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AUTOMATIC ESTIMATION OF PARAMETER TRANSFER FUNCTIONS FOR DISTRIBUTED HYDROLOGICAL MODELS

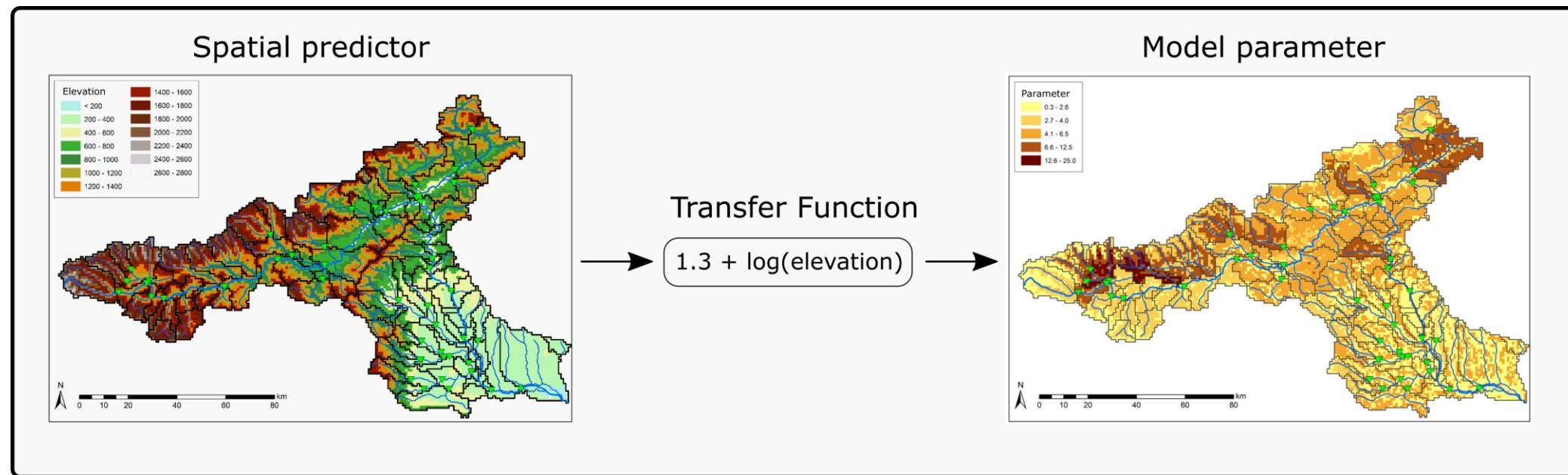
FUNCTION SPACE OPTIMIZATION APPLIED TO THE mHM MODEL

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Transfer Functions

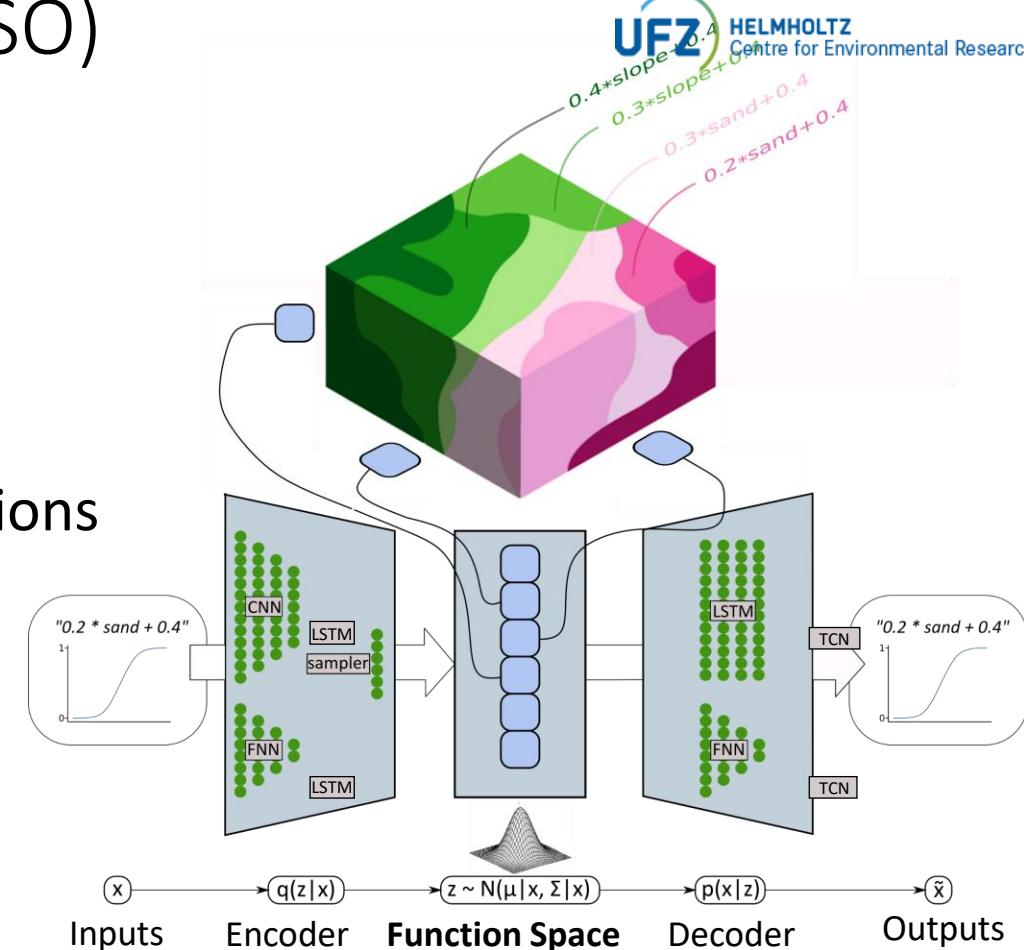


Transfer functions map geophysical catchment properties to distributed model parameters

Function Space Optimization (FSO)

