

Peer review status:

This is a non-peer-reviewed preprint submitted to EarthArXiv.

1	Operational Flood Forecasting System in Denmark – Integrating
2	Groundwater and Surface-water
3	
4	Jun Liu *, Julian Koch, Simon Stisen, Lars Troldborg, Raphael J. M. Schneider *
5	Department of Hydrology, Geological Survey of Denmark and Greenland, Copenhagen 1350, Denmark
6	
7	Corresponding author email (juliu@geus.dk; rs@geus.dk)
8	
9	Keywords: Operational Flood Forecasting; Flood Warning; Integrated Surface Water-Groundwater Modelling;
10	National Hydrological Model; Rising Groundwater Levels.
11	
12	Abstract
13	Most operational flood forecasting systems provide predictions of pluvial and fluvial floods, often neglecting
14	groundwater flooding processes. Groundwater flooding occurs when natural drainage system cannot drain water
15	away quick enough, causing the water table to rise above ground. This study presents an operational integrated
16	flood forecasting system that combines surface water and groundwater components. It provides key variables -
17	such as river discharge and shallow groundwater levels - for assessing flood risk in Denmark. During the winter of
18	2024, the system predicted relative shallow groundwater levels with an average error of 0.27 m at a 5-day lead
19	time at selected wells. Moreover, the system effectively captures peak flows in rivers alongside high groundwater
20	levels, as exemplified for a specific local flood event in Varde in January 2024. This work demonstrates how
21	groundwater flooding, often neglected in operational forecasting, can be included at the national scale.
22	
23	
24	
25	
26	
27	
28	
29	
30	
31	

1 Introduction

2 Floods are among the most devastating natural disasters worldwide, causing substantial socio-3 economic damage (Tellman et al. 2021). While fluvial and pluvial floods are well-recognized and 4 investigated for decades due to their immediate and visible impacts (Merz et al. 2021), flooding 5 caused by rising groundwater is often overlooked (Behzad and Nie 2024; Kreibich et al. 2009). 6 Groundwater flooding occurs when natural drainage systems cannot remove water quickly 7 enough, causing the water table to rise above the ground (Parkin 2024; Becker et al. 2022). It can 8 develop more gradual, often persists for weeks or months, feedback to surface waters, and can 9 also cause long-term damage - particularly to below-ground infrastructure such as basements, 10 tunnels, and sewage systems. The British Geological Survey estimates that groundwater flooding is responsible for approximately £530 million in damages annually in the United Kingdom, 11 12 representing around 30% of the country's total economic loss due to flooding (Allocca et al. 2021).

Flood forecasting systems play a critical role in reducing the societal impacts of floods by 13 14 supporting early warning, emergency planning, and climate adaptation strategies (L. J. Speight 15 et al. 2021). Many countries have developed national and regional forecasting frameworks, such as the National Water Prediction Service in the United States (NOAA 2016), the European Flood 16 17 Awareness System (Smith et al. 2016), the Australian Flood Warning Services (Pagano et al. 2016), and the British Flood Alerts and Warnings (L. Speight et al. 2025). These systems vary in 18 19 capabilities and tend to focus on surface water processes, with limited attention to integrated 20 surface water - groundwater interactions. Such forecasts may underestimate the full extent and 21 severity of a flood event without accounting for groundwater processes (e.g., low infiltration 22 buffer in saturated soils and groundwater exfiltration). This is particularly problematic in low-lying 23 areas or places with shallow water tables, where groundwater can emerge to the surface even 24 without heavy surface runoff (Becker et al. 2022). Conventional surface water-focused systems 25 often miss slow-onset groundwater floods. These events may not trigger standard warning thresholds, leading to delayed or absent alerts for communities vulnerable to prolonged flooding 26 27 of basements, roads, or critical infrastructure. Rising groundwater level induced floods can persist 28 for a longer duration and emergency planning and allocation measures will be inefficient when 29 groundwater contributions are neglected in flood warning systems (Parkin 2024).

