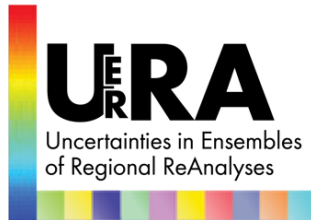




Schweizerische Eidgenossenschaft
Confédération suisse
Confederazione Svizzera
Confederaziun svizra

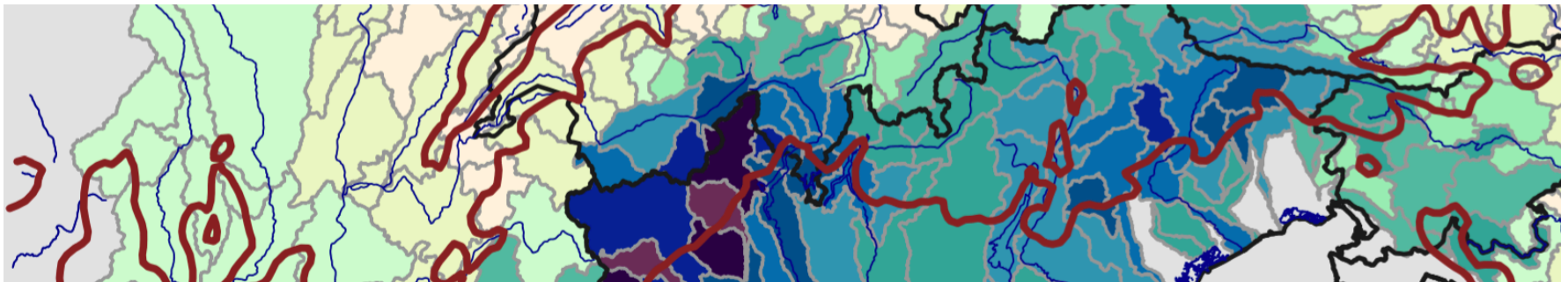
Federal Department of Home Affairs FDHA
Federal Office of Meteorology and Climatology MeteoSwiss



A Multi-Year Probabilistic Precipitation Dataset for the Alps: Results and Evaluation

Christoph Frei and Francesco Isotta

Federal Office of Meteorology and Climatology MeteoSwiss, Zürich





Motivation

- Spatial precipitation analyses are subject to error
 - Limited observations, limited predictability
 - Representativity errors (ambiguities about scale)
- Classical treatment of uncertainty is unsatisfactory
 - summary measures only
 - no information about scale dependence
- Ensembles are attractive
 - User can trace uncertainties through applications



Ensembles in Gridding

- Stochastic simulation

- ... based on "Gaussian random fields"
- ... 2nd order stationary covariance
- ... conditioned on observations

Ahrens & Jaun 2005
Bellerby & Sun 2005
Moulin et al. 2009
...

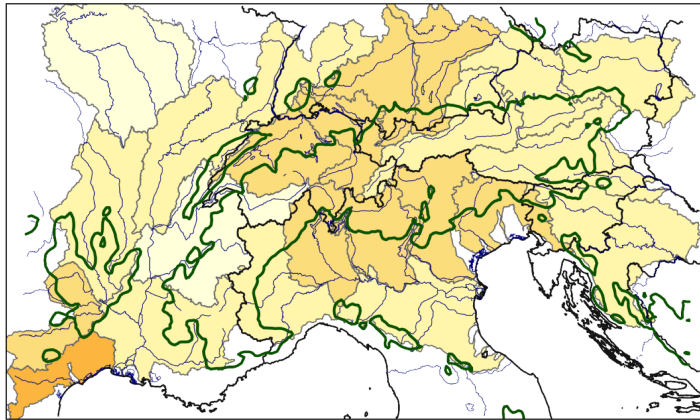
- Issues

- ★ • Gaussian model is unrealistic for precipitation.
- ★ • Spatial stationarity is unrealistic for large domains.
- ★ • "Frequentist" simulation neglects parameter uncertainty.
 - Measurement errors not accounted.
- ★ • Verification. Reliability of probabilistic estimates?
- ★ • Better than "poor man deterministic ensemble"?

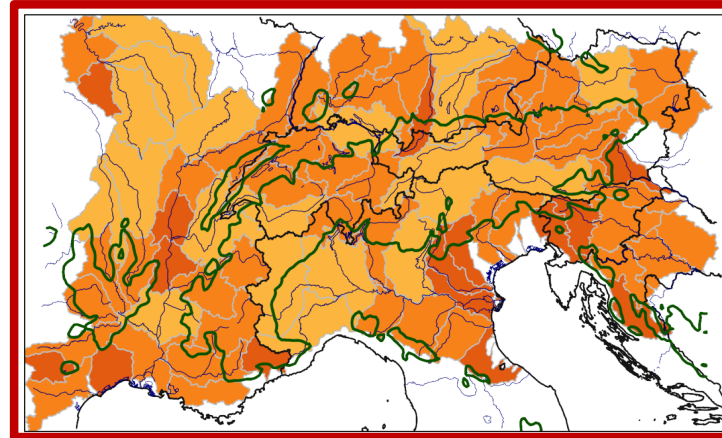


Hydrological Units

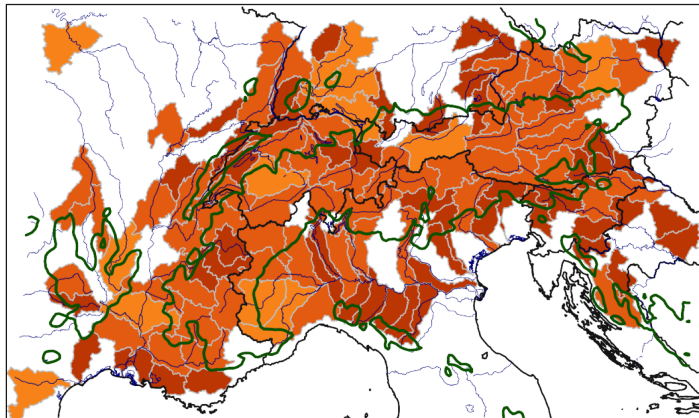
A: 14'000-44'000 km²



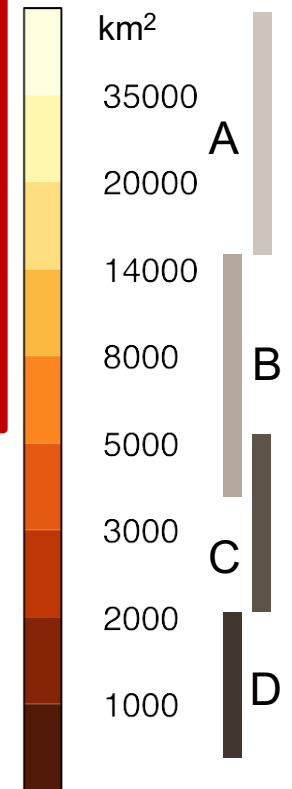
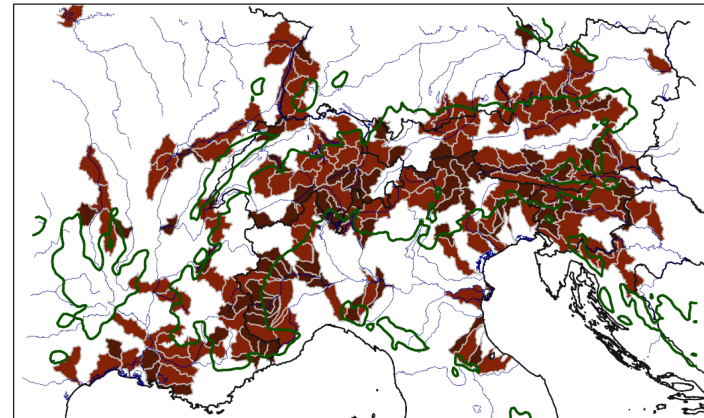
B: 3'500-14'000 km²



