

INTERPOLATION OF WIND VARIABLES USING THE MISH SYSTEM: A DATA-DRIVEN CLIMATOLOGICAL APPROACH

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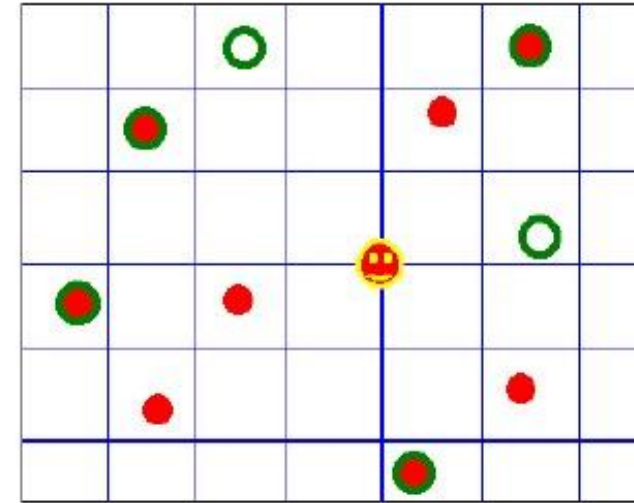
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Introduction - why meteorological interpolation is needed

- Meteorological stations are sparse → no continuous coverage
- Gridded data must be derived from point observations
- Climate variables depend on terrain, elevation, and distance from sea
- Long-term data enable space–time statistical modelling
- Goal: spatially consistent and physically meaningful fields

Szentimrey & Bihari, 2014



- : Closed old manual station with long data series (Sample in space and in time!)
- : New automatic station with short data series (predictor)
- : Closed old manual station and a new automatic station (predictor) (Sample in space and in time!)
- ⊗ : Optional location without data (predictand)
- ⊕ : Grid points with background information, e.g. forecast, satellite, radar data

Limitations of traditional GIS-based interpolation methods

- Use only one moment in time → ignore temporal variability
- Climate information (mean, variance,...) is lost
- No distinction between additive (temperature) and multiplicative (precipitation) variables
- Cannot easily integrate background fields (e.g. models, radar, satellite)
- Often less accurate for meteorological applications

MISH (Meteorological Interpolation based on Surface Homogenized Data Basis, *Szentimrey & Bihari, 2007, 2014*)

Modeling subsystem:

- Based on long, homogenized time series and deterministic variables (e.g. topography, elevation)
- Additive or multiplicative model, depending on the variable type (e.g. temperature vs. precipitation)
- High-resolution grid (e.g. $0.5' \times 0.5'$)

Interpolation subsystem:

- Uses the modeled parameters to estimate values at any location
- Allows integration of background information (e.g. radar, satellite, forecast data)
- Missing data completion (daily, monthly)

Why wind requires special attention

- The wind field has specific statistical and physical properties.
- Special modeling, interpolation, completion and gridding procedures were developed in MISH for wind speed and direction together.
- u and v components are interpolated separately, then recombined.
- Strongly influenced by surface roughness and orography.

Why wind requires special attention

Wind vector series: $\mathbf{v}(t) = [u(t), v(t)]^T$ ($t = 1, \dots, n$)

$u(t)$: components towards east,

$v(t)$: components towards north.

The components can be expressed by the wind speed $z(t) = \|\mathbf{v}(t)\|$ (m/s) and meteorological wind direction $\phi_{MET}(t)$ in degree, according to the following formulas:

$$u(t) = -z(t) \cdot \sin\left[\frac{\pi}{180} \cdot \phi_{MET}(t)\right], \quad v(t) = -z(t) \cdot \cos\left[\frac{\pi}{180} \cdot \phi_{MET}(t)\right]$$

$$z(t) = \sqrt{u^2(t) + v^2(t)}, \quad \phi_{MET}(t) = \frac{180}{\pi} \cdot \text{atan2}[-u(t), -v(t)] \quad (\text{if } \mathbf{v}(t) \neq \mathbf{0})$$

