

Summer School on Extreme Temperature and Precipitation Events

Session 3: Floods

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Organization: Justus Liebig University Giessen - JLU

Cluj-Napoca, Romania 10 September 2025



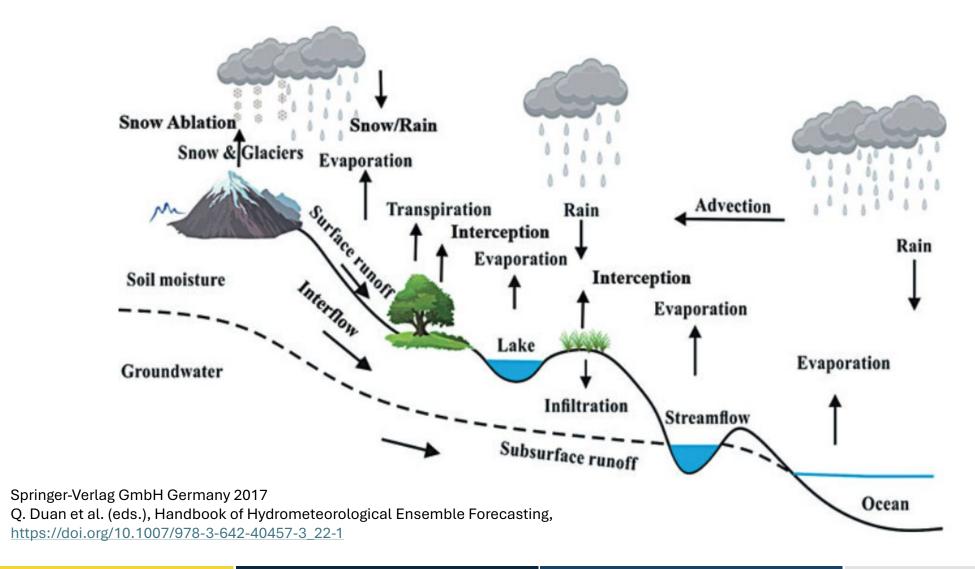








scewero 1. Water cycle







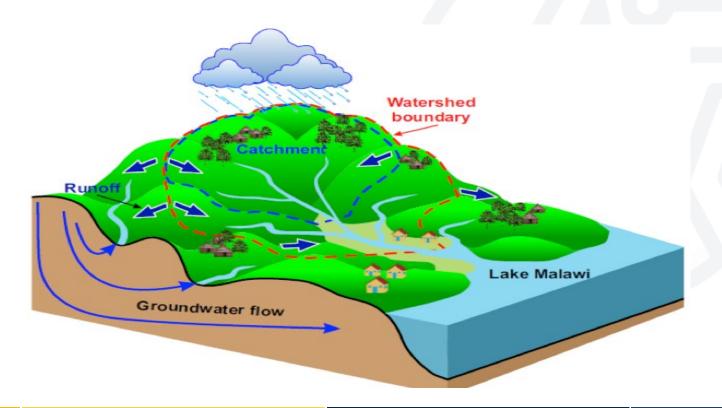


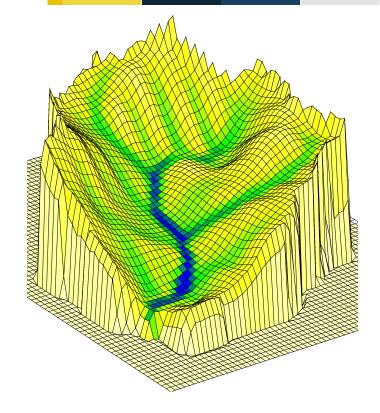


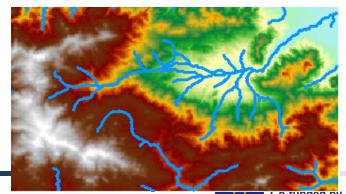


Catchment

Catchment is area where all water, drains to same outlet.











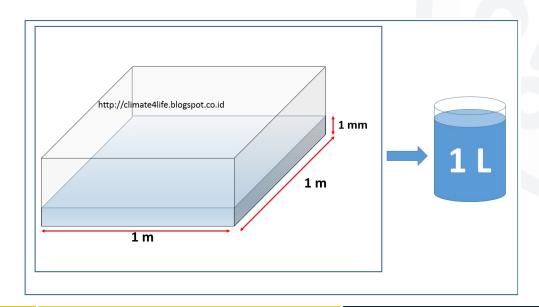






Precipitation

Measuring **rainfall** as a depth is an international standard mainly because it is very easy to convert depth to volume; The conversion factor is **1 mm** of **rainfall** = **1 Liter** of water/ \mathbf{m}^2





1 mm -> 1 mm/day

$$i = \frac{\Delta P}{\Delta t} \quad \left[\frac{mm}{hr}\right]$$

$$i = \frac{5 mm}{1 day} = 0.208 \frac{mm}{hr}$$



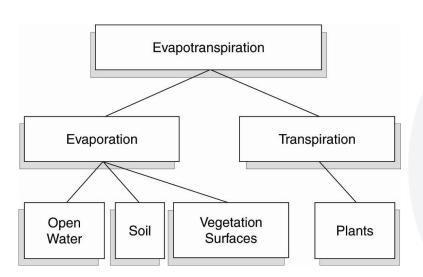




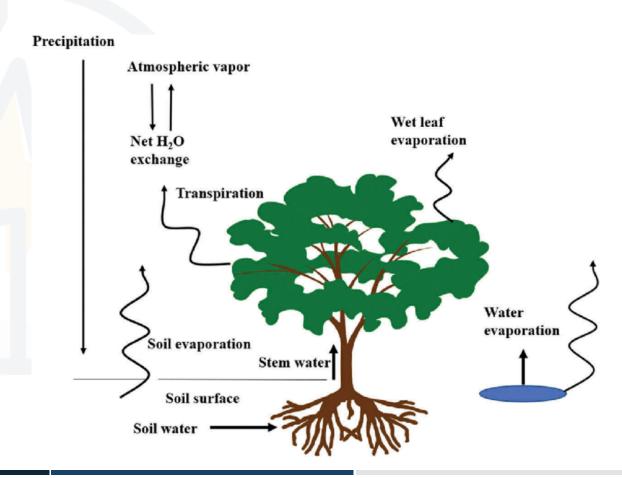




Evaporation, transpiration, and evapotranspiration









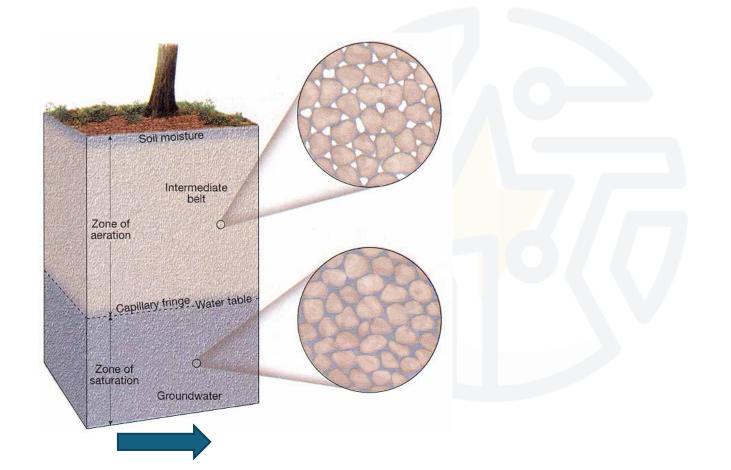








Infiltration















evapotranspiration = transpiration + evaporation

grass

groundwater

recharge

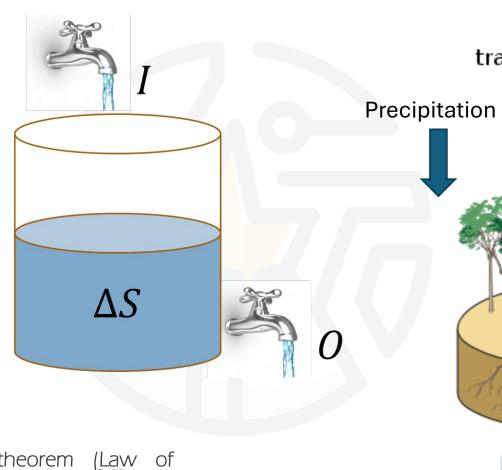
evaporation

transpiration

trees

Water Balance

$$I - O = \Delta S$$



On basis Reynolds transport theorem (Law of conservation mass, law of conservation of linear momentum, law of conservation energy)









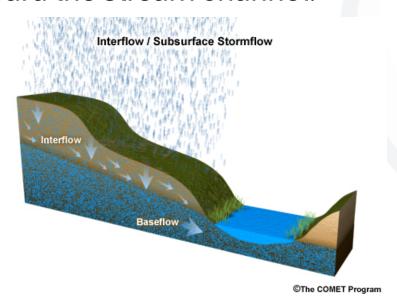




runoff

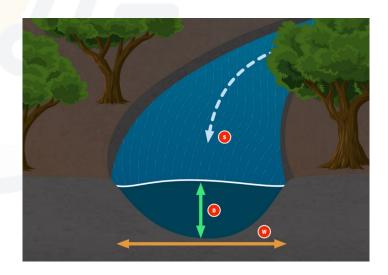
Streamflow

Streamflow, discharge, or runoff is the portion of raifall, snowmelt or irrigation water that runs over the soil surface toward the stream channel.



$Q = A \bar{u}$

- **Q** is the discharge ($[L^3T^{-1}]$; m³/s or ft³/s)
- A is the cross-sectional area of the portion of the channel occupied by the flow ($[L^2]$; m^2 or ft^2)
- **u** is the average flow velocity ($[LT^{-1}]$; m/s or ft/s)









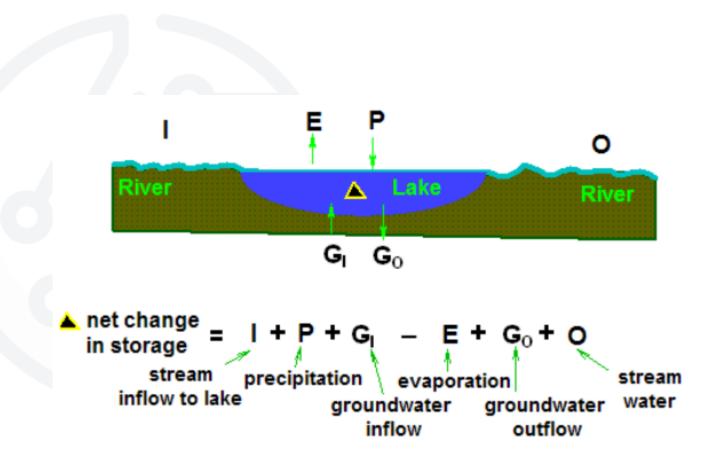






Water Balance















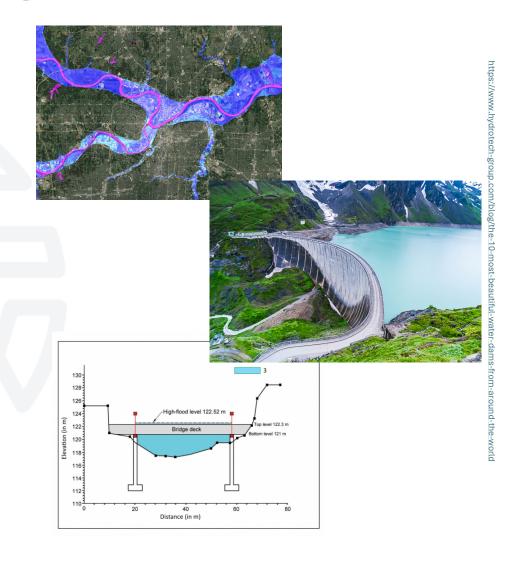


The use of models allows the stakeholder and decision makers to take descions on basis of limited measurements and qualitative predictions in space and time of the hydrological processes.

Modelling is a fundamental and quantitative way to understand and analyze complex systems and phenomena.

Modelling is a complement to theory and experiments.

For example: evaluating the water availability for human consumption, floods, droughts, irrigation, water contamination and others.









