

Koninklijk Meteorologisch Instituut

Institut Royal Météorologique

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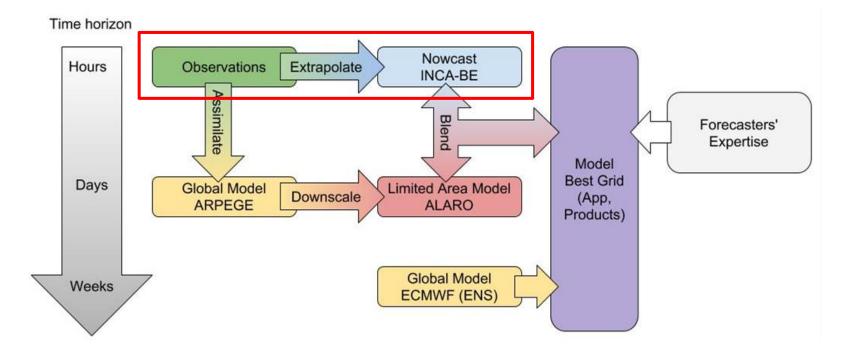
### The Seamless Prediction Programme at the Royal Meteorological Institute of Belgium

Maarten Reyniers, Michiel Van Ginderachter, Lesley De Cruz, and many others

EUMETSAT Integrated nowcasting workshop, 23-25 Jan 2024, Darmstadt, DE

### Existing RMI forecasting systems

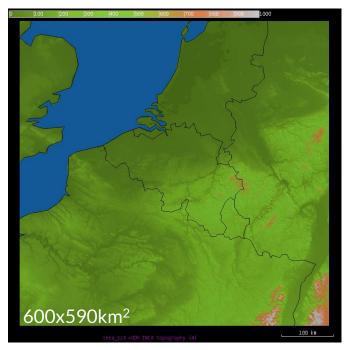
### **Existing system:** a treasure of forecast products



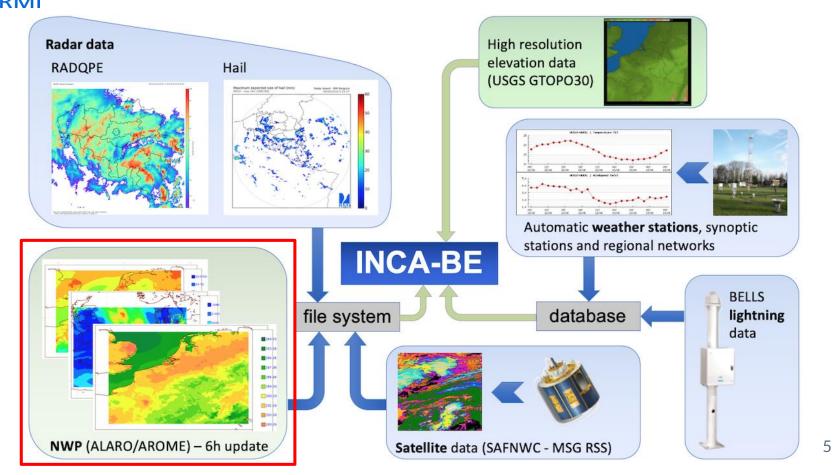
### **Operational nowcasting: INCA-BE**

- INCA = <u>Integrated Nowcasting through Comprehensive Analysis</u>
- **Nowcasting system** of several meteorological fields: temperature, humidity, wind, cloudiness, precipitation, precipitation type
- High resolution: 1 km
- Base code from GeoSphere, Austria
- Implemented at RMI since 2012 and heavily **adapted/improved** since then
- Literature:

"The INCA-BE system: ten years of operational nowcasting and its applications at the national meteorological service of Belgium" <u>https://zenodo.org/record/5798952</u>