Denmark, where both surface water and groundwater flooding pose recurring risks, experiences 1 2 considerable annual flood-related economic losses (Halsnæs, Larsen, and Drenck 2022). 3 Responding to this need, the Geological Survey of Denmark and Greenland (GEUS) has developed 4 a real-time, operational flood forecasting system that integrates both surface water and 5 groundwater processes. This system is designed to provide short-, medium-, and long-term forecasts of fluvial and groundwater floods at a national scale. In this study, we present the 6 7 development and initial performance of this integrated forecasting system under operational conditions. We carry out an initial event-based evaluation of its predictive capabilities with focus 8 9 on shallow groundwater dynamics, discuss limitations, and explore future directions. By 10 highlighting Denmark's experience, this work aims to contribute to the global advancement of integrated flood forecasting and to offer guidance for similar initiatives in other countries that 11 12 are exposed to groundwater flood risk.



15 16

13 14

10

Overview of the forecasting system

18 GEUS' flood forecasting system is an operational surface water - groundwater modelling system
 19 that integrates weather forecast downloading and preprocessing with pre-trained forecasting

20 models, e.g., physically based hydrological model and machine learning (ML) models, and data

management. The overall architecture of the forecasting system is illustrated in Figure 1. The process begins with downloading various near real-time and forecasted weather data from meteorological archives, which are then processed to the input formats required by the hydrological forecasting models. The models are then updated and run to predict water dynamics. Upon completion, the flood-relevant variables (discharge, depth to phreatic surface, etc.) are delivered to a database for data management, which serves as the backend for web viewer and sharing to users. The following sub-sections provide detailed descriptions of these components.

8

9

Climate data

10 The climate forcings, including precipitation, air temperature, and potential evapotranspiration, 11 used in the system consist of both observational data and weather forecasts. Observation-based 12 historical and real-time climate data are sourced from national gridded datasets provided by the 13 Danish Meteorological Institute (DMI). These datasets, available at daily resolution, extend back 14 to 1989 (Scharling 1999a; 1999b; DMI, n.d.). Real-time data is available via the DMI Open Data 15 API (https://opendatadocs.dmi.govcloud.dk/DMIOpenData). The variables are based on data 16 from a network of in situ weather stations distributed throughout Denmark. Air temperature is measured at a height of 2 meters above ground, and potential evapotranspiration is estimated 17 18 using a modified Makkink's equation (Plauborg et al. 2002). Operational precipitation gauges are 19 typically installed 1–1.5 meters above ground level and are subject to wind-induced turbulence, 20 which causes a systematic undercatch of precipitation. To correct for this bias, an empirical 21 correction is dynamically applied based on rainfall intensity, wind speed, and temperature, 22 distinguishing between solid and liquid precipitation (Simon Stisen et al. 2011; Allerup, Madsen, 23 and Vejen 1997; S Stisen et al. 2012).

Climate forecasts used in the system are sourced from two providers: DMI's Weather Model (HARMONIE) for DINI (Denmark, Iceland, the Netherlands and Ireland) and the European Centre for Medium-Range Weather Forecasts (ECMWF, (Palmer et al. 1990). The HARMONIE model delivers local high-resolution short-term forecasts with a lead time of up to 56 hours. ECMWF provides a suite of forecasts at various temporal scales: medium-range forecasts with lead times of 10 to 15 days, extended-range forecasts up to 46 days, and long-range seasonal forecasts extending up to 7 months. These forecasts support both short- and long-term hydrological
 modeling and enable comprehensive flood risk assessments.

3

4 Forecast models

5 The forecasting system uses two types of models: the physically based National Hydrological 6 Model (DK-model) and data-driven machine learning (ML) models. The DK-model is an integrated 7 surface water – groundwater model that covers most of Denmark's land area - approximately 8 43,000 km² and has been under continuous development over almost three decades (H J 9 Henriksen et al. 2021; Højberg et al. 2013; Koch et al. 2021; Schneider et al. 2022; Hans Jørgen 10 Henriksen, Ondracek, and Troldborg 2023). It is implemented in the MIKE SHE modelling 11 framework (Abbott et al. 1986; DHI 2020), which fully couples a finite-difference 3D subsurface 12 flow model with 2D overland flow, a simplified two-layer representation of the unsaturated zone, and 1D kinematic streamflow routing. The model has been calibrated for the period 2000 to 2010 13 14 using 304 daily streamflow time series and groundwater head observations from approximately 15 40,000 intakes across the country, with focus on shallow groundwater and streamflow in the 16 version (H J Henriksen et al. 2020) used as part of the forecasting system.

