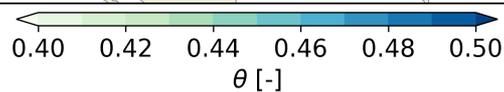
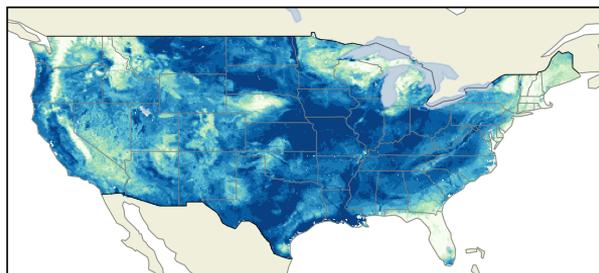
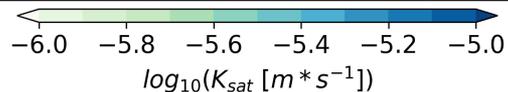
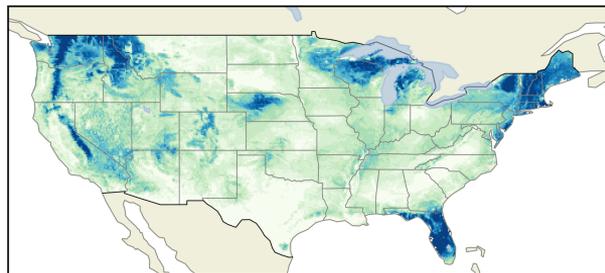


Application with Noah-MP

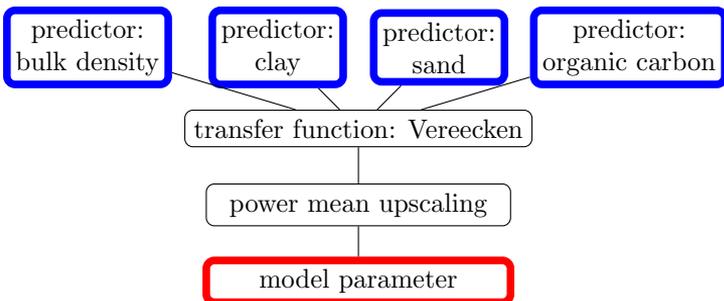
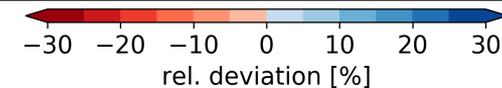
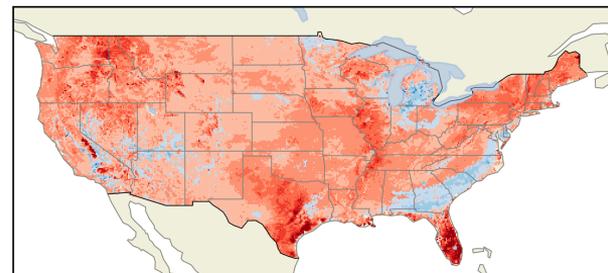
