

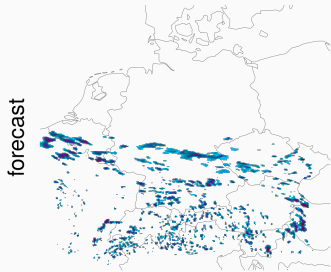
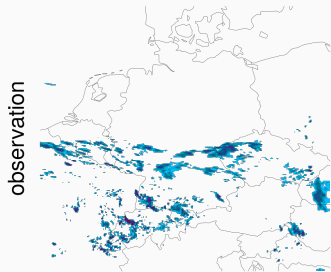
# Spatial forecast verification with wavelets

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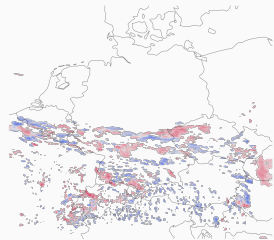
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8.7.2019

Institute of Geoscience, University of Bonn



difference image



*Double penalty: Displacement punished twice*

- Use wavelets to extract a field's structure
- Verify structure, disregard location
- Test the procedure using synthetic rain fields

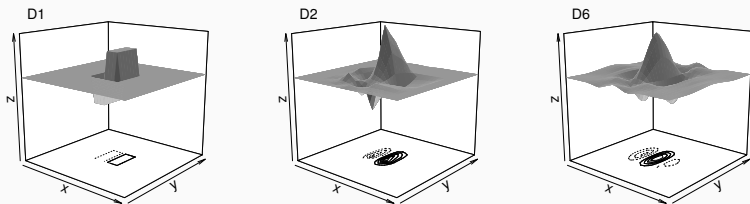
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# Method

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# The 2D redundant discrete wavelet transform



Transform field to a new basis, generated by a **mother wavelet**  $\psi(\mathbf{r})$

Rotation:  $\psi_{\text{vert}}(\mathbf{r}), \psi_{\text{horiz}}(\mathbf{r}), \psi_{\text{diag}}(\mathbf{r})$

Translation:  $\psi_{\mathbf{u}}(\mathbf{r}) = \psi(\mathbf{r} - \mathbf{u})$

Scaling:  $\psi_j(\mathbf{r}) = \psi(\mathbf{r}/2^j)$

→ **daughter wavelets**  $\psi_{\text{dir},\mathbf{u},j}(\mathbf{r})$

# Procedure

1.  $2^J \times 2^J$  intensities  $\xrightarrow[\text{transform}]{\text{redundant}}$   $2^J \times 2^J$  local wavelet-spectra

[www.youtube.com/embed/4fEkw76GXLk?autoplay=1](http://www.youtube.com/embed/4fEkw76GXLk?autoplay=1)

# Procedure

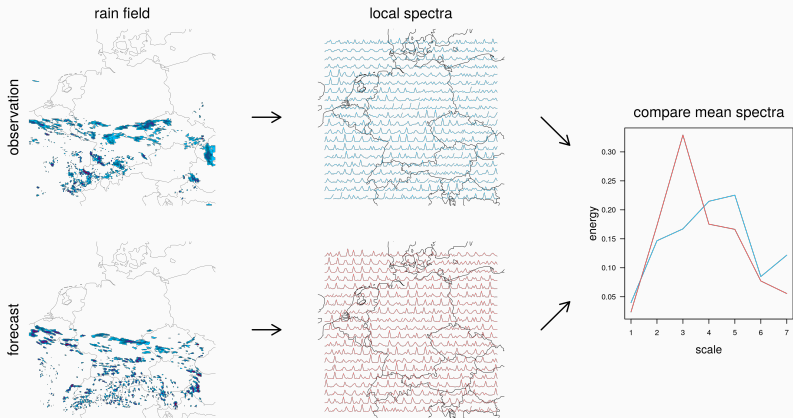
1.  $2^J \times 2^J$  intensities  $\xrightarrow[\text{transform}]{\text{redundant}}$   $2^J \times 2^J$  local wavelet-spectra

[www.youtube.com/embed/4fEkw76GXLk?autoplay=1](http://www.youtube.com/embed/4fEkw76GXLk?autoplay=1)

