

Model–data comparison with permutation entropy: Moving beyond summary statistics

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Background

- Soil water participates in terrestrial water and energy cycles.
- Soil matric potential is an important driving force for soil water flow in unsaturated soil.
- Movement of soil water is often modelled by the Richards equation.
- Model quality assessed by summary statistics (root mean square error, Nash–Sutcliffe coefficient)

Does the model capture the complexity of the underlying soil processes?

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Does the model capture the complexity of the underlying soil processes?

Complexity of a time series

- wide range of complexity: regular (low complexity) to random (high complexity)
- caused in natural time series by nonlinearity of underlying processes and their interactions
- soil matric potential
 - signal propagation from precipitation to throughfall to infiltration
 - influenced by soil hydraulic properties, evapotranspiration and possible measurement errors

Permutation entropy (PE)

Definition

$$H(n) = - \sum p(\pi) \log p(\pi)$$

$$PE = H(n) / \log(n!)$$

- works on ranks (order of values) (Bandt and Pompe, 2002)
- suitable for an arbitrary series of observations

Example

$$x = (3.1; 5.2; 7.9; -3.1; 4.0; 11.9; 0.3)$$

$$n = 2$$

$$A = \{(x_t, x_{t+1}) | x_t < x_{t+1}\}$$

$$p(A) = 4/6$$

$$B = \{(x_t, x_{t+1}) | x_t > x_{t+1}\}$$

$$p(B) = 2/6$$

$$H(2) = -4/6 \log(4/6) - 2/6 \log(2/6) \approx 0.92$$

$$PE = H(2) / \log(2!) \approx 0.92$$

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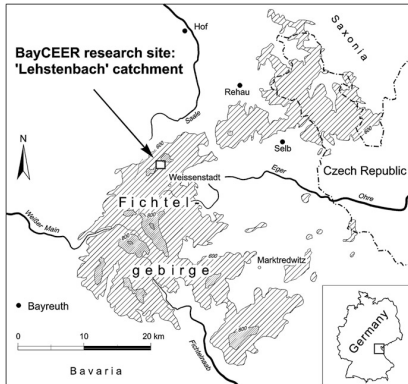
Weighted permutation entropy (WPE)

- for tracking abrupt changes
- weights the frequency of permutations by variance of values (Fadlallah et al., 2013)
- better suited for time series with large amplitudes in the signal compared to noise
- small WPE: small number of different patterns (regular/monotonic signal)
- large WPE: large number of different patterns (typical for noise)

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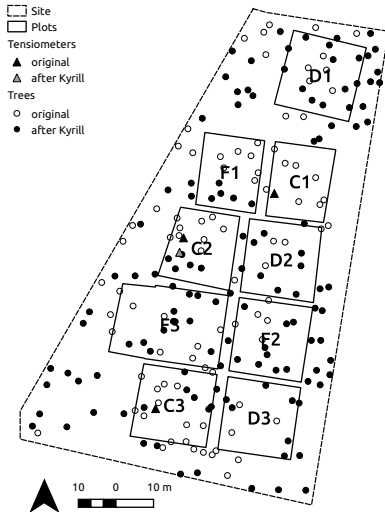
Study area – Lehstenbach catchment



Gerstberger et al., 2004

- 4.5 km², 695–877 m a.s.l
- annual precip 1 162 mm, mean temp 5.3 °C
- Norway spruce (*Picea abies*), age 65 years

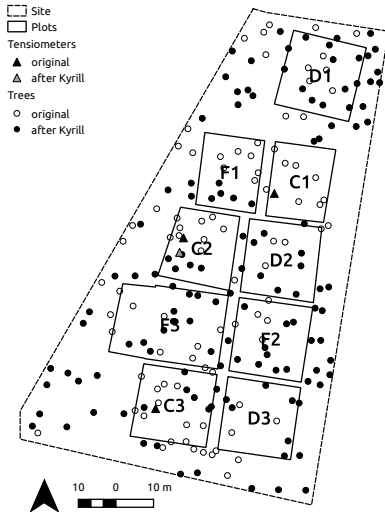
Plots



- Cambisols and Podzols
- sandy to loamy texture
- up to 15-cm thick mor-type organic layer

Bogner et al., 2017

Plots



Storm Kyrill



Bogner et al., 2017

Modelling the soil matric potential

WHNSIM

- Water Heat and Nitrogen Simulation Model (Huwe and Totsche, 1995)
- solves one-dimensional Richards equation
- K_{sat} measured with constant head method
- soil-water retention curve measured in the lab on undisturbed soil cores
- atmospheric conditions measured at nearby sites (Foken et al., 2017)
- model run in forward mode on the daily basis