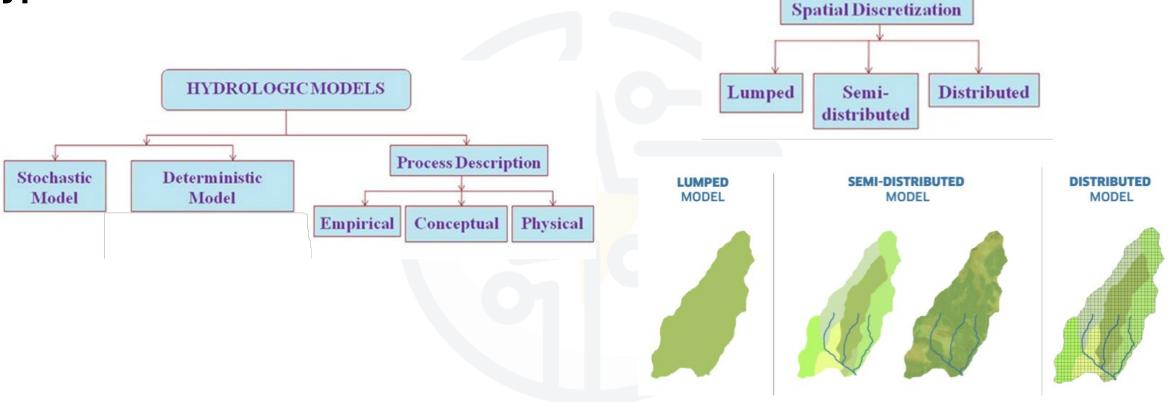






SCEWERO 2. Hydrological modelling

Type of models



Ochoa-Tocachi, Cuadros-Adriazola, Arapa, Aste, Ochoa-Tocachi, & Bonnesoeur. (2022). Guide to Hydrologic Modeling. Forest Trends Association. chrome-extension://efaidnbmnnnibpcajpcglclefindmkaj/https://www.forest-trends.org/wpcontent/uploads/2022/08/Guide-to-Hydrologic-Modeling-of-NI.pdf













SCEWERO 2. Hydrological modelling

The GR4J Hydrological model

Si
$$P_k \ge E$$
, alors $P_n = P_k - E$ et $E_n = 0$,
Si $P_k < E$, alors $P_n = 0$ et $E_n = E - P_k$

$$P_s = \frac{x1 \cdot \left(1 - \left(\frac{S_k}{X1}\right)^2\right) t anh\left(\frac{P_n}{X1}\right)}{1 + \frac{S_k}{X1} t anh\left(\frac{P_n}{X1}\right)} \text{ et } E_s = \frac{S_k \cdot \left(2 - \frac{S_k}{X1}\right) t anh\left(\frac{E_k}{X1}\right)}{1 + \left(1 - \frac{S_k}{X1}\right) t anh\left(\frac{E_k}{X1}\right)}$$

$$S' = S_k + P_s - E_s,$$

$$Perc = S' \left\{1 - \left[1 + \left(\frac{4}{9} \cdot \frac{S'}{X1}\right)^4\right]^{-\frac{1}{4}}\right\}, S_{k+1} = S' - Perc$$

$$Perc + (P_n - P_s)$$

$$Si \ 0 \le j \le X_4 \cdot SH1(j) = \left(\frac{j}{X_4}\right)^{\frac{1}{2}}; Si \ j > X_4 \cdot SH1(j) = 1$$

$$Si \ 0 \le j \le X_4 \cdot SH2(j) = \frac{1}{2} \left(\frac{j}{X_4}\right)^{\frac{1}{2}};$$

$$Si \ X_4 \le j \le 2X_4 \cdot SH2(j) = 1 - \frac{1}{2} \left(2 - \frac{j}{X_4}\right)^{\frac{1}{2}};$$

$$Si \ j > 2X_4 \cdot SH2(j) = 1$$

$$UH1(j) = SH1(j) - SH1(j-1)$$

$$UH2(j) = SH2(j) - SH2(j-1)$$

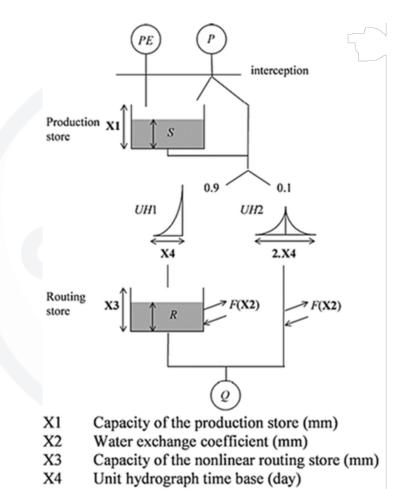
$$Q_1(k) = 0, 1 \cdot \sum_{j=1}^{m} UH1(j) \cdot Pr(k-j+1),$$

$$Q_1(k) = 0, 1 \cdot \sum_{j=1}^{m} UH2(j) \cdot Pr(k-j+1)$$

$$F = X2 \cdot \left(\frac{R_k}{X3}\right)^{\frac{7}{2}}; \quad R' = \max(0; R_k + Q_9(k) + F)$$

$$\left[Q_r = R' \cdot \left\{1 - \left[1 + \left(\frac{R'}{R_1}\right)^4\right]^{-\frac{1}{4}}\right\}; \quad R_{k+1} = R' - Q_r\right\}$$

O(k) = Or + Od



Perrin, C., Michel, C., & Andréassian, V. (2003). Improvement of a parsimonious model for streamflow simulation. *Journal of Hydrology*, *279*(1–4), 275–289. https://doi.org/10.1016/S0022-1694(03)00225-7







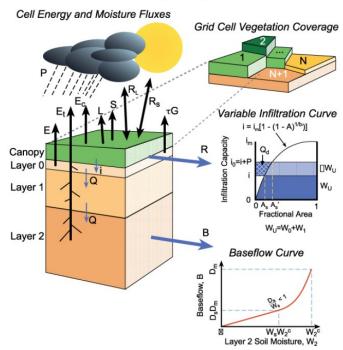




VIC-WUR

https://vicwur.readthedocs.io/en/latest/index_vicwur/

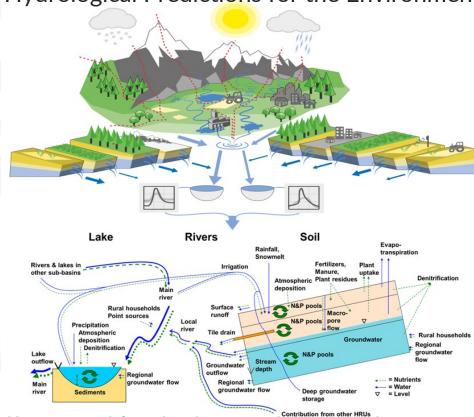
Variable Infiltration Capacity (VIC) Macroscale Hydrologic Model



https://www.wur.nl/en/research-results/chairgroups/environmental-sciences/earth-systems-and-globalchange-group/research/water-climate-food/vic-wur.htm

HYPE

The Hydrological Predictions for the Environment (HYPE)



https://www.smhi.se/en/research/researchunits/hydrology/hype---our-hydrological-model







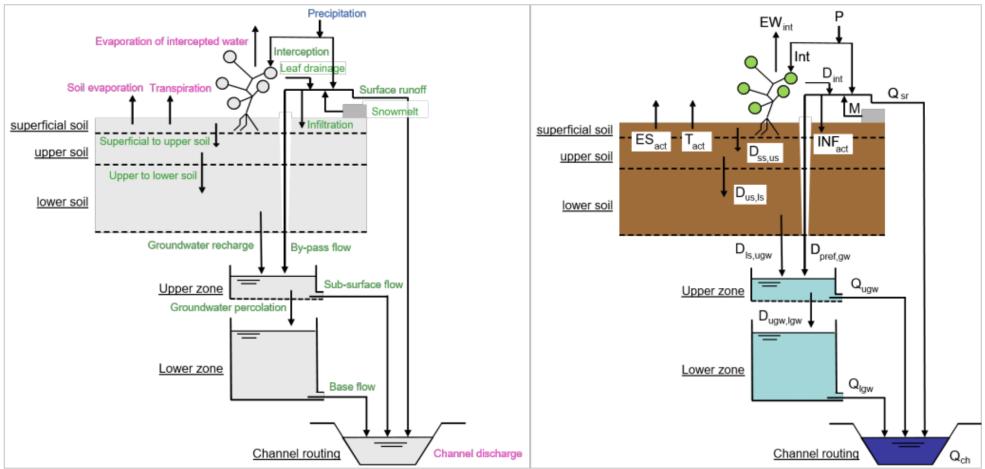






SCEWERO 2. Hydrological modelling

LISFLOOD



https://ec-jrc.github.io/lisflood-model/2_01_stdLISFLOOD_overview/





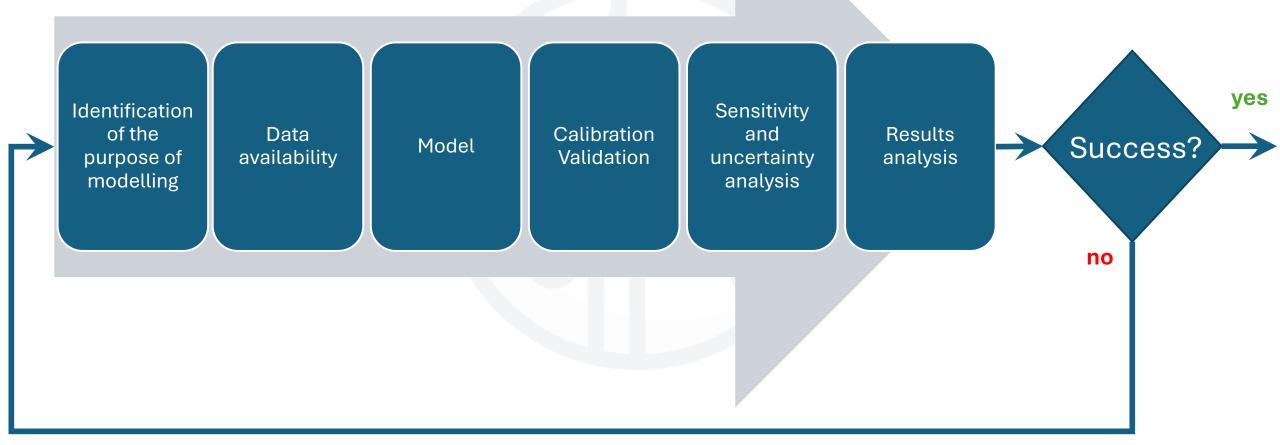






scewero 2. Hydrological modelling

Protocol







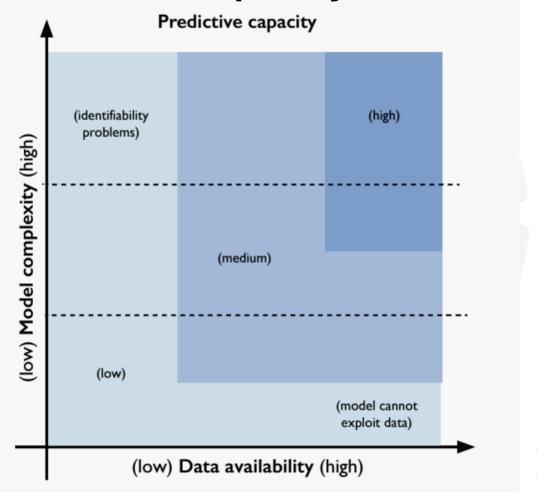






SCEWERO 2. Hydrological modelling

Data availability vs model complexity



Ochoa-Tocachi, Cuadros-Adriazola, Arapa, Aste, Ochoa-Tocachi, & Bonnesoeur. (2022). Guide to Hydrologic Modeling. Forest Trends Association. https://www.forest-trends.org/wp-content/uploads/2022/08/Guide-to-Hydrologic-Modeling-of-Nl.pdf



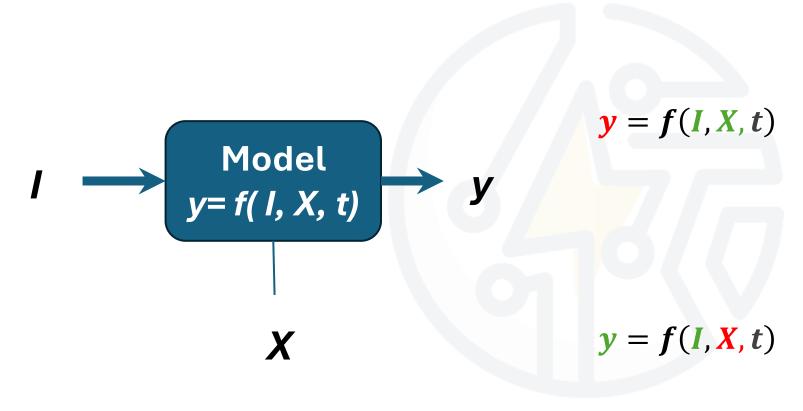








The inverse problem



We know X, and the inputs so we are interested to know the outputs

$$y = x_1 I + x_2$$

We know the inputs I, and the outputs y, so we are interested in the parameters.

$$y = x_1 I + x_2$$



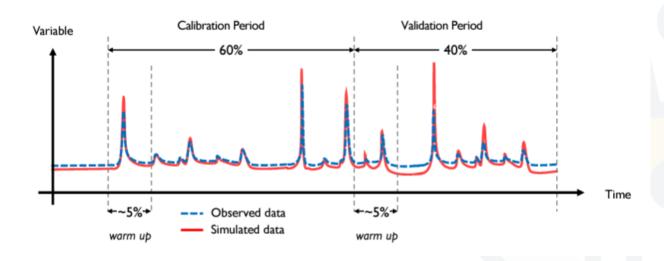








Calibration, validation and performance evaluation



Ochoa-Tocachi, Cuadros-Adriazola, Arapa, Aste, Ochoa-Tocachi, & Bonnesoeur. (2022). Guide to Hydrologic Modeling. Forest Trends Association. chromeextension://efaidnbmnnnibpcajpcglclefindmkaj/https://www.forest-trends.org/wp-content/uploads/2022/08/Guide-to-Hydrologic-Modeling-of-NI.pdf