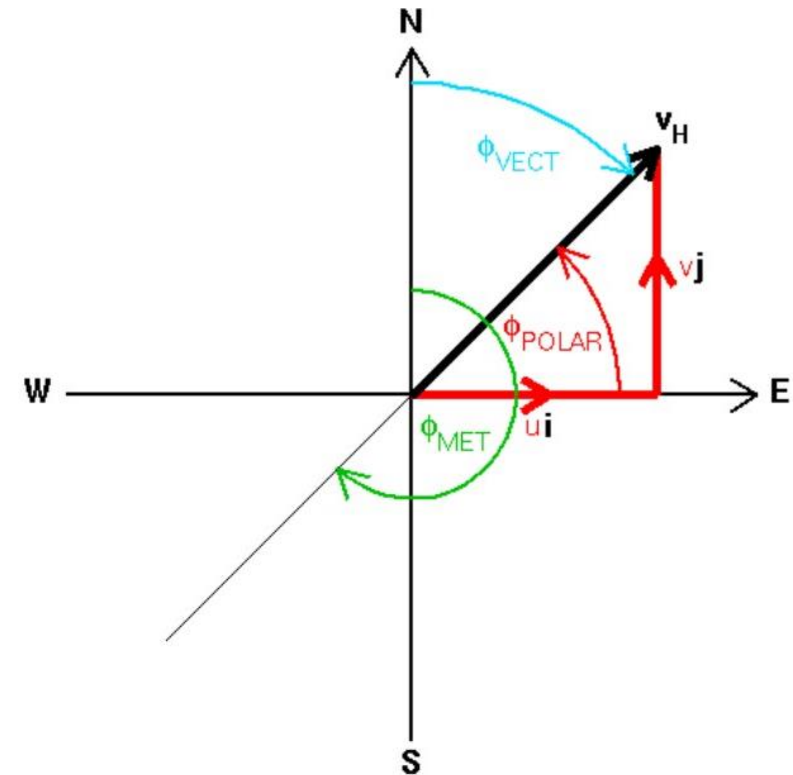
Then, the mean wind speed: $\bar{z} = \frac{1}{n} \sum_{t=1}^n z(t)$

The mean wind vector: $\bar{\mathbf{v}} = \frac{1}{n} \sum_{t=1}^n \mathbf{v}(t) = [\bar{u}, \bar{v}]^T$,

where $\bar{u} = \frac{1}{n} \sum_{t=1}^n u(t)$, $\bar{v} = \frac{1}{n} \sum_{t=1}^n v(t)$.

The “vectorial mean” meteorological wind direction in degree:

$$\bar{\phi}_{MET} = \frac{180}{\pi} \cdot \text{atan2}[-\bar{u}, -\bar{v}] \quad (\text{if } \bar{\mathbf{v}} \neq \mathbf{0})$$



Steps of modeling in case of wind speed (z) and direction (Φ)

- Calculation of wind components u , v data from z and Φ data.
- Wind speed (z) is modeled in MISH using a **multiplicative** model.
The modeling includes special predictors related to observation height (wh) and surface roughness (r).
The procedure outputs monthly wind-profile parameters (α and γ).
- The u and v components are converted to standard reference conditions ($wh = 10$ m, $r = 0.1$) by applying the wind-profile parameters $\rightarrow U_0, V_0$.
- The transformed components u_0 and v_0 are then modeled in MISH using an **additive** model.