Feigl et al., 2020

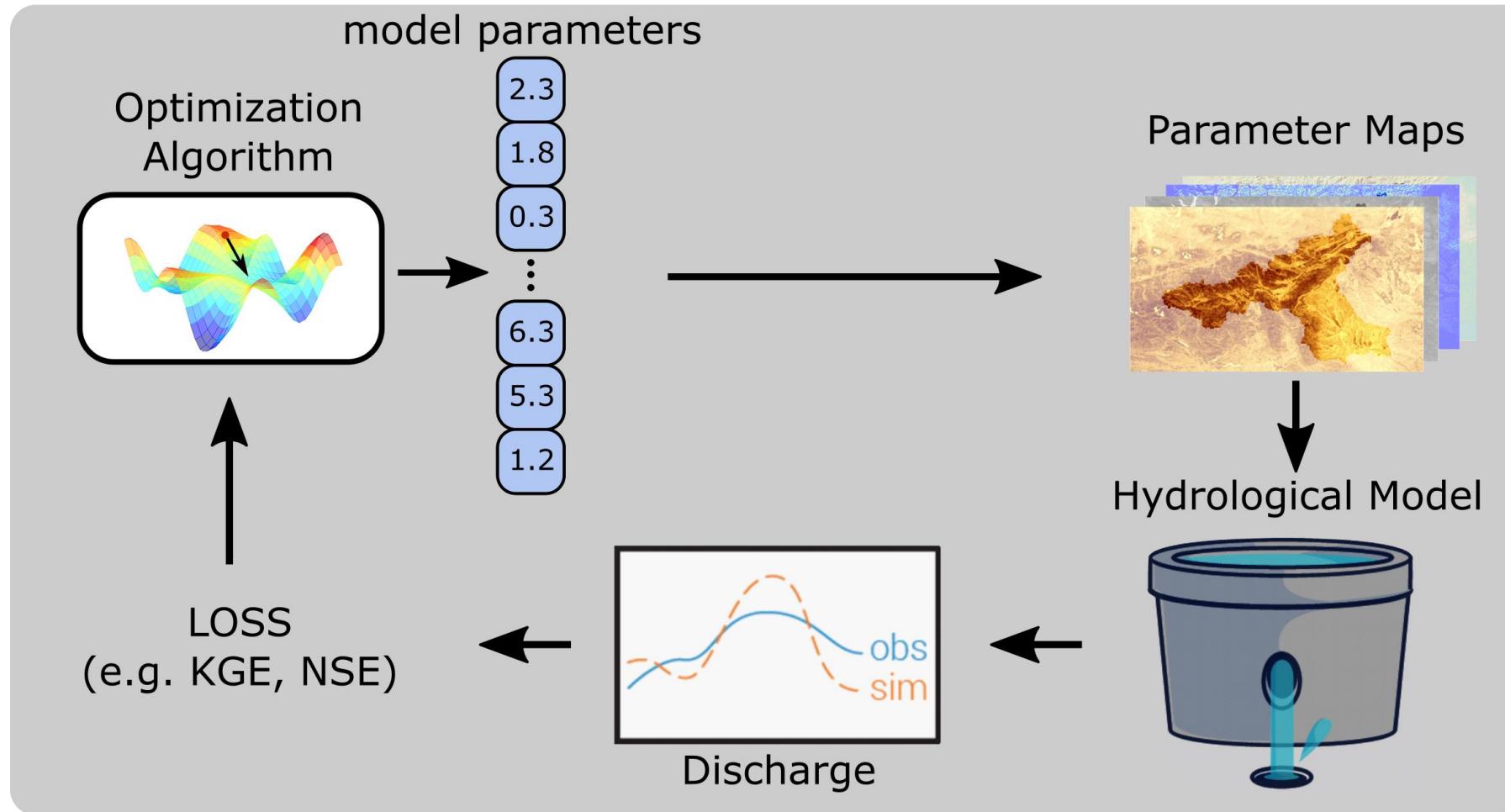
- FSO: optimization method for transfer functions
- Uses a text generating Neural Network
- Transforms search into continuous problem
- Successfully tested on a single catchment



FSO variational autoencoder
Feigl et al. (2020)

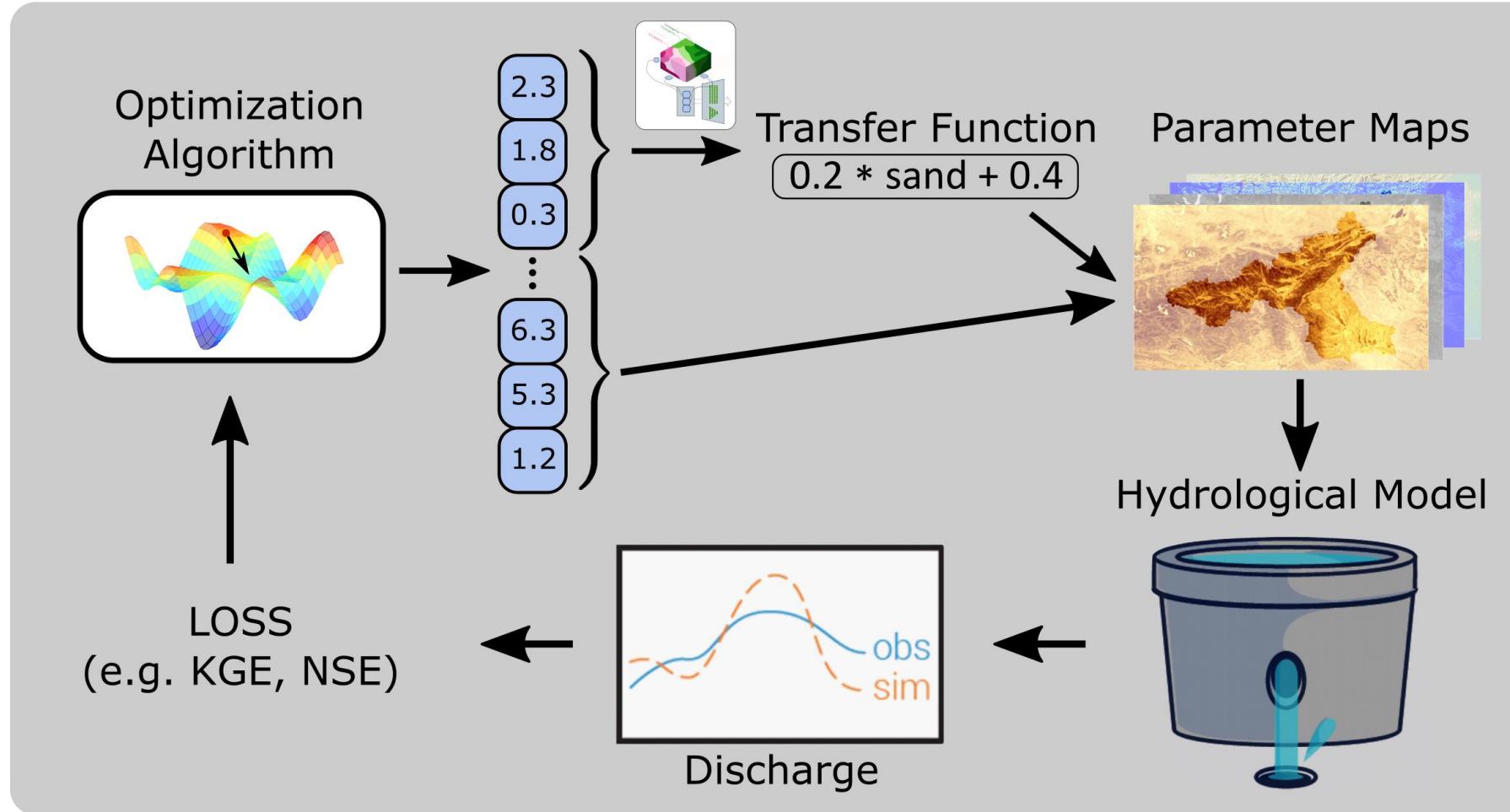
Function Space Optimization (FSO)