C: 2'000-5'000 km²



D: 500-2'000 km²



European River Catchments Dataset of EEA
altogether 534 hydrological units



Probabilistic Method

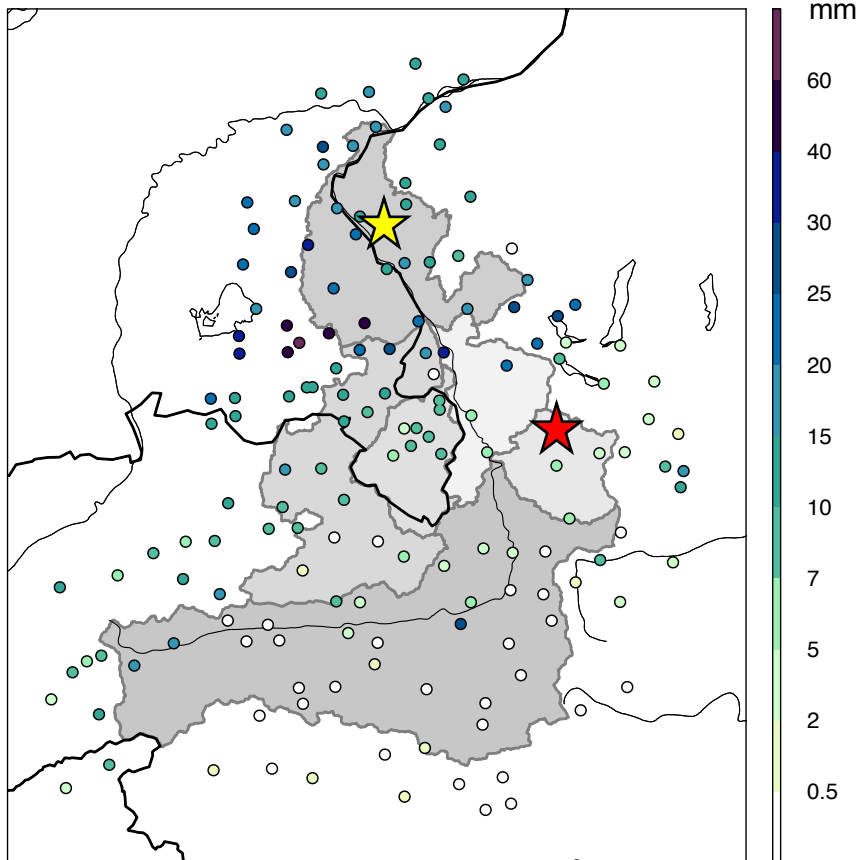
- Stochastic Model: Trans-Gaussian Random Fields
 - Box-Cox power transform
 - exponential spatial covariance & nugget effect
 - 5 parameters: transf. exponent, intercept, nugget, sill, range
- Inference: Bayesian
 - posterior (joint) distribution of model parameters
 - MCMC sampling with locally adaptive jump proposals
- Ensemble of Catch. Means: Conditional Simulation
 - at points of high-resolution grid within catchment (≤ 1 km)
 - conditioned on rain gauge data
 - average over all points in catchment (upscaling)

Extending ideas from: Ahrens & Jaun 2007, Moulin et al. 2009, Pappenberger et al. 2009, Erdin et al. 2012, Wilson et al. 2014

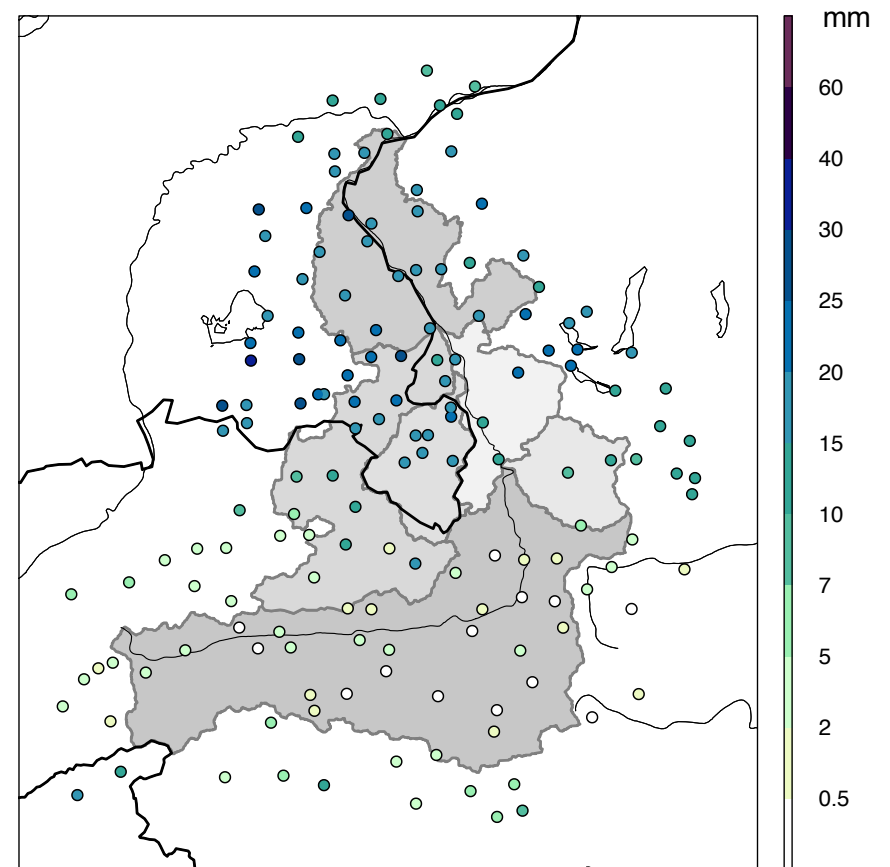


Example – Data

1990.06.30, convective



1990.10.29, stratiform



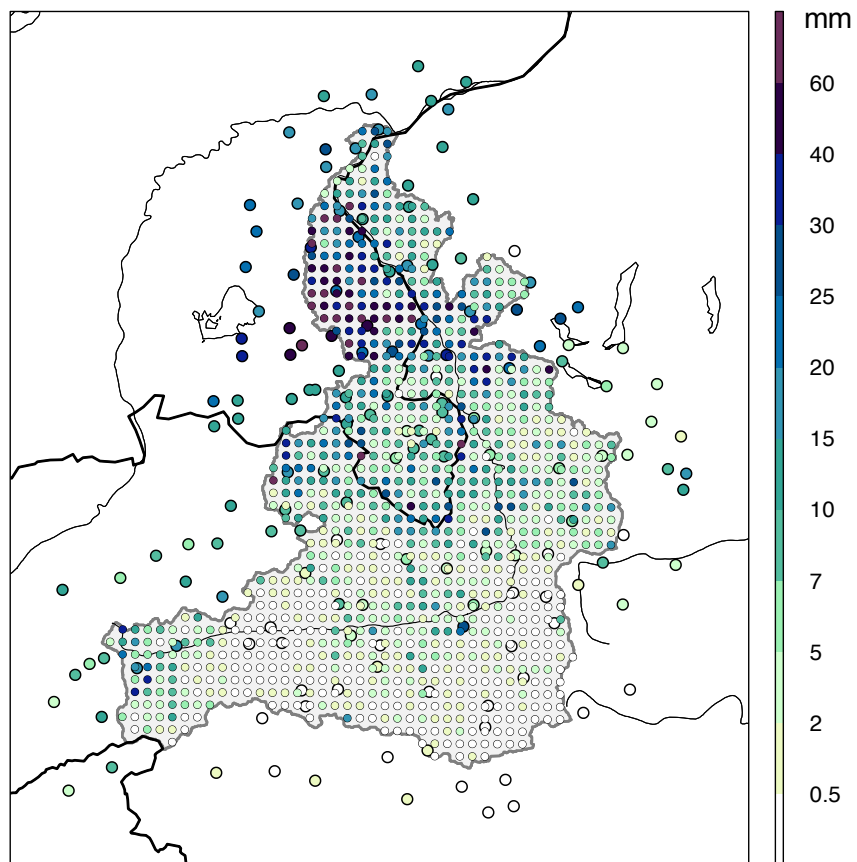
★
Salzach (6738 km²), Lower Salzach (1086 km²), Lammer (395 km²)