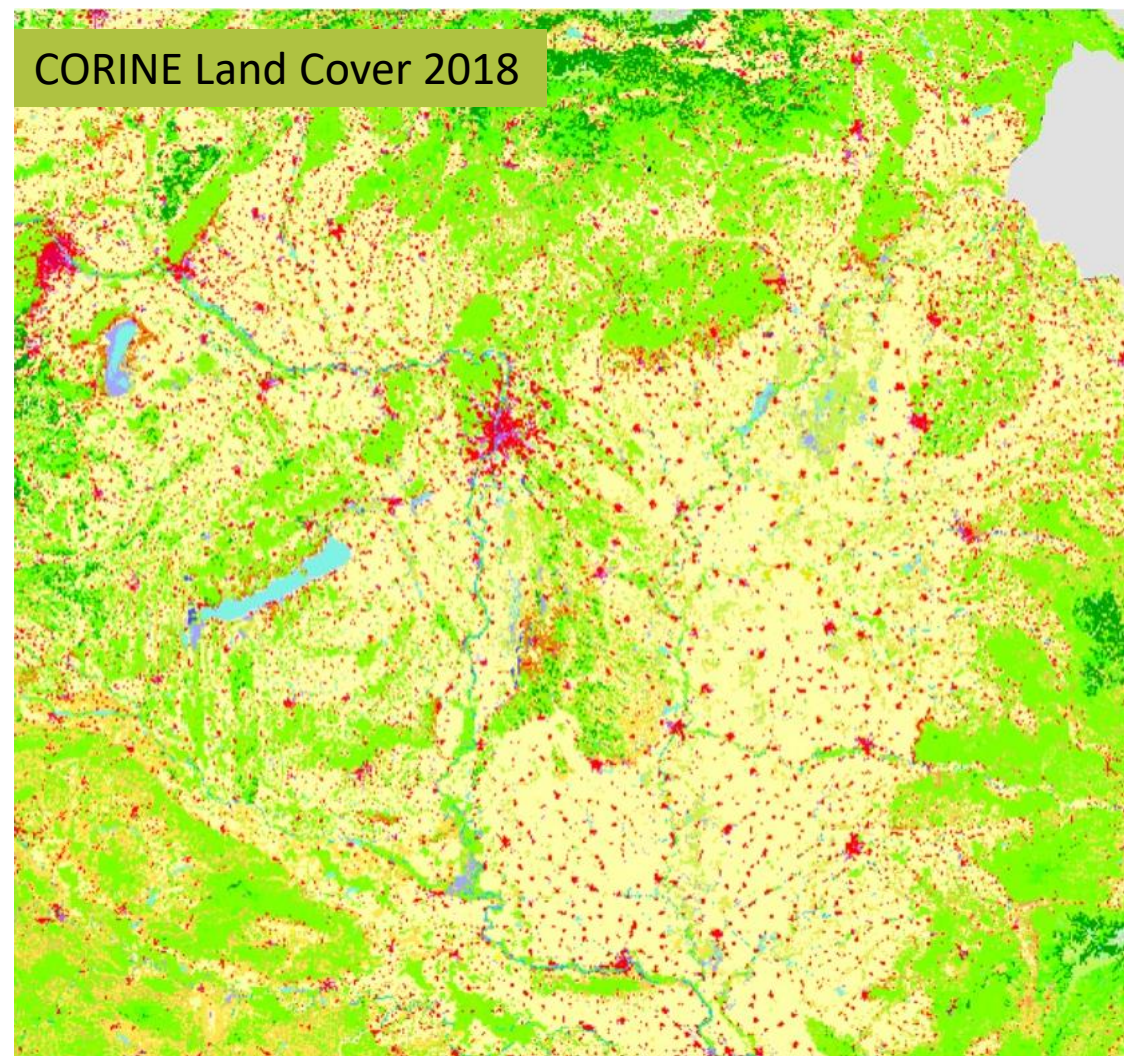
Steps of interpolation in case of wind speed (z) and direction (Φ)

- Calculation of wind components u , v data from z and Φ data.
- Interpolation applications for wind speed z by MISH.

Special input is the modeled wind profile parameters α and γ by months.

- Transformation of the components u , v data for $wh=10m$, $r=0.1$ by using the wind profile parameters α and γ : u_0 , v_0 .
- Interpolation applications for components u_0 and v_0 by MISH.
- Calculation of interpolated wind direction Φ data from the interpolated components u_0 and v_0 data.