D C	
Performance metric	Expression
Mean Absolute Error (MAE)	$F_1 = \frac{1}{N} \sum_{t=1}^{N} y_{s,t} - y_{o,t} $
Mean Square Error (MSE)	$F_2 = \frac{1}{N} \sum_{t=1}^{N} y_{s,t} - y_{o,t} ^2$
Root Mean Square Error (RMSE)	$F_3 = \left[\frac{1}{N}\sum_{t=1}^{N} y_{s,t}-y_{o,t} ^2\right]^{1/2}$
Minimax objective function	$F_4 = \frac{1}{N} \max y_{s,t} - y_{o,t} $
Average Absolute Percentage Error (AAPE)	$F_5 = 100 \frac{1}{N} \sum_{t=1}^{N} \left \frac{y_{s,t} - y_{o,t}}{y_{o,t}} \right $
Mean Square Relative Error (MSRE)	$F_6 = 100 \frac{1}{N} \sum_{t=1}^{N} \left \frac{y_{s,t} - y_{o,t}}{y_{o,t}} \right ^2$
Coefficient of determination (R^2)	$F_7 = \left\{ \frac{\sum_{t=1}^{N} (y_{o,t} - \overline{y_o}) (y_{s,t} - \overline{y_s})}{\left[\sum_{t=1}^{N} (y_{o,t} - \overline{y_o})^2 \right]^{0.5} \left[\sum_{t=1}^{N} (y_{s,t} - \overline{y_s})^2 \right]^{0.5}} \right\}^2$
Index of agreement (D)	$F_8 = 1 - \frac{\sum_{t=1}^{N} (y_{s,t} - y_{o,t})^2}{\sum_{t=1}^{N} (y_{s,t} - \overline{y_o} + y_{o,t} - \overline{y_o})^2}$
Nash-Sutcliffe Efficiency coefficient (NSE)	$F_9 = 1 - \frac{\sum_{t=1}^{N} (y_{s,t} - y_{o,t})^2}{\sum_{t=1}^{N} (y_{o,t} - \overline{y_o})^2}$

Biondi, D., Freni, G., Iacobellis, V., Mascaro, G., & Montanari, A. (2012). Validation of hydrological models: Conceptual basis, methodological approaches and a proposal for a code of practice. Physics and Chemistry of the Earth, Parts A/B/C, 42–44, 70–76. https://doi.org/10.1016/j.pce.2011.07.037

Modified Kling-Gupta efficiency $KGE' = 1 - \sqrt{(\gamma - 1)^2 + (\beta - 1)^2 + (r - 1)^2}$



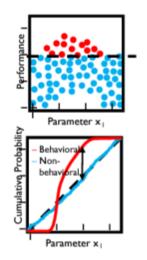


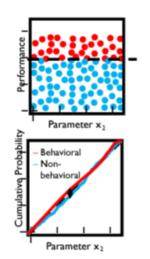


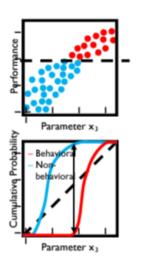


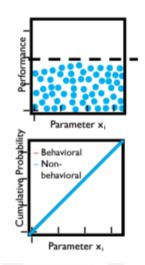


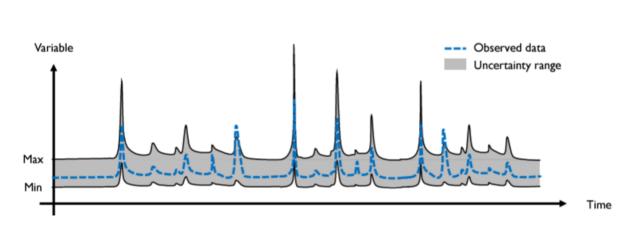
Sensitivity and uncertainty analysis











Ochoa-Tocachi, Cuadros-Adriazola, Arapa, Aste, Ochoa-Tocachi, & Bonnesoeur. (2022). Guide to Hydrologic Modeling. Forest Trends Association. chromeextension://efaidnbmnnnibpcajpcglclefindmkaj/https://www.forest-trends.org/wp-content/uploads/2022/08/Guide-to-Hydrologic-Modeling-of-NI.pdf







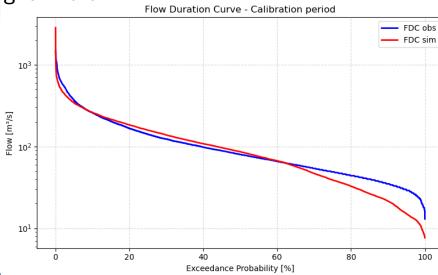




Flow duration curves

A cumulative frequency curve that shows the percentage of time that specified discharges are equaled or exceeded

- Arrange flows in chronological order
- Find the number of records (N)
- Sort the data from highest to lowest
- Rank the data (m=1 for the highest value and m=N for the lowest value)
- Compute exceedance probability for each value using the following formula
- Plot p on x axis and Q (sorted) on y axis















SCEWERO 2. Hydrological modelling

PDF fitting: Gumbel

Extreme Value Distribution: Extreme value distributions are widely used in hydrology. Storm rainfalls are most commonly modeled by the Extreme Value Type I distribution. Below is the "cumulative" probability distribution version of the function.

$$F(x) = e^{-e^{\left[\frac{x-u}{\alpha}\right]}} - \infty \le x \le \infty$$

$$\alpha = \frac{\sqrt{6}s_x}{\pi}$$

$$u = x - 0.5772\alpha$$

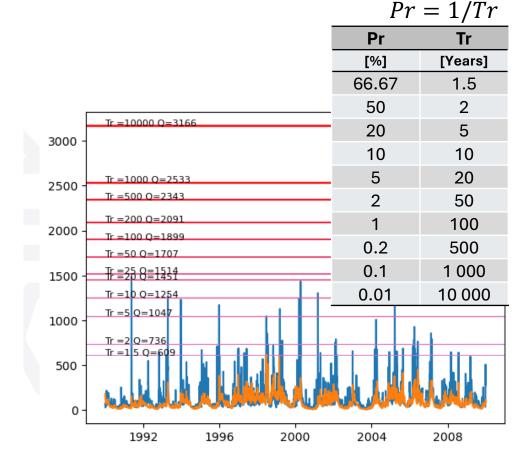
$$Let y = \frac{x-u}{\alpha}$$

 Correlating "y" with Return Period (T), the Extreme Value Distribution can be simplified to:

$$x_{T} = u + \alpha y_{T}$$

$$y_{T} = -\ln \left[\ln \left(\frac{T}{T - 1} \right) \right]$$

$$(x-u) / \alpha = -\ln(-\ln(F(x)))$$
$$x = -\ln(-\ln(F(x))) \cdot \alpha + u$$



 $\mu_{\rm y} = 0.5772$; $\sigma_{\rm y} = 1.2825$

$$\alpha = \underline{s_x} / \sigma_y$$

$$u = \overline{x} - \mu_y \cdot \alpha$$







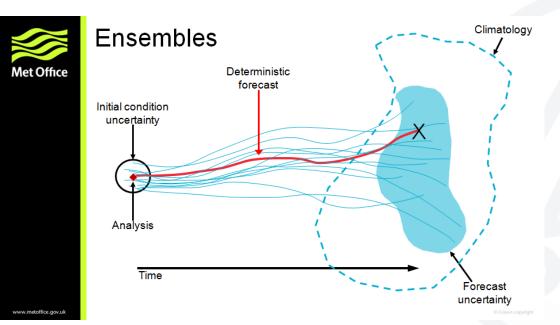




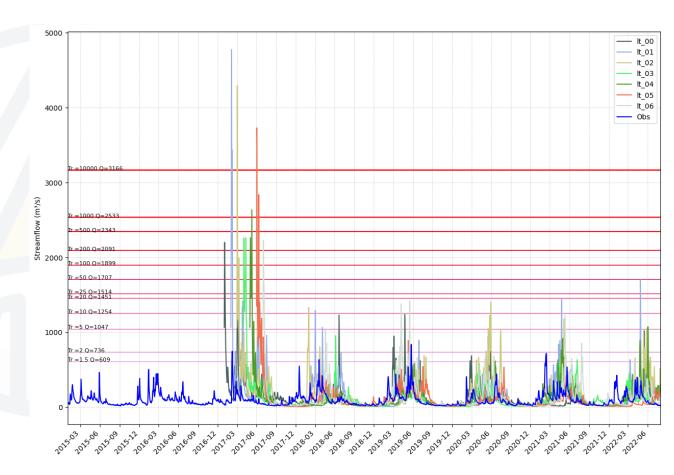


SCEWERO 2. Hydrological modelling

Ensemble forecasting



https://www.rmets.org/metmatters/ensembles-how-forecast-comes-together







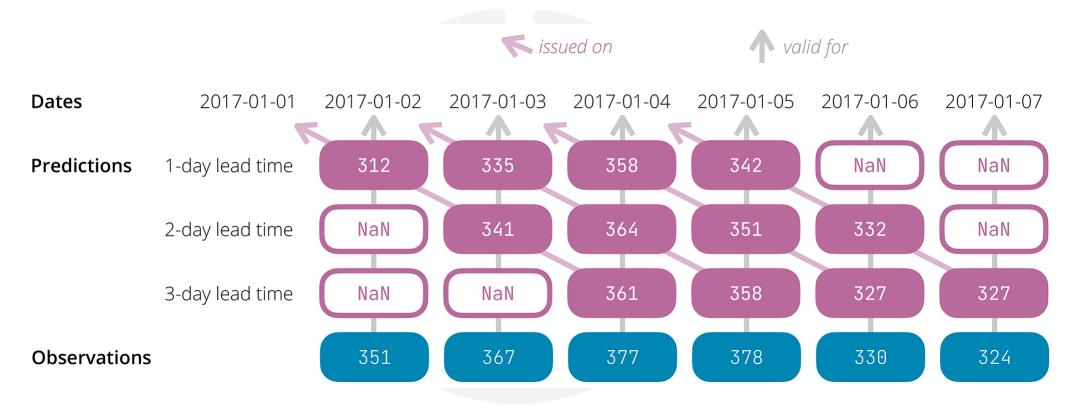








Understanding the lead time



Hallouin, T., Bourgin, F., Perrin, C., Ramos, M.-H., & Andréassian, V. (2024). EvalHyd v0.1.2: A polyglot tool for the evaluation of deterministic and probabilistic streamflow predictions. Geoscientific Model Development, 17(11), 4561-4578. https://doi.org/10.5194/gmd-17-4561-2024



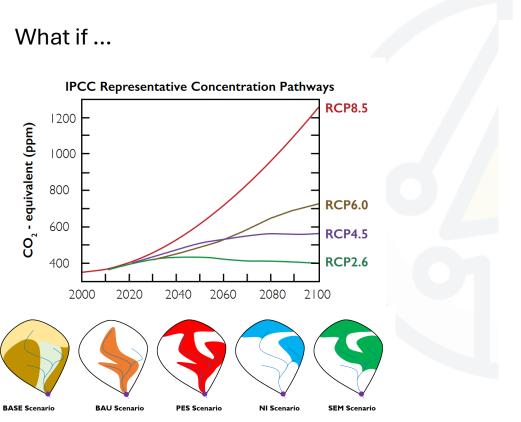


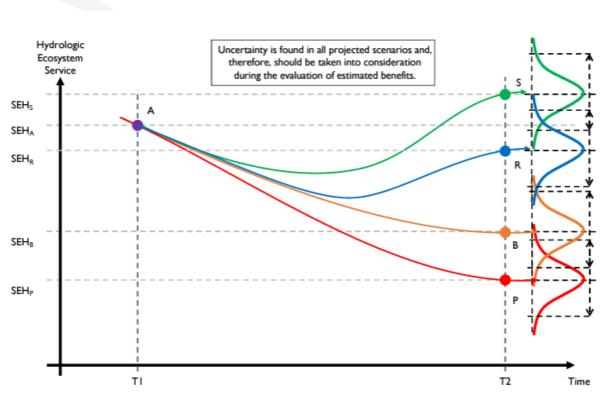






Results analysis





Ochoa-Tocachi, Cuadros-Adriazola, Arapa, Aste, Ochoa-Tocachi, & Bonnesoeur. (2022). Guide to Hydrologic Modeling. Forest Trends Association. https://www.forest-trends.org/wp-content/uploads/2022/08/Guide-to-Hydrologic-Modeling-of-NI.pdf













Performance metrics for ensemble forecasting

Brier Score (BS)

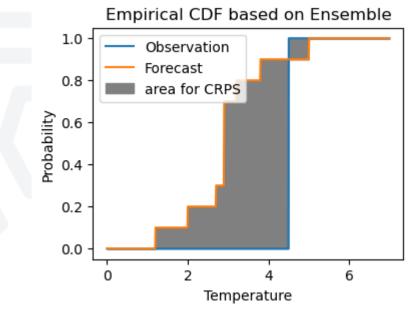
$$BS = \overline{(p-o)^2}$$

Brier Skill Score (BSS)

Rank Probability Scores (RPS)

Continuous Ranked Probability Scores (CRPS)

Continuous Ranked Probability Skill Score (CRPSS)



https://confluence.ecmwf.int/display/FUG/Section+12.B+Statistical+Concepts+-+Probabilistic+Data





