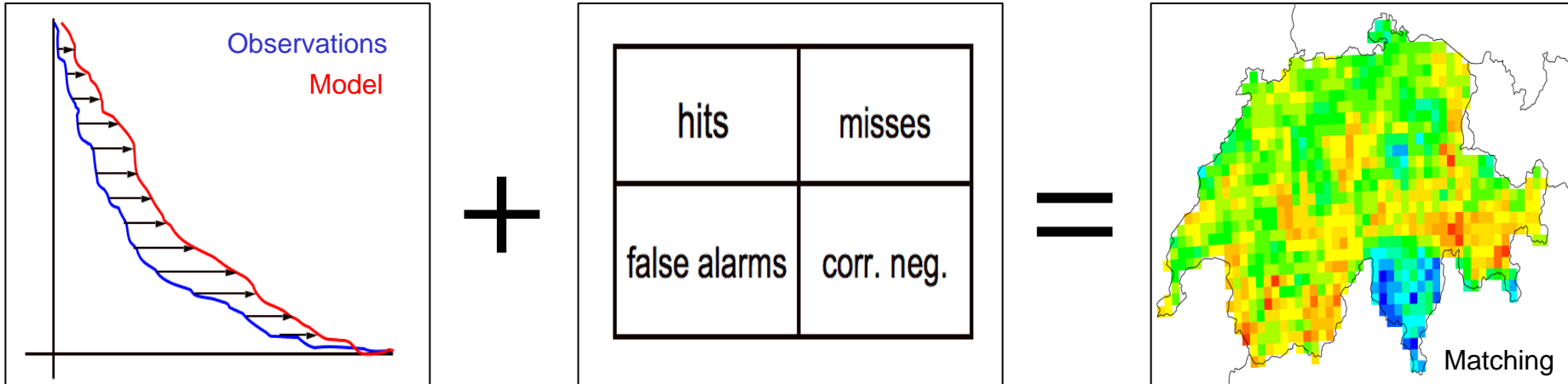


Quantile based point-to-point validations

Pros and Cons of empirically bias corrected contingency tables



Johannes Jenkner (ETH Zurich Switzerland)

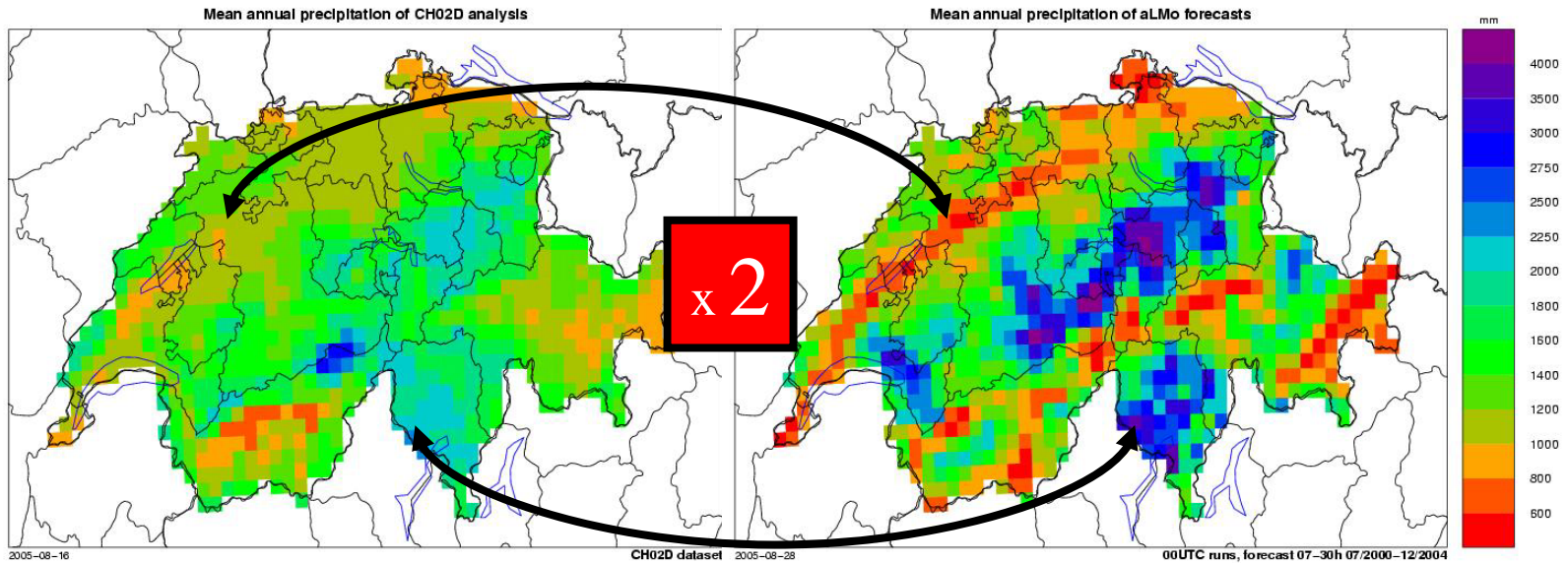
Cornelia Schwierz (University of Leeds UK)