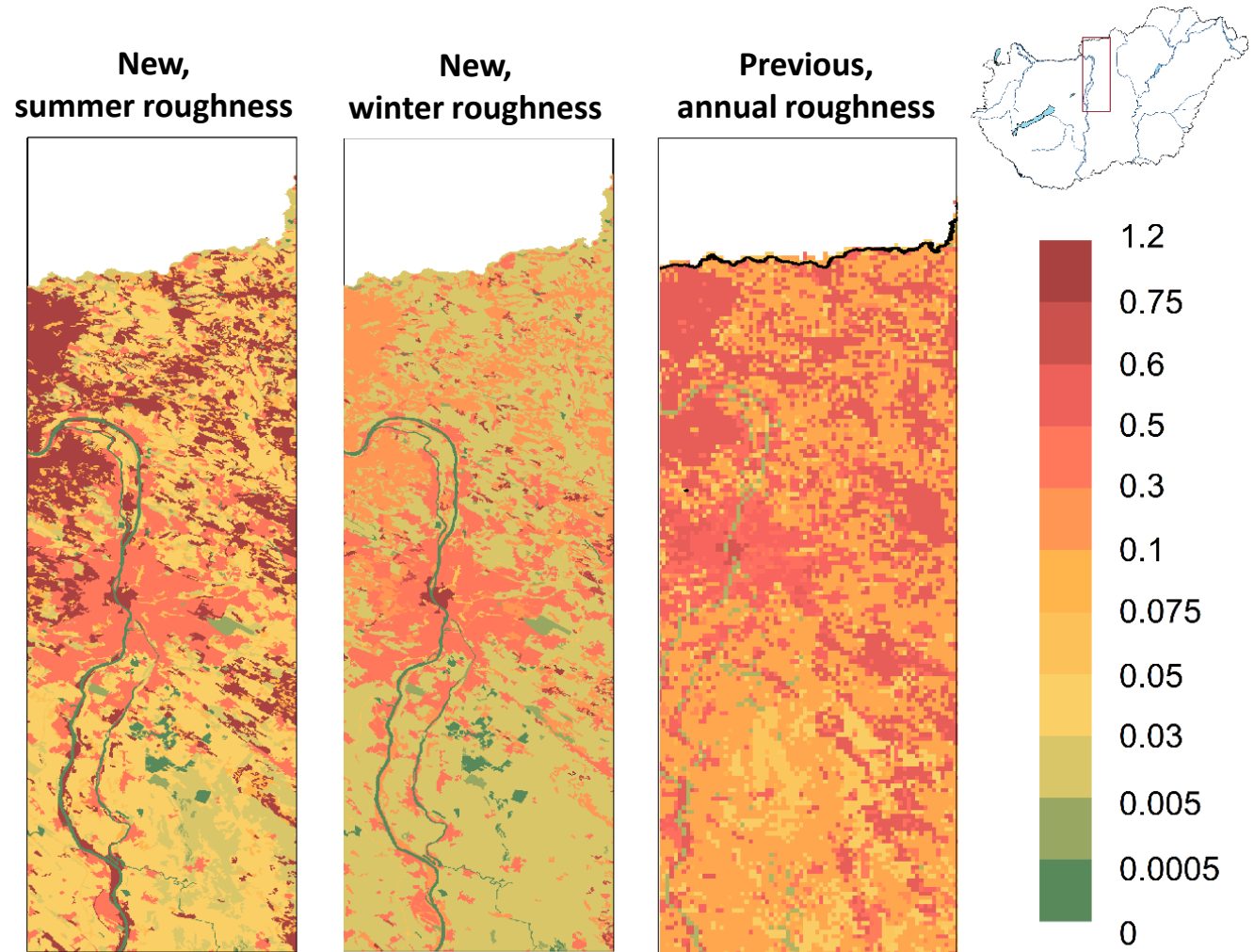
Renewal of the surface roughness database



Development of wind modeling

Innovations:

- Higher spatial resolution
- **Seasonal** roughness maps
- **Summer** (May–October) and **Winter** (November–April) roughness
- Considering vegetation conditions following *MacArthur & Haines* (1982)
- Expected outcome: **more accurate wind estimation**



Wind profile model in MISH

- derived from the power law and logarithmic profile models.

$$z_{2j} = \left(\frac{\ln\left(\frac{wh_2}{r}\right) * \ln\left(\frac{wh_1}{0.1}\right)}{\ln\left(\frac{wh_2}{0.1}\right) * \ln\left(\frac{wh_1}{r}\right)} \right)^\gamma * \left(\frac{wh_2}{wh_1} \right)^\alpha * z_{1j}$$

wh_1 : original height of the measurement

wh_2 : target height for conversion, e.g. 10 m, 30 m, etc.