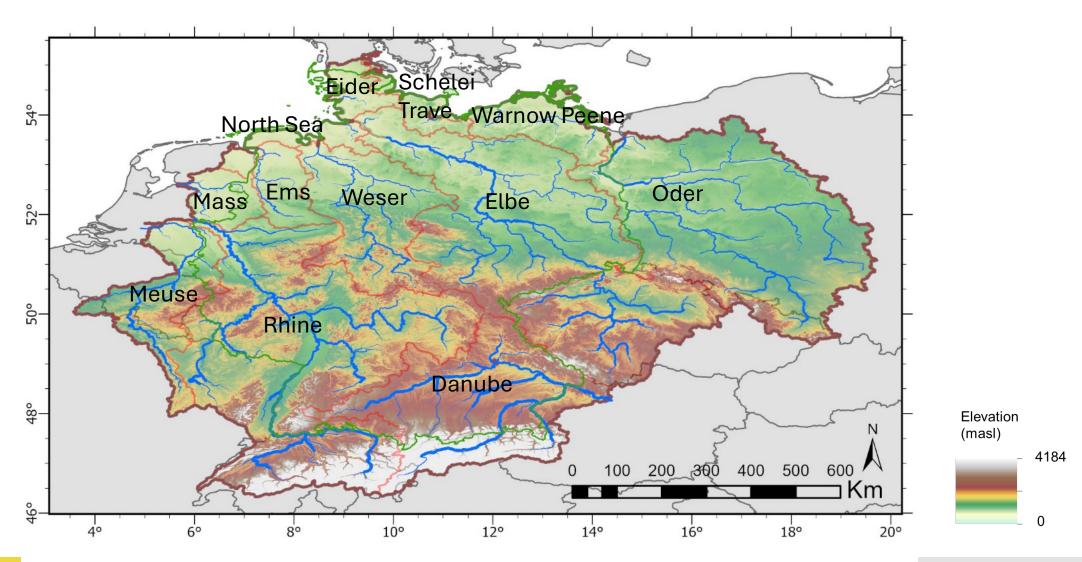








SCEWERO 3. Example – DAKI-FWS



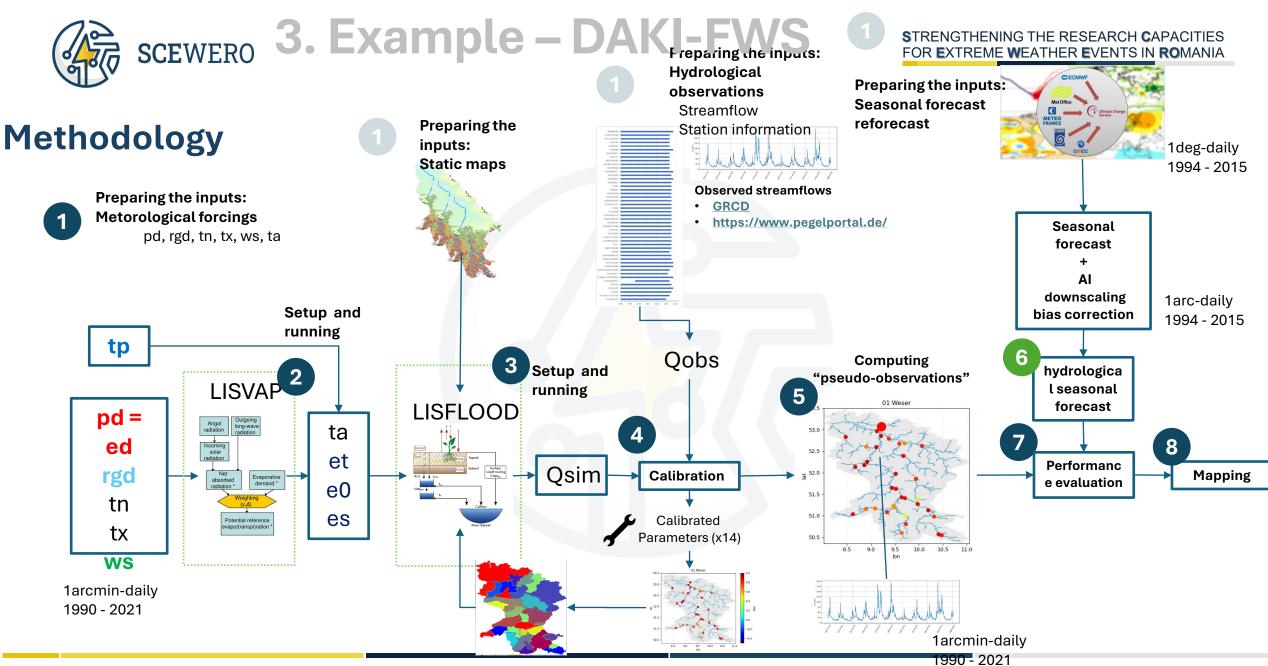


















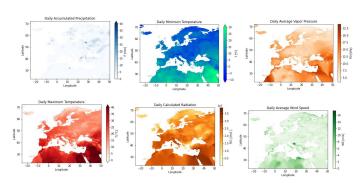






Meteorological forcing's

Observations

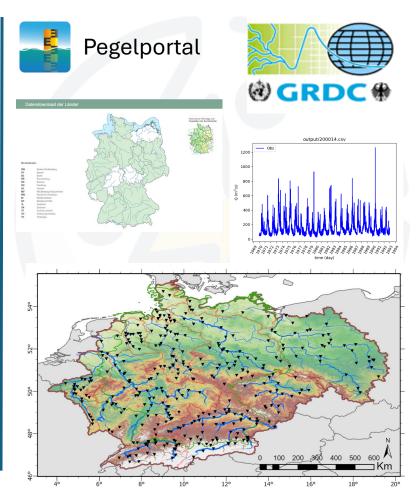


EMO - 1 Spatial resolution 1arcmin Time resolution daily Period 1990-2021

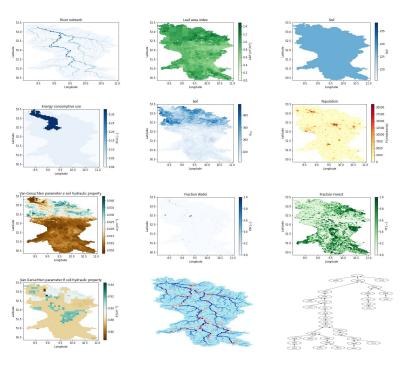
Seasonal forecast reforecast



Hydrological



Morphological, physical, soil, and land use properties



EFAS

Francesca Moschini; Margarita Choulga; Cinzia Mazzetti; Disperati, Juliana; Grimaldi, Stefania; Salamon, Peter; Prudhomme, Christel (2023): LISFLOOD static and parameter maps for Europe. European Commission, Joint Research Centre (JRC) [Dataset] PID: http://data.europa.eu/89h/f572c443-7466-4adf-87aa-c0847a169f23







EFAS





Pre-processor LISVAL

LISVAP Installed and tested in Levante HPC

 Installing/initializing git for large file git 1fs install

2. clone LISVAP from the respository

```
git clone https://github.com/ec-jrc/lisflood-lisvap.git
cd lisflood-lisvap
```

3. Creating a conda environment for lisvap

```
conda create --name lisvap python=3.8 --no-default-packages
source activate lisvap
```

4. Installing and checking lisvap

```
pip install -r requirements.txt
```

Installing and checking pcraster

```
conda install -c conda-forge pcraster
```

6. Running LISVAP from scr folder

```
python lisvap1.py /my system/lisflood-lisvap/tests/data/tests efas.xml -v -t
```

```
TIME-RELATED CONSTANTS
<textvar name="CalendarDayStart" value="01/01/1995 00:00">
       calendar day number of 1st day in model run
       e.g. 1st of January: 1; 1st of June: 151 (or 152 in leap year)
<textvar name="DtSec" value="86400">
       time step [seconds] ALWAYS USE 86400!!
<textvar name="StepStart" value="01/01/1995 06:00">
       Number of first time step in simulation
<textvar name="StepEnd" value="31/12/1995 06:00">
       Number of last time step
   </comment>
<textvar name="ReportSteps" value="1..14">
```











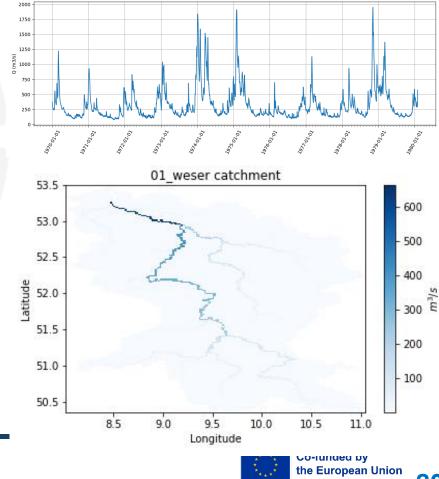
Hydrological Model LISFLOOD

LISFLOOD, LISFLOOD-Utilities and LISFLOOD-Calibration tools installed and tested in Levante HPC.

(Working in serie and in parallel).

```
# 2023-05-05 Installing Lisflood Calibration tool
# installation for solving the multiprocessing problems attemp 1
module load python3
module load git
conda create --name lisflood calibration 2 python=3.7 -c conda-forge
git clone --single-branch https://github.com/ec-jrc/lisflood-calibration.git
source activate lisflood calibration 2
conda install -c conda-forge pcraster gdal
pip install lisflood-model==4.1.2
# pip show lisflood-model # allows you to know where is the source code of lisflo
# /work/bb1245/hydromodelling/conda venvs/lisflood calibration 2/lib/python3.7/si
# After installing lisflood compile the cython module kinematic wave parallel to
# To compile this Cython module to enable OpenMP multithreading (parallel kinemat
# Delete the files *.so (if any) in directory hydrological-modules
# Inside the hydrological modules folder, execute "python compile kinematic wave
# For get rid of the errors in related with hdf5
conda remove hdf5
conda install -c conda-forge hdf5==1.12.1
# other tools
pip install pandas
pip install numpy
pip install matplotlib
pip install netCDF4
pip install deap
conda install -c conda-forge deap
pip install -r requirements.txt # run it in the lisflood code folder
pip install . # run it in the lisflood calibration folder
pip install tbb # For parallelization
```

```
<lfsettings>
   <setoption choice="1" name="gridSizeUserDefined"/>
       # options to turn hydrological modules on/off
       <setoption choice="0" name="cropsEPIC"/>
       <setoption choice="1" name="wateruse"/>
       <setoption choice="1" name="TransientWaterDemandChange"/>
       <setoption choice="0" name="useWaterDemandAveYear"/>
       <setoption choice="1" name="wateruseRegion"/>
       <setoption choice="0" name="groundwaterSmooth"/>
       <setoption choice="1" name="riceIrrigation"/>
       <setoption choice="0" name="indicator"/>
       <setoption choice="1" name="openwaterevapo"/>
       <setoption choice="0" name="simulateLakes"/>
       <setoption choice="0" name="simulateReservoirs"/>
       <setoption choice="0" name="simulateReservoirs"/>
       <setoption choice="1" name="SplitRouting"/>
       <setoption choice="0" name="inflow"/>
       # option to initialize Lisflood
```



Calibration - Validation

Non-dominated Sorting Genetic Algorithm-II (NSGA-II)

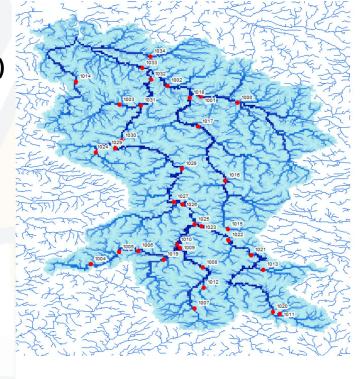
Modified Kling-Gupta Efficiency

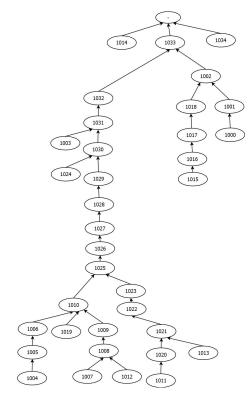
$$\mathit{KGE}$$
ʻ $=1-\sqrt{\left(r-1
ight)^2+\left(eta-1
ight)^2+\left(\gamma-1
ight)^2}$

$$r_{xy} = rac{\sum_{i} x_{i} y_{i} - n ar{x} ar{y}}{\sqrt{\sum_{i} x_{i}^{2} - n ar{x}^{2}} \, \sqrt{\sum_{i} y_{i}^{2} - n ar{y}^{2}}}$$

$$eta = rac{\mu_s}{\mu_o} \qquad \quad \gamma = rac{\sigma_s/\mu_s}{\sigma_o/\mu_o}$$

KGE'	Description
1.0 - 0.8	Very good
0.8 - 0.6	Good
0.6 - 0.4	Medium
0.4 - 0.2	Poor
< 0.2	Bad





