## Interpolating intraday precipitation data with radar background information

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- Precipitation data are interpolated with MISH in OMSZ. Daily, monthly, seasonal annual etc...
- hourly? ten-minute?
- Accurate rainfall estimates are needed for hazard warning, for insurance companies to settle claims and of course for the public to justify the damage.
- Plan:

MISH, Radar and MISH with radar background information to be operationally displayed hourly





## 1.Problem

It is a common phenomenon that a small but significant precipitation cell or supercell passes between two measuring stations, resulting in an inadequate record of the daily precipitation amounts of up to 100 mm falling on a small area between the measurements and observations.





Measurements



Radar 48 44 40 36 34 30 28 26 24 22 20 - 4 - 2

Measurements+Radar



Correlation= 0.559



Measurements



21.06.2023

Radar

Correlation= 0.919

Measurements+Radar



## Interpolating precipitation data



deterministic and stochastic (geostatistical) methods

?

### Different methods give different results!



### SPATIAL INTERPOLATION

According to the interpolation problem the unknown predictand  $Z(\mathbf{s}_0, t)$  is estimated by use of the known predictors  $Z(\mathbf{s}_i, t)$ , (i = 1, ..., M), where the location vectors s are the elements of the given space domain D and t is the time. The type of adequate interpolation formula depends on the probability distribution of the meteorological element.

Precipitation: Log-normal-, Gamma-, Gumbel, Weibull-,

Log-Pearson type-III distribution.....

Assuming quasi-lognormal distribution (e.g. precipitation sum) the multiplicative formula is adequate, that is, the estimate may be written as

$$\begin{split} & \stackrel{\wedge}{Z}(\mathbf{s}_{0}) = \vartheta \cdot \left( \prod_{q_{i} \cdot Z(\mathbf{s}_{i}) \geq \vartheta} \left( \frac{q_{i} \cdot Z(\mathbf{s}_{i})}{\vartheta} \right)^{\lambda_{i}} \right) \cdot \left( \sum_{q_{i} \cdot Z(\mathbf{s}_{i}) \geq \vartheta} \lambda_{i} + \sum_{q_{i} \cdot Z(\mathbf{s}_{i}) < \vartheta} \lambda_{i} \cdot \left( \frac{q_{i} \cdot Z(\mathbf{s}_{i})}{\vartheta} \right) \right) \\ & \vartheta > 0 , \ q_{i} > 0, \qquad \lambda_{i} \geq 0 \ (i = 1, \dots, M) \qquad \sum_{i=1}^{M} \lambda_{i} = 1 \end{split}$$

Interpolation parameters :

$$\vartheta = m(\mathbf{s}_0), \quad q_i = \frac{m(\mathbf{s}_0)}{m(\mathbf{s}_i)}$$
 Interpolating  
intraday  
precipitation data

Also in this case, the optimal interpolation parameters are clearly determined by certain climate statistical parameters - local statistical parameters, stochastic relationships!

# Interpolating precipitation data with MISHv.1.03

#### I. Modelling system for climate statistical parameters in space

(expected values, standard deviations, spatiotemporal correlations)

- Based on long homogenized data series and model variables.
- Modelling procedure must be executed only once before the interpolation applications.

#### **II. Spatial interpolation system**

- Additive (e.g. temperature) or multiplicative (e.g. precipitation) model and interpolation formula can be used depending on the climate elements.
- Daily, monthly, annual values and many years' means can be interpolated.
- Capability for application of background information such as satellite, radar, forecast data.
- Capability for gridding of data series.

Szentes, O., Lakatos, M., and Pongrácz, R.: *Long-term homogenized and gridded precipitation data for Hungary*, EMS Annual Meeting 2023, Bratislava, Slovakia, 4–8 Sep 2023, EMS2023-376, https://doi.org/10.5194/ems2023-376, 2023.