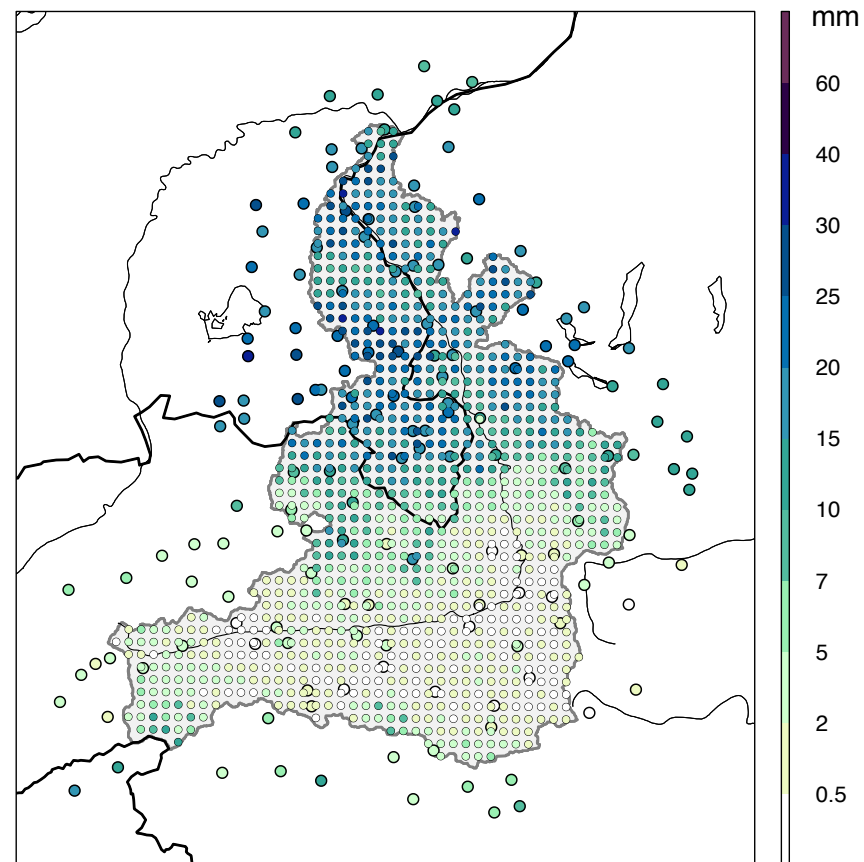


Example – Simulation

1990.06.30, convective



1990.10.29, stratiform

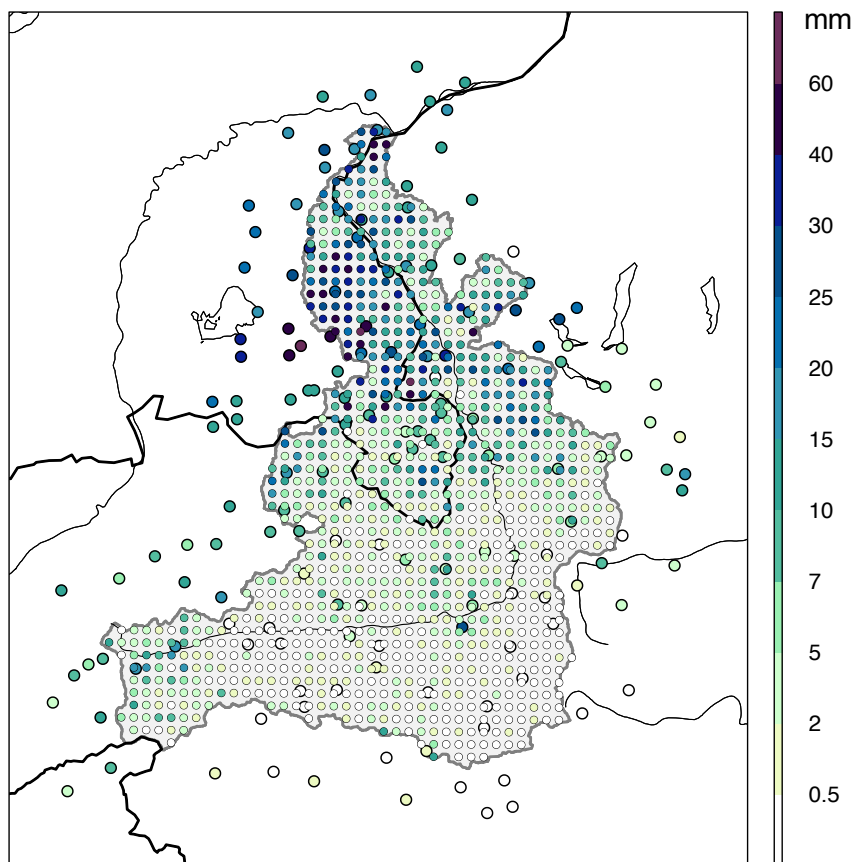


Ensemble Member

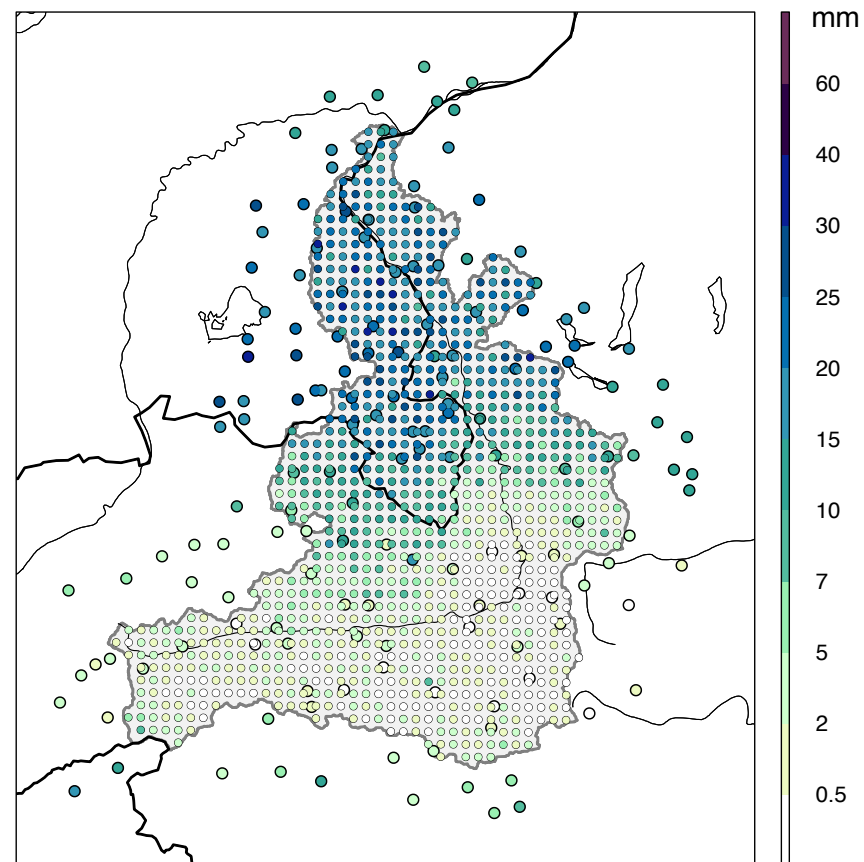


Example – Simulation

1990.06.30, convective



1990.10.29, stratiform

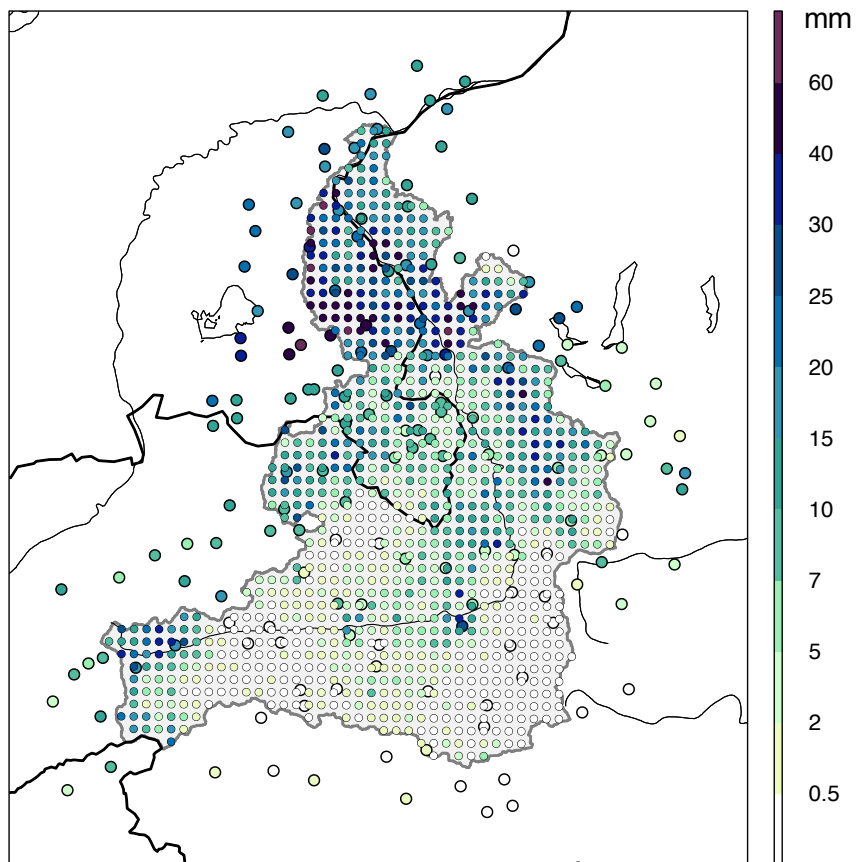


Ensemble Member

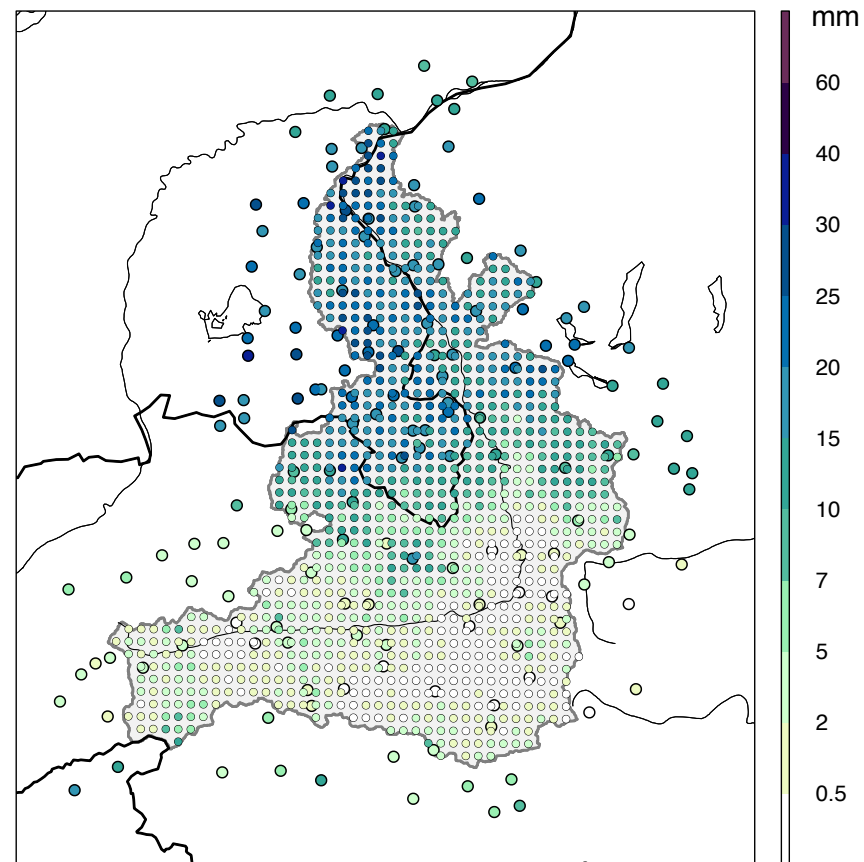