Classical Model Optimization



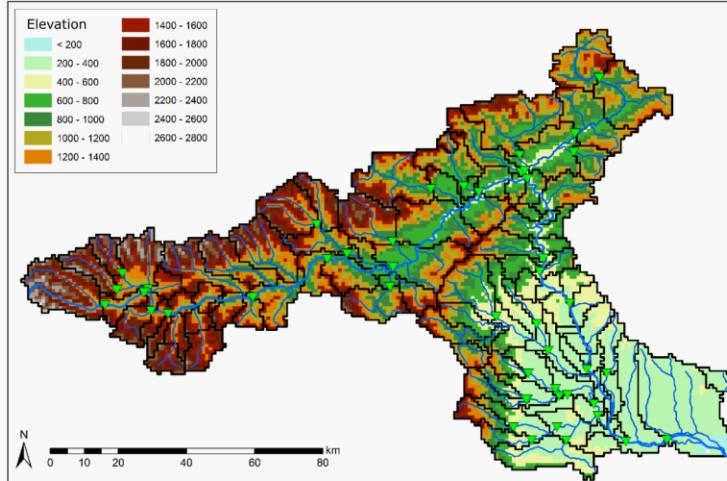
Function Space Optimization (FSO)

Function Space Optimization



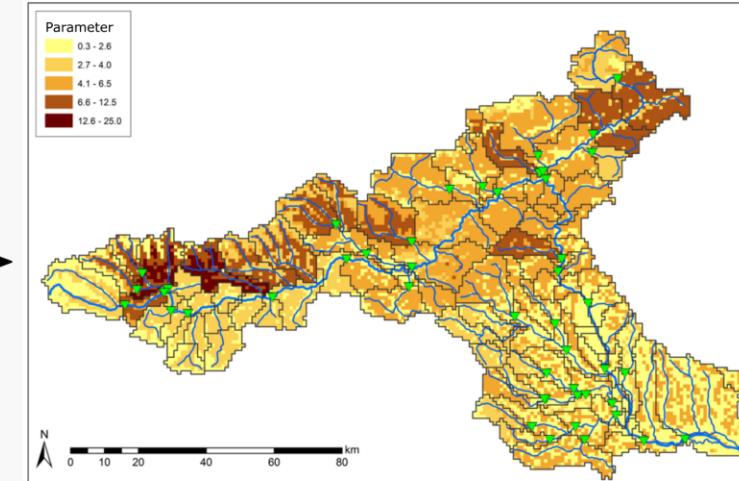
FSO parameter scaling

Spatial predictor



value range: e.g. elevation [0, 3000]

Model parameter



value range: e.g. [0, 25]

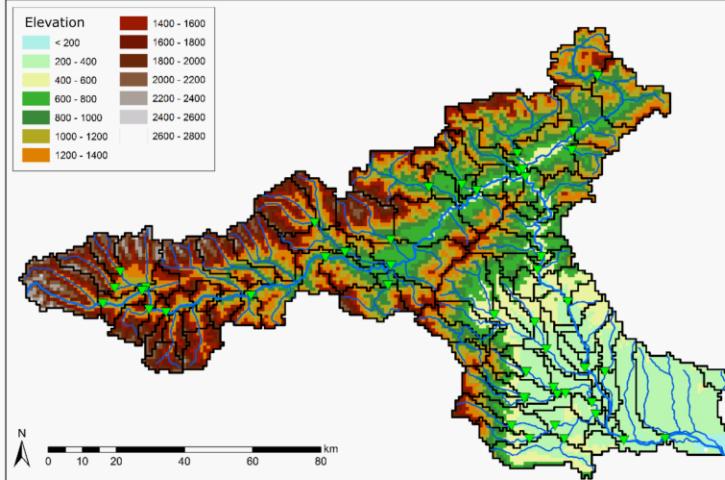
Transfer Function

$$1.3 + \log(\text{elevation})$$

?

FSO parameter scaling

Spatial predictor



value range: e.g. elevation [0, 3000]

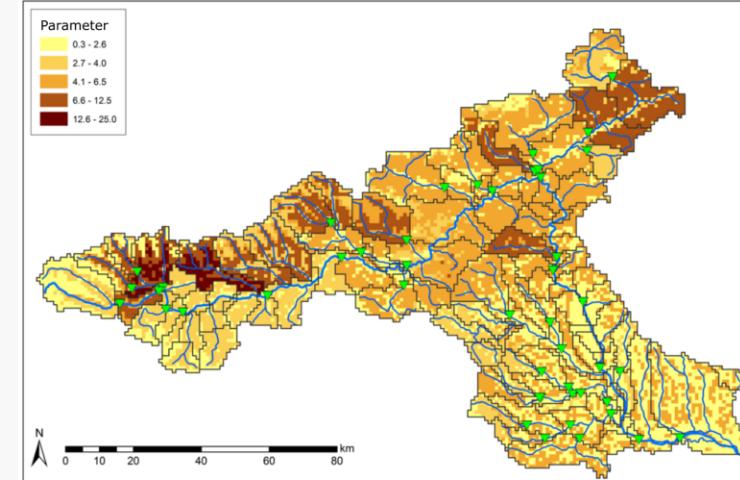


scale to range: [0,1]