Silke Dierer (MeteoSwiss Switzerland)

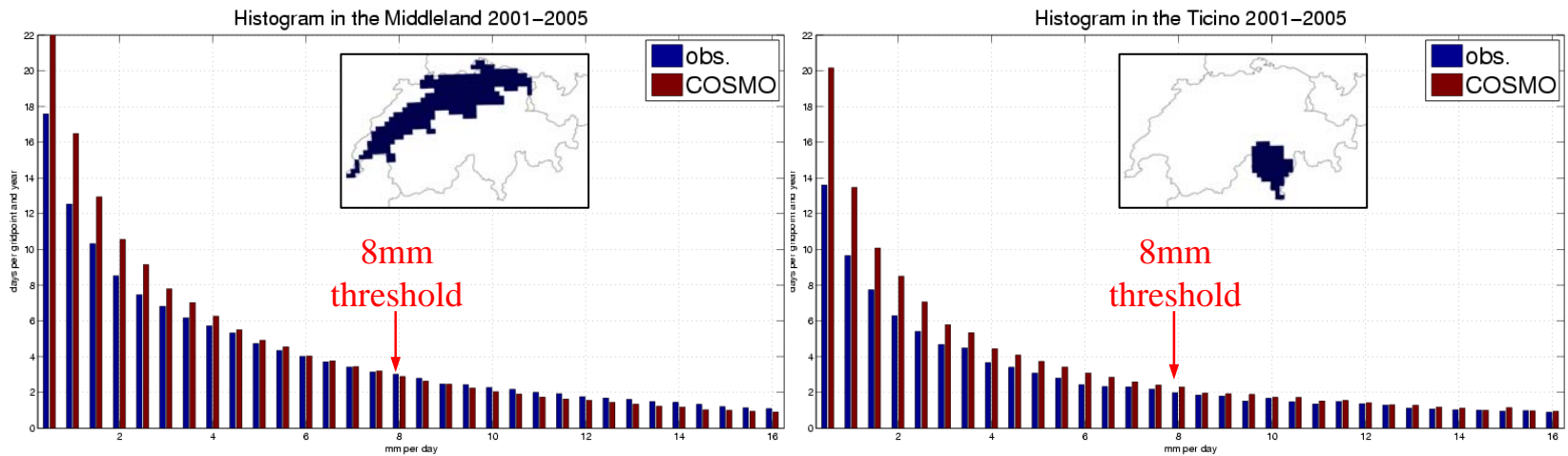
Outline

- Motivation
- Methodology and Proceeding
- Present benefits and drawbacks
- Application to QPF in Switzerland
- Summary

Motivation



Perfect matching possible at any gridpoint ?