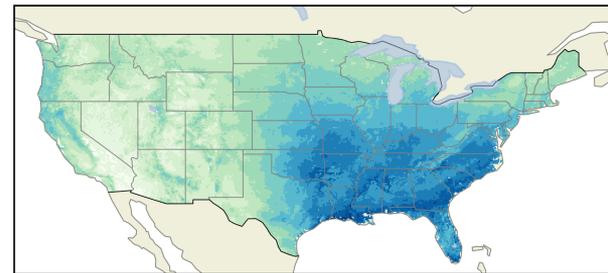
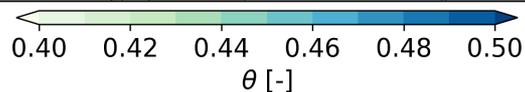
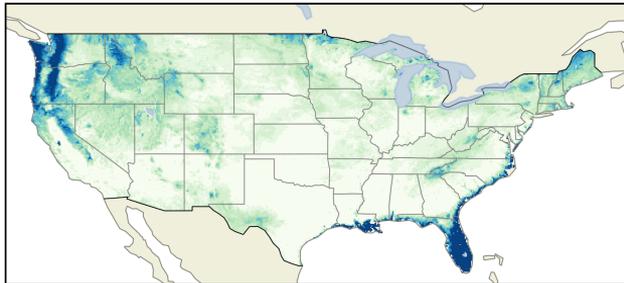
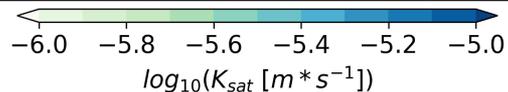
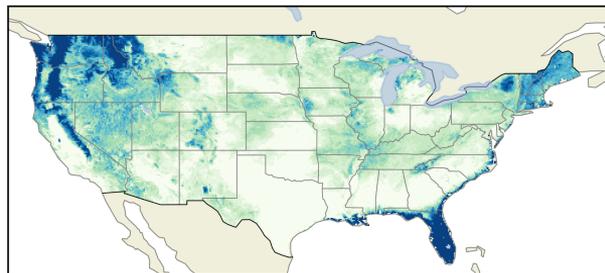


Ma, 2017 (JGRA)

Application with Noah-MP

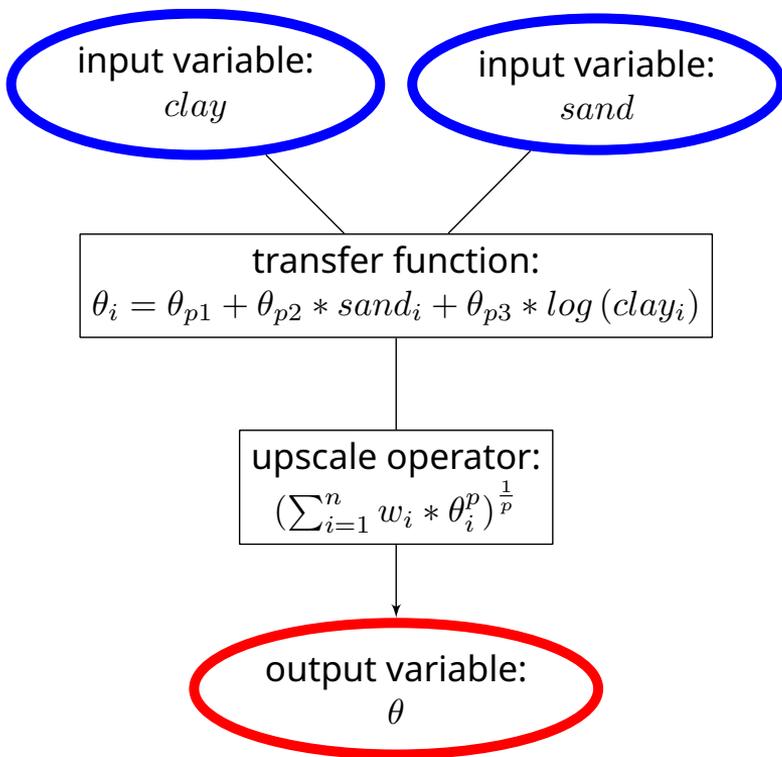


Application with Noah-MP