01_wesser









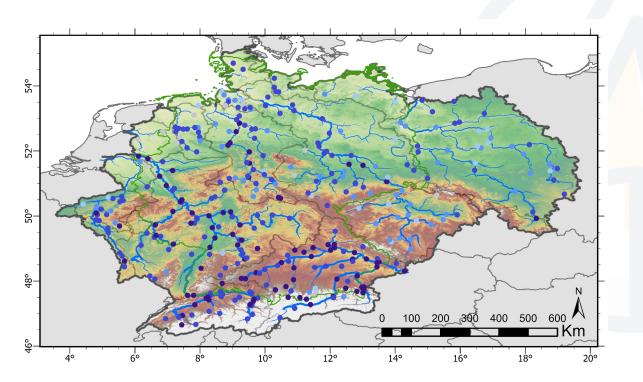


scewero 3. Example – DAKI-FWS

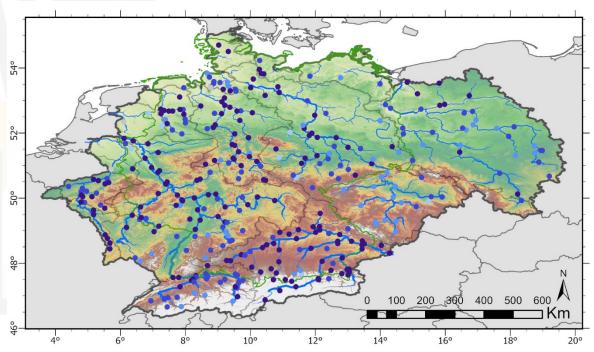
Calibration - Validation

KGE' Description 1.0 - 0.8 Excellent 0.8 - 0.6 Good 0.6 - 0.4 Acceptable 0.4 - 0.2 Low very low

Calibration



Validation







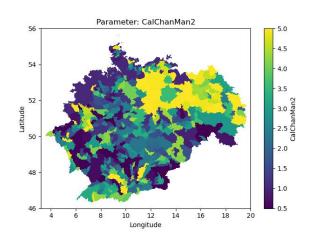


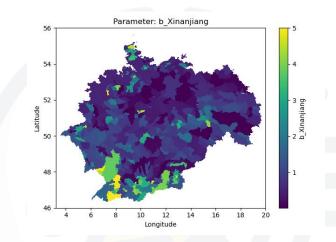


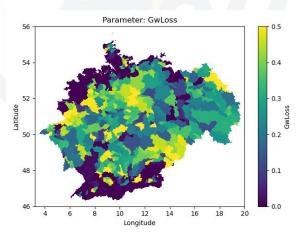


Parameters maps for the study region

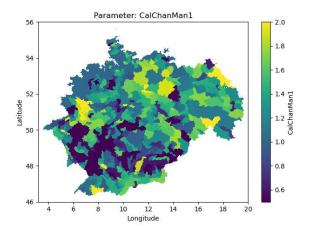
Parameter: adjust Normal Flood 14

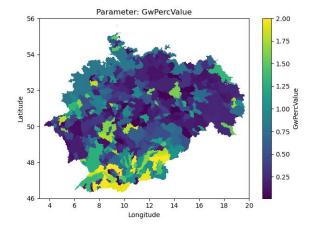






... a total of 14 maps















total_bill

scewero 3. Example – DAKI-FWS

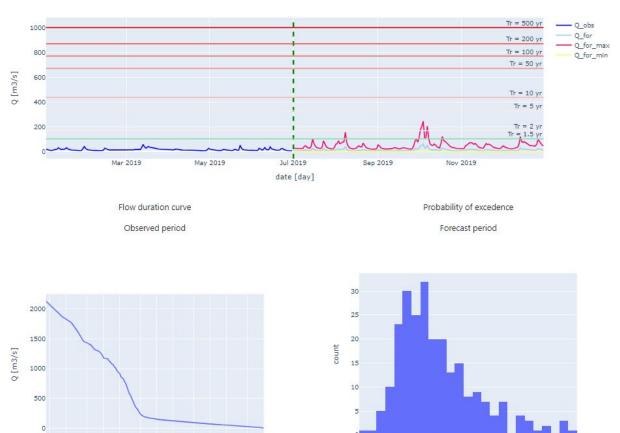
Dashboard

Hydrological seasonal forecast DAKI-FWS

Catchments of interest

Hydrograph







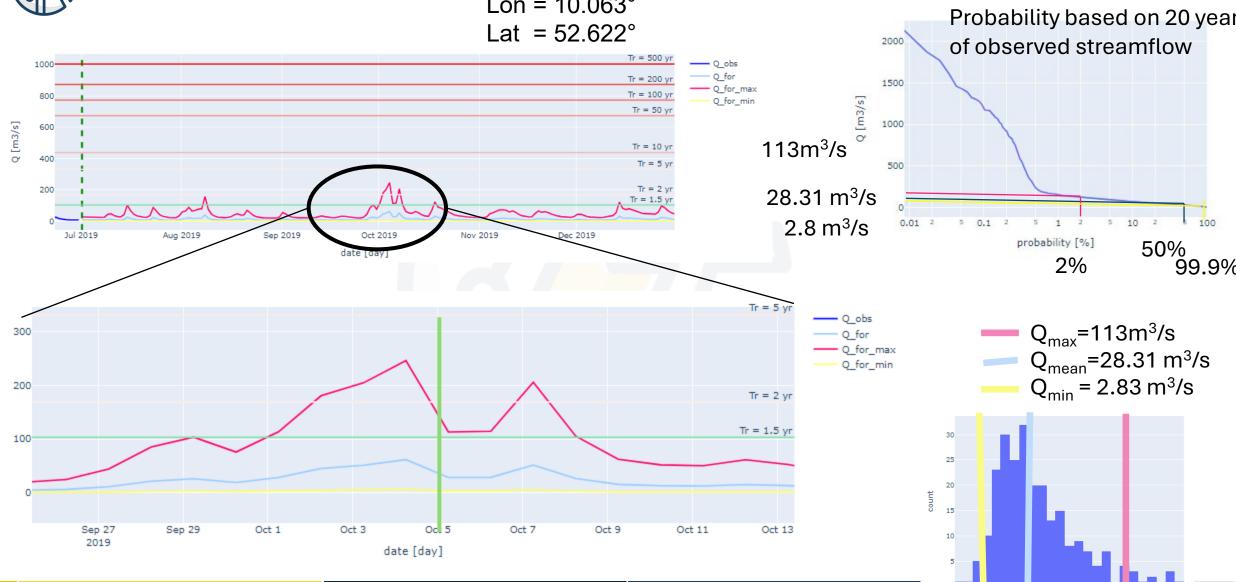
Selected catchment: None

Information about the calibration

(12.47, 65.30) means pr(Q>65.30 m3/s) = 12.78 %

probability [%]

(p, q) means pr(Q>q)=p





Q [m3/s]



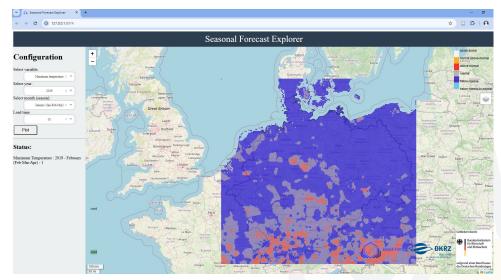




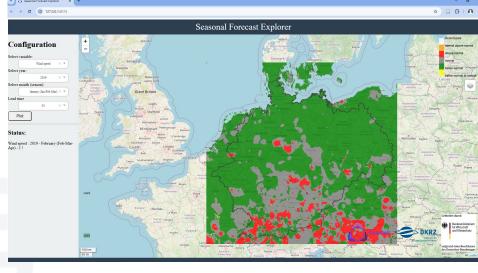


total_bill

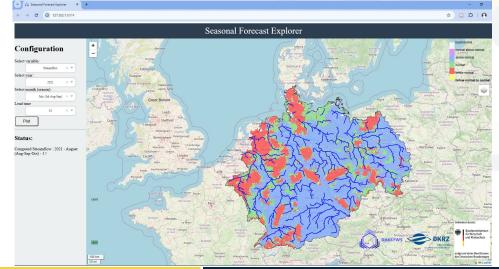
Temperature



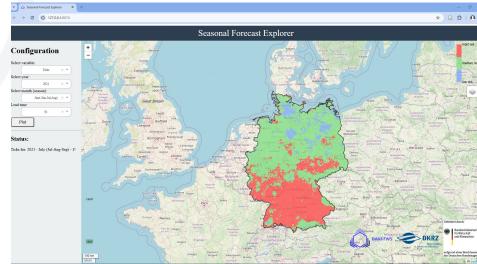
Wind speed



Streamflows



Ticks











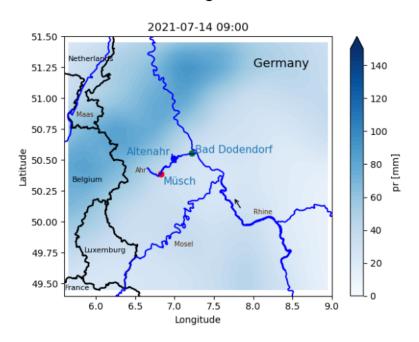


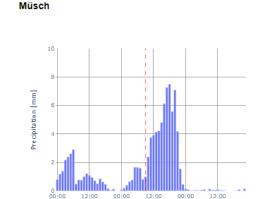
Bad Dodendorf

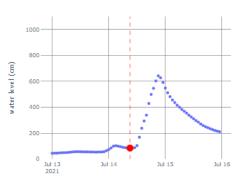
scewero 3. Example – DAKI-FWS

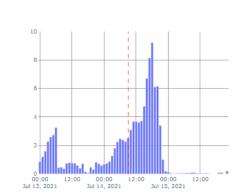
Propagation of the Ahrtal flood in 2021

Cumulative rainfall in the region:

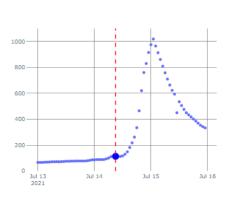


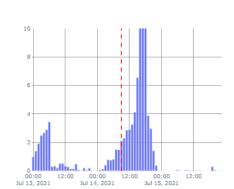


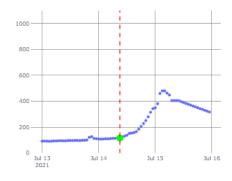




Altenahr





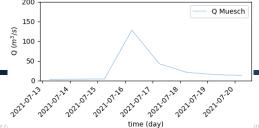


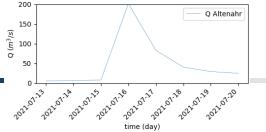
Observed precipitation and water level

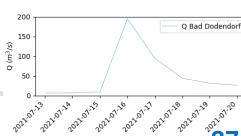










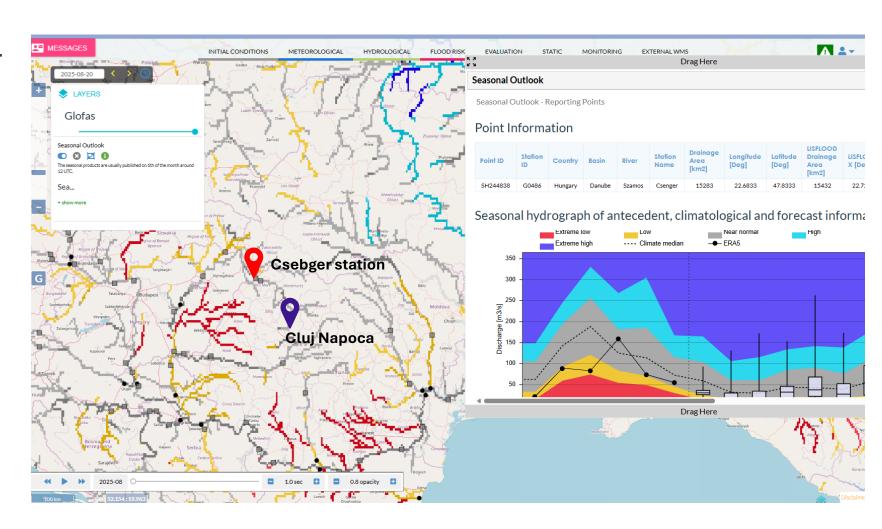


Global Flood Awareness System - GLOFAS

https://global-flood.emergency.copernicus.eu/

European Flood Awareness System - EFAS

https://european-flood.emergency.copernicus.eu/en









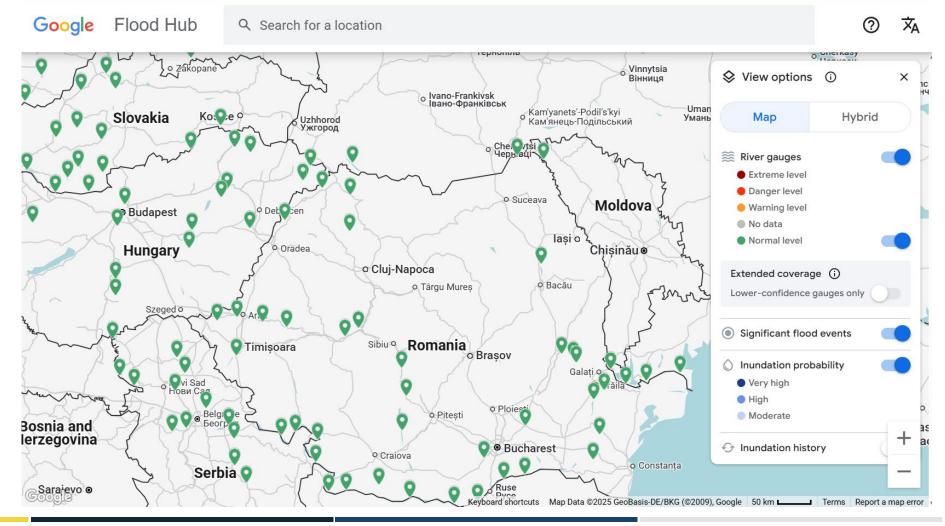




AI - Google Flood Hub

https://g.co/floodhub

Nearing, G., Cohen, D., Dube, V. *et al.* Global prediction of extreme floods in ungauged watersheds. *Nature* **627**, 559–563 (2024). https://doi.org/10.1038/s41586-024-07145-1













Hydrological cycle
Hydrological modelling
Models
Protocol
Application













Thank you!