Transfer Function

$$1.3 + \log(\text{elevation})$$

Model parameter

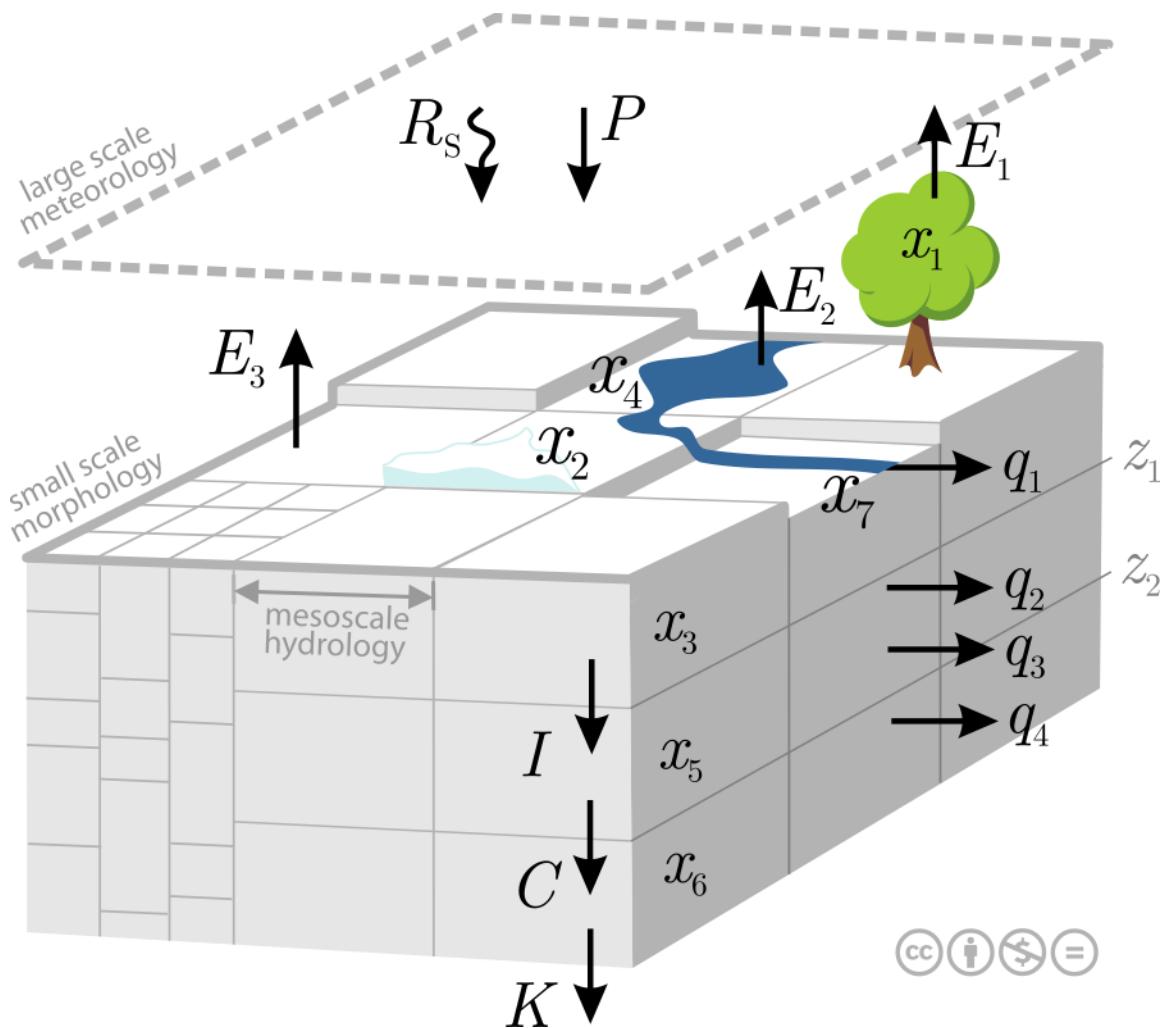


value range: e.g. [0, 25]



scale to parameter range: [0,25]

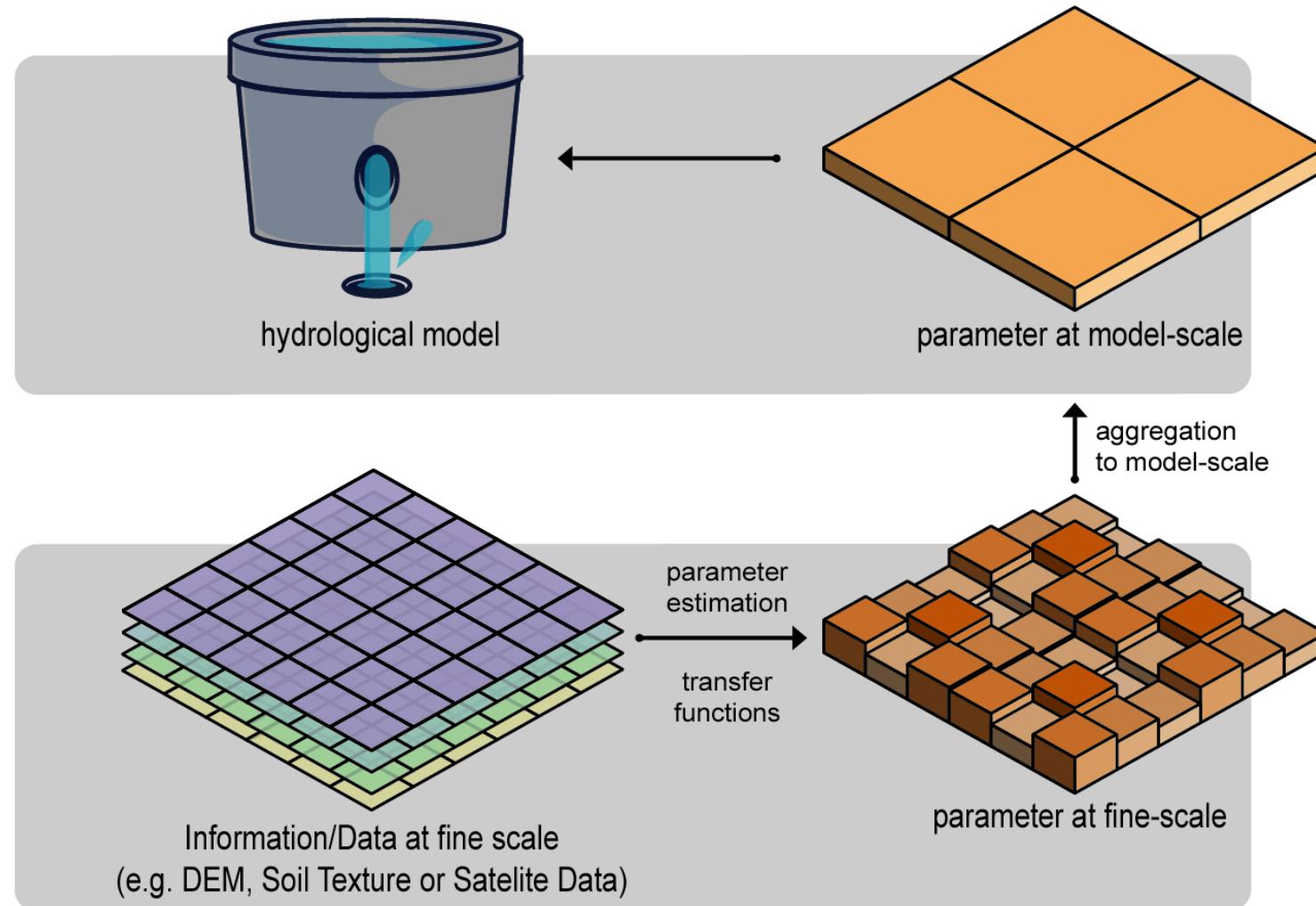
The mesoscale Hydrological Model (mHM)