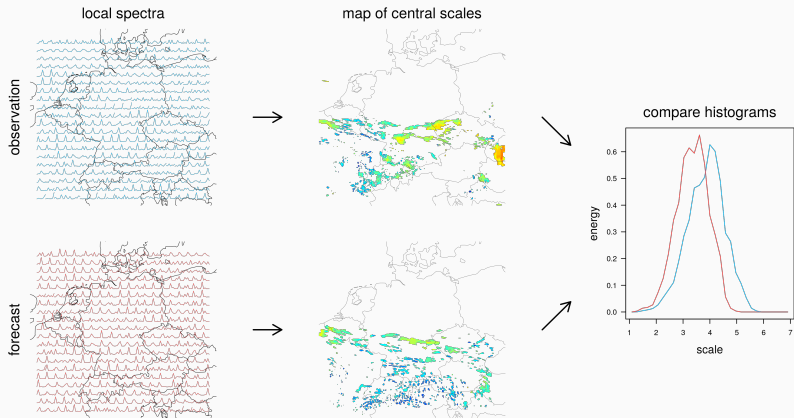
2. Average over the three directions (*for now ...*)
3. remove bias towards large scales (Eckley et al. 2010), normalize
4. aggregate in space ... *but how?*
5. verify ... *but how?*



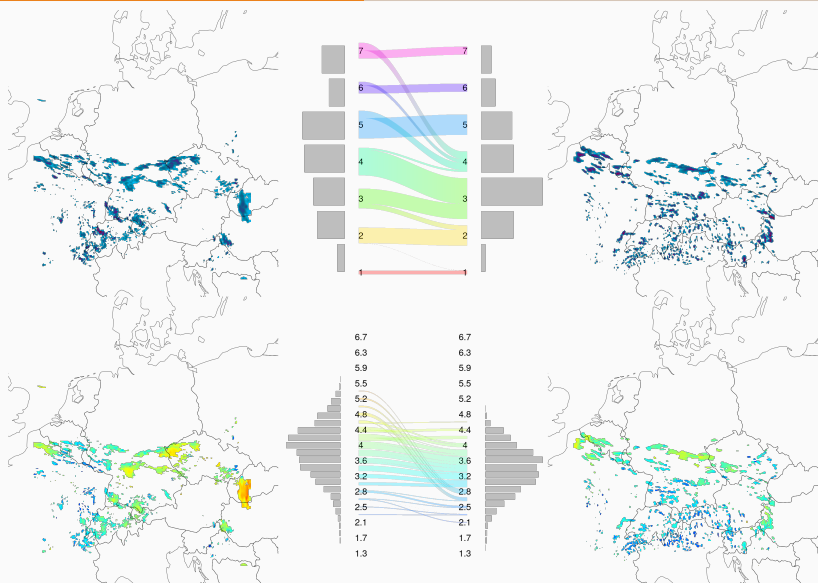
# Aggregation #1: Spatial mean spectra



# Aggregation #2: Map of central scales



# Verification: Earth Mover's distance



EMD bounded from below by the difference in centre → obtain **sign**

# Random rain fields

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# A stochastic model for precipitation

Hewer (2018): “Stochastisch-physikalische Modelle für Windfelder und Niederschlagsextreme”

$$\text{Precip} = \max \left( \underbrace{E}_{\text{evap.}} - \underbrace{T}_{\text{thresh.}} - \underbrace{\mathbf{v} \cdot \nabla q}_{\text{advection}} - \underbrace{q \nabla \cdot \mathbf{v}}_{\text{convergence}}, 0 \right),$$
$$\mathbf{v} = \nabla \times \Psi + \nabla \chi$$