### **Operational nowcasting: INCA-BE input**



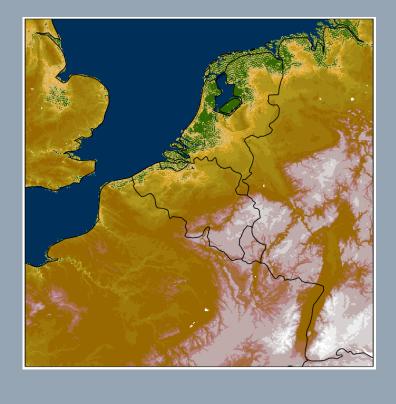
### NWP models at RMI

### Two parallel INCA-BE versions

- INCA-ALARO: background model ALARO
- INCA-AROME: background model **AROME**

### 2 ACCORD MODELS (ALARO | AROME)

- coupled to ARPEGE (MF) | IFS (ECMWF)
- *in-house* models (= full control)
- 4 runs/day (0, 6, 12, 18 UTC)
- 3h data assimilation cycle (AROME)
- +48h forecast range
- 45s timestep
- 1.3 km horizontal resolution



### **NWP models at RMI: ALARO and AROME**

- Dynamics part of the models are identical ( = equations of the atmosphere and how to solve them)
- Difference in the physics parameterisations
   (= source and sinks in the equations of the atmosphere)

	ALARO	AROME
deep convection	scale-aware parameterisation	no parameterisation
microphysics	less sophisticated (water, ice, snow, rain)	sophisticated (+ graupel and hail)
surface interactions	less sophisticated	sophisticated
developments	contributions by RMI	no contributions by RMI
applicability	~ between 15 - 1 km	~ < 3 km

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FORTHCOMING DEVS	ALARO	AROME
deep convection	scale-aware parameterisation	no parameterisation
microphysics	<b>more</b> sophisticated (water, ice, snow, rain, <b>graupel</b> )	sophisticated (+ graupel and hail)
surface interactions	sophisticated	sophisticated
developments	contributions by RMI	no contributions by RMI
applicability	~ between 15 - 1 km	~ < 3 km

# Operational nowcasting: INCA-BE output

### Basic fields forecast up to +12h, 1h step

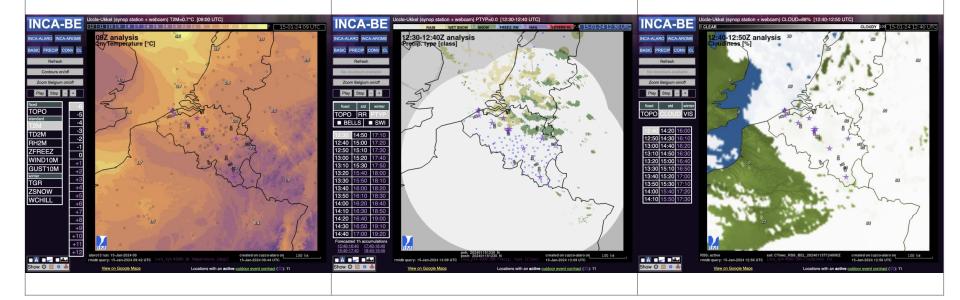
Temperature, dewpoint, relative humidity, snowfall level, freezing level, wind, wind gust speed, wind chill, ground temperature

### Precipitation fields forecast up to +4h, 10' step

Precipitation, precipitation type, lightning, severe weather index

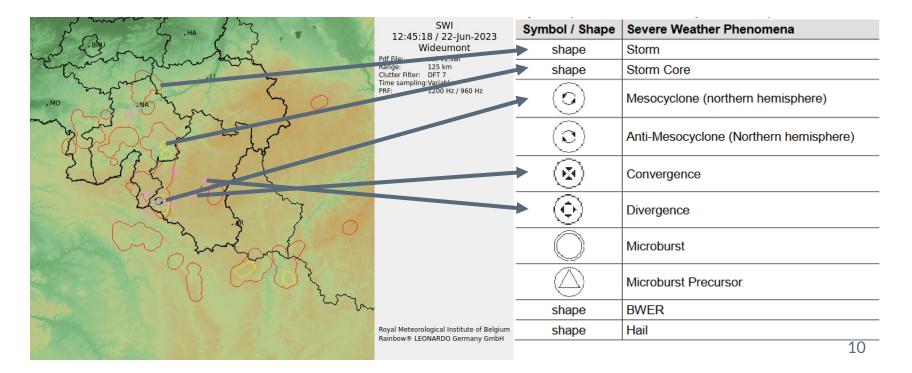
### Cloudiness fields forecast up to +2h, 10' step