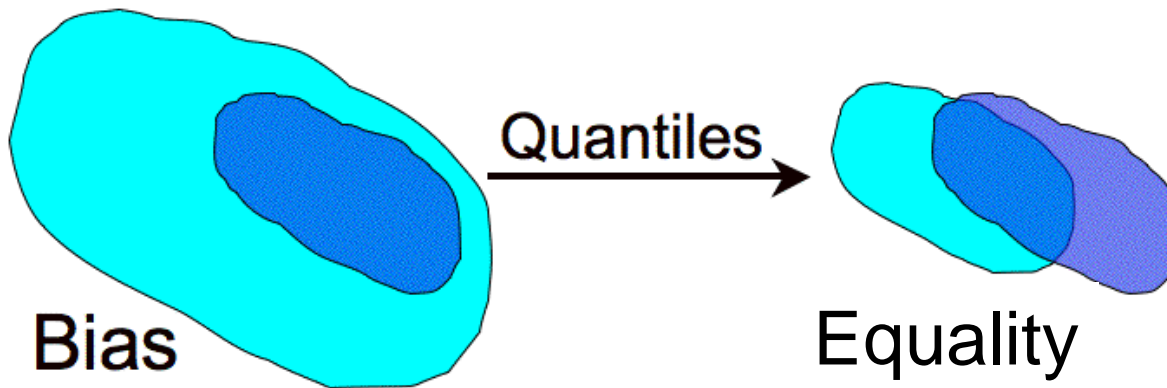
Interpretation of (equitable) categorical scores ?

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Methodology and Proceeding

- 1.) Computation of local precipitation quantiles
- 2.) Bias equal to quantile difference
- 3.) Advanced contingency table
- 4.) Determination of pattern overlap

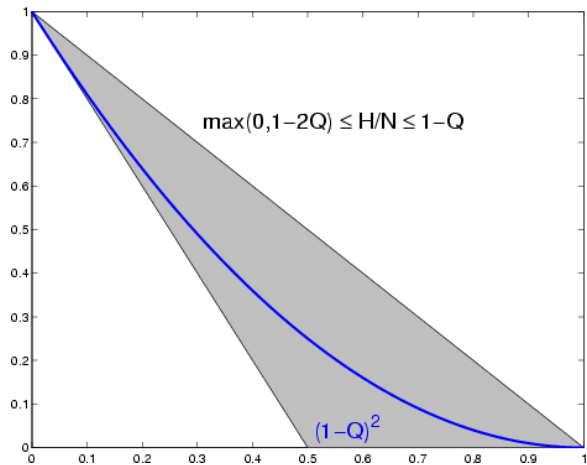


| | predicted yes | predicted no |
|--------------|---------------|--------------|
| observed yes | H | M |
| observed no | F | Z |

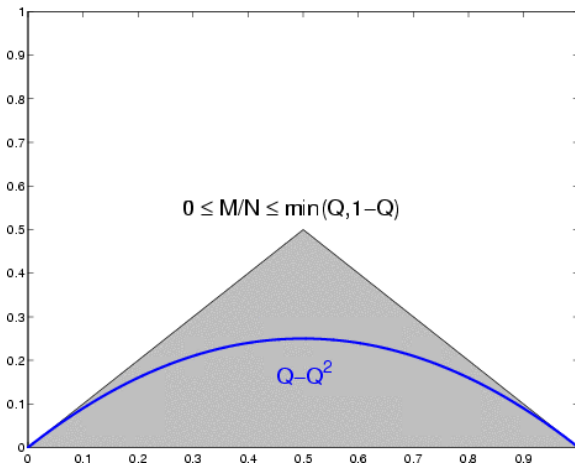
$$Q = \frac{M + Z}{H + 2M + Z} \quad M = F \quad \rightarrow \quad Z = \frac{Q}{1 - Q}H + \frac{2Q - 1}{1 - Q}M$$