Example – Simulation

1990.06.30, convective



1990.10.29, stratiform

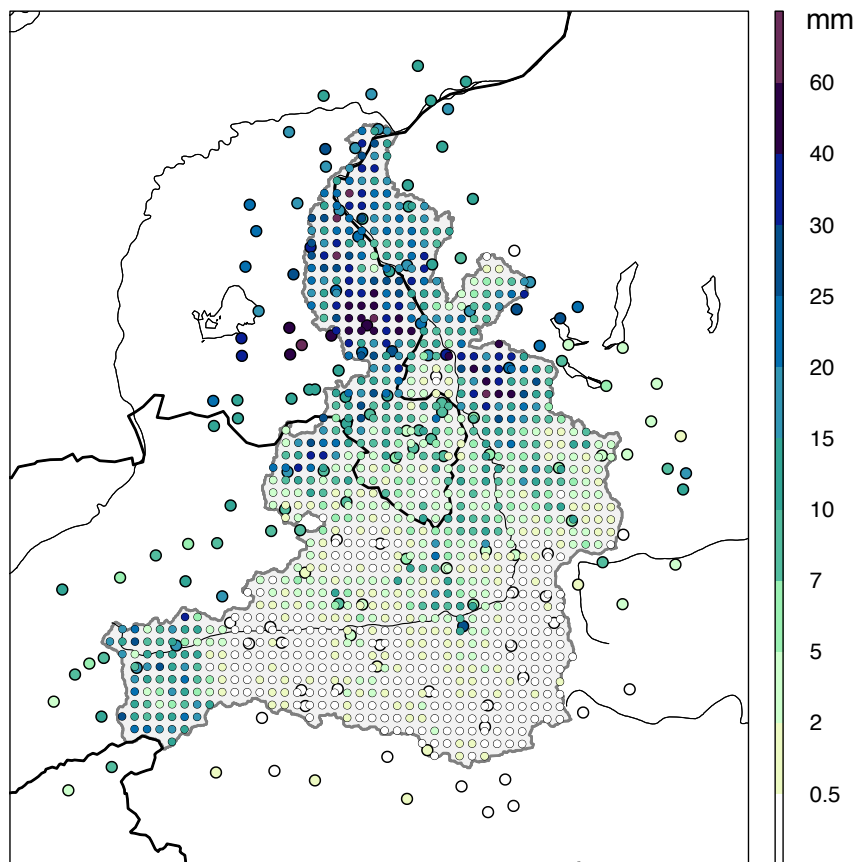


Ensemble Member

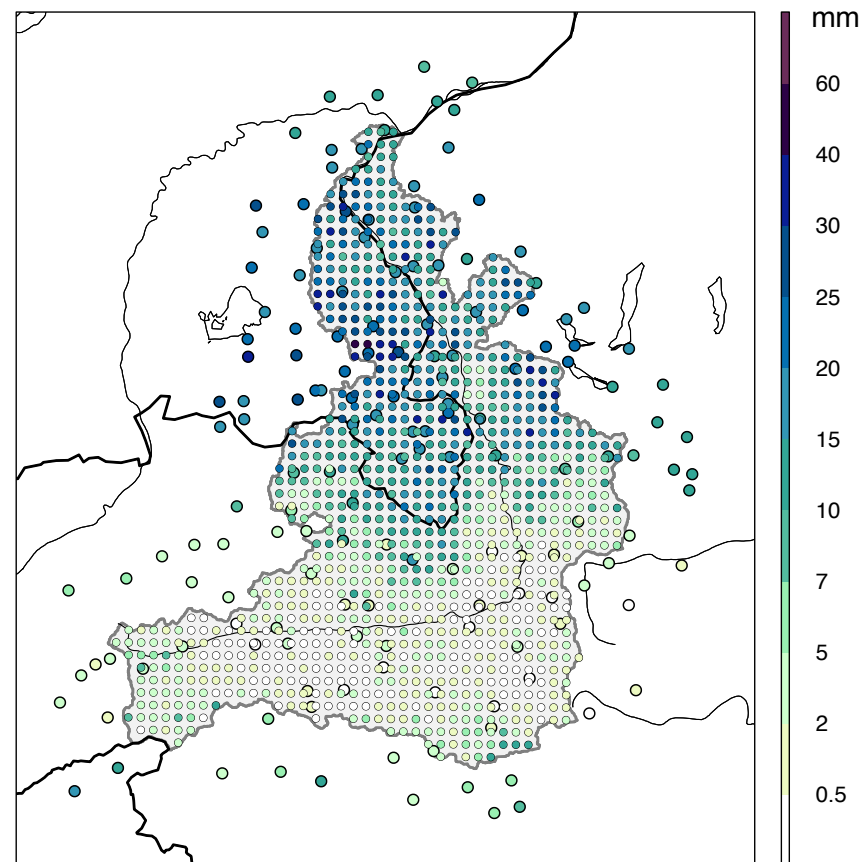


Example – Simulation

1990.06.30, convective



1990.10.29, stratiform



Ensemble Member

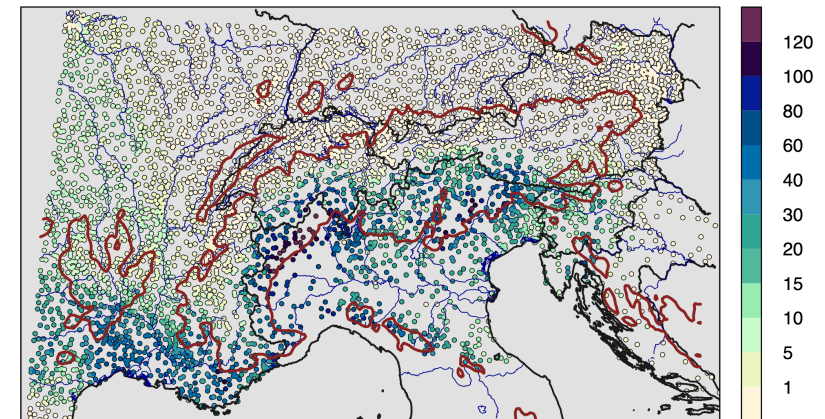


Pan-Alpine Probabilistic Dataset

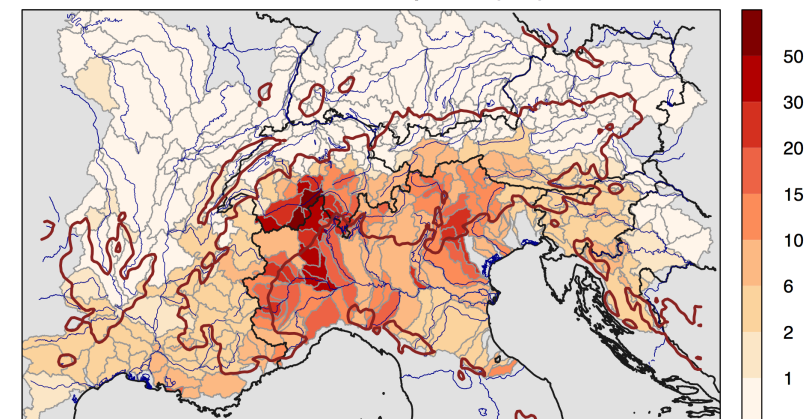