ML models have demonstrated promising potential for flood forecasting (Nearing et al. 2024). To 17 18 enhance both accuracy and computational speed of discharge predictions, we developed a hybrid 19 ML post-processing model based on Long Short-Term Memory (LSTM) neural networks (Liu et al. 20 2024). Trained on historical data, this model incorporates real-time climate information and DK-21 model simulations to generate 10-day discharge forecasts, taking into account weather 22 predictions during the forecasting period. The current ML system provides discharge forecasts for 23 approximately 3,000 catchment outlets across Denmark (Liu et al. 2025). A similar ML post-24 processing model for groundwater level forecasting is currently under development.

25

26 Data management

27 Operationally, the models begin with real-time simulations that incorporate a 30-day look-back 28 period each day. These simulations provide the most up-to-date estimates of shallow 29 groundwater levels and discharges, ensuring comparability with historical runs (also backup in the database) in terms of climate forcings and model configurations. The 30-day look-back period
 is set based on GEUS' experience to maintain stability of the DK-model and real-time climate data.
 The real-time simulations are submitted to the database but also serve as initial conditions for
 forecast runs.

5 The forecasts start after real-time simulations. Short-range forecasts (54 hours) and medium-6 range forecasts (10, 15, and 46 days) are generated once per day. Long-term forecasts (up to 7 seven months) are produced monthly, which will be available at the start of each month but 8 depend on the availability of ECMWF monthly weather forecasts. The medium-range forecasts 9 use ensemble modeling with 51 ensemble members. All simulations provide streamflow estimates 10 for up to 60,000 river points and shallow groundwater levels (represented by depth to the top 11 phreatic surface) at 100m and 500m resolutions. As of the completion of this paper, short-range 12 and 10-day ahead deterministic forecasts are available, while operational forecasts with 13 ensembles and long-term forecasts are under development.

Once the real-time simulations and forecasts are completed, the results are uploaded to a centralized database called HydroDB, which supports data storage, statistical analysis, visualization, and user sharing. HydroDB stores historical simulations spanning 30 years (including river discharge and depth to top phreatic surface, additional data is available upon request), updated annually to reflect newly available data. Real-time simulations and forecast outputs are updated daily and backed up for around one year. The current model structure data, e.g., river network, geological- and calculation layers, is stored in the database.

21 Using historical records, along with the most recent real-time and forecast data, statistical 22 analyses are conducted to assess current and forecasted conditions relative to historical baselines 23 (e.g., how wet or dry a period is). Through a REST api, the selected part of the data can be accessed 24 and visualized, eg. the webviewer at GEUS Hydromonitor (https://data.geus.dk/hydromonitor/) 25 and through an Structured Query Language (SQL) -gateway registrated users, such as Danish Agency for Climate Data (KDS) can access HydroDB. KDS uses access to stores a local copy for 26 27 direct external access to real-time data through the Hydrological Information and Prognosis 28 System (HIP, https://hipdata.dk/).

29



Figure 2. Example of forecasted absolute shallow groundwater (left) and discharge (right) on January 23, 2025. Shallow groundwater levels are presented as a raster map at 100 m resolution. The system also provides discharge forecasts at 62,728 points (from the 100m model, shown as red dots with size indicating discharge values) along nearly all river channels in Denmark. The up-to-date forecasting are available via GEUS Hydromonitor (https://data.geus.dk/hydromonitor/)

8 Performance and case application

9 Accuracy of groundwater level forecasts

10 The forecasting system has been operational since October 2024, providing 10-day ahead forecasts of shallow groundwater levels and river discharge based on ECMWF's 10-day weather 11 12 forecasts. Figure 3a shows changes in groundwater levels between two dates, i.e., October 15, 13 2024, and January 9, 2025. On average, groundwater levels increased by 1.70 meters across 14 Denmark during this period, with spatial variation (Figure 3a). We initially evaluated the relative 15 groundwater forecasts by comparing them with measurements from 7 randomly distributed wells 16 in Denmark. The subplots in Figure 3 display the 10-day groundwater level forecasts (blue curves) alongside observed measurements (dotted black curves) relative to the groundwater levels on 17 October 15, 2024. Overall, the simulated relative groundwater dynamics align well with the 18 19 observations. The root-mean-square-error (RMSE) between observations and 5-day ahead 20 forecasts (marked by red points in the subplots) ranges from 0.11 to 0.64 meters among the 7 21 wells. Discrepancies persist in absolute groundwater levels, suggesting that further model 22 improvement or post-processing is necessary.