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

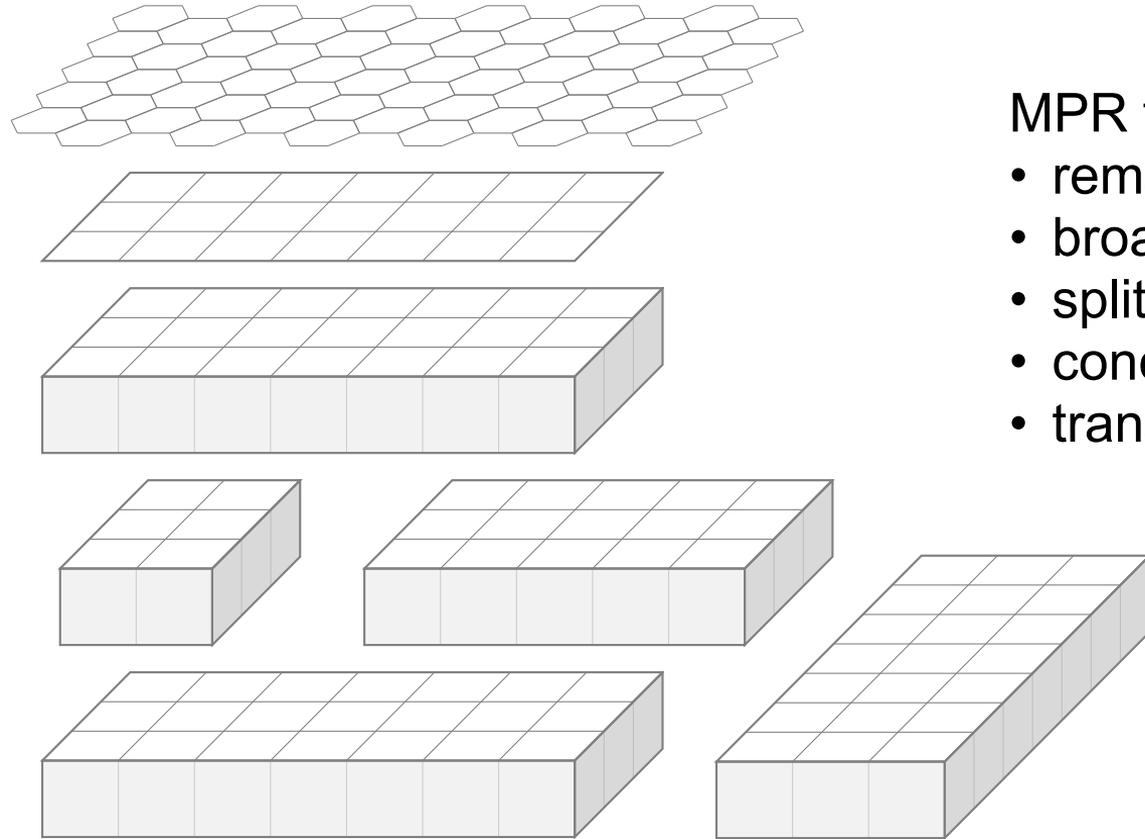
&Parameters
parameter_names(1:3) = "theta_p1", "theta_p2", "theta_p3"
parameter_values(1:3) = 0.505, -0.00142, -0.00037
/
    
```

- 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"
- 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