- Area-mean precipitation over hydrological units in the Alps
- Input data as for APGD
 - Isotta et al. 2014
 - ~6000 rain gauge obs per day
- 100 ensemble members
- 534 hydrological units,
 - based on EEA catchment DS
 - four hierarchical scales
- daily, 1971-2008
- 2003-2008 processed so far

2008.11.04

station measurements (mm)



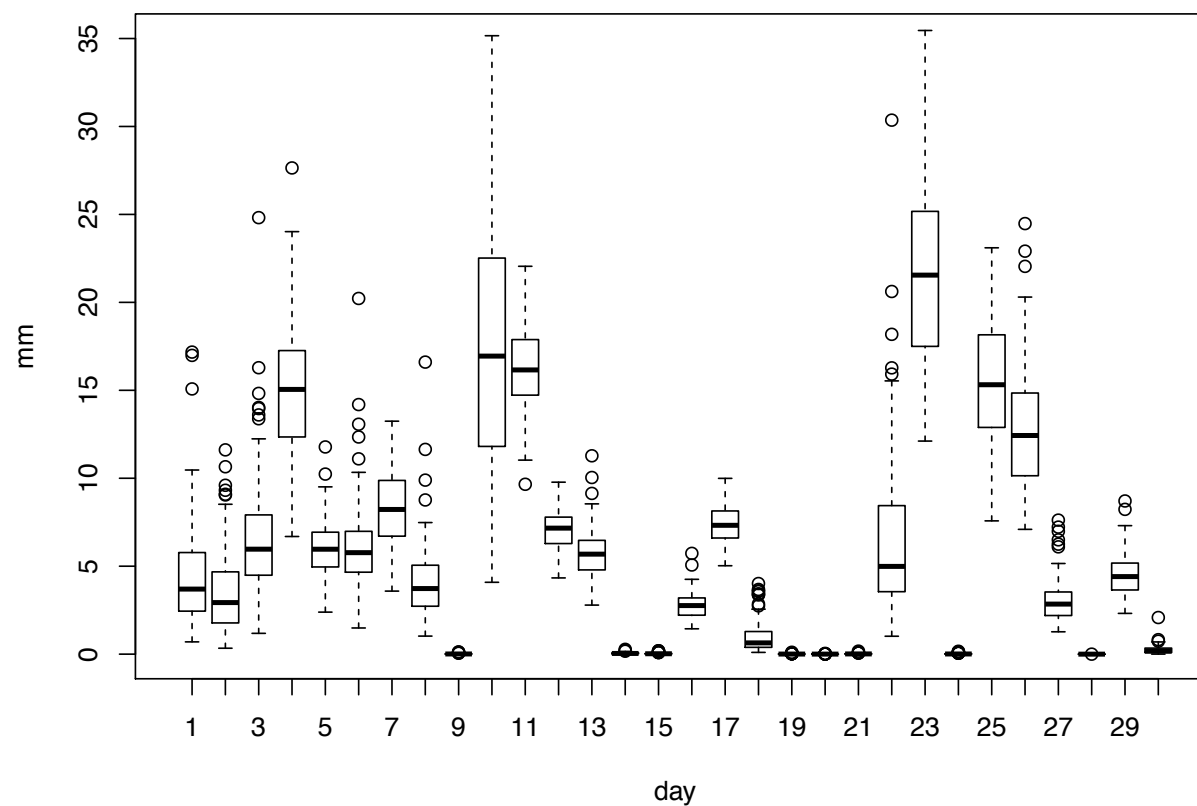
90% ensemble inter-quantile (mm)