r : surface roughness at station z_j

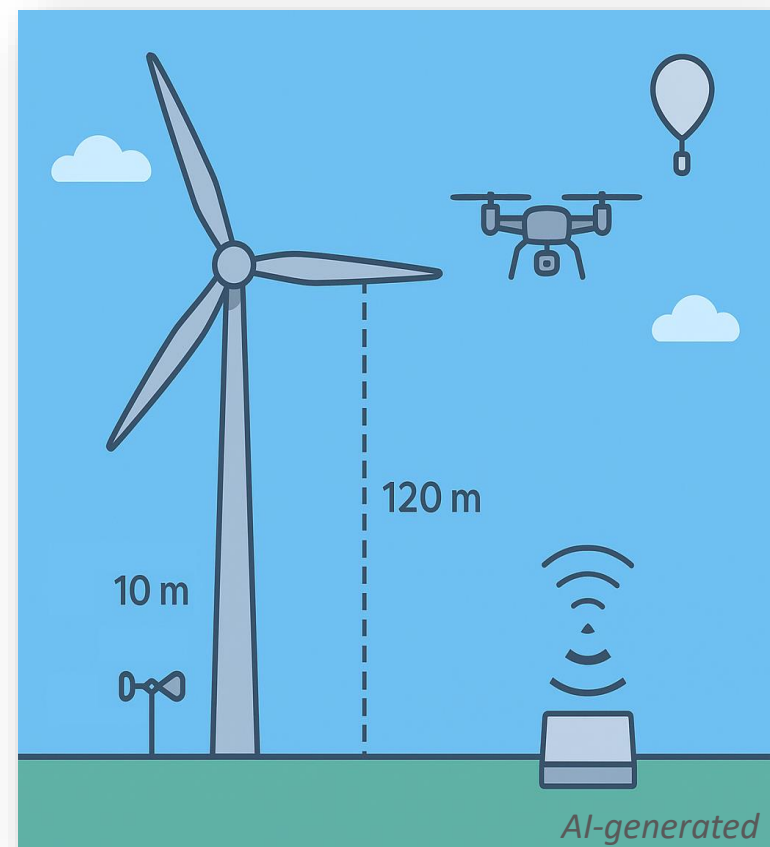
z_{1j} : wind speed measured at height wh_1

z_{2j} : wind speed converted to height wh_2

α, γ : modeled parameters, depending on month and variable — i.e., modeled separately for wind maxima, wind averages or wind direction

Wind profile model in MISH

- MISH allows wind speed scaling to arbitrary heights
 - energy, infrastructure!
- However: above ~100–120 m
 - Dynamic processes dominate (not purely statistical behaviour) (*Dobi, 2006*)
 - Requires dedicated validation (LIDAR, SODAR, drone, weather balloon measurements)
 - Ongoing research direction



Integration of background information

Predictand: $Z(\mathbf{s}_0, t)$, Predictors: $Z(\mathbf{s}_i, t), (i = 1, \dots, M)$

Interpolation without background information (multiplicative interpolation formula):

$$\hat{Z}(\mathbf{s}_0, t) = \vartheta \left(\prod_{q_i Z(\mathbf{s}_i, t) \geq \vartheta} \frac{q_i Z(\mathbf{s}_i, t)^{\lambda_i}}{\vartheta} \right) \left(\sum_{q_i Z(\mathbf{s}_i, t) \geq \vartheta} \lambda_i + \sum_{q_i Z(\mathbf{s}_i, t) \geq \vartheta} \lambda_i \frac{q_i Z(\mathbf{s}_i, t)}{\vartheta} \right)$$

where $\vartheta > 0, q_i > 0, \lambda_i \geq 0$ and $\sum_{i=1}^M \lambda_i = 1$ (interpolation parameters)

Background information (e.g. radar, forecast, satellite data etc.) on a dense grid:

$$\mathbf{B} = \{B(\mathbf{s}, t) \mid \mathbf{s} \in \mathbf{D}\} \quad (\mathbf{D}: \text{space domain})$$