Calculating the entropy

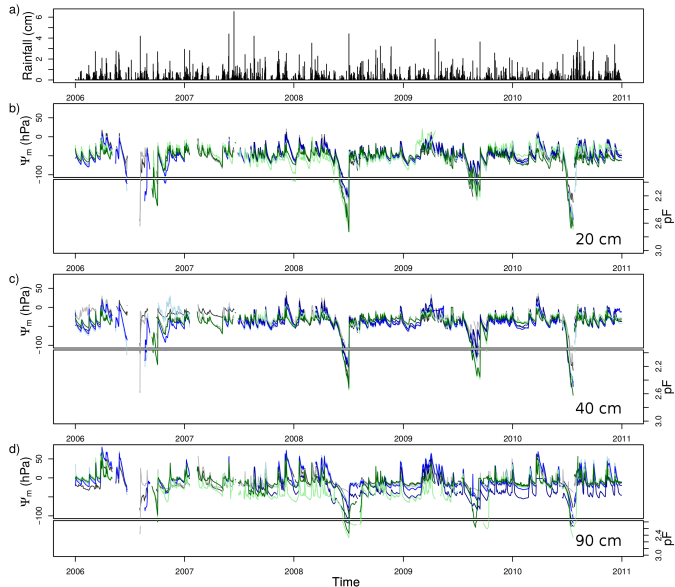
Data pretreatment

- measured matric potential at C1 (the completest series)
- daily median values (from hourly data)
- gap filling with Singular Spectrum Analysis (Golyandina and Osipov, 2007, e.g.)

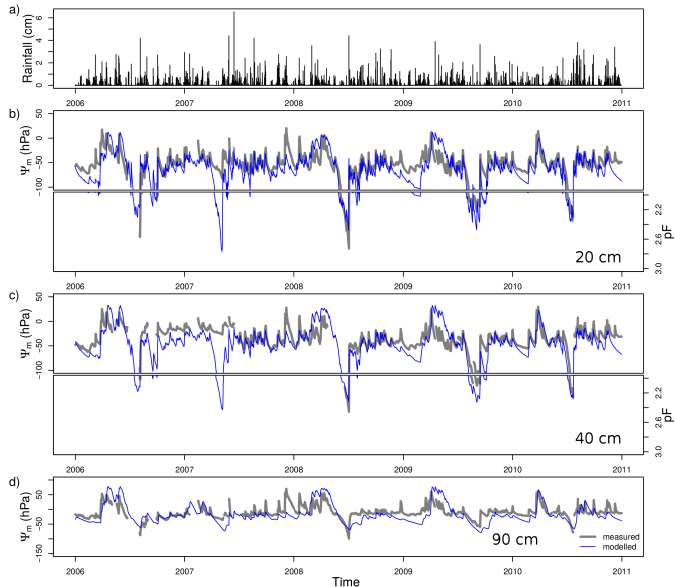
Weighted permutation entropy

- pretreated measured and modelled matric potential
- window length 90 days, slid by 1 day
- order $n = 4$

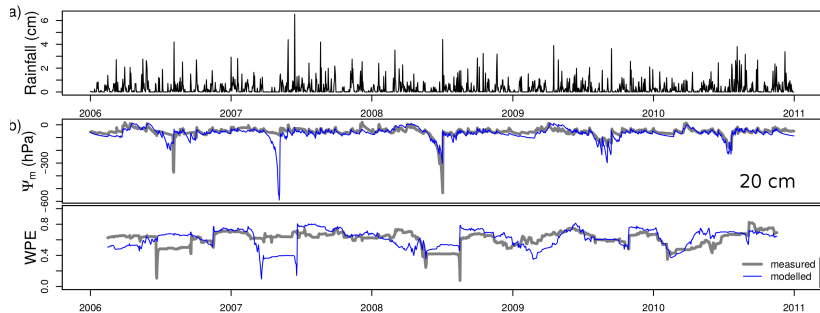
Measured matric potential



Modelled matric potential

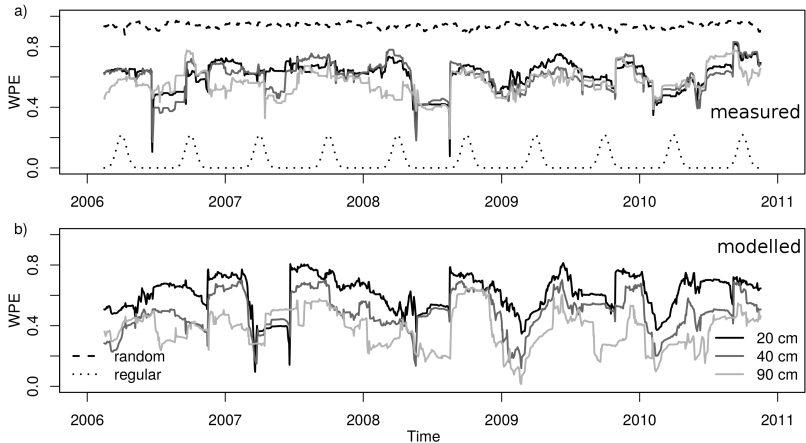


Weighted permutation entropy



Bogner et al., 2017

Entropy in different depths



Bogner et al., 2017

Lessons learned

- WHNSIM captures the complexity of measured matric potential in the upper soil.
- Synchronicity of matric potential in different depths is reproduced.
- The modelled signal in 90 cm is less complex (damped) than the measurements.
- Some process might be missing from the model (in the deeper soil).

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