- Developed by Samaniego et al. (2010)
- Spatially explicit distributed model
- Uses grid cells as primary units
- Defines parameter fields with the Multiscale Parameter Regionalization method (MPR)



Multiscale Parameter Regionalization (MPR)



Regionalization method by Samaniego et al. (2010)

Benchmark Study – Zink et al. (2017)



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doi:10.5194/hess-21-1769-2017
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Hydrology and
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A high-resolution dataset of water fluxes and states for Germany accounting for parametric uncertainty

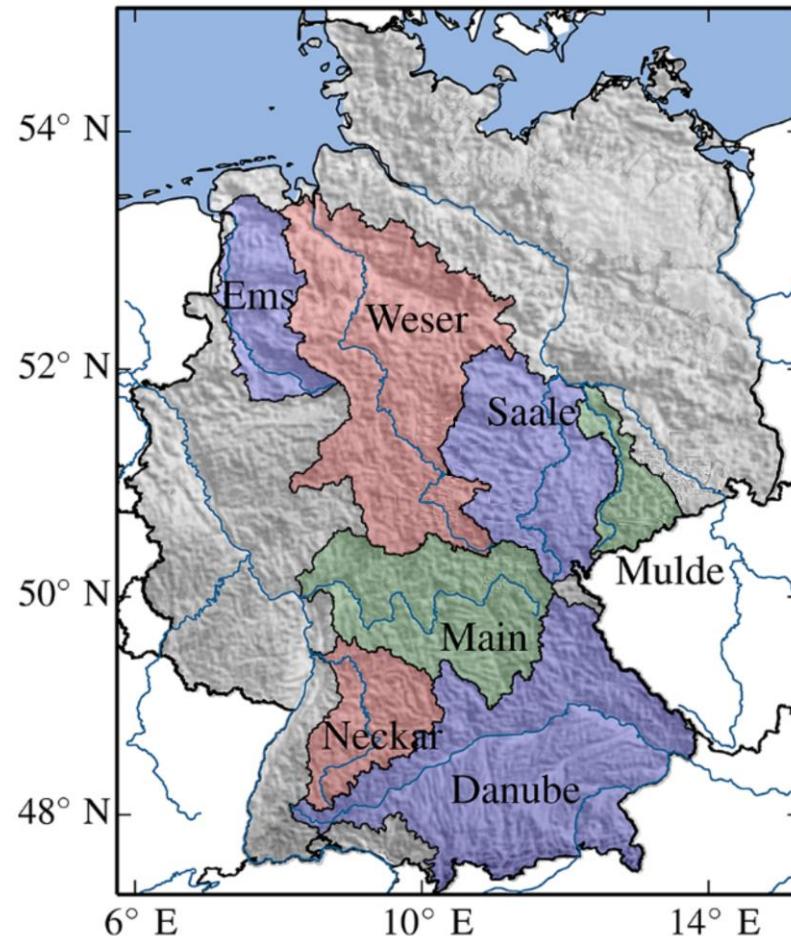
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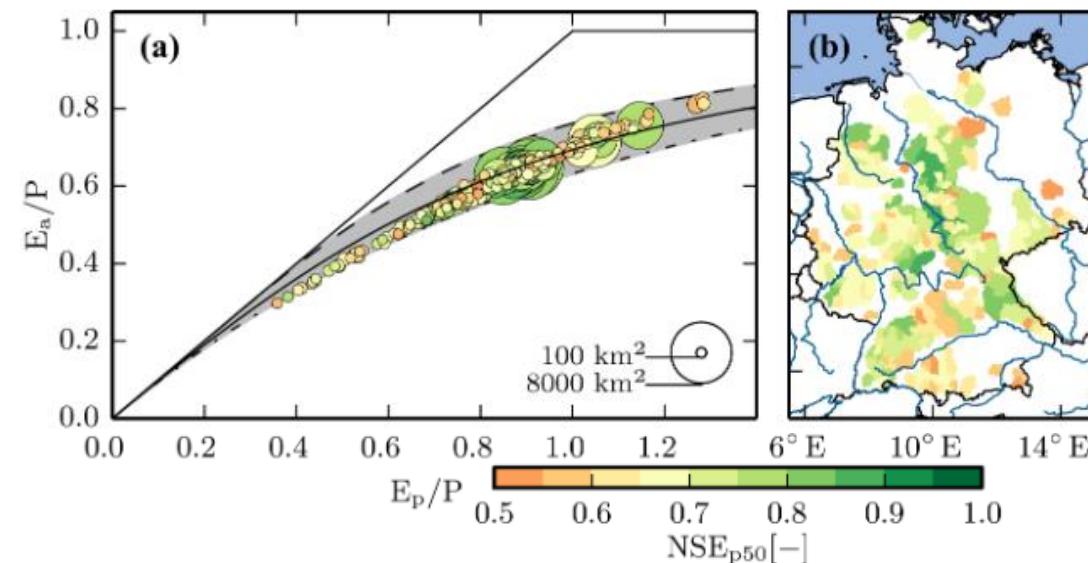
Benchmark Study – Zink et al. (2017)



Training basins

(Zink et al., 2017)

- Optimizing mHM 100 times with 2000 iterations
- Using 7 gauging stations
- Validate 100 parameter sets on 220 Basins



Validation basins

(Zink et al., 2017)

Study Objectives

1. Apply FSO using a wide range of catchments
2. Simultaneously optimize 2 transfer functions and all other numerical parameters
3. Optimize: Saturated Hydraulic Conductivity, Field Capacity
4. Analyze performance and transferability in a prediction in ungauged basins (PUB) setting
5. Compare original mHM tranfer functions with FSO estimates

Case study – study basins

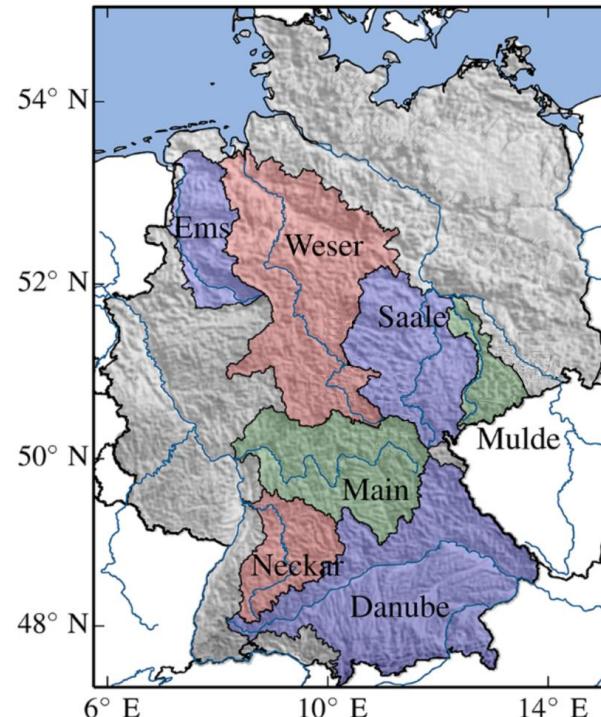
7 Training basins, 220 Validation Basins