Cloudiness, visibility



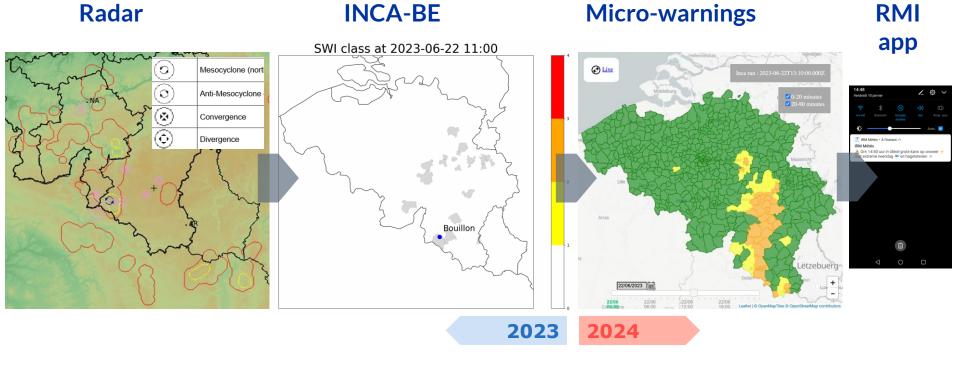
# Operational nowcasting: INCA-BE devs 2023

- New INCA-field "Severe Weather Index" (SWI)
- Starting point: radar-based severe weather contours and rotation detection (MC)

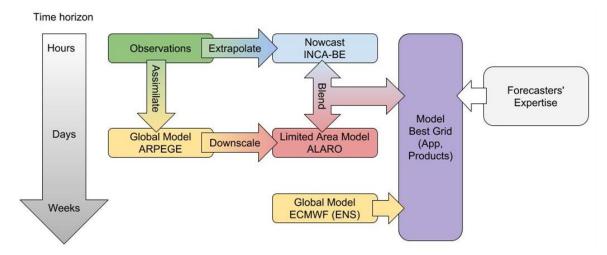


# Operational nowcasting: INCA-BE devs 2024

### SWI = input for Wind Gust notifications in RMI smartphone app (2024)



### **Existing systems:** a treasure of forecast products



### Works well, but :

- heterogeneous in terms of time horizon, resolution, update frequency, ...
  - $\rightarrow$  unclear which product is best for which lead time/application
- mostly **deterministic** 
  - $\rightarrow$  lacks reliable uncertainty information

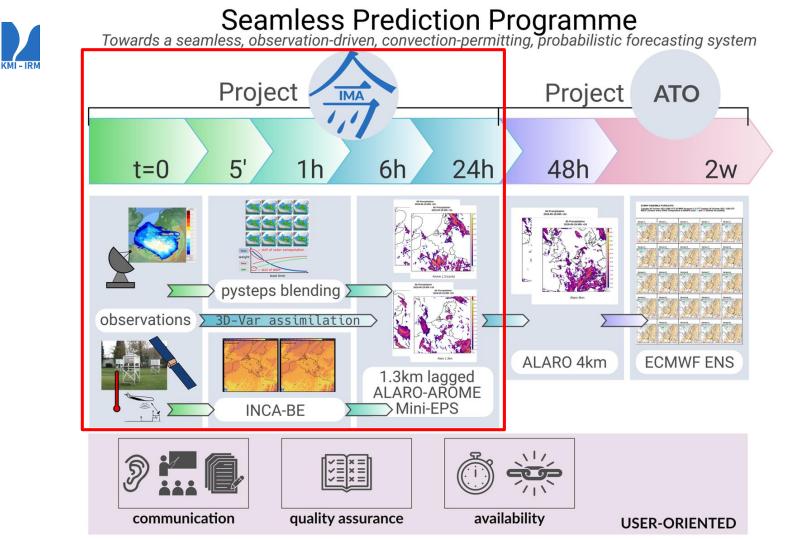
### The next-generation forecasting system