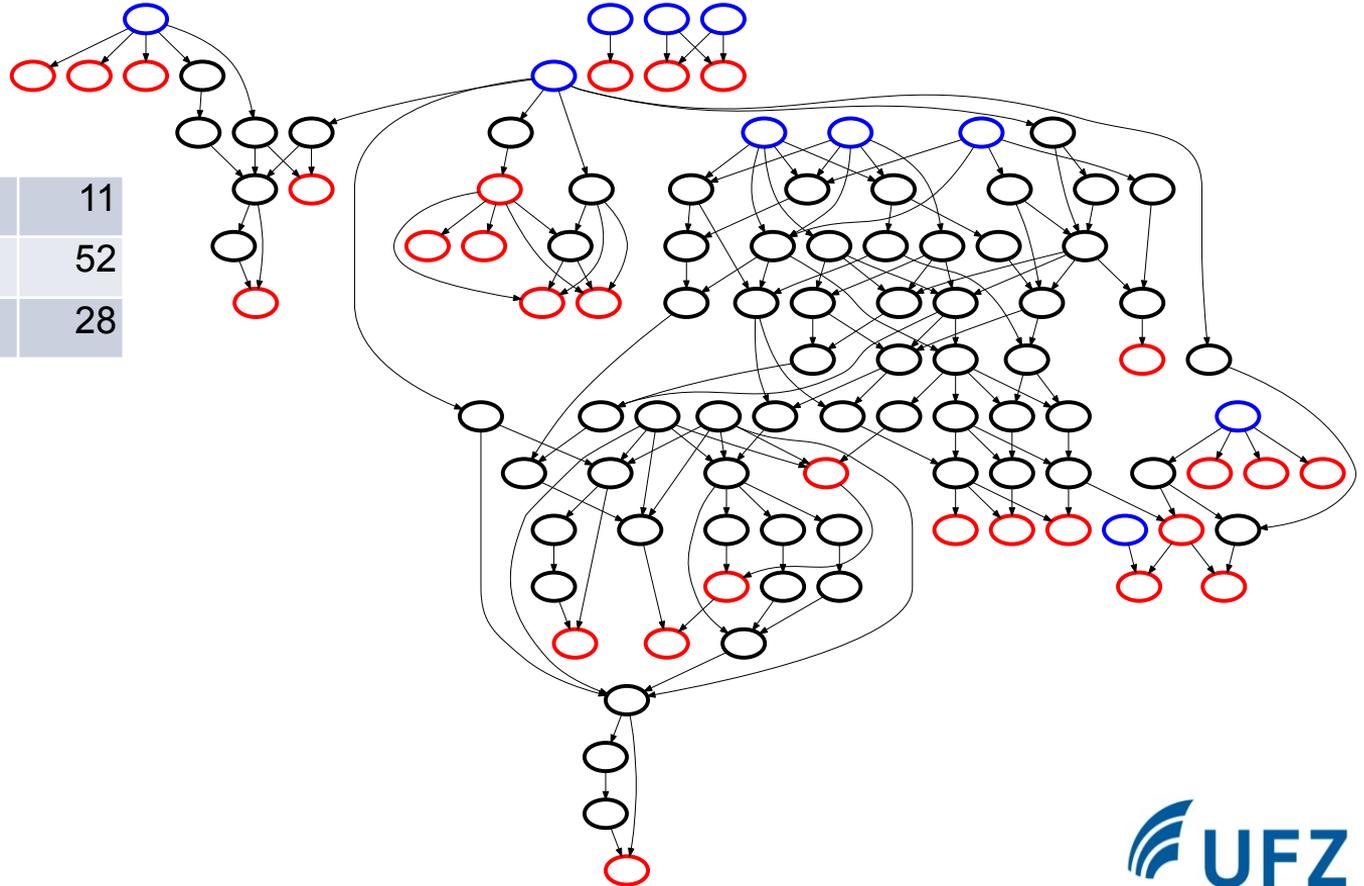


MPR features dimension:

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

MPR configuration is modular

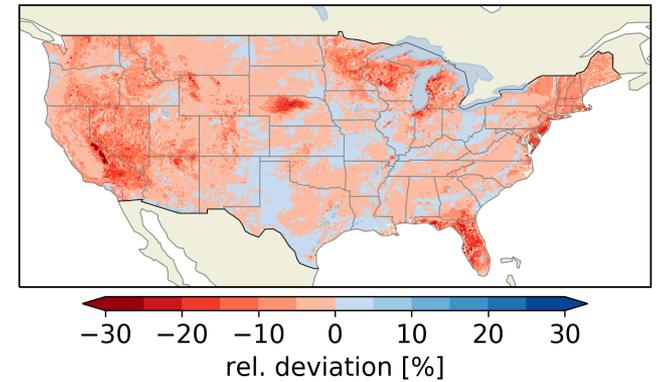
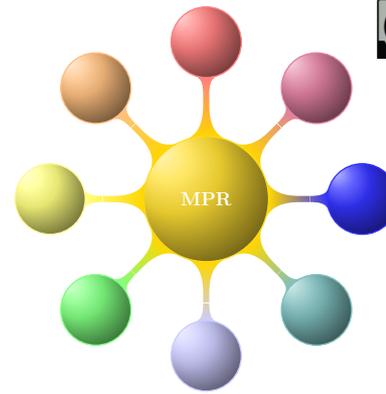
predictor variables	11
global parameters	52
model parameters	28



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