Deb, K., Pratap, A., Agarwal, S., & Meyarivan, T. A. M. T. (2002). A fast and elitist multiobjective genetic algorithm: NSGA-II. IEEE transactions on evolutionary computation, 6(2), 182-197.

Kratzert, Frederik, Daniel Klotz, Sepp Hochreiter, and Grey S. Nearing. "A Note on Leveraging Synergy in Multiple Meteorological Data Sets with Deep Learning for Rainfall–Runoff Modeling." *Hydrology and Earth System Sciences* 25, no. 5 (May 20, 2021): 2685–2703. https://doi.org/10.5194/hess-25-2685-2021.

Yang, Shuyu, Dawen Yang, Jinsong Chen, Jerasorn Santisirisomboon, Weiwei Lu, and Baoxu Zhao. "A Physical Process and Machine Learning Combined Hydrological Model for Daily Streamflow Simulations of Large Watersheds with Limited Observation Data." *Journal of Hydrology* 590 (November 2020): 125206. https://doi.org/10.1016/j.jhydrol.2020.125206.

Besaw, Lance E., Donna M. Rizzo, Paul R. Bierman, and William R. Hackett. "Advances in Ungauged Streamflow Prediction Using Artificial Neural Networks." *Journal of Hydrology* 386, no. 1–4 (May 2010): 27–37. https://doi.org/10.1016/j.jhydrol.2010.02.037.

Lees, Thomas, Marcus Buechel, Bailey Anderson, Louise Slater, Steven Reece, Gemma Coxon, and Simon J. Dadson. "Benchmarking Data-Driven Rainfall–Runoff Models in Great Britain: A Comparison of Long Short-Term Memory (LSTM)-Based Models with Four Lumped Conceptual Models." *Hydrology and Earth System Sciences* 25, no. 10 (October 21, 2021): 5517–34. https://doi.org/10.5194/hess-25-5517-2021.

Kratzert, Frederik, Daniel Klotz, Claire Brenner, Karsten Schulz, and Mathew Herrnegger. "Rainfall–Runoff Modelling Using Long Short-Term Memory (LSTM) Networks." *Hydrology and Earth System Sciences* 22, no. 11 (November 22, 2018): 6005–22. https://doi.org/10.5194/hess-22-6005-2018.

Hu, Youchuan, Le Yan, Tingting Hang, and Jun Feng. "Stream-Flow Forecasting of Small Rivers Based on LSTM," 2020. https://doi.org/10.48550/ARXIV.2001.05681.

Ochoa-Tocachi, Cuadros-Adriazola, Arapa, Aste, Ochoa-Tocachi, & Bonnesoeur. (2022). Guide to Hydrologic Modeling. Forest Trends Association. https://www.forest-trends.org/wp-content/uploads/2022/08/Guide-to-Hydrologic-Modeling-of-NI.pdf

Biondi, D., Freni, G., Iacobellis, V., Mascaro, G., & Montanari, A. (2012). Validation of hydrological models: Conceptual basis, methodological approaches and a proposal for a code of practice. Physics and Chemistry of the Earth, Parts A/B/C, 42–44, 70–76. https://doi.org/10.1016/j.pce.2011.07.037













Hands on: Hydrological modelling









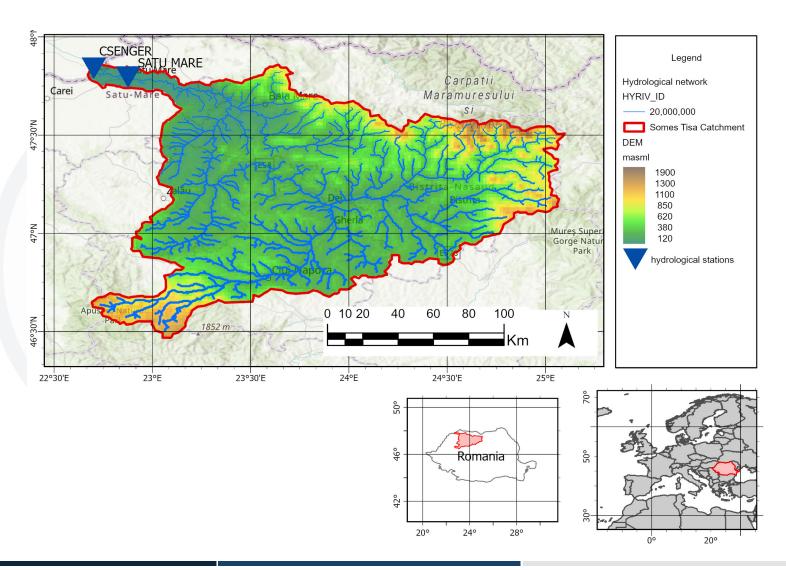


SCEWERO 1. Define model goals

Somes-Tisa Watershed Area

Somes-Tisa watershed area is located to the North and North-Western part of the country, being delineated to the North by the natural border – Tisa river with Ukraine on a length of 61 km, to the West by the border with the Hungarian Republic, and on the national territory, it borders on Siret Basin to the East, Mures Basin to the South and Crisuri Basin to the South-West.

https://inundatii.ro/en/river-basins/somes-tisa-watershed-area/













1. Define model for decision-making

Flood records

The maximum quantities of precipitations are recorded during spring-summer, in particular during the period March-June and in August. One can remember several maximum historical values of precipitations occurred during serious floods across the entire river basin as those recorded in May 1970 (over 150 mm/24 hours and even 200 mm/24 hours in Baia Sprie).

Serious floods, with significant maximum discharges, come from rainfalls recorded on large areas of the basin or from mixed source (snowmelt and rain). The maximum historical flood over the last 50-60 years was that occurred in 1970, having a mixed source. An analysis of the series of floods over around the last 50 years shows that the most significant floods occurred in: 1970, 1974, 1975, 1978, 1979, 1980, 1981, 1989, 1993, 1995, 1998, 2000, 2001, 2005 and 2008.

The first half of 1970 was characterized by catastrophic floods. The most serious floods occurred on the rivers Somes, Tisa, Tur, Viseu, Iza, Lapus, Sieu, Crasna, Almas, Somesul Mic and their tributaries. Discharges exceeding the insurance of 1% were recorded on the rivers Somes (downstream Dej) and Viseu, on the other water courses, there were discharges recorded with the insurance ranging between 5- 10%. There was a breach of dikes on rivers Somes and Tur (Satu Mare county), which caused the flooding of many urban and rural localities. Floods were caused by long-lasting heavy rainfalls accompanied by snowmelt resulting in floods that exceeded the flood level, on most rivers in Somes-Tisa river basin. Important urban localities were affected, such as Nasaud, Beclean, Gherla, Dej, Zalau, Jibou, Simleu, Satu Mare, as well as many rural localities (over 200).

https://inundatii.ro/en/river-basins/somes-tisa-watershed-area/













2. Implementing, validating and analyzing models

Floods

Meteorological forcings

https://jeodpp.jrc.ec.europa.eu/ftp/jrc-

opendata/CEMS-

EFAS/meteorological_forcings/EMO-

1arcmin/

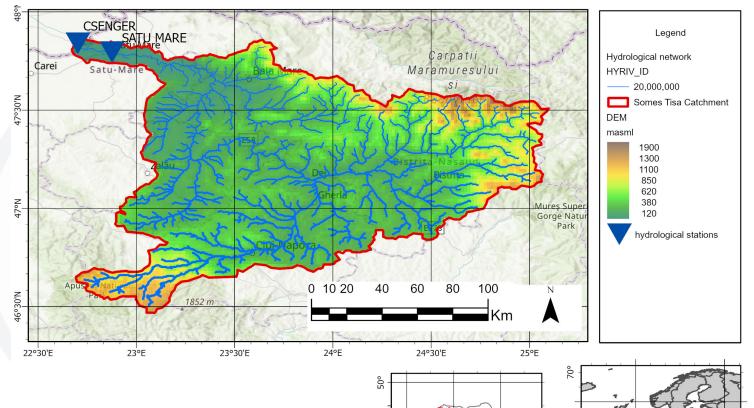
Discharge

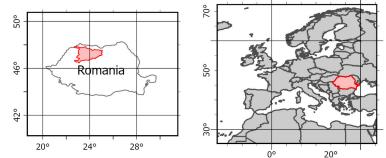
https://portal.grdc.bafg.de/applications/p ublic.html?publicuser=PublicUser#dataD ownload/Stations

Geographical information DEM 30 meter resolution 1arcmin EFAS

https://jeodpp.jrc.ec.europa.eu/ftp/jrc-opendata/CEMS-EFAS/LISFLOOD_static_and_parameter_maps_for_EFAS/

https://inundatii.ro/en/river-basins/somes-tisa-watershed-a











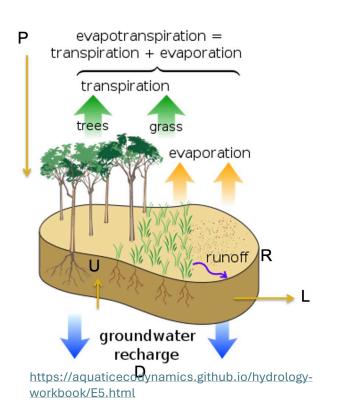


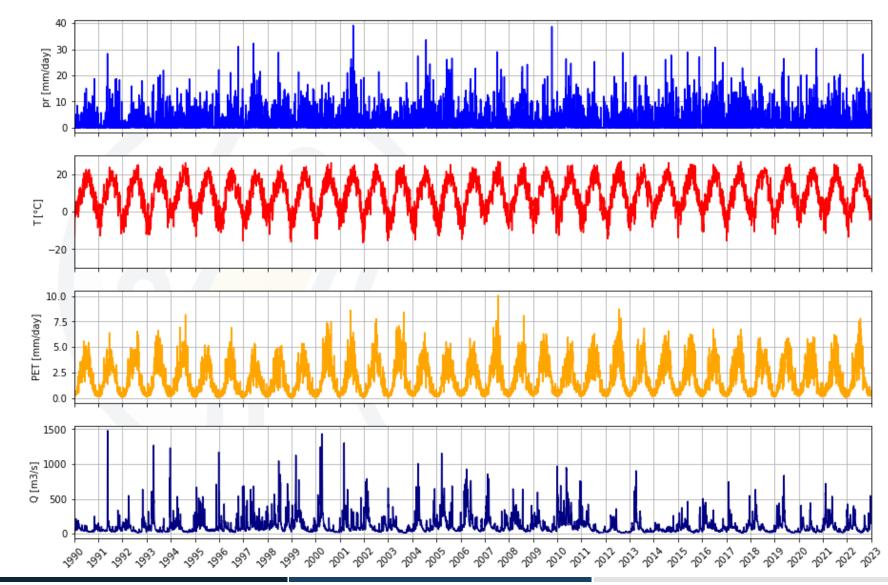




1. Data preparation

01_data_preparation_example

















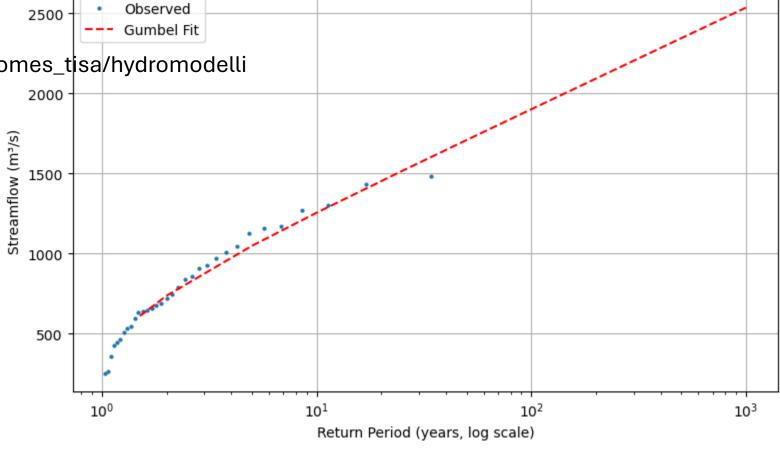
Extreme events

Return Period Analysis using Gumbel Distribution

/work/bb1480/b381840/scewero/somes_tisa/hydromodelling/somes_tisa_extrems.ipynb

Gumbel D Return pe	istribution riod	Streamflow
Tr	Q	
[Years]	[m³/s]	
1.5	610	
2	736	
5	1048	
10	1254	
25	1515	
50	1708	
100	1900	
200	2091	
500	2343	