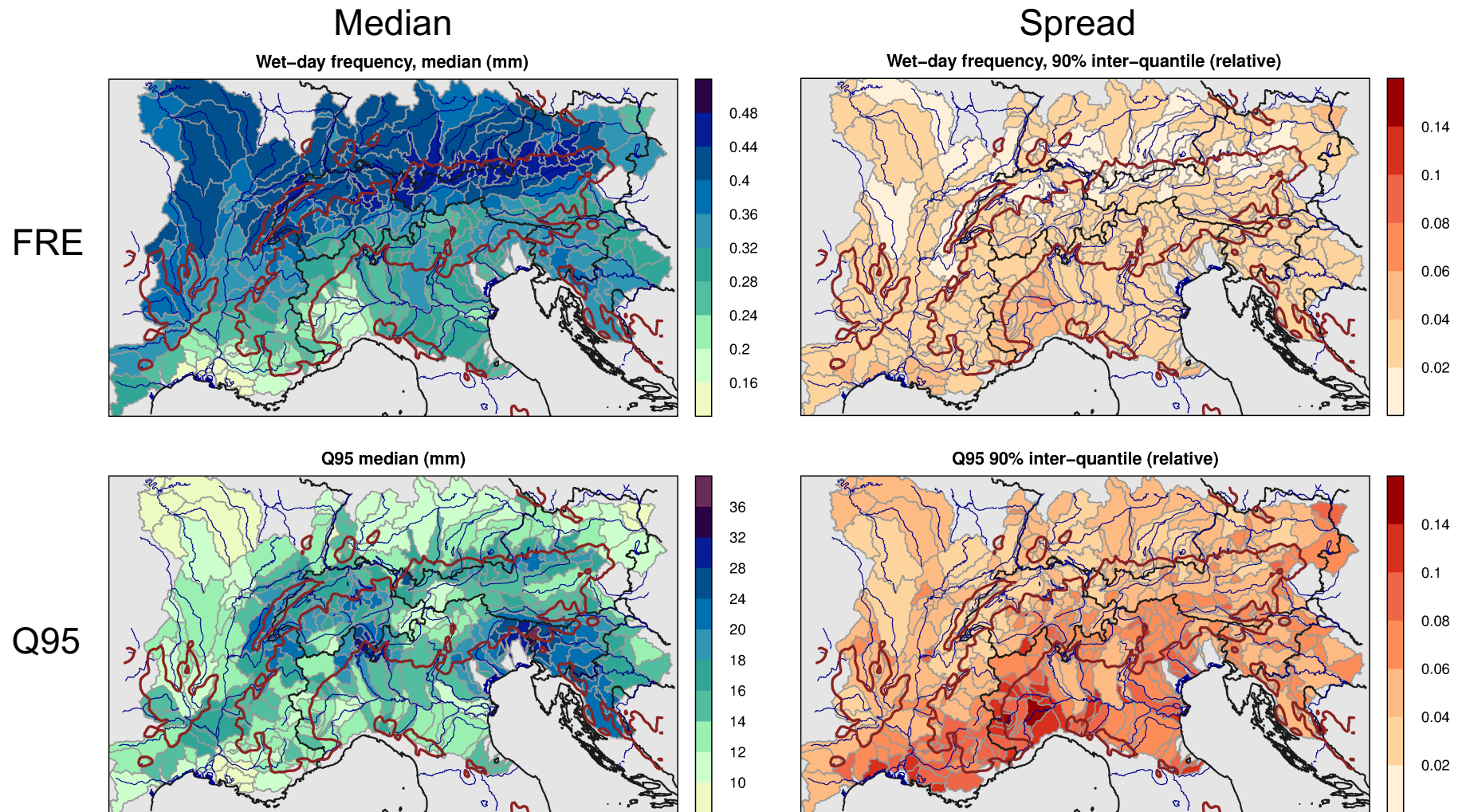
Magnitude of Uncertainties

Lammer (395 km², 3 stations within)
June 2008





Climate Indices



2004 – 2008, annual

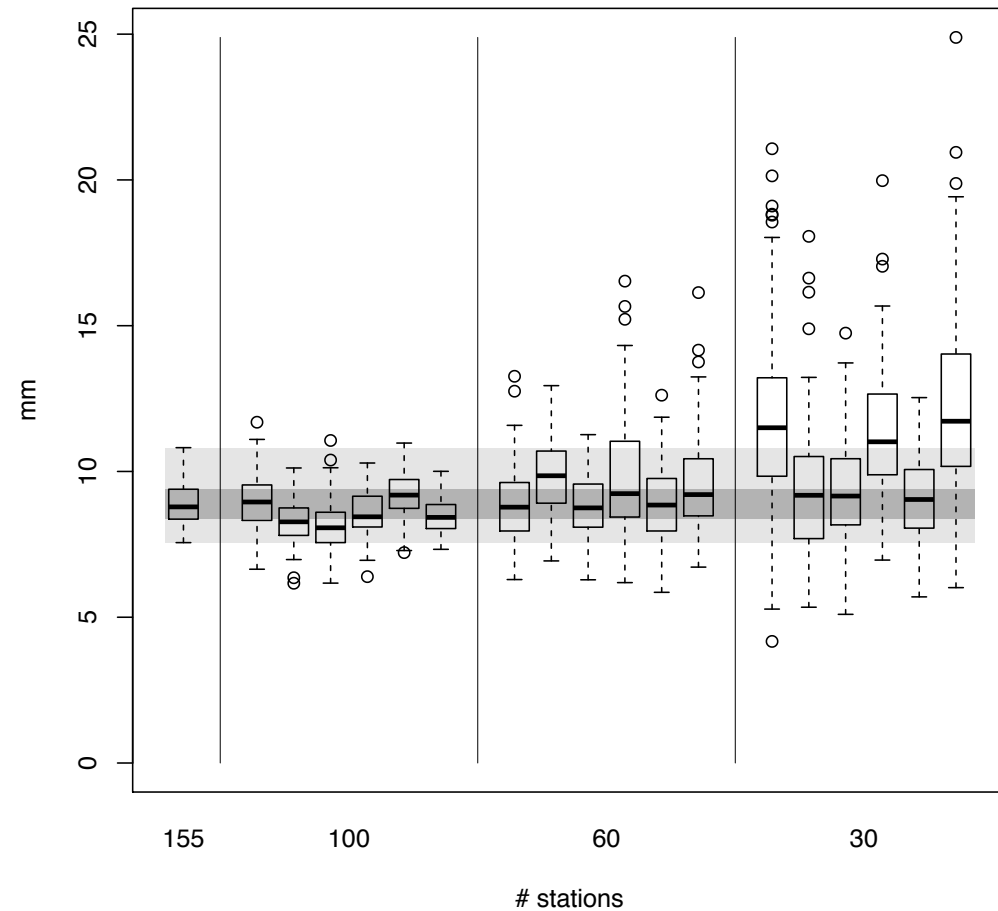


Uncertainty and Station Density

Experiment with
randomly “diluted”
station data

Box Plot:
Probabil. Ensembles

Salzach, 1990.06.30





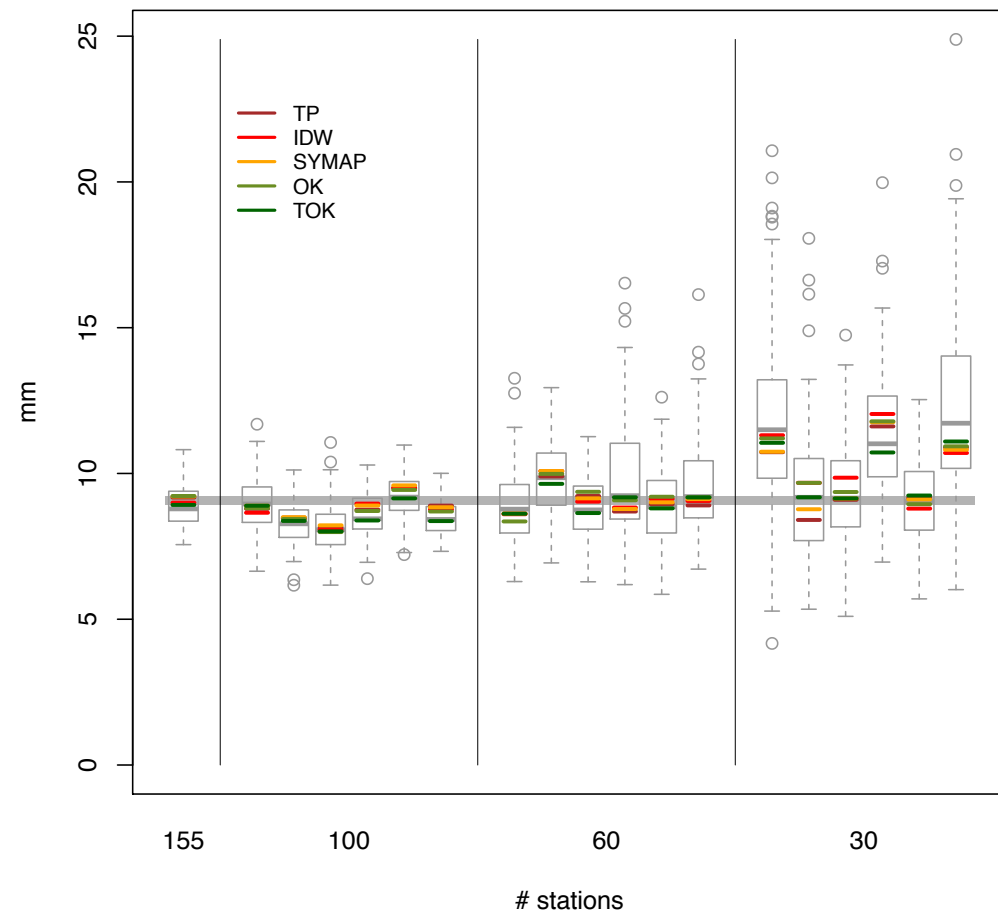
Uncertainty and Station Density

Experiment with
randomly “diluted”
station data

Box Plot:
Probabil. Ensembles

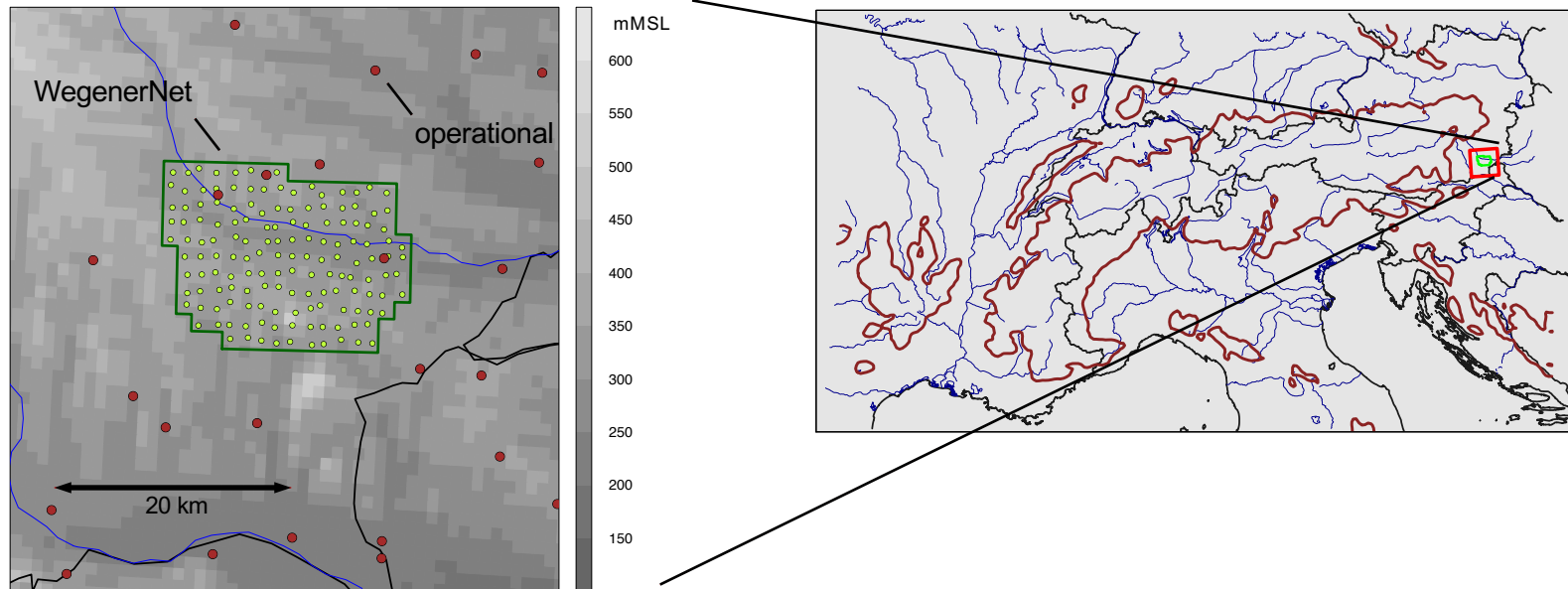
Colour:
Determin. Ensembles
“poor man ensemble”

