

Flood forecasting using sensor network and Support Vector Machine model

Jakub Langhammer

Charles University in Prague, Faculty of Science
Department of Physical Geography and Geoecology



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Motivation

Hydrological modeling in mid-latitude montane basins

Specific conditions

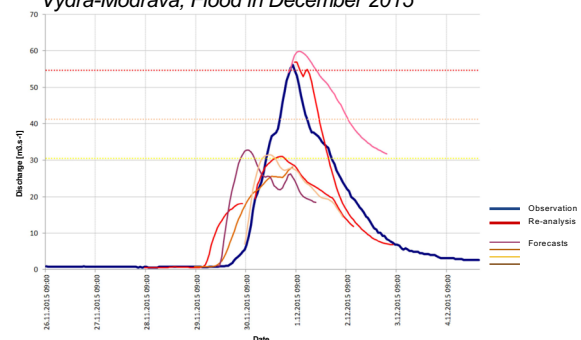
- Decisive area for flood generation
- Complex physiography
- Rapid runoff response

Limitations

- Limited data sources
- Uncertainties in forecasting by process-based models

=> Potential for data-driven models

Example of the volatility of hydrological forecasts
Vydra-Modrava, Flood in December 2015



Rain-on-snow flood - issued hydrological forecasts, reanalysis and discharge observations at Vydra-Modrava station (CHMI)



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Study area

Vydra River basin, Sumava mts.

- Mid-latitude montane basin
- Source area of frequent flooding
- Irregular rainfall distribution
- Significant share of peatbogs
- Extensive forest disturbance by bark beetle
=> **Rapid flood generation**

Effects of climate change

- = rise of air temperature
- = changing runoff seasonality
- => **rising frequency of high flows**
- => **prolonged low flow periods**



LANGHAMMER, J., BERNSTEINOVÁ, J., 2020. Which Aspects of Hydrological Regime in Mid-Latitude Montane Basins Are Affected by Climate Change?. *Water*, 12(8), 2279.

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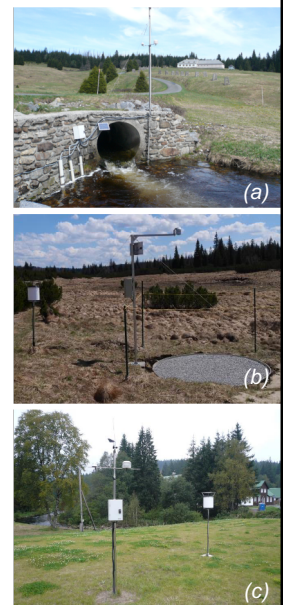
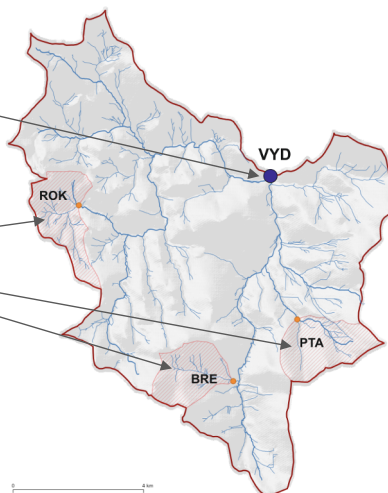
Monitoring and input data

Long-term monitoring

- **CHMI gauging station at outlet**
- Precipitation, snow cover (Churáňov, Filipova huť)
-> outside the catchment

Experimental catchments

- **Experimental catchments**
- Charles University, since 2006
- Basin headwaters
- Automated sensor network
 - gauging stations (a)
 - meteo stations (b)
 - snow pillow, snow scale (c)



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Input data and modeling scenarios

Observations

- Gauging station at basin outlet
 - daily discharge
- Sensor network in headwaters PTA, BRE, ROK
 - water stage
 - precipitation
 - snow cover

Calculated indices

- Sensor network data
 - API30, API 7
 - baseflow index (Digital recursive filter)
 - Potential Evapotranspiration (Oudin method)

Data processing

- Time span 2011 - 20
- Filling the gaps in time series
- Unified to daily values