2534





1000





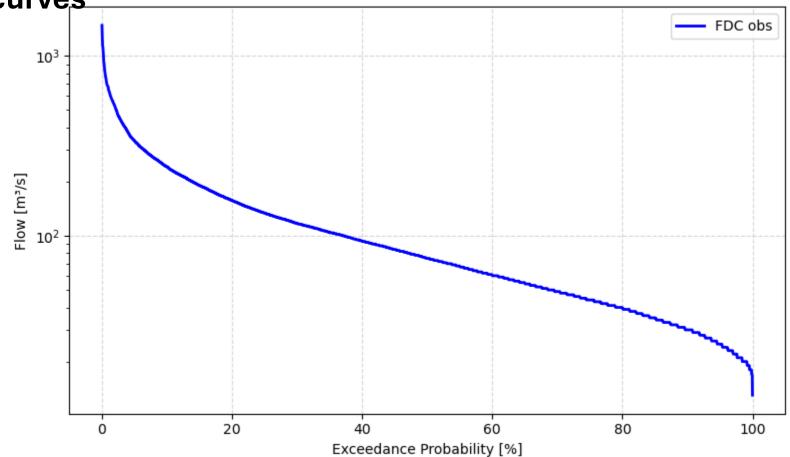






Flow duration curves

















Hydrological model

GR4J

Si
$$P_k \ge E$$
, alors $P_n = P_k - E$ et $E_n = 0$,
Si $P_k \le E$, alors $P_n = 0$ et $E_n = E - P_k$

$$Ps = \frac{X1 \cdot \left(1 - \left(\frac{S_k}{X1}\right)^2\right) \cdot \tanh\left(\frac{Pn}{X1}\right)}{1 + \frac{S_k}{X1} \cdot \tanh\left(\frac{Pn}{X1}\right)} \text{ et } Es = \frac{S_k \cdot \left(2 - \frac{S_k}{X1}\right) \cdot \tanh\left(\frac{E_k}{X1}\right)}{1 + \left(1 - \frac{S_k}{X1}\right) \cdot \tanh\left(\frac{E_k}{X1}\right)}$$

S'=
$$S_k+Ps-Es$$
,
 $Perc = S' \left\{ 1 - \left[1 + \left(\frac{4}{9} \cdot \frac{S'}{X1} \right)^4 \right]^{-\frac{1}{4}} \right\}, S_{k+1} = S'-Perc$

Perc + (Pn - Ps)

Si
$$0 \le j \le X_4$$
. SH1 $(j) = \left(\frac{j}{X_1}\right)^{\frac{5}{2}}$; Si $j > X_4$. SH1 $(j) = 1$

Si
$$0 \le j \le X_4$$
, $SH2(j) = \frac{1}{2} \left(\frac{j}{X_4}\right)^{\frac{1}{2}}$;

Si
$$X_4 \le j \le 2X_4$$
. $SH2(j) = 1 - \frac{1}{2} \left(2 - \frac{j}{X_4}\right)^{\frac{5}{2}}$;
Si $j > 2X_4$. $SH2(j) = 1$

$$UH1(j) = SH1(j) - SH1(j-1)$$

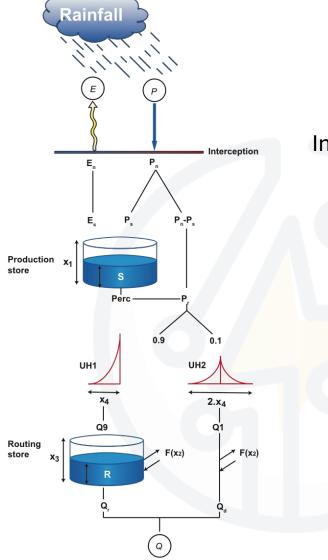
 $UH2(j) = SH2(j) - SH2(j-1)$

$$Q9(k) = 0.9.\sum_{j=1}^{l} UH1(j).Pr(k-j+1),$$

$$Q1(k) = 0.1. \sum_{j=1}^{m} UH2(j).Pr(k-j+1)$$

$$F = X2. \left(\frac{R_k}{X3}\right)^{\frac{1}{2}}; \quad R = \max(0; R_k + Q9(k) + F)$$

(DP)
$$\begin{cases} Qr = R' \cdot \left\{ 1 - \left[1 + \left(\frac{R'}{X3} \right)^4 \right]^{-\frac{1}{4}} \right\}; & R_{k+1} = R' - Qt \\ Qd = max(0; Q1(k) + F); \\ Q(k) = Qr + Qd \end{cases}$$



Perrin, C., Michel, C., & Andréassian, V. (2003). Improvement of a parsimonious model for streamflow simulation. Journal of Hydrology, 279(1-4), 275-289. https://doi.org/10.1016/S0022-1694(03)00225-7

Implementation in python

```
import numpy as np
This class modelate a proccess rainfall-runoff
Perrin, C., C. Michel, et al. (2003). "Improvement of a parsimonious model for streamflow simul
http://www.cemagref.fr/webgr/Modelesgb/gr4j/fonctionnement_gr4jgb.htm
class GR4J:
    'Class hydrological model GR4J'
    def __init__(self, PEQ, A, X, S0, R0):
        self.set_PEQ(PEQ)
        self.a = np.copy(A) # Area of basin in km2
        n = PEQ.shape[0] # Get the lenght of data series
        self.X = np.copy(X) # Parameters of model
        self.Pn = np.zeros((n))
        self.En = np.zeros((n))
        self.R_ = np.copy(R0)
        self.S = np.copy(S0)
    # Define the imputs
    def set PEQ(self, PEQ):
        self.P = np.copy(PEQ[:,1])
        self.E = np.copy(PEQ[:,2])
        self.Q = np.copy(PEQ[:,3])
```







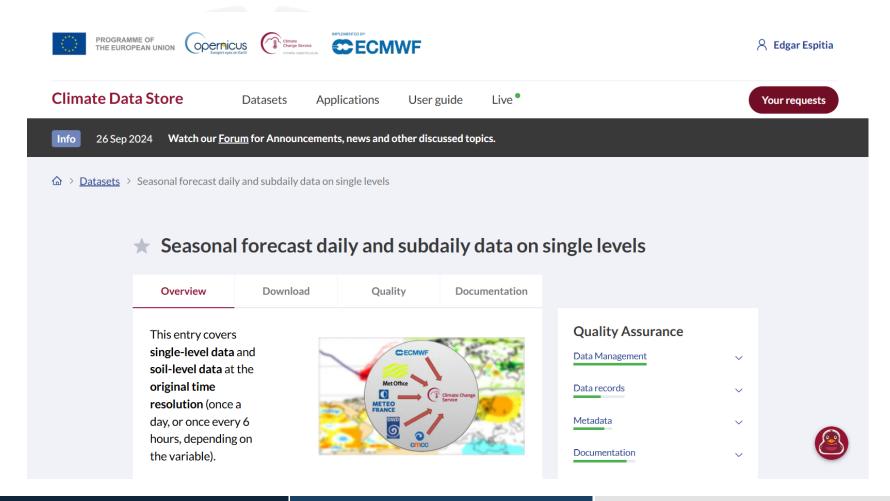






https://cds.climate.copernicus.eu/datasets/seasonal-original-single-levels?tab=overview

Meteorological forcings





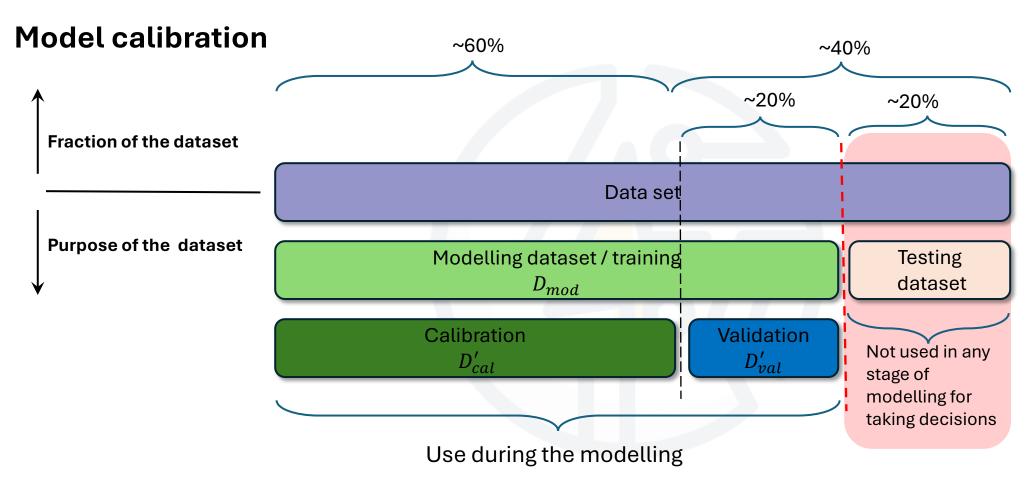












% based on modele experience

Based on: Shen, H., Tolson, B. A., & Mai, J. (2022). Time to Update the Split-Sample Approach in Hydrological Model Calibration. *Water Resources Research*, 58(3), e2021WR031523. https://doi.org/10.1029/2021WR031523













Model calibration

$$KGE^{\varsigma} = 1 - \sqrt{\left(r-1
ight)^2 + \left(eta-1
ight)^2 + \left(\gamma-1
ight)^2}$$

$$r_{OS} = rac{\sum_{i=1}^{n} \left(O_i - ar{O}
ight) \left(S_i - ar{S}
ight)}{\sqrt{\sum_{i=1}^{n} \left(O_i - ar{O}
ight)^2} \, \sqrt{\sum_{i=1}^{n} \left(S_i - ar{S}
ight)^2}}$$

bias ratio

$$\beta = \frac{\mu_s}{\mu_o}$$

Variability ratio

$$\gamma = rac{\sigma_s/\mu_s}{\sigma_o/\mu_o}$$

	KGE'	Description
	0.8 - 1.0	Excellent
١	0.6 - 0.8	Good
١	0.4 - 0.6	Aceptable
1	0.2 - 0.4	Low
	< 0.2	Very low

r
0.8 - 1.0
0.6 - 0.8
0.4 - 0.6
0.2 - 0.4
< 0.2

bias ratio β
> 1.5
1.3 - 1.5
1.1 - 1.3
0.9 - 1.1
0.7 - 0.9
0.5 - 0.7
< 0.5

variability ratio γ
> 1.5
1.3 - 1.5
1.1 - 1.3
0.9 - 1.1
0.7 - 0.9
0.5 - 0.7
< 0.5





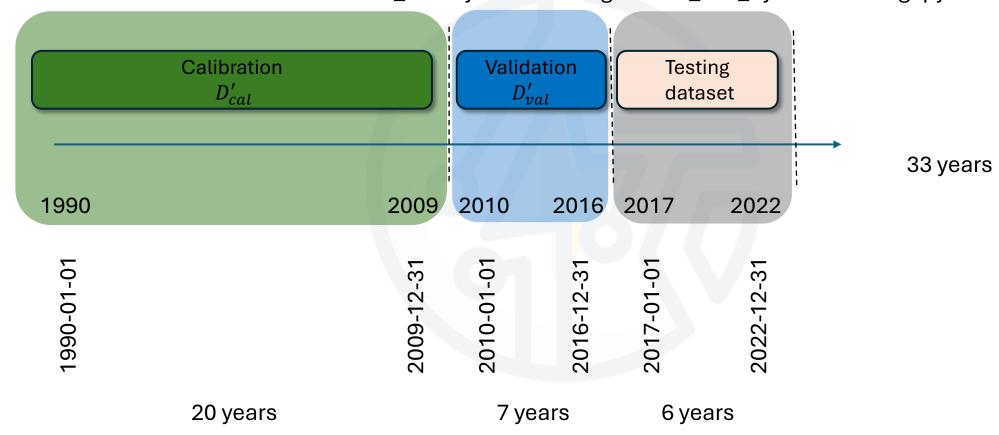






Model calibration

/work/bb1480/b381840/scewero/somes_tisa/hydromodelling/somes_tisa_hydromodelling.ipynb