1st MPR workshop

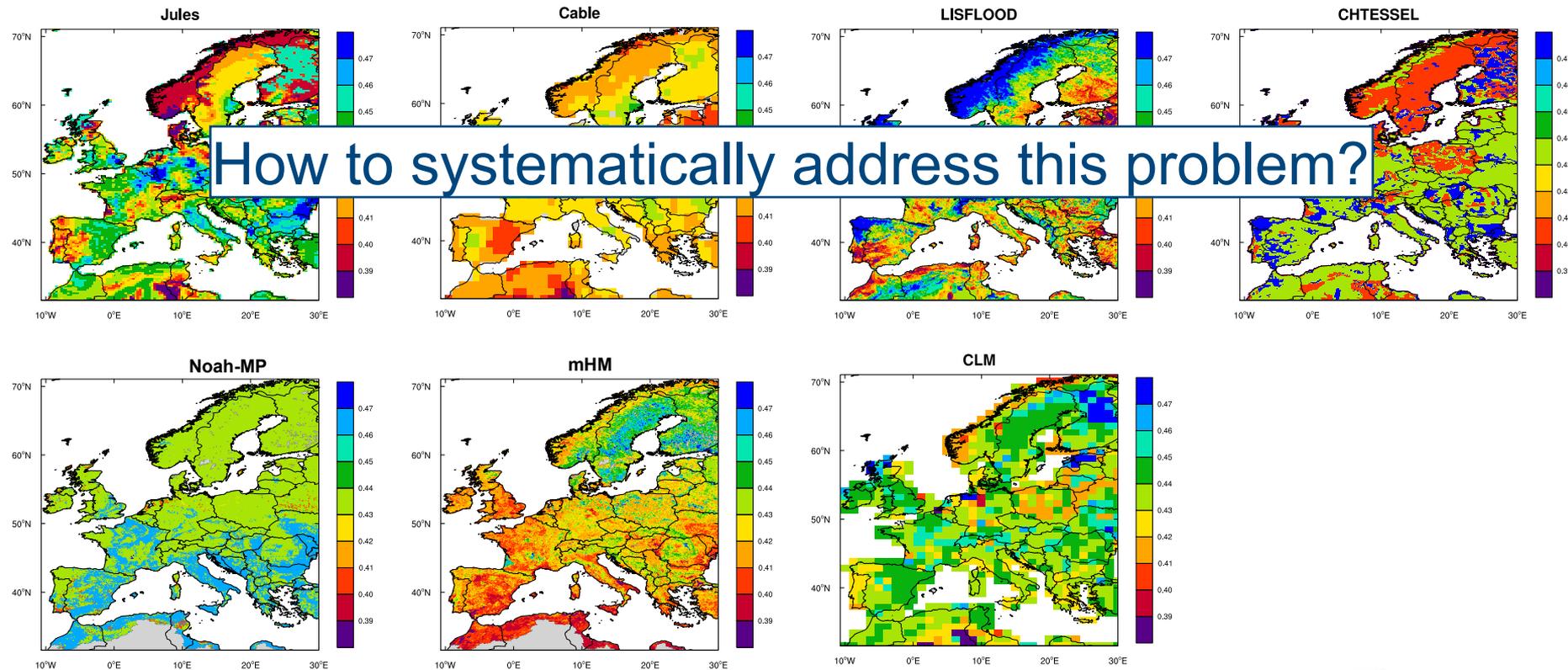


Thank you!

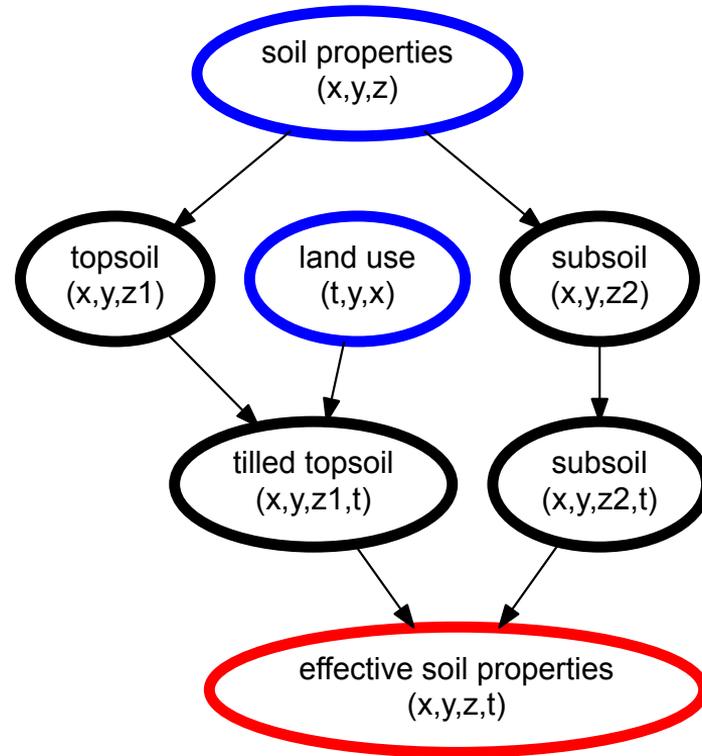


Appendix

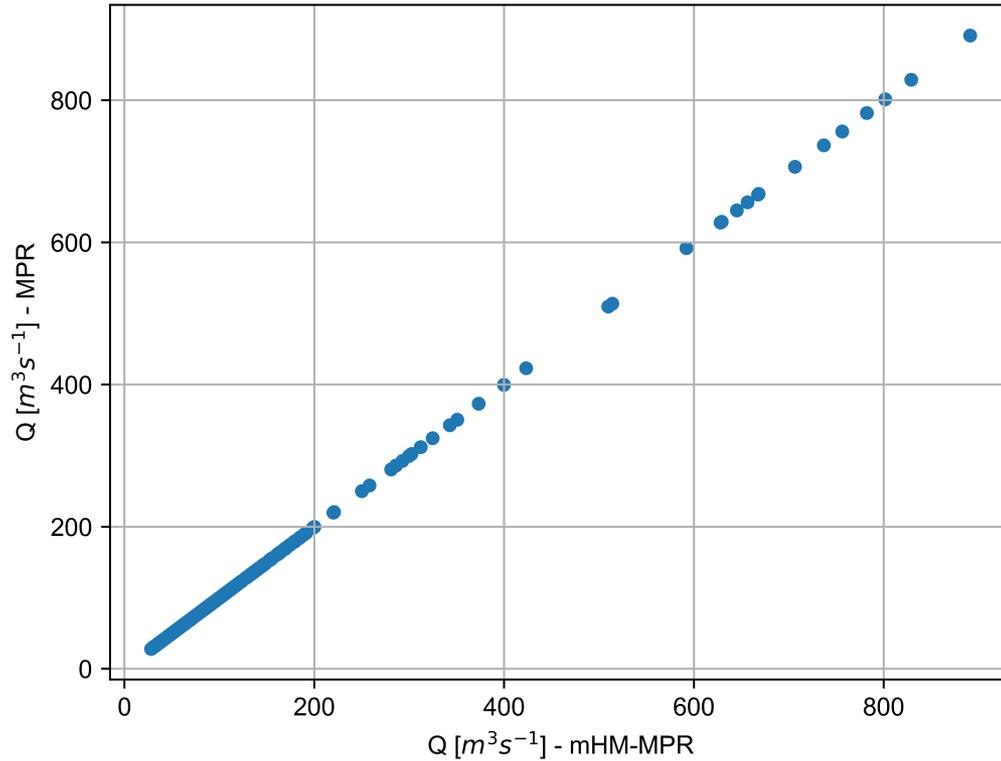
Porosity in different models



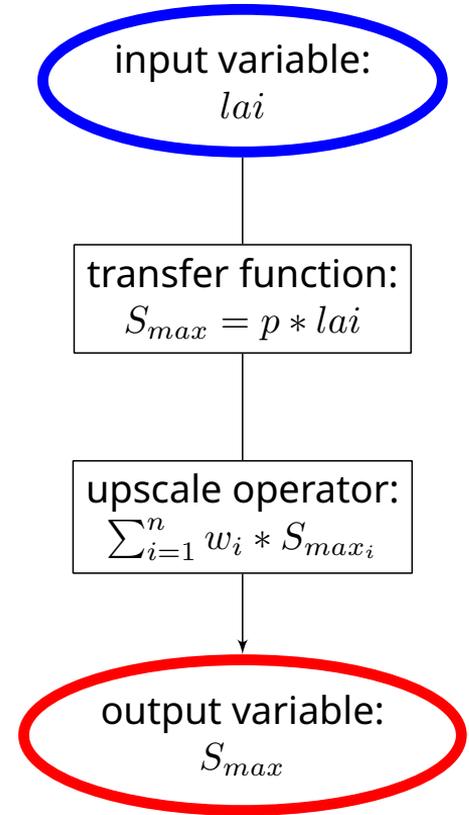
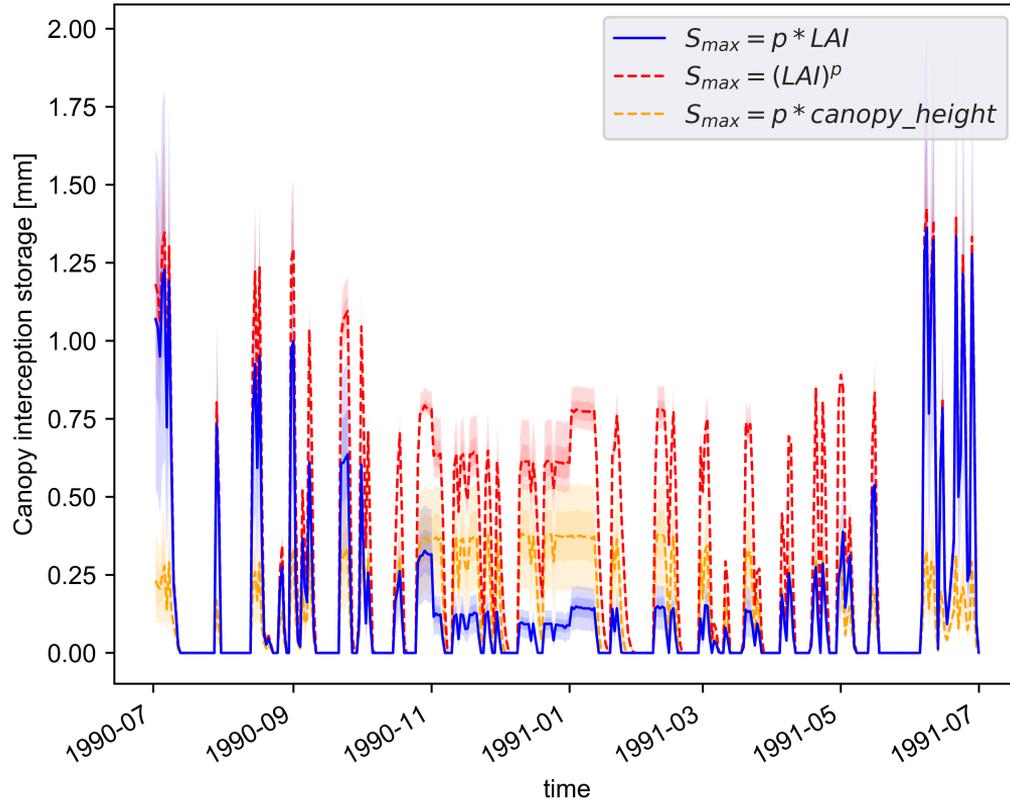
MPR configuration is flexible