Methodology and Proceeding

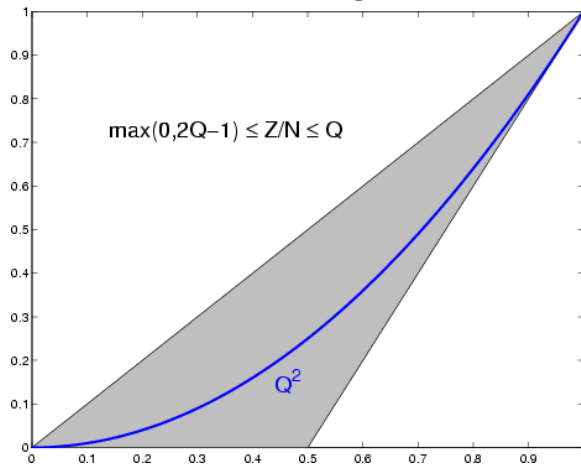
hits



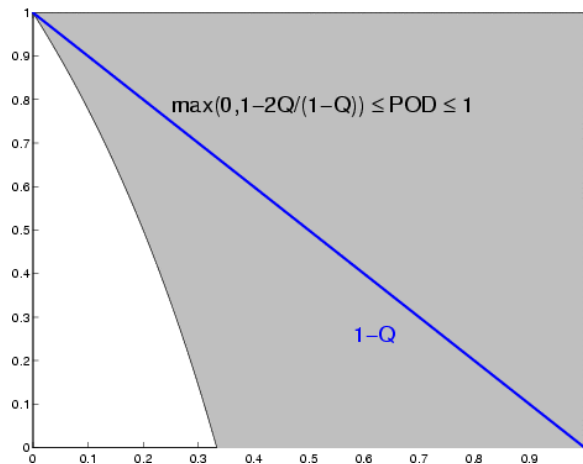
misses / false alarms



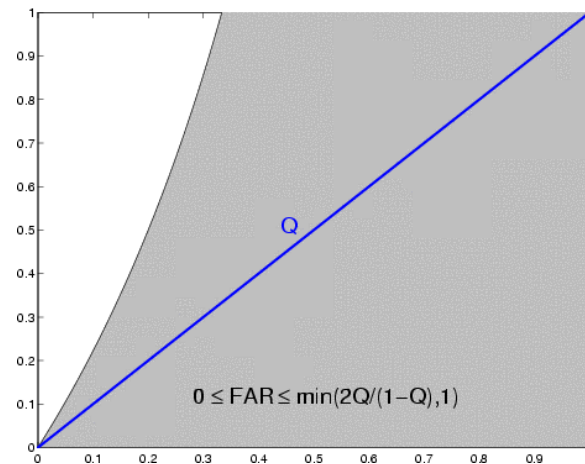
correct negatives



POD



FAR



Methodology and Proceeding

Hanssen-Kuipers discriminant equal to Heidke skill score:

$$HK = \frac{H}{H+M} + \frac{Z}{Z+M} - 1$$
$$= 1 - \frac{M}{M_{rand}}$$

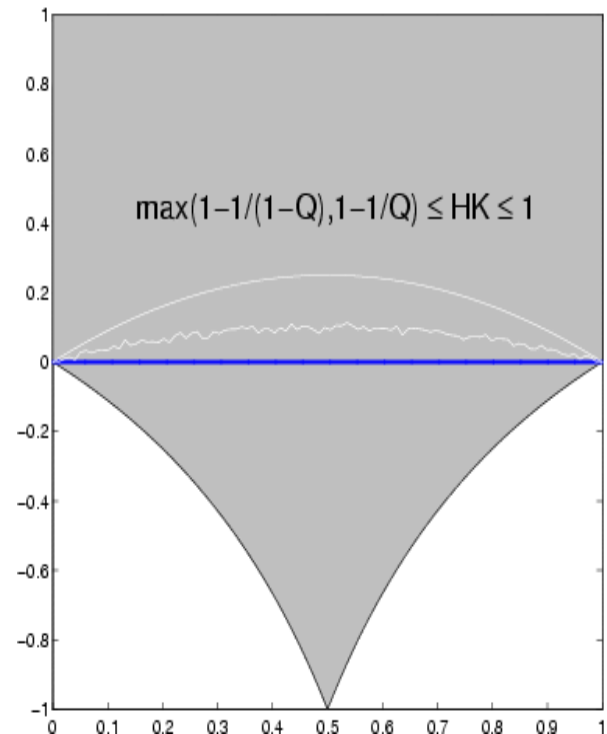
Binomial distribution of misses:

$$var(M) = \begin{cases} NQp(1-p) & Q \leq 0.5 \\ N(1-Q)p(1-p) & Q \geq 0.5 \end{cases}$$