Resolution:

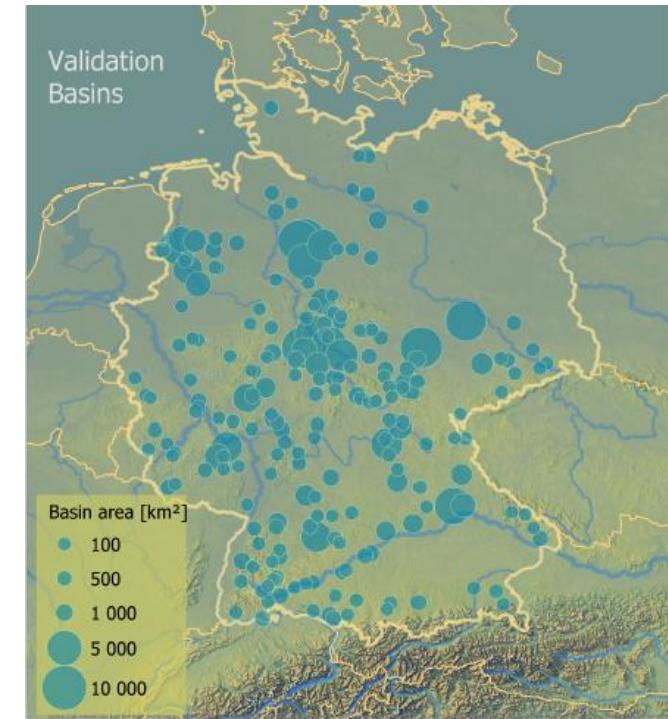
Spatial predictors: 100 x 100 m
Model grid: 4 x 4 km

Spatial predictors:

Mean sand percentage (sand)
Mean clay percentage (clay)
Mineral bulk density
Aspect
Terrain slope
Elevation



7 Training basins
(Zink et al., 2017)

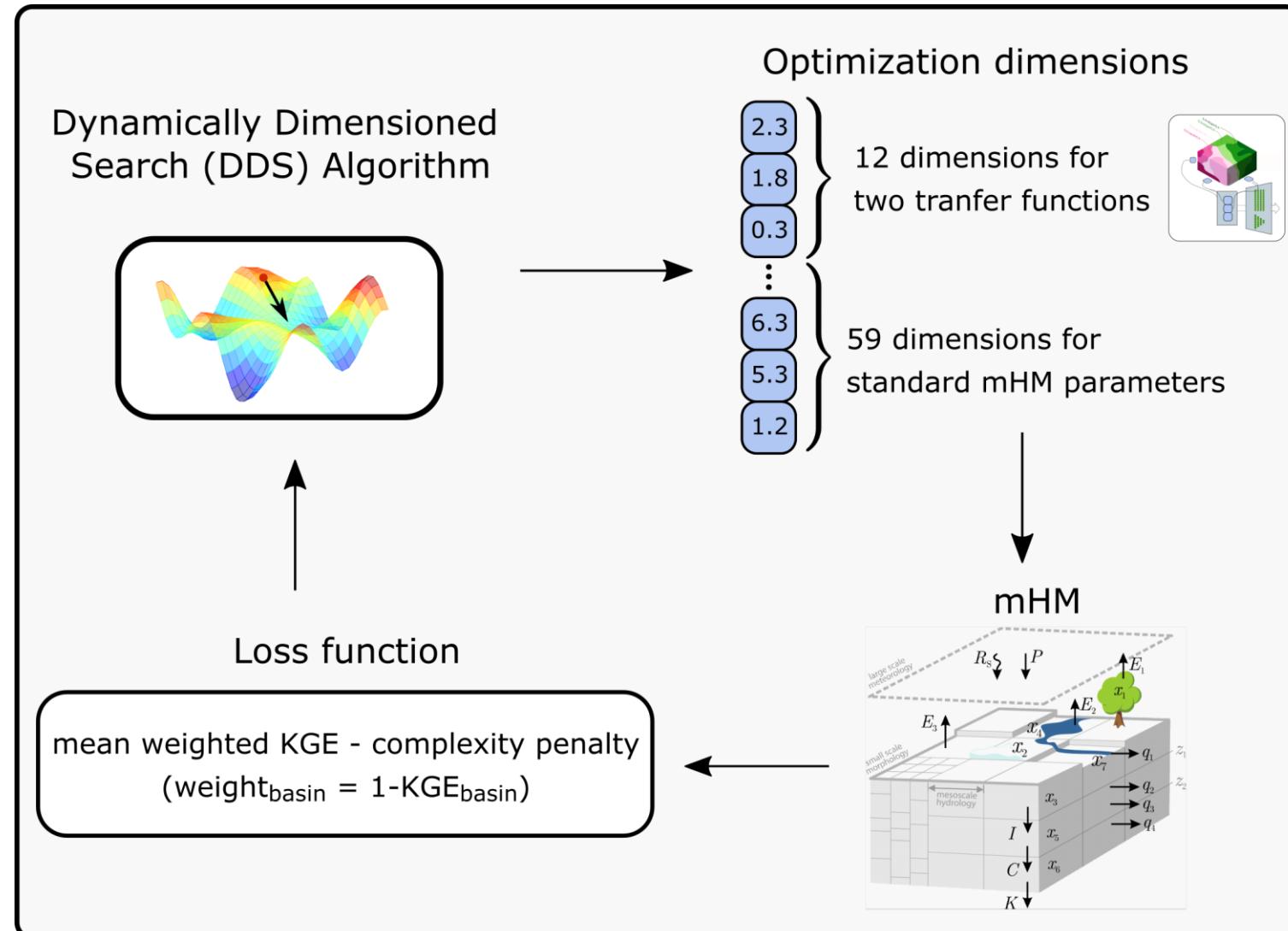


220 Validation basins

Time series:

7 & 220 gauging stations
Calibration: 2000-2004
Validation: 1965-1999
Spin-up: 5 years

Case study – Optimization



FSO optimization using the DDS algorithm (Tolson & Shoemaker, 2007)