Salzach, 1990.06.30





Wegener Net



An experimental network of high-resolution climate stations

20 x 15 km domain
150 stations, regularly spaced

5-min time resolution
2007 – present

wegenernet.org
Kirchengast et al. 2014; Sungmin et al. 2016

Evaluation

mean over all WN-Stations = reference
112 days

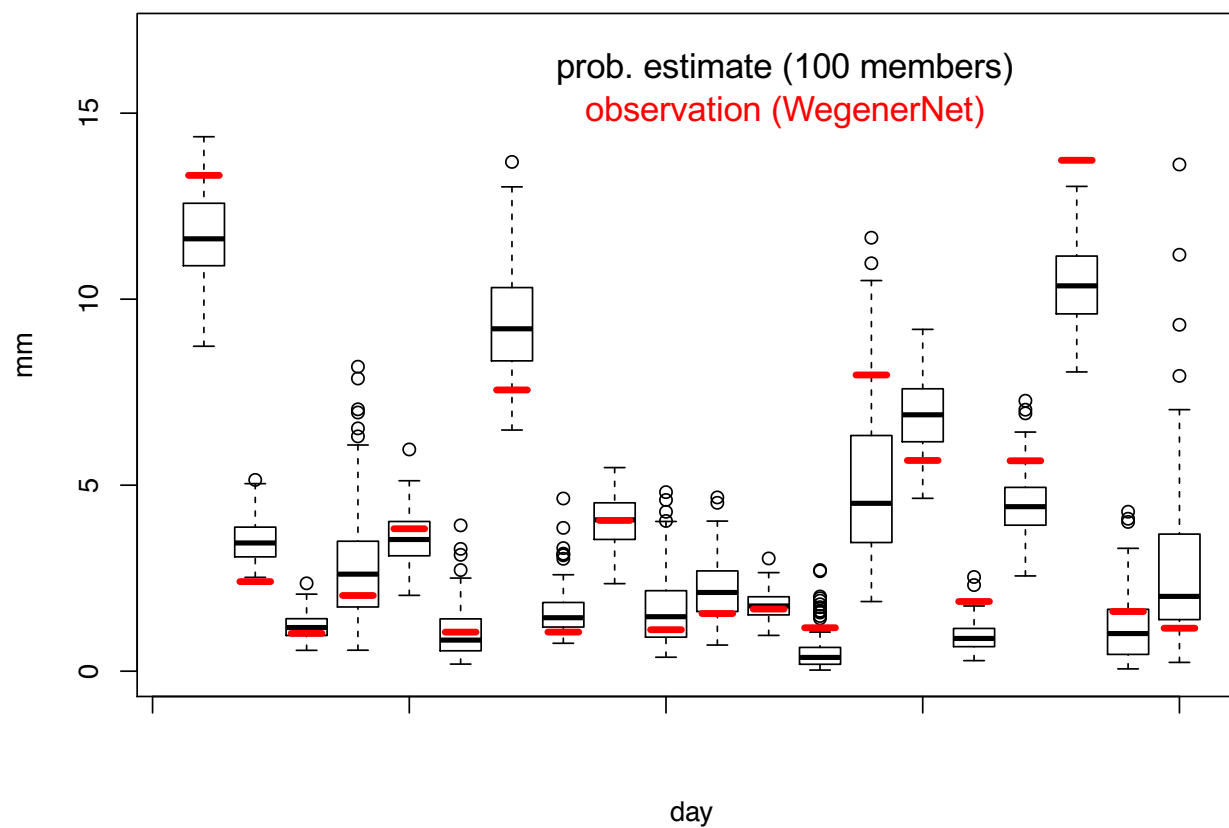
2007-2008, ≥ 1 mm, no missings, no snowfall

probab. estimate of area mean in WN-domain
from operational data in 30-km neighbhd.
(49 stations)



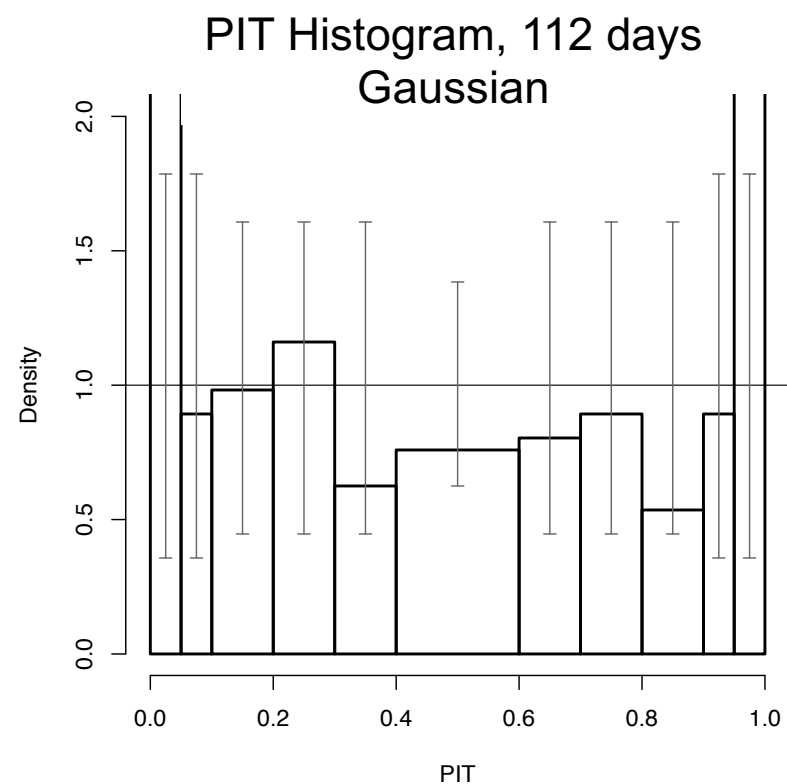
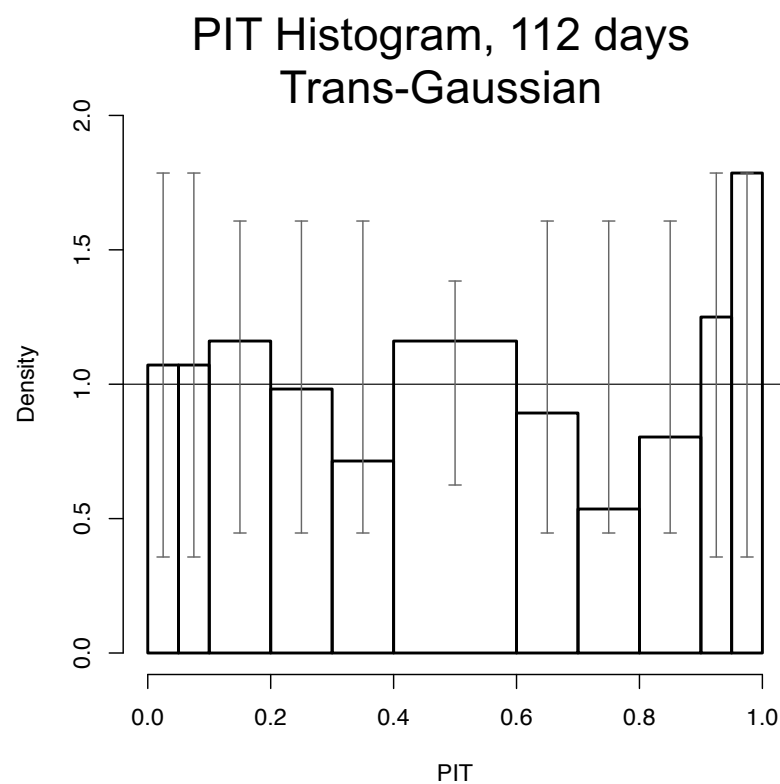
Evaluation