Variance of HK proportional to 1/N:

$$var(HK) = \begin{cases} \frac{1}{N} \frac{p(1-p)}{Q(1-Q)^2} & Q \leq 0.5 \\ \frac{1}{N} \frac{p(1-p)}{Q^2(1-Q)} & Q \geq 0.5 \end{cases}$$

→ multiple uncertainty for extreme quantiles!!!



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Present benefits and drawbacks

- + fair comparisons possible
- + useful information about model behaviour
(**model developers**)
- + only misses are counted
- + accuracy limit can be implemented
(pixels with a negligible absolute difference are omitted)
- definition of quantiles:
samples with non-rain data not easy to handle /
cut-off for low intensities necessary
- quantiles less intuitive than fixed thresholds (**end-users**)
- extreme quantiles ($Q < 0.05$ and $Q > 0.95$)
suffer from small sample sizes

Outline

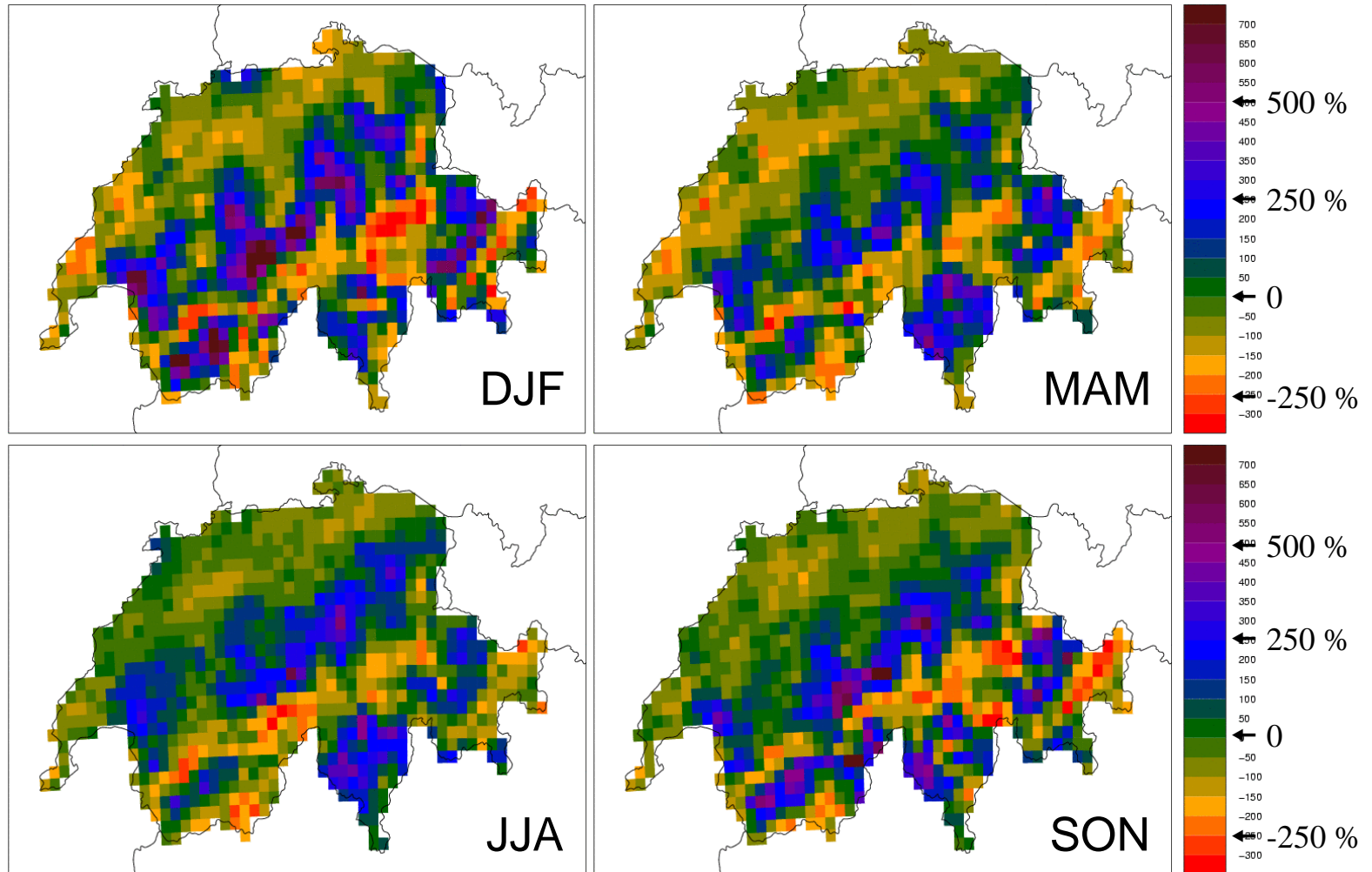
- Motivation
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Comparison of daily precipitation sums:

- 1.) Observational analysis (provided by Christoph Frei)
 - based on ~ 450 pluviometers
 - gridded in connection with a monthly climatology
- 2.) Operational forecasts of the COSMO model
(provided by MeteoSwiss)
 - nonhydrostatic and fully elastic dynamics
 - 7km grid spacing

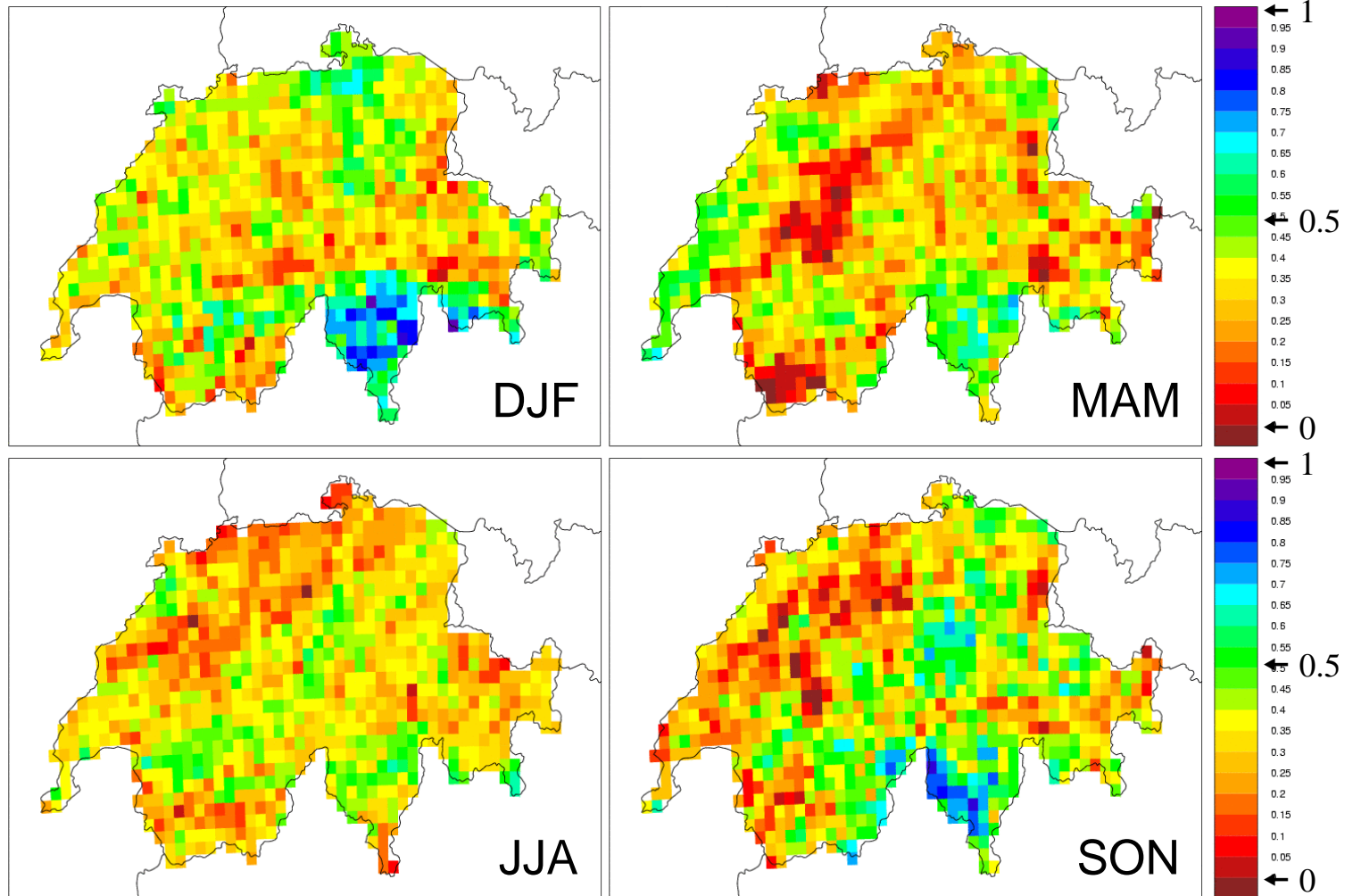
→ period of 2001 - 2005

Application to QPF in Switzerland



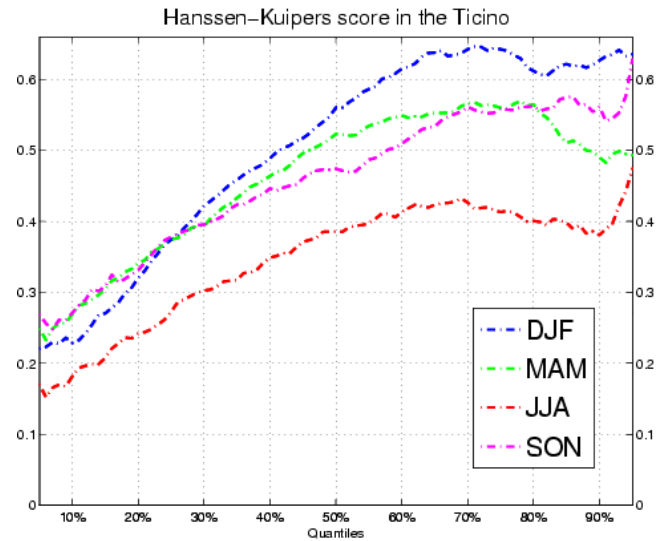
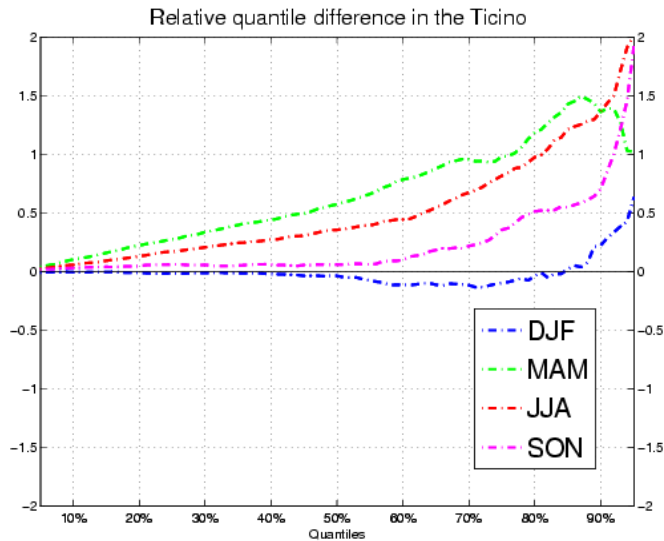