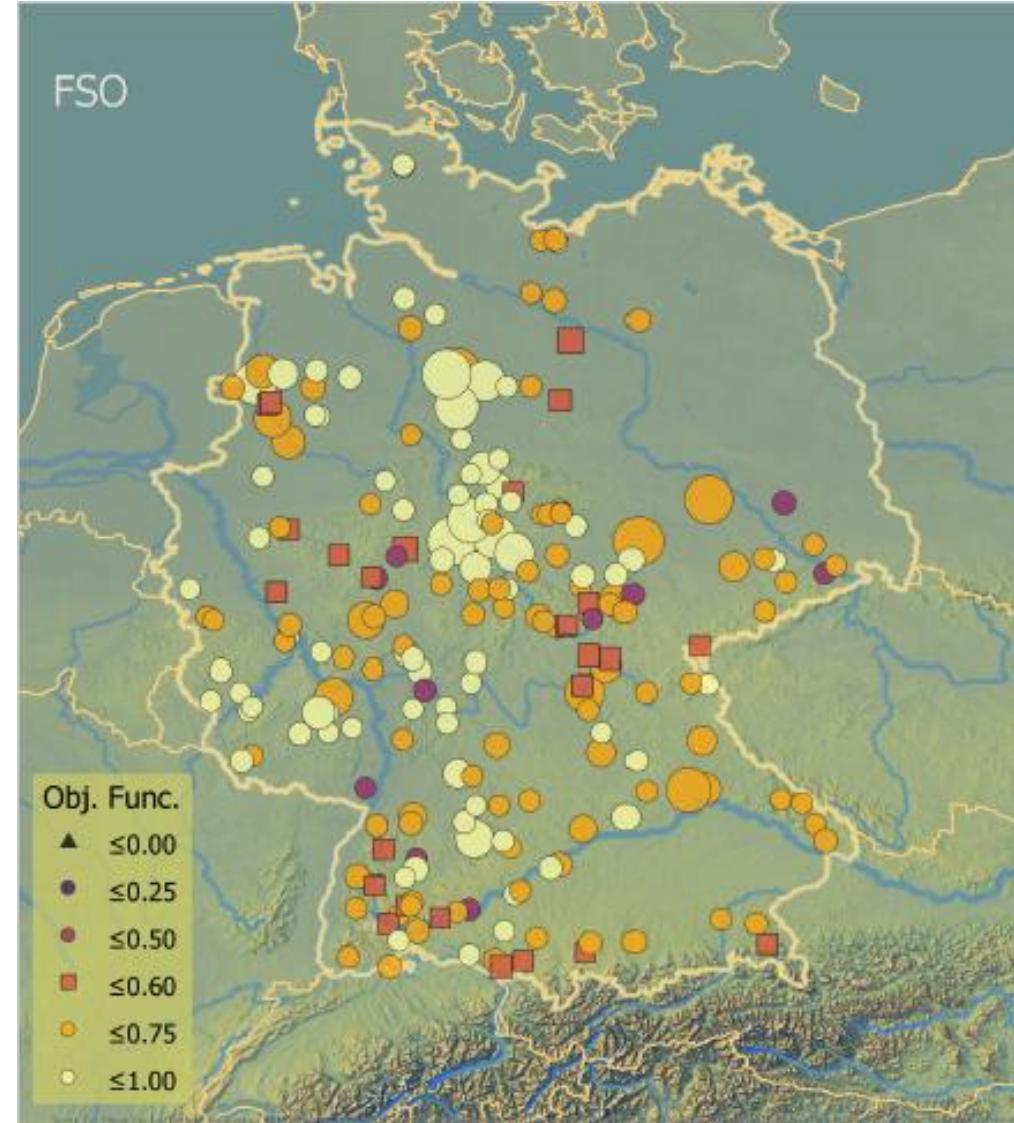
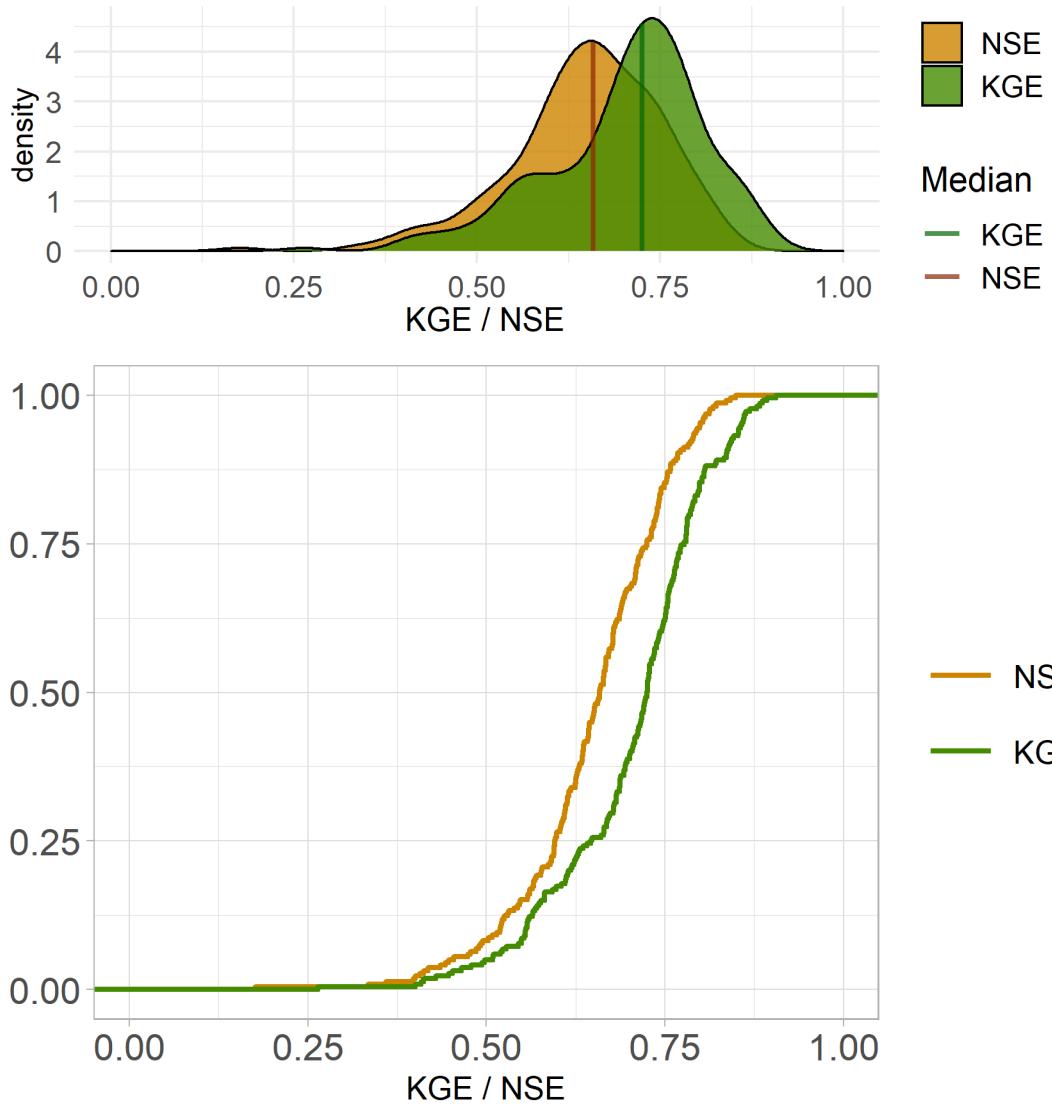
Preliminary results – 7 Training Basins