**Requirements** of a next-generation forecasting system:

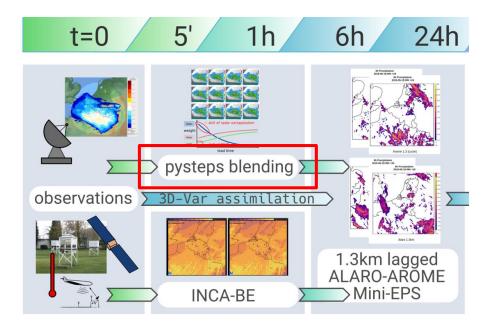
- **Ensemble** prediction system → uncertainty propagation
- High spatial and temporal resolution
- Observation-driven and rapidly updating (nowcasting)
- Based on state-of-the-art numerical weather prediction models
- Seamlessly combining available forecasts to maximize skill at all times
- Robust and user-oriented

 $\rightarrow$  Creation of the RMI's transversal Seamless Prediction Programme





Short-term Seamless Probabilistic Prediction: Project IMA

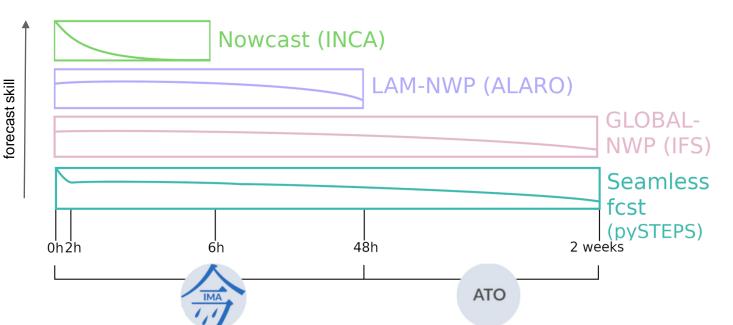




### Seamless forecast:

# forecast without discontinuities, using an **optimal combination** of all information sources

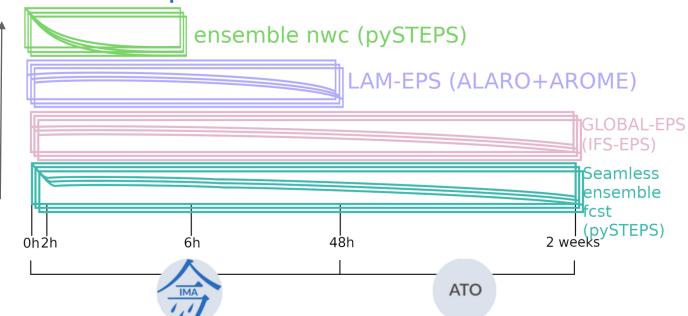
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# Project IMA: Seamless from 5' to 24h

### Seamless forecast:

# **probabilistic** forecast without discontinuities, using an **optimal combination** of all information sources



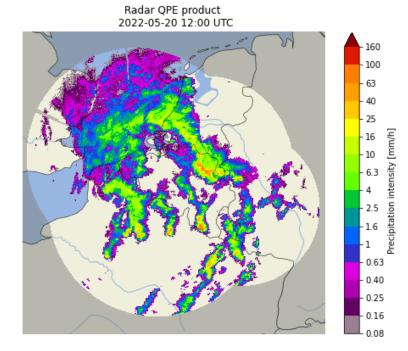
forecast skill

# Nowcasting and blending in pySTEPS



# Rainfall patterns show scaling behaviour:

- rain fields are not organized as a collection of cells but are a continuum of structures over a broad range of scales (~ 200 km - 100m)
- each scale has a characteristic lifetime (limits predictability)