## Modelling of Monthly Climate Statistical Parameters for a half minutes grid (Szentimrey, 2014)

#### First step of modelling by using model variables

The monthly climate statistical parameters belonging to the stations  $\mathbf{S}_k$  (k = 1,..,K) can be used for modelling. The basic principle is as follows. Let  $P(\mathbf{s}), Q(\mathbf{s}), r(\mathbf{s}_1, \mathbf{s}_2)(\mathbf{s}, \mathbf{s}_1, \mathbf{s}_2 \in D)$  be certain model functions depending on different model variables with the following properties: (a) Modelling of correlations:  $r(\mathbf{S}_{j_1}, \mathbf{S}_{j_2}) \approx \operatorname{corr}(Z(\mathbf{S}_{j_1}, t), Z(\mathbf{S}_{j_2}, t))$  ( $j_1, j_2 = 1, .., K$ ) (b) Modelling of difference of means (E):  $P(\mathbf{S}_{j_1}) - P(\mathbf{S}_{j_2}) \approx E(\mathbf{S}_{j_1}) - E(\mathbf{S}_{j_2})$ (c) Modelling of st. deviations (D):  $\frac{Q(\mathbf{S}_{j_1})}{Q(\mathbf{S}_{j_2})} \approx \frac{D(\mathbf{S}_{j_1})}{D(\mathbf{S}_{j_2})}$ 

The model variables may be distance, height, topography.

Szentimrey, T. & Bihari, Z. (2014) Manual of interpolation software MISHv1.03. Hungarian Meteorological Service, 60.

#### Second step of modelling by interpolation

Predictand location:  $\mathbf{s}_0$ , predictor station locations:  $\mathbf{S}_{0i}$  (i = 1, ..., M).

The weighting factors can be calculated where  $\mathbf{r}$ ,  $\mathbf{R}$  contain the modelled predictandpredictors, predictors-predictors correlations.

Modelling of means, expected values (E) by additive interpolation:

$$E(\mathbf{s}_0) = \sum_{i=1}^M \lambda_i \left( P(\mathbf{s}_0) - P(\mathbf{S}_{0i}) \right) + \sum_{i=1}^M \lambda_i E(\mathbf{S}_{0i})$$

Modelling of st. deviations (*D*) by multiplicative interpolation:

$$D(\mathbf{s}_0) = \prod_{i=1}^{M} \left( \frac{Q(\mathbf{s}_0)}{Q(\mathbf{s}_{0i})} \cdot D(\mathbf{s}_{0i}) \right)^{\lambda}$$

#### **Interpolation with Background Information**

Background information can decrease the interpolation error. For example: forecast, satellite, radar data

 $Z(\mathbf{s}_{0}, t): \text{ predictand}$  $\hat{Z}(\mathbf{s}_{0}, t) = \lambda_{0} + \sum_{i=1}^{M} \lambda_{i} Z(\mathbf{s}_{i}, t): \text{ interpolation}$  $\mathbf{G} = \left\{ G(\mathbf{s}, t) \mid \mathbf{s} \in \mathbf{D} \right\}: \text{ background information on a dense grid}$ 

$$\hat{Z}_{G}(\mathbf{s}_{0},t) = \hat{Z}(\mathbf{s}_{0},t) + \mathbf{E}\left(Z(\mathbf{s}_{0},t) - \hat{Z}(\mathbf{s}_{0},t) \mid \mathbf{G}\right)$$
  
where  $\mathbf{E}\left(Z(\mathbf{s}_{0},t) - \hat{Z}(\mathbf{s}_{0},t) \mid \mathbf{G}\right)$  is the conditional

expectation of  $Z(\mathbf{s}_0, t) - \hat{Z}(\mathbf{s}_0, t)$ , given **G**.

## Correlation between measurements and radar information



2015-2022



## January average (2015-2022)









## July average (2015-2022)



## Flash flood





The spatial variability of the radar data was found to be much smaller than that of the measurements (about 480 stations are available.) This resulted in a regression coefficient of 1.134 (the coefficient of the spatial trend of the precipitation data on conditional the expected value of the radar data.)

### 05.06.2021

#### Measurements+Radar





105

r = 0.889

#### EXAMINATION OF BACKGROUND INFORMATION 15 UTC

Correlation:**0.981**Constant:0.000Coefficient:1.247



#### 

## Conclusion

- Interpolation of rainfall data based on short period measurements is also possible with MISH software thanks to the quasi-multiplicative formula.
- Radar data cannot replace measurements!
- In fact, the MISH interpolation of measurements with radar background information can be understood as the transformation of radar data into precipitation data, which is called downscaling.
- In the following years (in test mode), not only daily data will be interpolated with radar background information, but also intra-day (one hour) precipitation sums.
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## Thank you for your attention!