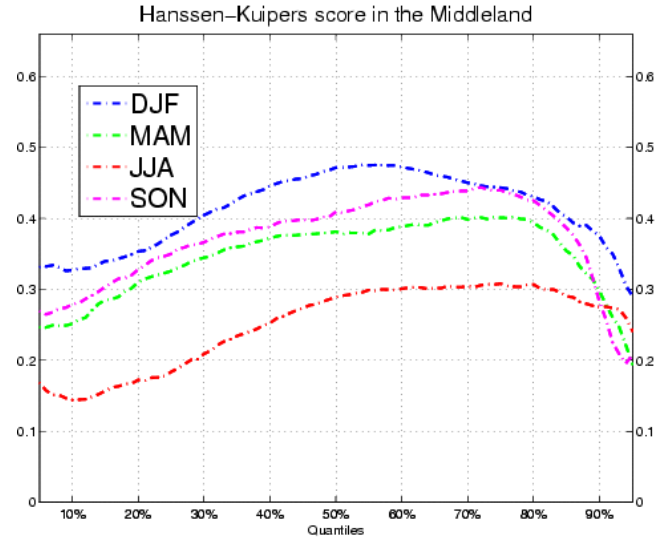
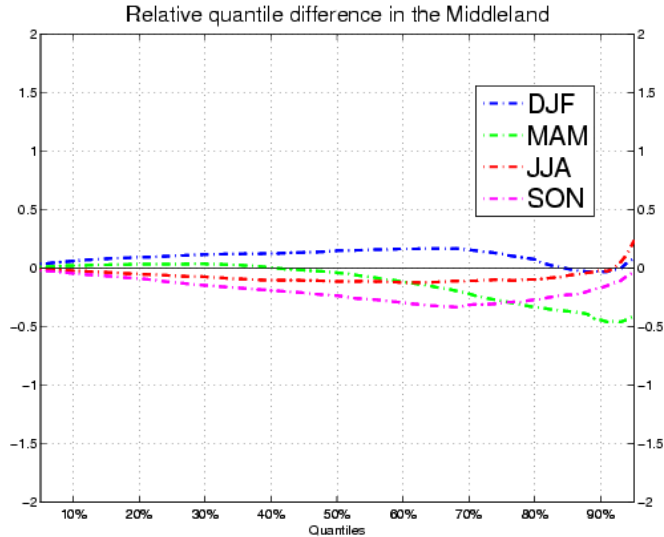
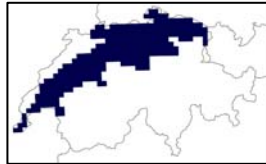
Relative quantile difference $(Q90_{\text{COSMO}} - Q90_{\text{obs.}}) / \text{Median}_{\text{obs.}}$

Application to QPF in Switzerland



Hanssen-Kuipers score for 90%.

Application to QPF in Switzerland



Spatial means of intensity spectrum

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Summary

1.)

Simplification of contingency table

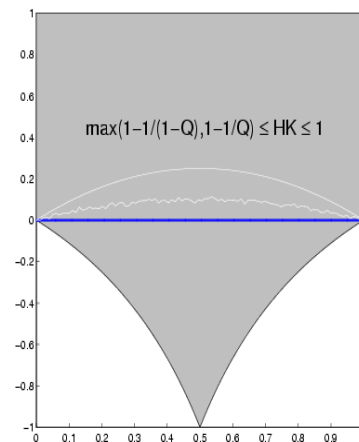
$$Q = \frac{M + Z}{H + 2M + Z} \quad M = F \quad \rightarrow \quad Z = \frac{Q}{1-Q}H + \frac{2Q-1}{1-Q}M$$

2.)

Amplitude and matching errors independent from each other

3.)

Symmetric setting of HK beneficial



$$\text{var}(HK) = \begin{cases} \frac{1}{N} \frac{p(1-p)}{Q(1-Q)^2} & Q \leq 0.5 \\ \frac{1}{N} \frac{p(1-p)}{Q^2(1-Q)} & Q \geq 0.5 \end{cases}$$

4.)

Straightforward application