Training period

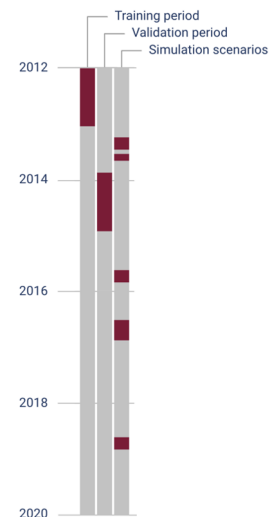
- Complex period covering all types of events
- Hydrological year 2012

Validation period

- Hydrological year 2015

Simulation scenarios

- Spring snowmelt (April 2016)
- Rain on snow event (December 2015)
- Flood from regional rain (Spring 2013)
- Summer storm flood (June 2013)
- Repeated storms (August 2018)



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Model setup

Data driven model

Support Vector Machine

Non-probabilistic classifier, based on the augmentation of dimensionality for classification

Network design

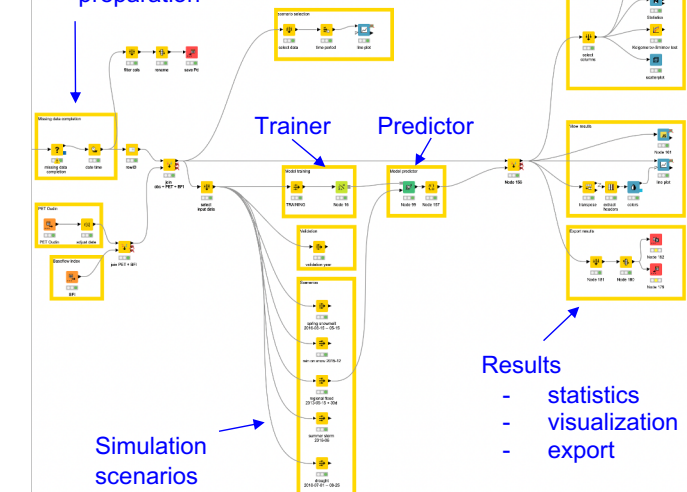
- Target
 - CHMI gauging station (Qd)
- Source
 - sensor network (H, P, SNW)
 - calculated indices (API, BFI, PET)

Computing environment

- KNIME 4.5
- LibSVM library, nu-SVR algorithm
- Python

Input data

- import
- preparation



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Results – Validation period

Complex hydrological year

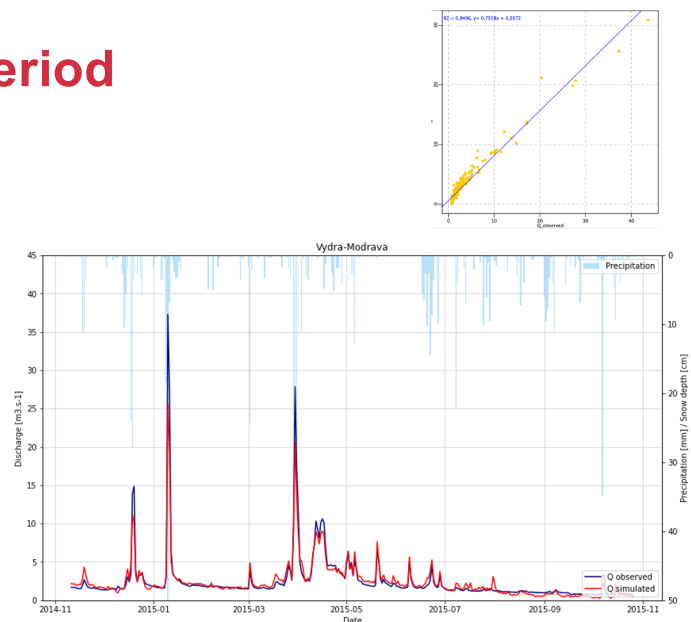
2014-11-01 -> 2015-12-01

Hydrological year covering key types of events

- spring snowmelt
- recurrent flooding from rainfall
- single storm floods
- multiple-peak storms
- low flow period