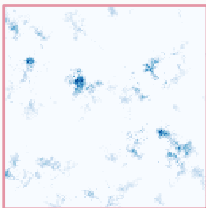
$$\Psi, \chi, q \sim \mathcal{N} \left( \mathbf{0}, \text{Matérn} \left( \underbrace{\|b\|}_{\text{scale}} (\mathbf{t} - \mathbf{s}), \underbrace{\nu}_{\text{smoothness}} \right) \right)$$

large  $\nu \rightarrow$  smooth fields

large  $b \rightarrow$  small features

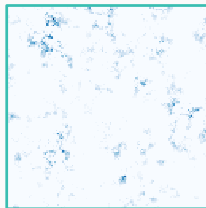
# Simulated rain fields

$v = 2.5, b = 0.1$



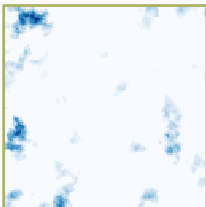
rough &  
large scale

$v = 2.5, b = 0.2$



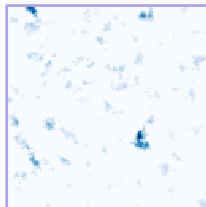
rough &  
small scale

$v = 3, b = 0.1$



smooth &  
large scale

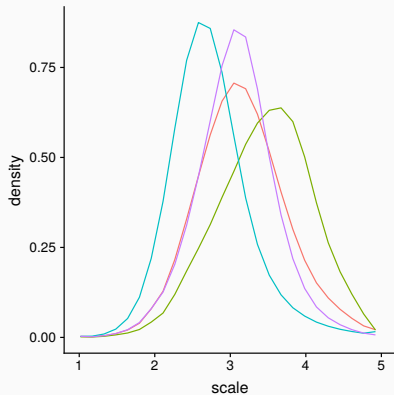
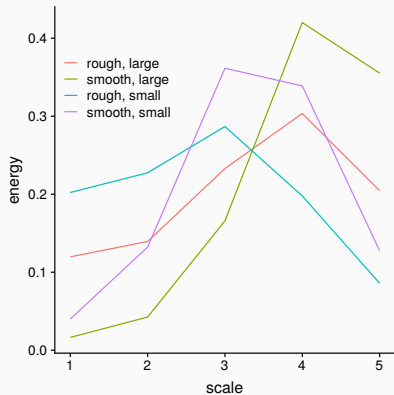
$v = 3, b = 0.2$



smooth &  
small scale

# Simulated rain fields

Average spectra and histograms over many realizations:



# Idealized experiments

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# Experiment set-up

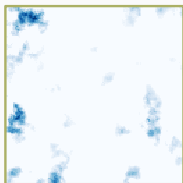
$v = 2.5, b = 0.1$



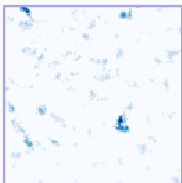
$v = 2.5, b = 0.2$



$v = 3, b = 0.1$



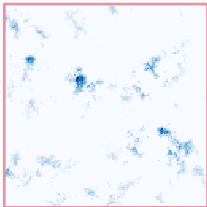
$v = 3, b = 0.2$



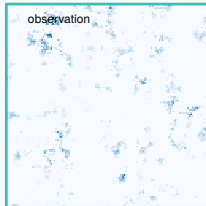
1. Draw an **“observation”** from one of the models
2. Draw one **“forecast”** from each model
3. Calculate **local wavelet spectra**
4. Verification using Earth mover's distance and distance in centre
5. Repeat **1000 times**

# Observations are rough and small-scaled

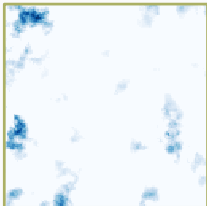
$v = 2.5, b = 0.1$



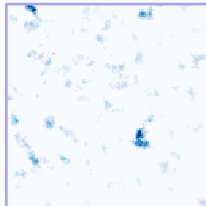
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$v = 3, b = 0.1$