Figure 3. Comparison of relative groundwater level forecasts (forced by ECMWF 10-day deterministic weather
 forecasts) and measurements at 7 wells during October 15, 2024, to February 15, 2025. Subplot (a) shows the
 differences between October 15, 2024, and January 9, 2025. The remaining subplots show the time series of
 groundwater level forecasts and measurements.

7 Case application

1

6

The DK-model is developed with moderate spatial resolution (up to 100 m), allowing flood 8 forecasting across scales but computationally manageable. To demonstrate its capabilities for 9 local applications, we present a flood event that occurred downstream of the Varde River in mid-10 11 western Jutland on January 23rd, 2024. The flood extent was captured by Sentinel-1 satellite imagery (Figure 4a), which reveals significant inundation across the region (Hansen at al. 2025 12 under review). River discharge increased from 17.81 m³/s on January 19th to 31.92 m³/s on 13 January 23rd (Figure 4a). Our model successfully captured this flood event at the river section, 14 15 with simulated discharge increasing from 16.07 m^3 /s to 32.47 m^3 /s. The RMSE of the simulated discharge was 3.99 m³/s for January 2024 compared to observations, demonstrating the system's 16 promising capability for river flood forecasting. 17 18 There are no groundwater well measurements available in this region at the time of interest. 19 Therefore, we compared the simulated groundwater level dynamics during the event to historical

20 averages (1989–2023) to illustrate groundwater behavior during the flood events. As shown in

1 Figure 4b, groundwater levels are significantly higher than the historical averages across most of

- 2 the area. Even in the river valley, where groundwater is typically close to the surface, levels were
- 3 elevated by 0.2–0.4 m compared to the historical baseline.
- Figures 4c and 4d present two profiles showing model topography, groundwater levels, flooded
 areas along the profiles, and the location of the main river channels. These profiles further confirm
 that during the event groundwater levels (green curves) are consistently higher than the historical
 averages (blue curves). The flooding is concentrated along the river channel but also extends
 across the adjacent plains, as indicated by the gray vertical lines representing flooded locations.
 Notably, the simulated groundwater levels exceed the surface topography in some sections of the
 profiles, corresponding well with the observed inundation.





12 Figure 4. Application of the flood forecasting system to a local case study (Varde River). (a) Inundation extent derived 13 from satellite imagery on January 23, 2024, and a comparison of simulated and observed discharge at the 14 hydrological station during January 2024; (b) Relative changes of the groundwater levels on the same date compared 15 to historical reference (mean values of 1989-2023) for the area; (c) and (d) are vertical profiles along two cross-16 sections across river valley. The curves show model topography, groundwater level dated January 23, 2024, and 17 groundwater level from historical reference. Flooded area from satellite data is indicated in grey, and the river 18 location in red. Please note that the real water surface elevation is unknown, so the grey and read bars do not indicate 19 water depth of the inundated area but the locations.

20

1 Perspectives and conclusions

2 This study presented the ongoing development of a national-scale operational flood forecasting 3 system for Denmark that explicitly integrates surface water and groundwater hydrological 4 processes. Groundwater is often neglected in operational flood forecasting; hence, its inclusion 5 represents a significant advancement. In groundwater-dominated regions such as Denmark, the 6 integration of groundwater into a forecasting framework enables the prediction of groundwater 7 flooding, a regularly occurring but often overlooked hazard (Parkin 2024). Additionally, the 8 explicit representation of slower-reacting delayed groundwater processes enhances streamflow 9 forecasts from more conventional rainfall-runoff type models (as e.g. shown by Liu et al. (2024)). 10 Ongoing and future developments of the system aim to further increase forecast skill and expand 11 lead time. These include the integration of ensemble and seasonal weather forecasts. Also, special 12 attention is being paid to improving the accuracy of absolute groundwater level predictions; 13 amongst others by developing Deep Learning post-processors to enhance groundwater level 14 predictions, comparable to the already applied post-processing of streamflow simulations (Liu et 15 al. 2024). Nevertheless, relative groundwater levels (e.g. as quantiles or return events) provide 16 relevant information, and such statistical indicators are more robustly simulated than absolute 17 *levels* (Seidenfaden et al. 2025). *Finally, derived local scale models are envisaged for particularly* 18 vulnerable areas, especially urban areas at the risk of being affected by compound events of flooding from rivers, sea and groundwater (Seidenfaden et al. 2024). 19

The Danish integrated system developed across a range of institutions can serve as a reference for other countries facing similar hydrological hazards. This work highlights that with appropriate data, modeling infrastructure and institutional collaboration, it is feasible to implement such integrated systems at the national scale – ultimately improving preparedness and resilience to a broader range of flood hazards.

