



A 20-year reforecast study combining high-resolution hydrological modelling, ensemble forecasting and data assimilation for the 12 largest tributaries of the Rhine





Deltares



PhD thesis

Public defense LIVE STREAM:

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Interpolate, Simulate, Assimilate operational aspects of improving hydrological forecasts in the Rhine basin

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Rhine basin

- ~160.000 km2
- Operational forecasting system for water levels and discharge at the Dutch border
- 12 subbasins (Moselle, Main, etc.) from which the discharge is fed into a hydrodynamic model of the Rhine
- The most upstream part (Rhine above Maxau) is characterized by the Alps and natural lakes
- Stations from which data is assimilated are in red dots.
- Assimilation / error correction is done for each subbasin separately





Why this study?

- Ongoing movement towards spatially distributed models on (very) high spatial and temporal resolutions
- Ongoing movement to implement ensemble forecasts
- Ongoing development to combine ensemble forecasts with state-updating data assimilation
- But only few studies that combine all three...
- And even less so over long time periods



Data

- Hourly 1.2x1.2 km forcing data set (1998-2016)
- ECMWF hindcast data set for 20 years of forecasts every 3-4 days.





 52^{-} 51^{-} 50^{-} 49^{-} 48^{-} 47^{-} 6 8 10 10 5 0 -5-10

Avg. daily mean temperatu

(†)

(CC)

Avg. yearly PET sum



van Osnabrugge, B. (Bart) (2017) Gridded precipitation dataset for the Rhine basin made with the genRE interpolation method. Deltares.

Dataset. https://doi.org/10.4121/uuid:c875b385-ef6d-45a5-a6d3-d5fe5e3f525d

van Osnabrugge, B. (Bart) (2018) Gridded Hourly Temperature, Radiation and Makkink Potential Evaporation forcing for hydrological modelling in the Rhine basin. Wageningen University & Research. Dataset. <u>https://doi.org/10.4121/uuid:e036030f-</u> c73b-4e7b-9bd4-eebc899b5a13



Model: wflow_hbv

- Based on the well known HBV concept
- Discretisized for each grid cell
- Flows are routed with kinematic wave
- Hourly time step
- 1.2x1.2 km
- Upper zone and Lower zone states are updated as well as the water level



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Methods and materials Data assimilation and Auto Regression

Asynchronous EnKF

• AEnKF in OpenDA (Ridler et al., 2014, see also Rakovec et al. 2015)

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AR

- ARMASA algorithm in Delft-FEWS (Broersen and Weerts, 2005)
- $X_t = c + e_t + \sum_{i=1}^p \varphi_i X_{t-i} + \sum_{i=1}^q \theta_i e_{t-i}$

c: constant

e: white noise error

- φ : auto regressive parameters
- θ : moving average parameters

Here: only AR part is used with maximum order 3



Historic state estimation

- KGE of modelled discharge as proxy for correct state
- KGE improves greatly for each basin with AEnKF state updating
- Especially when initial model results were very poor
- Calculated over a 20 year simulation period on a daily time step





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Shown as hydrograph

- Effect of updating is shown in the hydrograph
- Correction is applied for whole biased periods

Year with lowest KGE for station Opladen in Wupper basin



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Results in forecast mode

Forecast verification scores. MCRPS (left) and MCRP **Skill Score** against climatology (middle) or the open loop simulation (right)

Solid line: open loop simulation

Dotted line: AEnKF

Stripe-dot: AR correction

Forecast verification score for station Dietersheim in Nahe basin

Comparison AR and AEnKF

- Below zero line: EnKF has better results than AR
- Colors indicate the quality of the original (open loop) simulation (colorbar on the right)
- AR outperforms EnKF for the first ~2 days, EnKF outperforms for longer forecasts
- Exceptions are poorly modelled basins

Conclusions

- Large scale data assimilation experiment combining current trends in hydrological forecasting: high resolution spatial and temporal model, with ensembles and state-updating but for a long continuous time period
- AR correction strong for 2 day period
- AEnKF improves longer forecasts
- AR can give 'odd' results, while AEnKF results are always consistent with the model

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