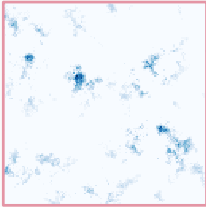


$v = 3, b = 0.2$

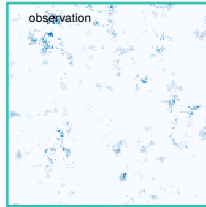


# Observations are rough and small-scaled

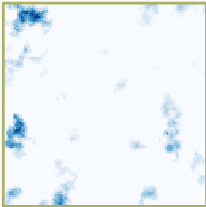
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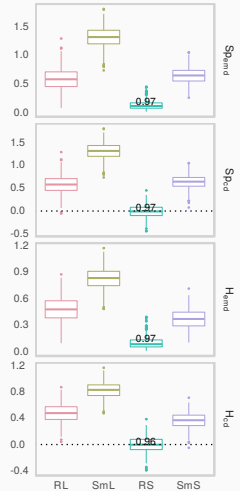
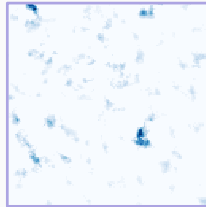
$v = 2.5, b = 0.2$



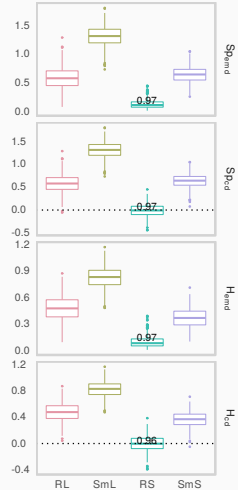
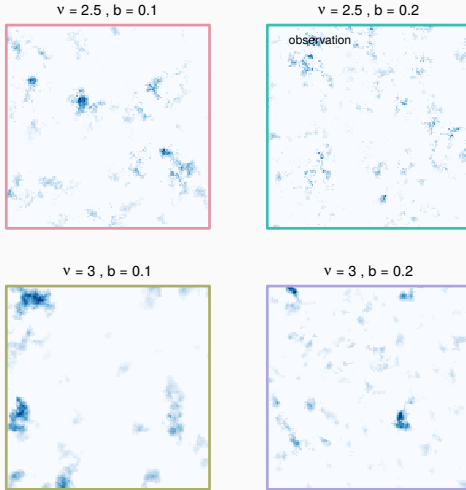
$v = 3, b = 0.1$



$v = 3, b = 0.2$



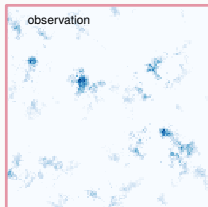
# Observations are rough and small-scaled



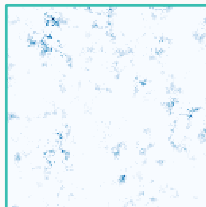
→ correct forecast gets best scores, all others are too large-scaled

# Observations are rough and large-scaled

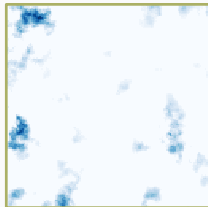
$v = 2.5, b = 0.1$



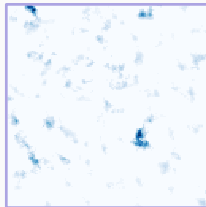
$v = 2.5, b = 0.2$



$v = 3, b = 0.1$

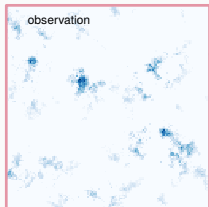


$v = 3, b = 0.2$

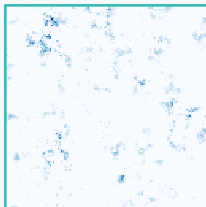


# Observations are rough and large-scaled

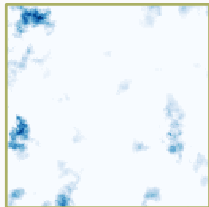
$v = 2.5, b = 0.1$



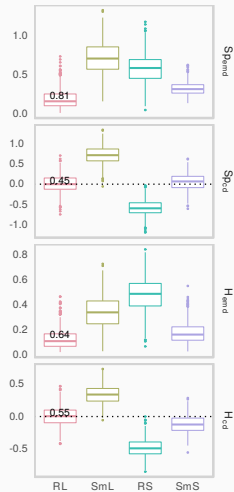
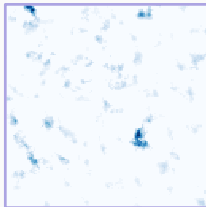
$v = 2.5, b = 0.2$