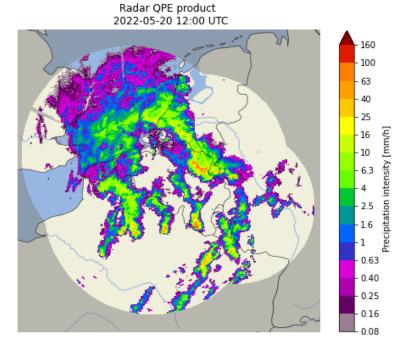


# pySTEPS: methodology

### Start from radar image/product

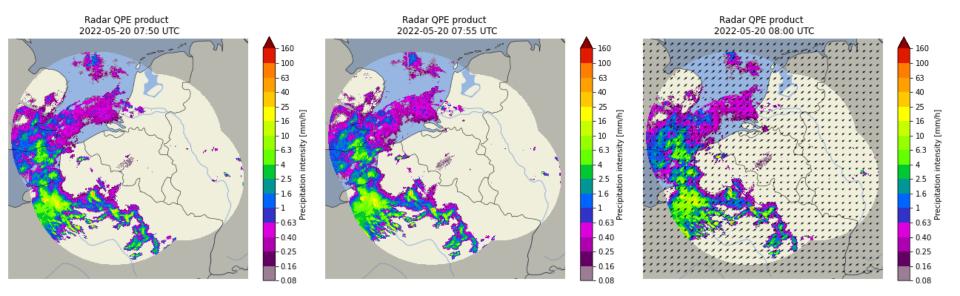
- 1. Derive advection vectors
- **2. Decomposition** into different scales (cascade)
- **3. Lagrangian evolution** of the rainfield and **noise generation**
- 4. Blending with NWP forecast and noise

Original concept by Seed, 2003, Journal of Applied Meteorology, 42, 381

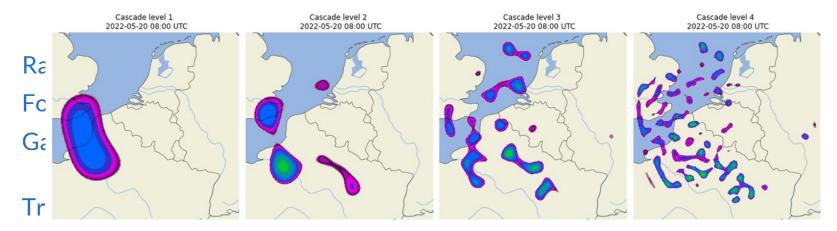


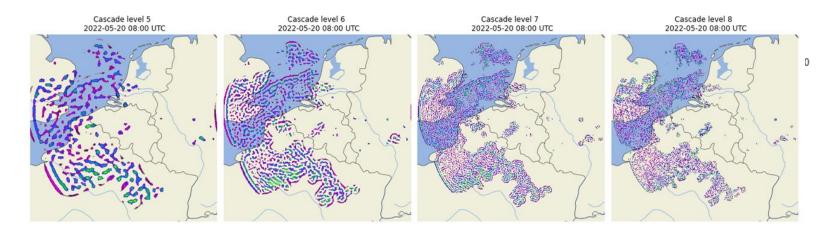
### (1) Derive advection vectors

# Using optical flow algorithms (Lucas-Kanade) on sequence of radar images/products



# (2) Decomposition into "cascade"





# (3) Lagrangian evolution & creation of noise

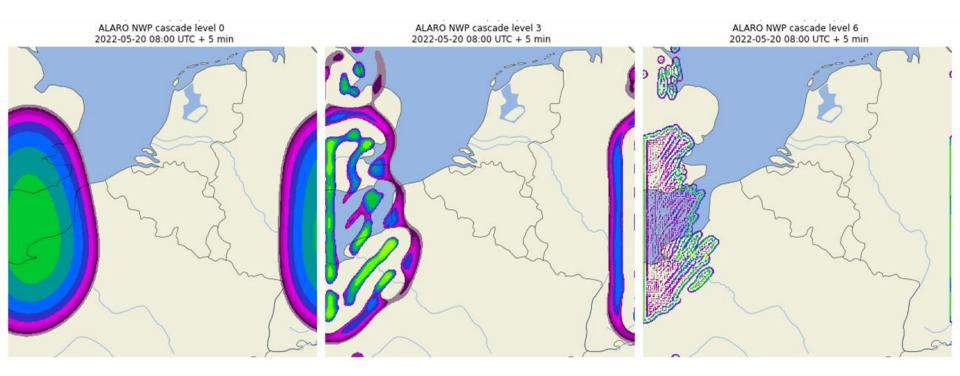
Evolution of each cascade level modelled with AR-2 process

$$Y_{k}(t+2\Delta t) = \Phi_{k,1}Y_{k}(t+\Delta t) + \Phi_{k,2}Y_{k}(t)$$

Parameters ( $\phi$ ) are function of Lagrangian autocorrelation and regressed to climatological values