Interpolation with background information:

$$\hat{Z}_B(\mathbf{s}_0, t) = \hat{Z}(\mathbf{s}_0, t) + \beta_1(t) \cdot \left(B(\mathbf{s}_0, t) - \hat{B}(\mathbf{s}_0, t) \right), \text{ where}$$

$\beta_1(t)$ is the estimated regression coefficient,

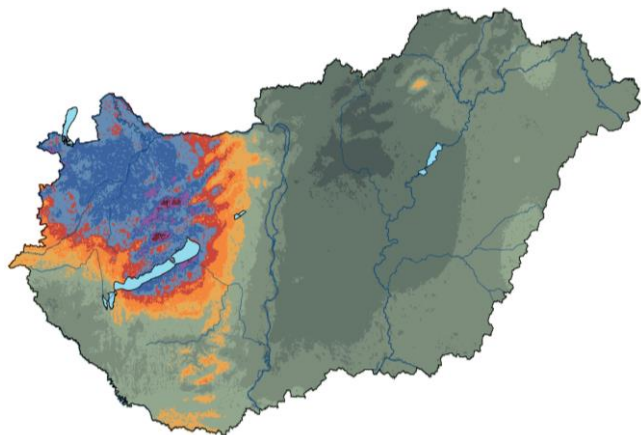
$\hat{B}(\mathbf{s}_0, t) = F_M(B(\mathbf{s}_1, t), \dots, B(\mathbf{s}_M, t); q_1, \dots, q_M, \lambda_1, \dots, \lambda_M)$ is the interpolation formula for the background information

Integration of background information

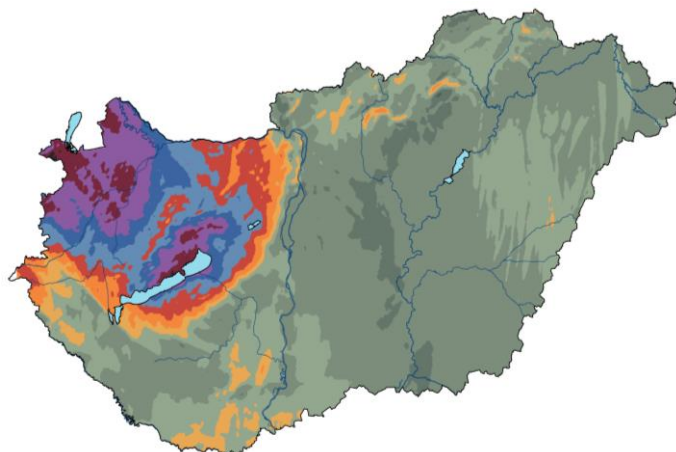
Case study: September 14, 2024, Storm *Boris* with heavy rains, stormy wind gusts

- Data: surface measurements of **wind gust** and AROME-RUC high-resolution numerical weather prediction model wind gust data
- Method: MISH interpolation software

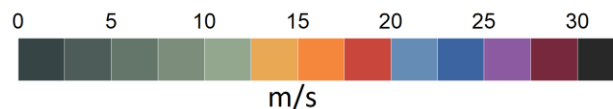
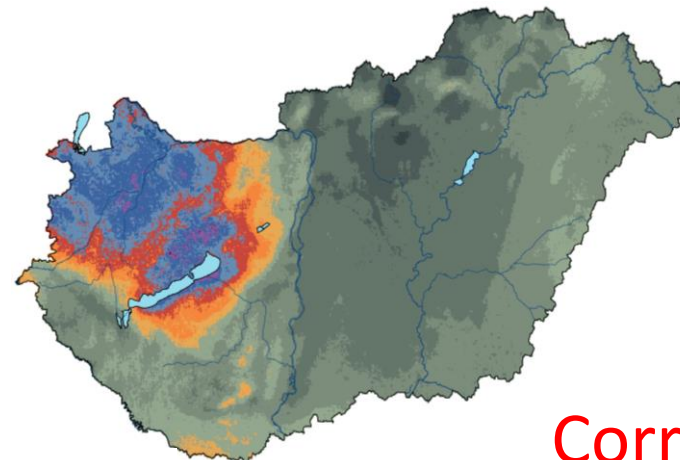
Interpolation of surface measurements (MISH)