Example: first 20 days



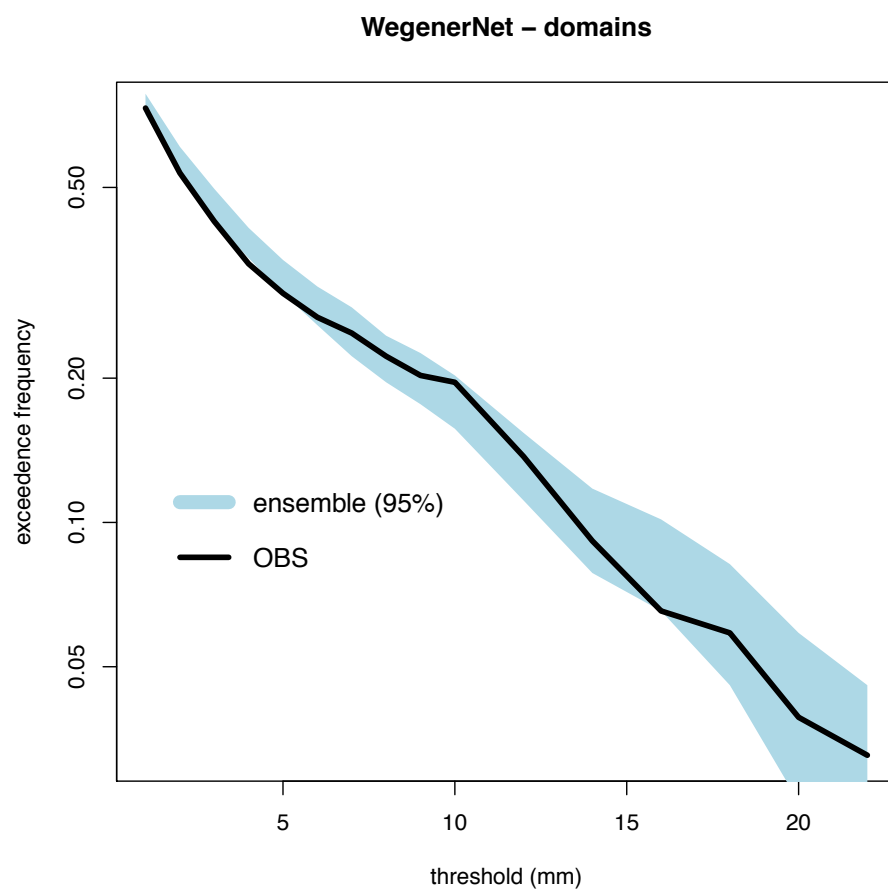


Evaluation





Evaluation



Exceedence frequency
of daily area mean over
WegenerNet domain

112 wet test days



Why not a Grid?

- Boundaries within domain are unavoidable
 - ‘global’ simulation assumes unrealistic stationarity
 - need modelling and simulation over sub-regions
 - ensembles are inconsistent across boundaries
 - hydrological units are a natural sub-division for users
- High-resolution grids are computationally demanding
 - conditional simulation always delivers point-scale precip.
 - need upscaling from much finer “primary” grid
 - for $\geq 400\text{-km}^2$ catches proper upscaling is affordable

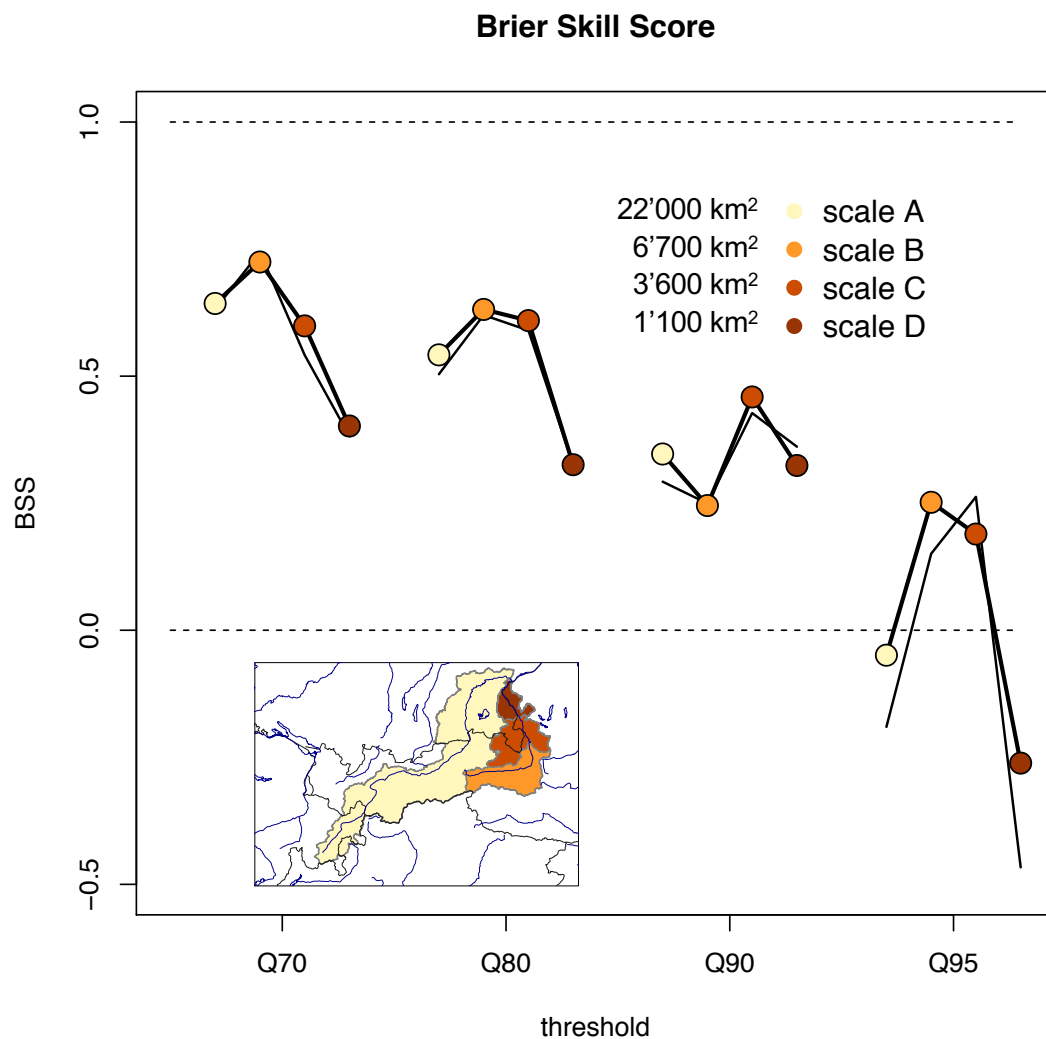


Conclusion

- A method for probabilistic precipitation analysis
- Addresses limitations of classical conditional simulation
 - Trans-Gaussian model > precipitation is not Gaussian
 - Bayesian inference > account for uncertainty of model
 - Sub-regional application > cope with non-stationarities
- Ensembles largely consistent with independent obs.
- Uncertainties ...
 - ... vary with nature of precip, station density, size of area
 - ... happen to be very large (factor of 5 for 400-km² means)
 - ... are larger than spread between deterministic analyses
- Use ensemble dataset for evaluation



Evaluation Regional Reanalyses



Brier Skill Score

Preliminary UKMO
Regional Reanalysis
20 ensemble members
~25 km resolution
Jan.-May 2008