Statistics

$R^2 = 0,9496$



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Results - Spring snowmelt

Spring snowmelt in April, 2016

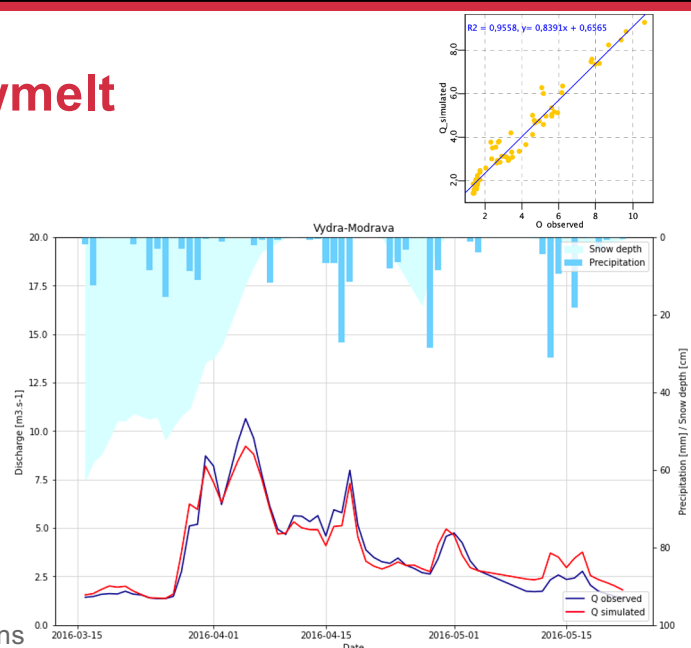
- 25 days of high flows
- Continuous snowmelt
- Recurrent but weak precipitation
- One wave, four peaks

Fit of simulation to observations

- Good fit of shape
- Good timing of peaks
- Slight underestimation of flows

Statistics

- NSE = 0,9405
- $R^2 = 0,9558$
- Kolmogorov-Smirnov test
p-value 0,21 > 0,05 = fit of distributions



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Results - Rain on snow flood

Flood in December, 2015

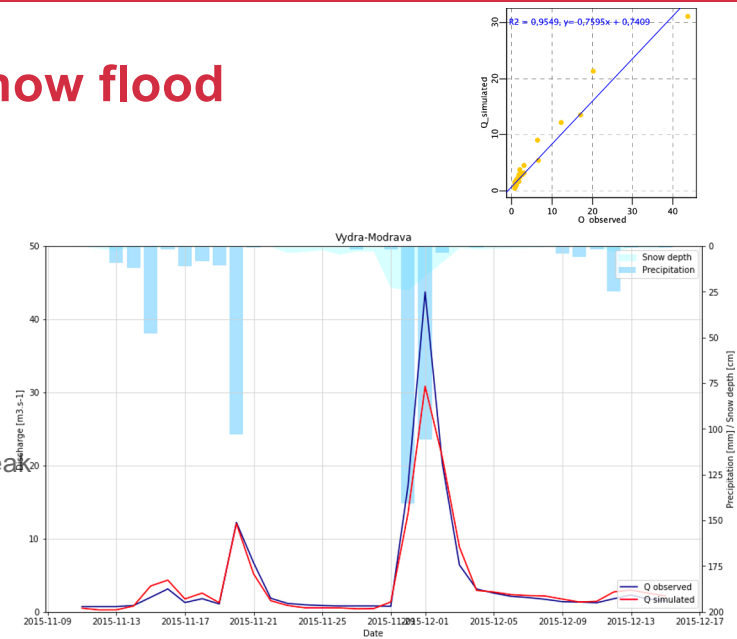
- Beginning of winter
- Fresh snowpack + heavy rainfall
- Rapid snowmelt
- 5-10-years flood

Fit of simulation to observations

- Good fit of shape
- Good timing of peaks
- Slight underestimation of major peak

Statistics

- NSE = 0,9103
- $R^2 = 0,9549$
- Kolgomorov-Smirnov test
p-value 0,21 = fit of distributions



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Results - Regional-scale flood