AROME-RUC



MISH + AROME-RUC



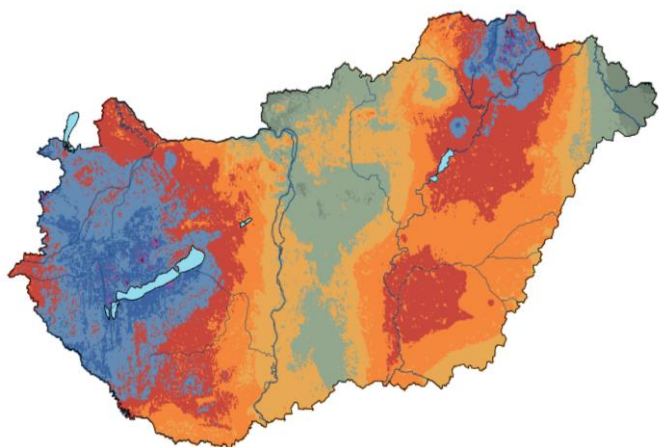
**Correlation:
0.89**

Bokros, Izsák, Sz. Gáspár, Lancz, Lakatos, Pongrácz (2025): Improving wind hazard assessment using high-resolution numerical weather prediction models and interpolation techniques in Hungary, 2nd Natural Hazard and Climate Change Conference (21-23, May, 2025, Szeged)

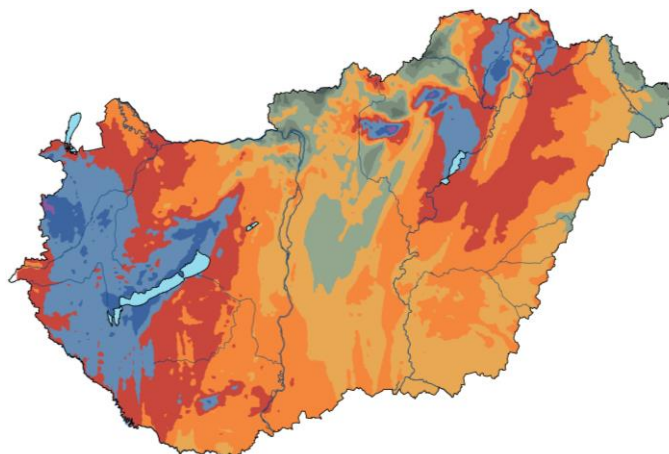
Integration of background information

Case study: December 24, 2024, Mediterranean Cyclone

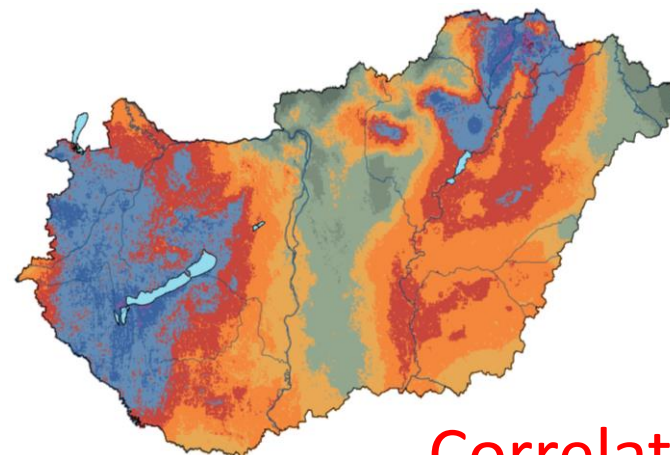
Interpolation of surface measurements (MISH)