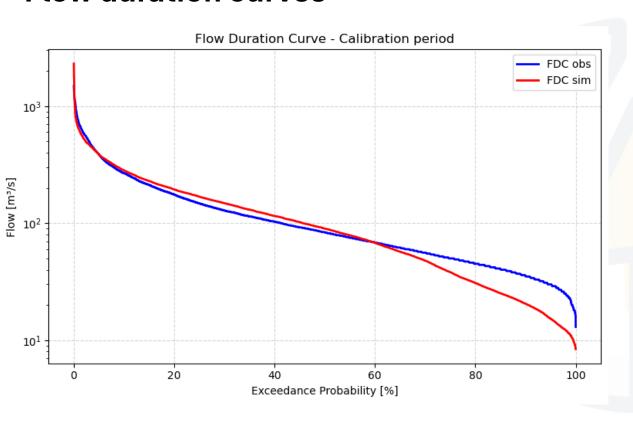


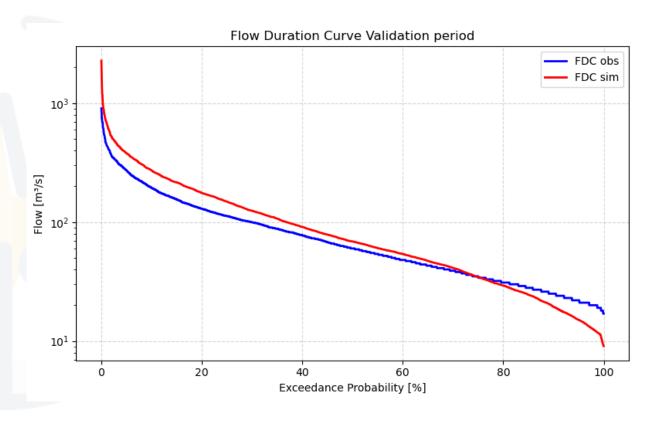






Flow duration curves









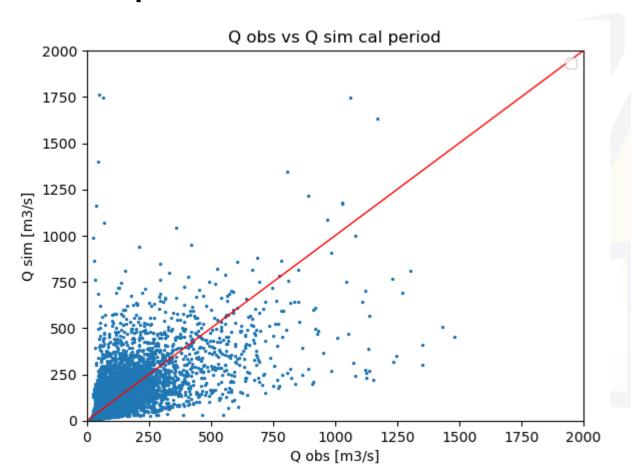


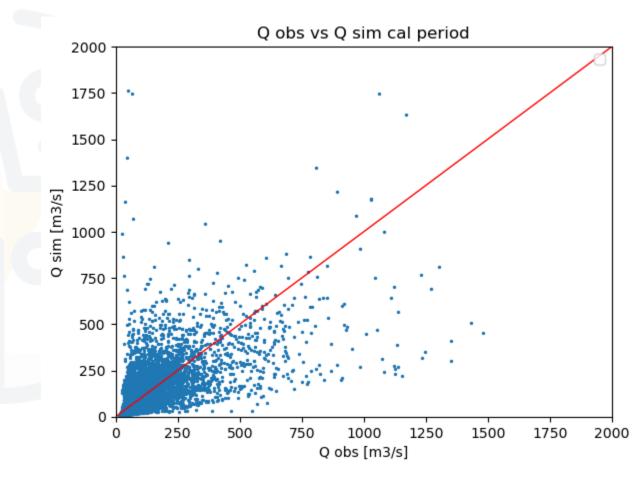






Scatter plots







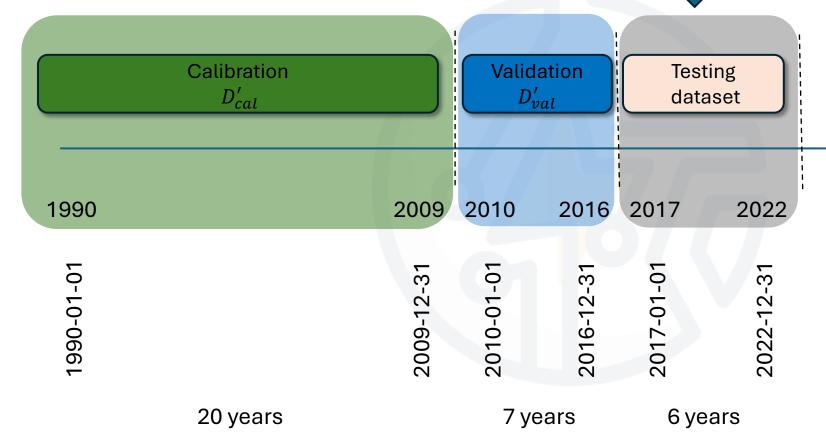






Testing

/work/bb1480/b381840/scewero/somes_tisa/hydromodelling/somes_tisa_generate_Q.ipynb



33 years





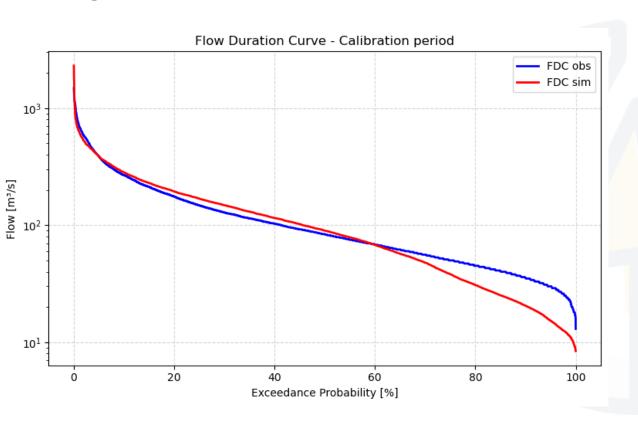


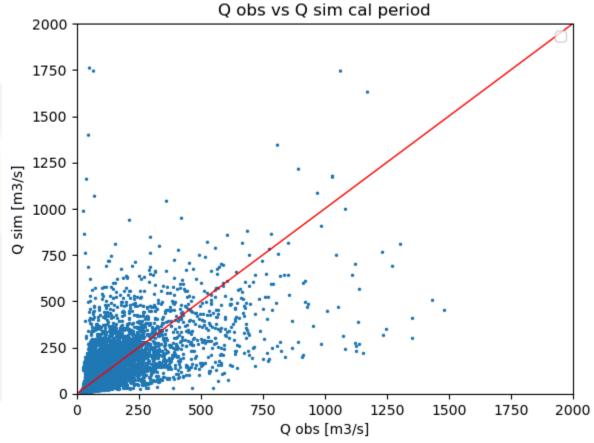






Testing









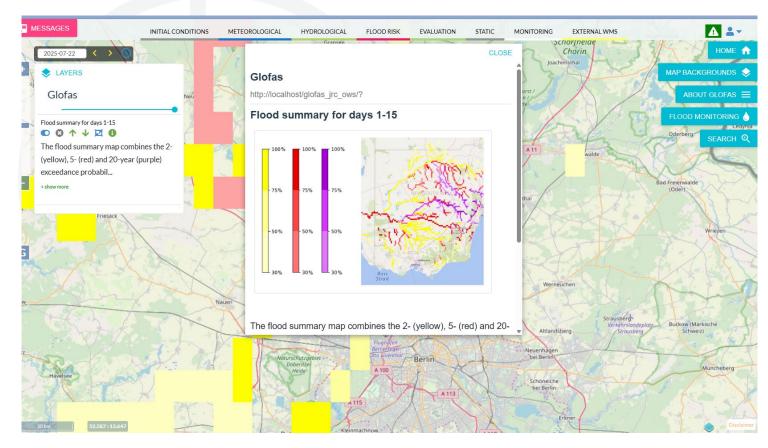






Short term flood summary

https://global-flood.emergency.copernicus.eu/







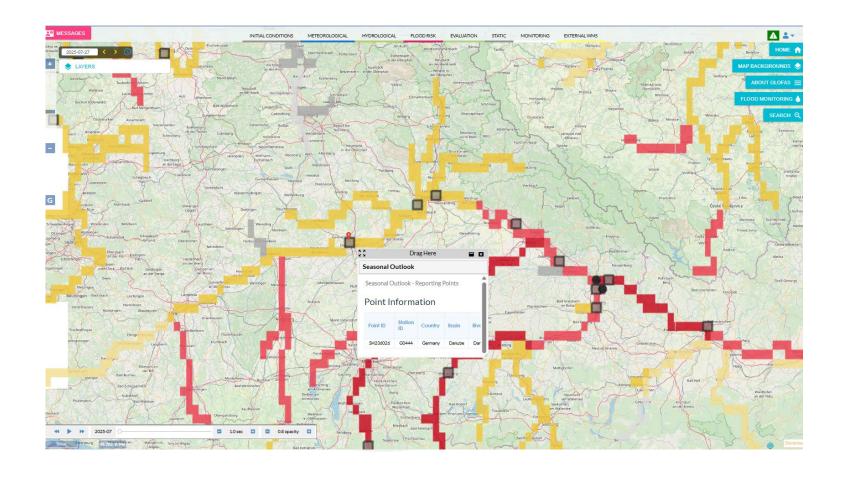








Seasonal forecast













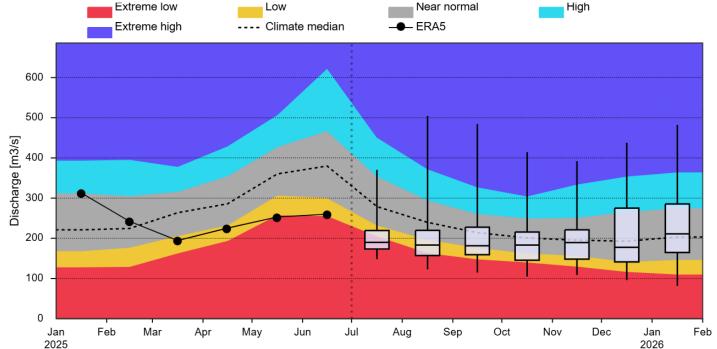
Seasonal forecast

Seasonal Outlook - Reporting Points

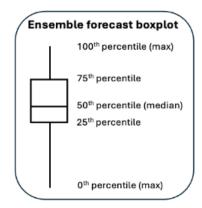
Point Information

oint ID	Station ID	Country	Basin	River	Station Name	Drainage Area [km2]	Longitude [Deg]	Latitude [Deg]	LISFLOOD Drainage Area [km2]	LISFLOOD X [Deg]	LISFLOOD Y [Deg]
SH236026	G0444	Germany	Danube	Danube	Ingolstadt	20001	11.4220	48.7540	20419	11.4250	48.7750

Seasonal hydrograph of antecedent, climatological and forecast information



Climatological thresholds From extreme low to extreme high (red to blue), derived from range-dependent model climatologies



	Anomaly	category	Rank	Description
4	Extren	ne low	1-10	Bottom 10% of the climatological distribution
	Lo	w	10-25	15% from the 1 st decile to the 1 st quartile
	Bit low		25-40	15% from the 1 st quartile to the 2 nd quintile
	Normal	Near normal (25-75)	40-60	20% from the 2 nd quintile to the 3 rd quintile
	Bit high		60-75	15% from the 3 rd quintile to the 3 rd quartile
	Hi	gh	75-90	15% from the 3 rd quartile to the 9 th decile
	Extrem	ne high	90-100	Top 10% of the climatological distribution







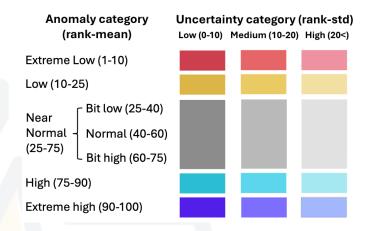






Seasonal forecast





Uncertainty category	Rank STD	Description									
Low uncertainty	<10	The standard deviation of the ensemble member ranks is less than 10									
Medium uncertainty	10-20	The standard deviation of the ensemble member ranks is between 10 and 20									
High uncertainty	20<	The standard deviation of the ensemble member ranks is higher than 20									

Seasonal forecast probability (numbers) and expected anomaly category (coloured cells) [2025-07-01]

			J	ıly 202	5					Au	gust 2	025			September 2025								October 2025								November 2025							December 2025							January 2026						
	EL	L	BL	N	ВН	н	ЕН	EL	L	BL	N	вн	н	ЕН	EL	L	BL	N	ВН	н	EH	EL	L	BL	N	ВН	н	ЕН	EL	L	BL	N	ВН	н	EH	EL	L	BL	N	ВН	н	EH	EL	L	BL	N	ВН	Н	ЕН		
July 2025	65	27	4	0	2	2	0	33	31	12	10	8	4	2	16	29	16	15	8	8	8	23	10	18	23	12	12	2	10	19	16	27	12	14	2	14	16	17	12	10	23	8	12	10	16	21	12	23	6		
June 2025	22	21	14	21	14	2	6	14	31	19	16	10	6	4	12	14	18	17	16	17	6	17	14	14	19	14	12	10	16	17	12	23	14	6	12	8	19	12	19	20	14	8									
May 2025	12	23	16	21	20	8	0	12	21	14	21	10	16	6	6	14	14	33	17	10	6	18	16	17	21	14	8	6	16	14	14	10	19	25	2																
April 2025	12	12	19	14	21	12	10	8	12	16	14	21	19	10	6	19	18	23	8	10	16	12	15	21	14	12	14	12																							







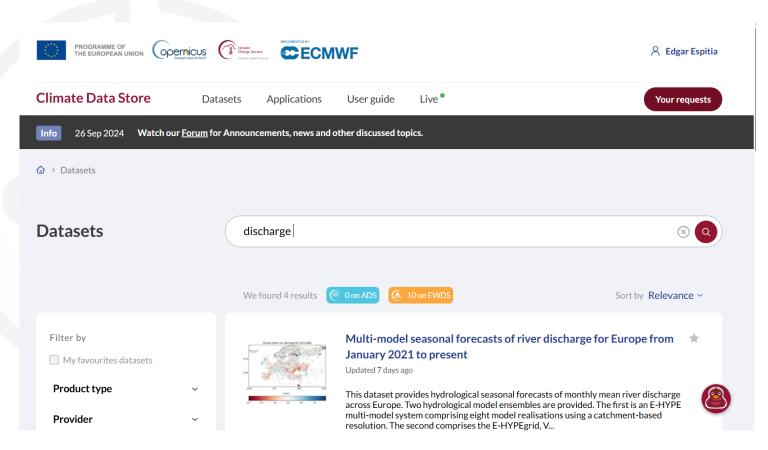






Seasonal forecast download

https://cds.climate.copernicus.eu/datasets















Thank you!

Questions?