25

26 Acknowledgements

This work is conducted as part of the Danish flood warning system (varslingssystem for
oversvømmelser) from 2023 to 2026. The project is being led by the Danish Meteorological
Institute (DMI) in cooperation with the Geological Survey of Denmark and Greenland (GEUS), the
Danish Agency for Climate Data (KDS), the Danish Environmental Protection Agency (MST), the
Page 10 of 15

Danish Coastal Authority (KDI), and the Danish Environmental Portal (DMP). We acknowledge the
 participants of the project, and further persons involved in the development of the Danish
 National Hydrological Model (DK-model) at GEUS.

4

5 References

- 6 Abbott, M B, J C Bathurst, J A Cunge, P E O'Connell, and J Rasmussen. 1986. "An Introduction to
- 7 the European Hydrological System Systeme Hydrologique Europeen, 'SHE', 1: History
- 8 and Philosophy of a Physically-Based, Distributed Modelling System." *Journal of Hydrology*
- 9 87 (1): 45–59. https://doi.org/https://doi.org/10.1016/0022-1694(86)90114-9.
- 10 Allerup, P, H Madsen, and F Vejen. 1997. "A Comprehensive Model for Correcting Point
- 11 Precipitation." *Hydrology Research* 28 (1): 1–20.
- 12 Allocca, Vincenzo, Mariano Di Napoli, Silvio Coda, Francesco Carotenuto, Domenico Calcaterra,
- 13 Diego Di Martire, and Pantaleone De Vita. 2021. "A Novel Methodology for Groundwater
- 14 Flooding Susceptibility Assessment through Machine Learning Techniques in a Mixed-Land
- 15 Use Aquifer." *Science of the Total Environment* 790:148067.
- 16 Becker, Bernhard, Frank Reichel, Daniel Bachmann, and Reinhard Schinke. 2022. "High
- 17 Groundwater Levels: Processes, Consequences, and Management." Wiley Interdisciplinary
- 18 *Reviews: Water* 9 (5): 1–21. https://doi.org/10.1002/wat2.1605.
- Behzad, Hamid M, and Yunpeng Nie. 2024. "Groundwater Flooding Risks Overlooked." *Science*384 (6695): 518–19.
- 21 DHI. 2020. "MIKE SHE User Guide and Reference Manual."
- 22 DMI. n.d. "Frie Data." https://www.dmi.dk/frie-data.
- 23 ———. 2024. "Weather Model (HARMONIE) for DINI and IG." 2024.
- https://opendatadocs.dmi.govcloud.dk/Data/Forecast_Data_%0AWeather_Model_HARM
 ONIE DINI IG.
- 26 Halsnæs, K, M A D Larsen, and K L Drenck. 2022. "Samfundsøkonomiske Konsekvenser Af
- Oversvømmelser Og Investeringer i Klimatilpasning." Kgs. Lyngby, Denmark: DTU for
 Miljøministeriet.
- 29 Henriksen, H J, S J Kragh, J Gotfredsen, M Ondracek, M van Til, A Jakobsen, R J M Schneider, et
- 30 al. 2021. Udvikling Af Landsdækkende Modelberegninger Af Terrænnære Hydrologiske Page 11 of 15