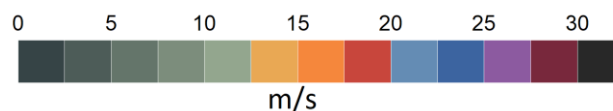
AROME-RUC



MISH + AROME-RUC



Correlation:
0.793



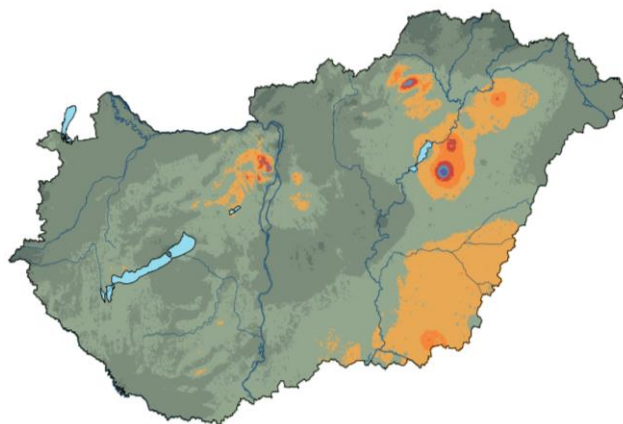
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Integration of background information

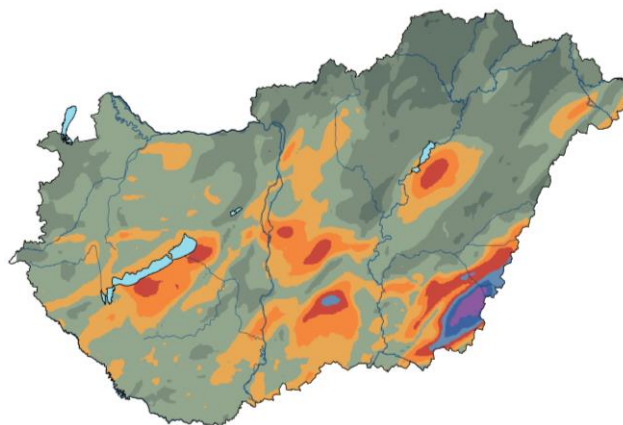
Case study: June 10, 2024 (Only AROME):

- Summer storms cause local strong gusts —background is crucial for interpolation.
- Low correlation between observations and model fields in case of convective events.

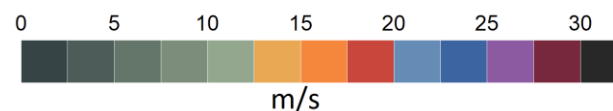
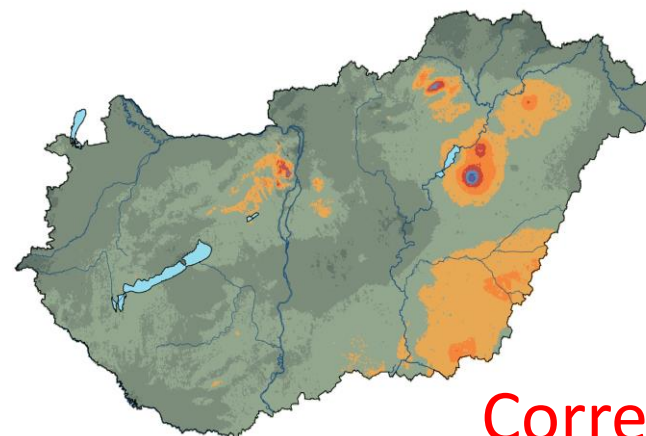
Interpolation of surface measurements (MISH)



AROME



MISH + AROME



Correlation:
0.1

Bokros, Izsák, Sz. Gáspár, Lancz, Lakatos, Pongrácz (2025): Improving wind hazard assessment using high-resolution numerical weather prediction models and interpolation techniques in Hungary, 2nd Natural Hazard and Climate Change Conference (21-23, May, 2025, Szeged)

Summary

- MISH provides a climatologically based, data-driven interpolation framework.
- Wind is treated with dedicated procedures due to its vector nature.
- We renewed the surface roughness database, including seasonal values.
- The wind-profile model enables height scaling using modeled α and γ . Upcoming sodar and drone data will support real-height validation and future development.
- Background fields further enhance interpolation performance.

Thank you for your kind attention!

References

- Bokros, Izsák, Sz. Gáspár, Lancz, Lakatos, Pongrácz* (2025): Improving wind hazard assessment using high-resolution numerical weather prediction models and interpolation techniques in Hungary, 2nd Natural Hazard and Climate Change Conference (21-23, May, 2025, Szeged)
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