Training Basins KGE Results

	Period	median KGE	Main	Neckar	Weser	Ems	Saale	Mulde	Donau
FSO-mHM	Calibration	0.83	0.90	0.85	0.90	0.82	0.81	0.77	0.82
	Validation	0.80	0.85	0.83	0.89	0.80	0.77	0.65	0.71

FSO results after approx. 900 iterations

Preliminary results – 220 validation basins



Imhof-Like Background Topography by @John_M_Nelson

Saturated hydraulic conductivity (cm/d):

mHM: $K_{Sat} = \gamma_1 * \exp(\gamma_2 + \gamma_3 * \text{sand} - \gamma_4 * \text{clay}) * \log(10)$

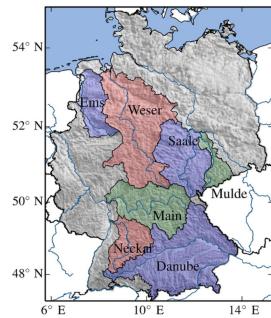
FSO-mHM: $K_{Sat} = \text{elevation} + \exp(\text{bulk density}) - 3.14$

Field Capacity (-):

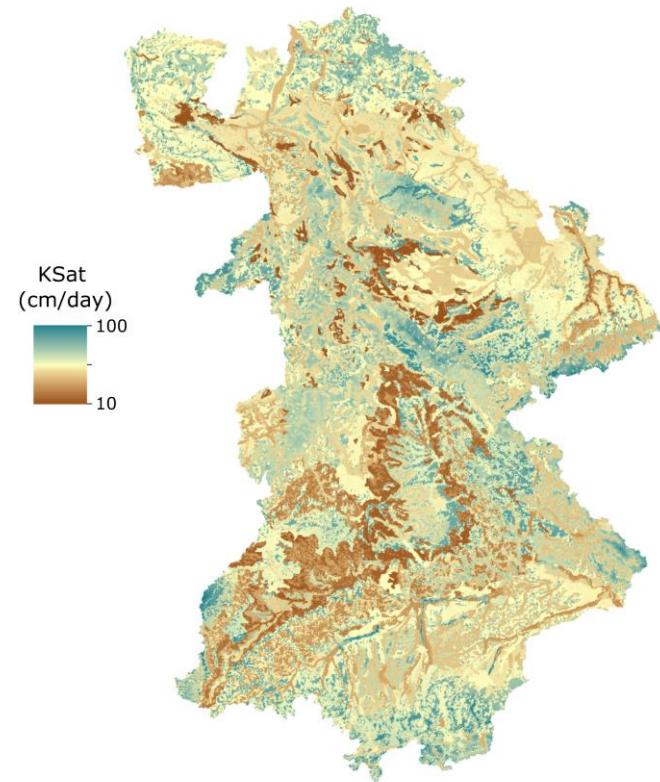
mHM: $\text{FieldCap} = \Theta_S * \exp(\gamma_5 * (\gamma_6 + \log_{10}(K_{Sat})) * \log(v_{Genu_n})$

FSO-mHM: $\text{FieldCap} = -0.336 \sqrt{0.333 / \sqrt{\text{bulk density}}}$

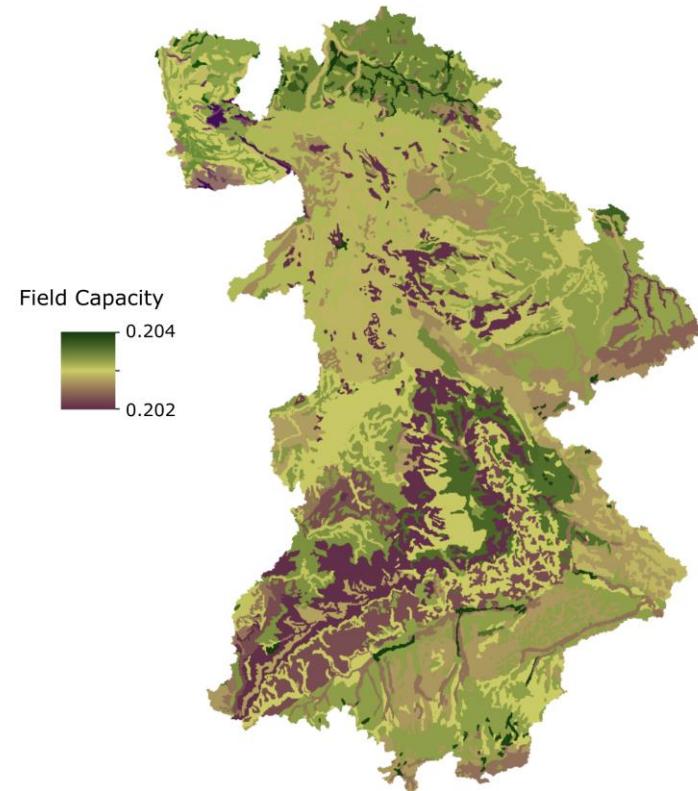
Preliminary results – estimated parameter fields



Saturated Hydraulic
Conductivity (cm/day)



Field Capacity (-)



Resulting parameter fields on the 100×100 m grid for the top layer of
the model (tillage layer, first 20 cm)

Summary, Discussion & Outlook

- FSO trained with 5 years data of 7 gauging stations:
 - training median KGE = 0.80
 - PUB median KGE = 0.73
- Preliminary results look promising → only 900 iterations
- Field Capacity is constant → most likely local minimum → continue optimization
- Multiple longer optimization runs needed for robust performance evaluation
- Compare validation basins results with performance of Zink et al. (2017)
- Comparison of final FSO parameter fields to geophysical properties

References

- Feigl, M., Herrnegger, M., Klotz, D., & Schulz, K. (2020).** Function Space Optimization: A symbolic regression method for estimating parameter transfer functions for hydrological models. *Water resources research*, 56(10), <https://doi.org/10.1029/2020WR027385>
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