Create noise cascade with appropriate spatial correlation for each scale (Gaussian filtering) and evolve using same AR-2 model

### (3) Lagrangian evolution & creation of noise

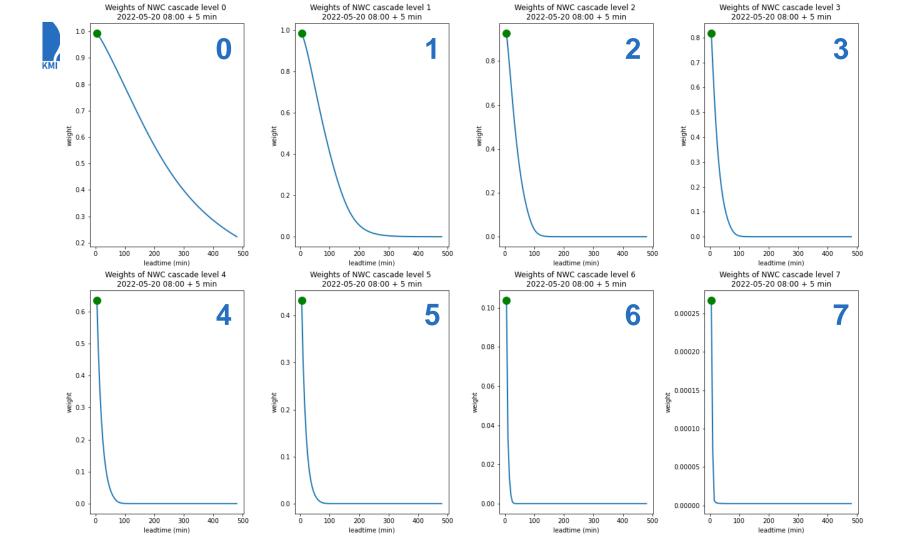


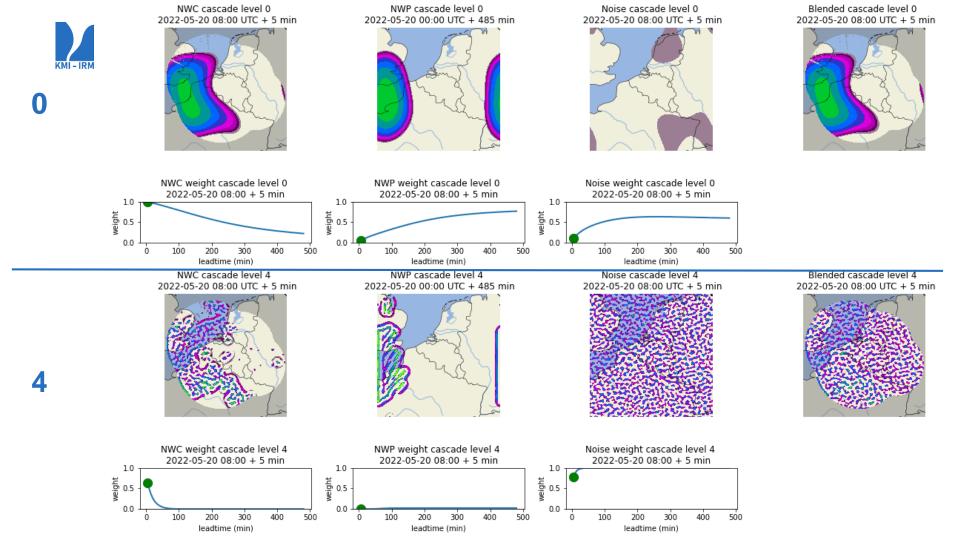
Combining the EXTRAPOLATION, NOISE and NWP cascades

$$Y_{k} = \omega_{k}^{e} Y_{k}^{e} + \omega_{k}^{n} Y_{k}^{n} + \omega_{k}^{m} Y_{k}^{m}$$