1 Forhold i 100m Grid Ved Anvendelse Af DK-Modellen: Dokumentationsrapport Vedr. 2 Modelleverancer Til Hydrologisk Informations- Og Prognosesystem. Udarbejdet Som En Del 3 Af Den Fællesoffen. GEUS. https://doi.org/10.22008/gpub/38113. 4 Henriksen, H J, S J Kragh, J Gotfredsen, M Ondracek, M van Til, A Jakobsen, R J M Schneider, J 5 Koch, L Troldborg, and P Rasmussen. 2020. "Dokumentationsrapport Vedr. 6 Modelleverancer Til Hydrologisk Informations-Og Prognosesystem." Copenhagen: GEUS. 7 Henriksen, Hans Jørgen, Maria Ondracek, and Lars Troldborg. 2023. Vandressourceopgørelse – 8 Datarapport. Baggrundsrapport Til Miljøstyrelsens Samlede Afrapportering Omkring 9 Forvaltning Af Fremtidens Drikkevandsressource. Metode, Resultater, Usikkerheder Og 10 Forventede Klimapåvirkninger. Vol. 2023. Danmarks Og Grønlands Geologiske Undersøgelse Rapport. GEUS. https://doi.org/10.22008/gpub/34675. 11 12 Højberg, Anker Lajer, Lars Troldborg, Simon Stisen, Britt B.S. Christensen, and Hans Jørgen 13 Henriksen. 2013. "Stakeholder Driven Update and Improvement of a National Water Resources Model." Environmental Modelling and Software 40:202–13. 14 15 https://doi.org/10.1016/j.envsoft.2012.09.010. 16 Koch, Julian, Jane Gotfredsen, Raphael Schneider, Lars Troldborg, Simon Stisen, and Hans 17 Jørgen Henriksen. 2021. "High Resolution Water Table Modeling of the Shallow 18 Groundwater Using a Knowledge-Guided Gradient Boosting Decision Tree Model." 19 Frontiers in Water 3 (September): 1–14. https://doi.org/10.3389/frwa.2021.701726. 20 Kreibich, H, A H Thieken, H Grunenberg, K Ullrich, and T Sommer. 2009. "Extent, Perception 21 and Mitigation of Damage Due to High Groundwater Levels in the City of Dresden, 22 Germany," 1247–58. 23 Liu, Jun, Julian Koch, Simon Stisen, Lars Troldborg, Anker Lajer Højberg, Hans Thodsen, Mark F T 24 Hansen, and Raphael J M Schneider. 2025. "CAMELS-DK : Hydrometeorological Time Series 25 and Landscape Attributes for 3330 Danish Catchments with Streamflow Observations from 304 Gauged Stations," 1551–72. 26 27 Liu, Jun, Julian Koch, Simon Stisen, Lars Troldborg, and Raphael J.M. Schneider. 2024. "A 28 National-Scale Hybrid Model for Enhanced Streamflow Estimation - Consolidating a

29 Physically Based Hydrological Model with Long Short-Term Memory (LSTM) Networks."

Page 12 of 15

Hydrology and Earth System Sciences 28 (13): 2871–93. https://doi.org/10.5194/hess-28 2871-2024.

3 Merz, Bruno, Günter Blöschl, Sergiy Vorogushyn, Francesco Dottori, Jeroen C J H Aerts, Paul 4 Bates, Miriam Bertola, Matthias Kemter, Heidi Kreibich, and Upmanu Lall. 2021. "Causes, Impacts and Patterns of Disastrous River Floods." Nature Reviews Earth & Environment 2 5 6 (9): 592–609. 7 Nearing, Grey, Deborah Cohen, Vusumuzi Dube, Martin Gauch, Oren Gilon, Shaun Harrigan, 8 Avinatan Hassidim, et al. 2024. "Global Prediction of Extreme Floods in Ungauged 9 Watersheds." Nature 627 (8004): 559–63. https://doi.org/10.1038/s41586-024-07145-1. 10 NOAA. 2016. "National Water Model: Improving NOAA's Water Prediction Services." Office of Water Prediction. 11 12 Pagano, T C, J F Elliott, B G Anderson, and J K Perkins. 2016. "Australian Bureau of Meteorology Flood Forecasting and Warning." In *Flood Forecasting*, 3–40. Elsevier. 13 Palmer, T N, C Brankovic, F Molteni, S Tibaldi, L Ferranti, A Hollingsworth, U Cubasch, and E 14 15 Klinker. 1990. "The European Centre for Medium-Range Weather Forecasts (ECMWF) Program on Extended-Range Prediction." Bulletin of the American Meteorological Society 16 17 71 (9): 1317–30. Parkin, Geoff. 2024. "Groundwater Flooding–a Hidden Hazard." In Proceedings of the Institution 18 19 of Civil Engineers-Civil Engineering, 177:50–52. Emerald Publishing Limited. 20 Plauborg, Finn Lars, Jens Christian Refsgaard, Hans Jørgen Henriksen, Gitte Blicher-Mathiesen, 21 and Claus Kern-Hansen. 2002. "Vandbalance På Mark-Og Oplandsskala." 22 Scharling, Mikael. 1999a. "Klimagrid Danmark - Nedbør, Lufttemperatur Og Potentiel 23 Fordampning 20X20 & 40x40 Km - Metodebeskrivelse." Danish Meteorological Institute, 24 Technical report (Danish Meteorological Institute), . 25 https://books.google.dk/books?id=El3bcQAACAAJ. ———. 1999b. "Klimagrid Danmark Nedbør 10x10 Km (Ver. 2) - Metodebeskrivelse." Danish 26 27 *Meteorological Institute*, 15–17. 28 Schneider, Raphael, Julian Koch, Lars Troldborg, Hans Jørgen Henriksen, and Simon Stisen. 29 2022. "Machine-Learning-Based Downscaling of Modelled Climate Change Impacts on