Spring flood in June, 2013

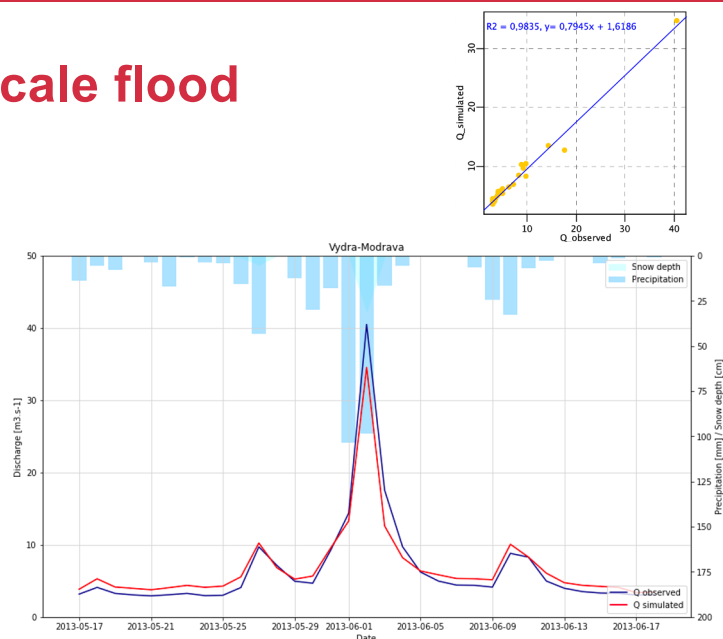
- Beginning of summer
- Long-lasting precipitation
- 5-10-year flood in headwater
- 50-year flood in lowland

Fit of simulation to observations

- Good fit of shape
- Good timing of peaks
- Good fit of peak flows
- Slight overestimation of low flows

Statistics

- NSE = 0,94524
- $R^2 = 0,9835$
- Kolgomorov-Smirnov test
p-value 0,13 = fit of distributions



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Results - Summer storm

Summer, 2016

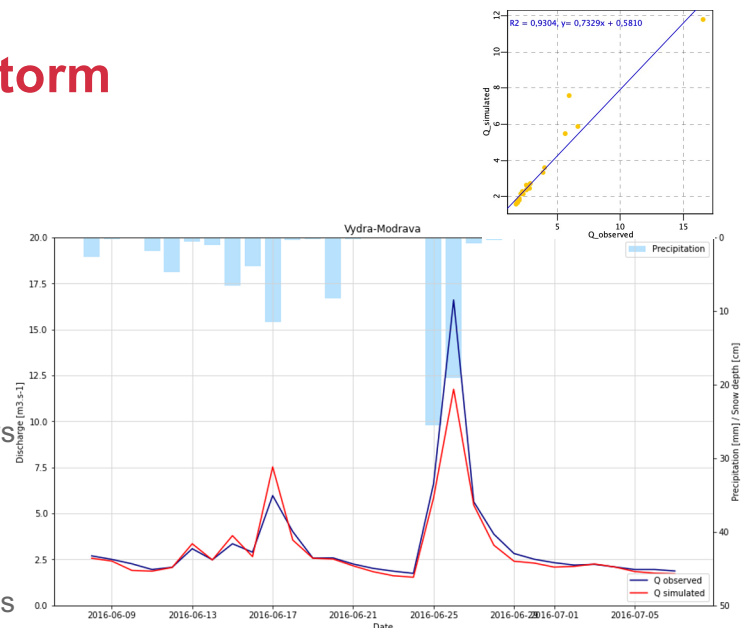
- 5-10-year flood in headwater
- Summer storm

Fit of simulation to observations

- Good fit of shape
- Good timing of peaks
- Good fit of peak flows
- Slight overestimation of low flows

Statistics

- NSE = 0,9161
- $R^2 = 0,9304$
- Kolgomorov-Smirnov test
p-value 0,67 = fit of distributions



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Conclusions

Potential of SVM for flood forecasting

- Good fit of simulations to observations even in basins with complex physiography
- Ability to include heterogeneous data sources (i.e. sensor networks)
- Improvements of fit by use of calculated indices of basin status (API, BFI, PET)
- High computational efficiency
- Instant delivery of the predictions

Limitations

Black-box approach

Sensitivity to responsible model setup and training

- Even data-driven model needs hydrologists
- Avoiding models based on statistically correlated elements with no physical correspondence
- Hydrologically valid design of model network selection of variables
- Complexity of training period



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**Thank you
for your attention!**

prof. Jakub Langhammer

Charles University in Prague, Faculty of Science

Department of Physical Geography and Geoecology

jakub.langhammer@natur.cuni.cz,

<http://langhammer.natur.cuni.cz/>



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LANGHAMMER, J., ČESÁK, J., 2016. Applicability of a Nu-Support Vector Regression Model for the Completion of Missing Data in Hydrological Time Series. *Water* 2016, 8(12), 560



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