MPR verification

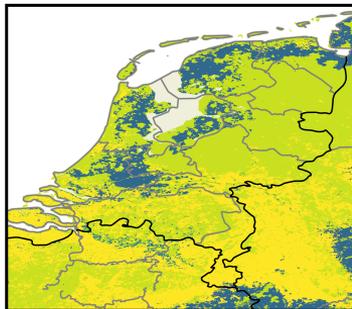


MPR coupling to mHM

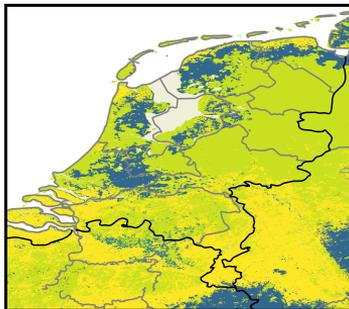


How MPR validation with EU-SoilHydroGrids

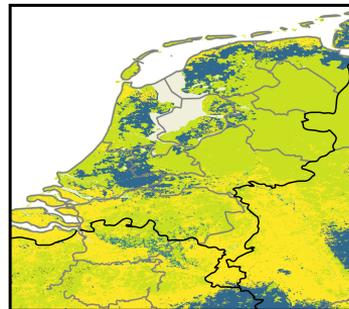
a) EU-SoilHydroGrids¹



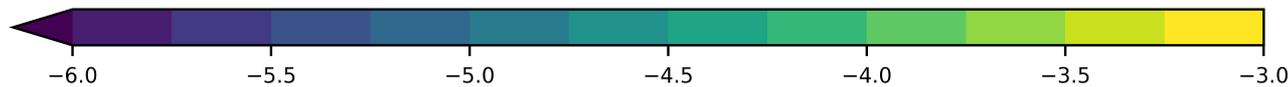
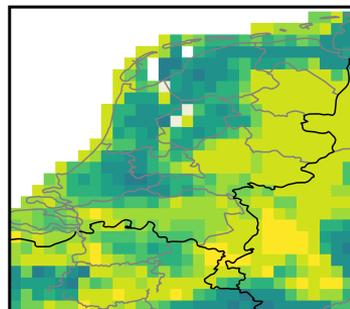
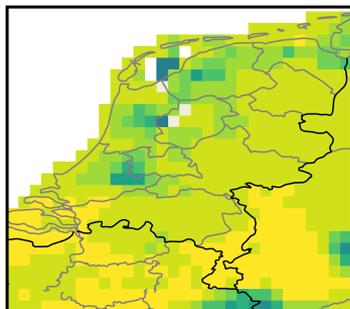
b) MPR applied on SoilGrids²



c) R eu-ptf³ applied on SoilGrids²



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



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

¹ Tóth et al., 2017 (HP) ² Hengl et al., 2017 (PLOS) ³ 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.