- 1 Groundwater Table Depth." *Hydrology and Earth System Sciences* 26 (22): 5859–77.
- 2 https://doi.org/10.5194/hess-26-5859-2022.
- 3 Seidenfaden, Ida Karlsson, Raphael Schneider, Bertel Nilsson, Julian Koch, Lars Troldborg, Klaus
- 4 Hinsby, Mark F T Hansen, et al. 2025. *Hydrologisk Tørke. Temarapport 2 i Det*
- 5 Tværorganisatoriske Samarbejde "Styrket Vidensgrundlag for Tørke" Mellem KDS, GEUS Og
- 6 DMI. Danmarks Og Grønlands Geologiske Undersøgelse Rapport. GEUS.
- 7 https://doi.org/10.22008/gpub/34762.
- 8 Seidenfaden, Ida Karlsson, Maria Rebekka Skjerbæk, Hans Jørgen Henriksen, and K Kristian.
- 9 2024. "Compound Flooding from Storm Surges, Rivers, and Groundwater Hydrodynamic
 10 Modelling in a Coastal Catchment."
- 11 Smith, P J, Florian Pappenberger, Fredrik Wetterhall, J Thielen Del Pozo, Blazej Krzeminski, Peter
- 12 Salamon, Davide Muraro, Milan Kalas, and Calum Baugh. 2016. "On the Operational
- 13 Implementation of the European Flood Awareness System (EFAS)." In *Flood Forecasting*,
- 14 313–48. Elsevier.
- Speight, Linda, Cathryn E Birch, Katherine Self, and Sally Brown. 2025. "Expert Perspectives on
 the next Generation of UK Surface Water Flood Warning Services." *Weather*.
- 17 Speight, Linda J., Michael D. Cranston, Christopher J. White, and Laura Kelly. 2021. "Operational
- 18 and Emerging Capabilities for Surface Water Flood Forecasting." Wiley Interdisciplinary
- 19 *Reviews: Water* 8 (3): 1–24. https://doi.org/10.1002/wat2.1517.
- 20 Stisen, S, A L Højberg, L Troldborg, J C Refsgaard, B S B Christensen, M Olsen, and H J Henriksen.
- 21 2012. "On the Importance of Appropriate Precipitation Gauge Catch Correction for
- Hydrological Modelling at Mid to High Latitudes." *Hydrology and Earth System Sciences* 16
 (11): 4157–76.
- 24 Stisen, Simon, Torben O. Sonnenborg, Anker L. Højberg, Lars Troldborg, and Jens Christian
- 25 Refsgaard. 2011. "Evaluation of Climate Input Biases and Water Balance Issues Using a
- 26 Coupled Surface-Subsurface Model." *Vadose Zone Journal* 10 (1): 37–53.
- 27 https://doi.org/10.2136/vzj2010.0001.
- 28 Tellman, Beth, Jonathan A Sullivan, Catherine Kuhn, Albert J Kettner, Colin S Doyle, G Robert
- 29 Brakenridge, Tyler A Erickson, and Daniel A Slayback. 2021. "Satellite Imaging Reveals

- 1 Increased Proportion of Population Exposed to Floods." *Nature* 596 (7870): 80–86.
- 2
- 3

4 Additional information

5 Please complete all fields of this table.

Funding statement	This study was supported by the national public budget
Author contributions	Writing – Original Draft: Jun Liu, Raphael J. M. Schneider
	Writing – Review & Editing: Julian Koch, Raphael J. M. Schneider, Simon Stisen,
	Lars Troldborg
	Conceptualization: Raphael J. M. Schneider; Julian Koch
	Project Administration: Lars Troldborg
Competing interests	The authors declare no competing interest.
Additional files	None

6