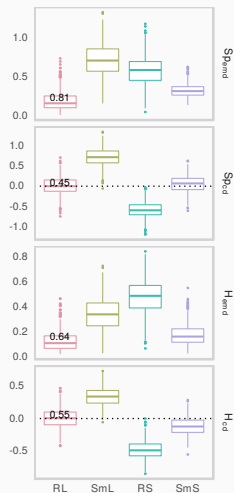
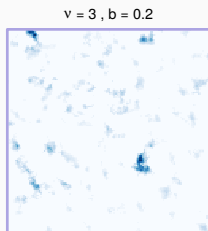
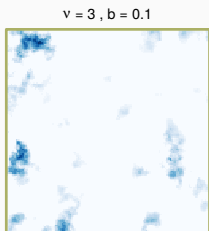
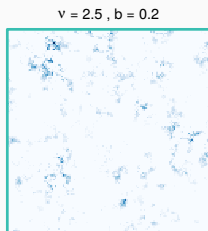
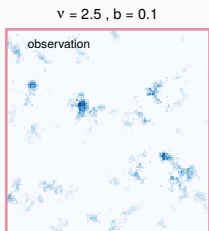
$v = 3, b = 0.1$



$v = 3, b = 0.2$



# Observations are rough and large-scaled

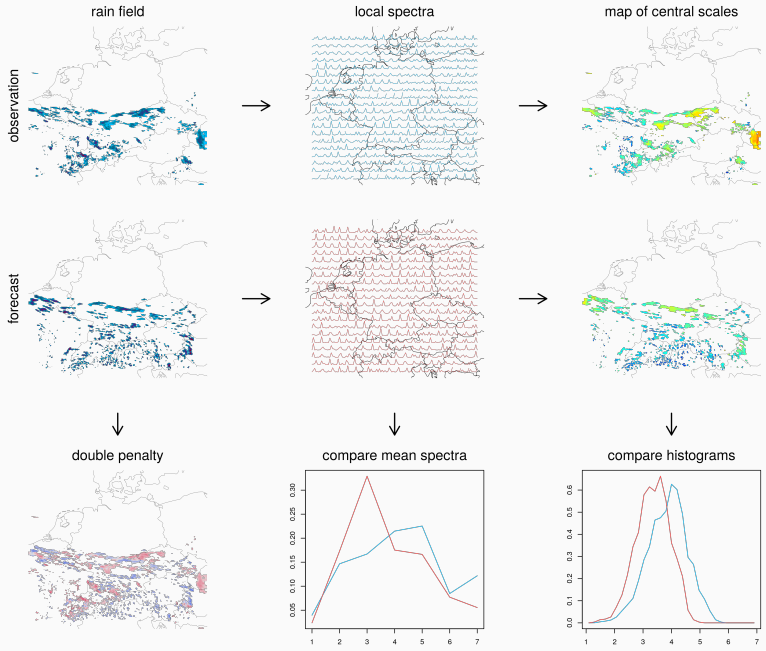


→ higher failure-rates, ( $\nu = 2.5, b = 0.1$ ) similar to ( $\nu = 3, b = 0.2$ )

## Summary and outlook

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- Wavelets extract the structure of rain fields
- Two ways of aggregating the information:  
Spatial mean and central scales
- Physically consistent random test cases:

Code for the random rain fields and wavelet-based verification:  
[https://github.com/s6sebusc/wv\\_verif](https://github.com/s6sebusc/wv_verif)

Buschow, S., Pidstrigach, J., and Friederichs, P.: *Assessment of wavelet-based spatial verification by means of a stochastic precipitation model (wv\_verif v0.1.0)*, Geosci. Model Dev. Discuss., <https://doi.org/10.5194/gmd-2019-90>, in review, 2019.

- Systematic study of real cases
- Include direction information
- Other variables are also possible

# References

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Idris A Eckley, Guy P Nason, and Robert L Treloar. “Locally stationary wavelet fields with application to the modelling and analysis of image texture”. In: *Journal of the Royal Statistical Society: Series C (Applied Statistics)* 59.4 (2010), pp. 595–616.



Rüdiger Hewer. “Stochastisch-physikalische Modelle für Windfelder und Niederschlagsextreme”. PhD thesis. University of Bonn, 2018.