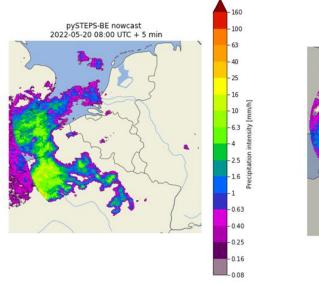
Weights are depended on the skill of each component

- Extrapolation: start from perfect skill and evolve according to AR-2 model
- NWP: start from skill w.r.t. most recent radar image and evolve towards skill of past days





# pySTEPS at RMI: pySTEPS-BE



Radar QPE product

2022-05-20 08:05 UTC

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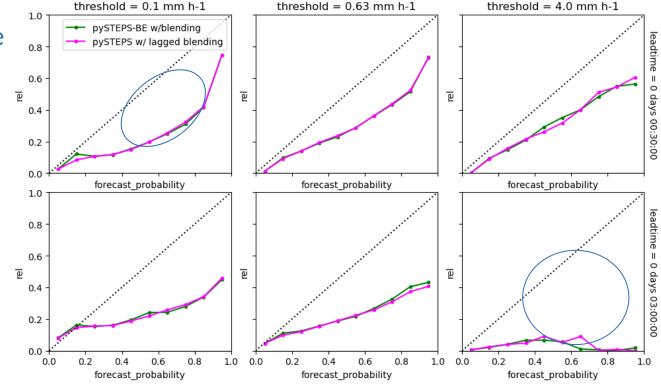
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- pySTEPS-BE every 10' in container
  - range: + 6h
  - timestep: 5'
  - 48 members
- Runtime ~ 6'
- Output: netCDF ~ 2 GB

# **pySTEPS-BE: verification**

### Quality indicators tailored to probabilistic forecasts:

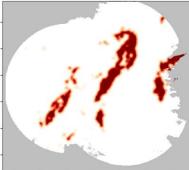
Reliability diagrams Fractional Skill Score Skillful Spatial Scale



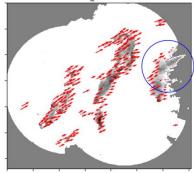
### pySTEPS: open source, community driven

### https://pysteps.github.io/

### **Probability forecast**

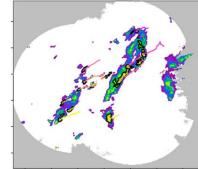


**Optical flow** 



# Downscaling

### Cell tracking



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Q Search the docs ...

#### FOR USERS

Installation Gallery

My first nowcast (Colab Notebook) 🖻 API Reference

Example data Configuration file (pystepsrc) Machine learning applications Bibliography

FOR DEVELOPERS Contributing Guide Importer plugins Testing Building the docs Packaging

Publishing to conda-forge GitHub repository Z

Theme by the Executable Book Project

Read the Docs v: stable •



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Optical flow









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Advection correction

Precipitation downscaling with RainFARM Probability forecasts



Extrapolation nowcast

Cascade

decomposition

Generation of T stochastic noise D

STEPS nowcast

Thunderstorm Detection and Tracking - T-DaTing





LINDA nowcasts

Ensemble verification





# What's next?

# Data assimilation in ALARO & AROME (Mode-S, GNSS, radar, ...)

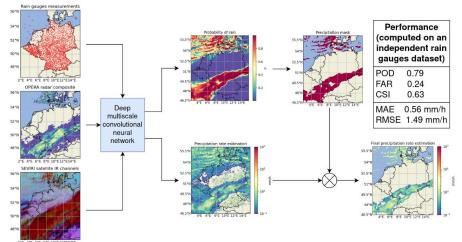
### Making project IMA operational

- How often to run the blended/extrapolation-only nowcast?
- How to treat the interval 6-24h?

Move to Al-based multi-modal QPE (Moraux et al., 2019; 2021)

Future research:

- Investigate AI-based "self-learning" blending and QPN
- Extend beyond the 24h time horizon by blending with ECMWF EPS



courtesy of Arthur Moraux

# Thank you!

Het Koninklijk Meteorologisch Instituut

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