#### River discharge extremes in Norwegian regulated catchments: simulations using a hydrologic model including human interventions

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# Motivation —> Objectives

- Increasing frequency and severity of extreme weather
- Impact prediction requires modelling human interventions

- Evaluate LISFLOOD hydrologic / water resources model in a heavily regulated catchment (Drammen)
- Use local data for model improvement w.r.t. extremes

# Outline

- Drammen catchment
- LISFLOOD hydrologic and water resources model
- Model input and calibration
- Results
- Conclusions and way forward

#### **Drammen catchment**





- Seasonal hydrologic regime dominated by snowmelt
- 54 reservoirs (only 4 in European LISFLOOD setup)
- Active storage ~ 35% of average annual streamflow
- Reservoirs crucial to reduce flood damage, especially when large snowmelt is predicted

#### **Active storage (% annual precipitation)**



Most regulating capacity (active storage / annual precipitation) is upstreams, where precipitation is also largest

#### LISFLOOD hydrologic and water resources model

- Developed by European Commission's JRC and ECMWF
- Operational use in flood prediction (EFAS) and drought monitoring (EDO)
- Recent applications in Water-Energy-Food-Ecosystem Nexus assessments

![](_page_5_Picture_4.jpeg)

![](_page_5_Picture_5.jpeg)

# LISFLOOD: hydrology

![](_page_6_Figure_1.jpeg)

#### **Space resolution:**

- Gridded: now 1'
- Sub-grid land cover tiles

#### **Time resolution:**

- Input & water balance: daily
- River routing: hourly

#### **River routing:**

- 1D kinematic wave: channel, floodplain
- Lakes
- Regulated reservoirs
- Human water use

### **LISFLOOD: water resources**

![](_page_7_Figure_1.jpeg)

Environmental flow requirement

- Water demands from multiple sectors
- Water abstractions based on water demand, availability and ecological flow requirements
- Sources: groundwater, rivers, lakes and reservoirs
- Simple reservoir model

## **LISFLOOD:** reservoirs

Outflow = f(storage;  $\alpha$ ,  $\beta$ )

![](_page_8_Figure_2.jpeg)

#### Simple model (+)

Useful for large data-scarce domains

#### Limitations (-)

- Real operation purposes (e.g. hydropower) not included
- Reservoirs are independent
- Cannot exploit detailed data

# **Model input**

• Land surface (vegetation, land cover, soil, river network, etc.): EFAS maps at 1' resolution

 Atmospheric forcing 1978-2020 (1 km): seNorge\_2018 (precipitation, temperature), HySN5 (radiation, humidity) and Klinogrid (wind)

• Reservoir active storage: NVE

### **Model calibration: 2 steps**

To avoid compensation errors arising when calibrating all parameters simultaneously

![](_page_10_Figure_2.jpeg)

### Naturalised streamflow calibration

![](_page_11_Figure_1.jpeg)

- Relatively high and robust KGE (> 0.75 at 60% stations) and correlation (> 0.8)
- North-western headwaters: lower KGE due to large negative bias

### **Streamflow underestimation**

![](_page_12_Figure_1.jpeg)

- Average measured streamflow exceeds precipitation by up to 270 mm/year
- Negative model bias partly due to underestimated precipitation input

### **Reservoir calibration**

![](_page_13_Figure_1.jpeg)

- Lower KGE than naturalised calibration, due to simplistic reservoir model
- Robust under validation

## **Regulated vs naturalised simulations**

![](_page_14_Figure_1.jpeg)

Comparison of naturalised and regulated (reservoirs) simulations against measured discharge:

- **Reservoirs** improve reproduction of discharge seasonality
- Improvement due to storage buffering effect rather than simplistic regulation model
- Naturalised simulation is a low benchmark

## **Regulated calibration: extremes**

#### Drammenselva at Mjøndalen bru

![](_page_15_Figure_2.jpeg)

- Underestimation during most low flow periods.
- Overestimation of most **spring annual maxima**, especially the largest ones (2013, 2018)
- Underestimation of 2015 and 2020 autumn maxima
- Fair reproduction of several flow peaks (2011, 2014, 2019)

### **Extremes: annual maxima**

Generalised extreme value (GEV) distribution fitted to annual maxima:

• Shape parameter determines upper tail thickness (a)

Example: GEV shape parameter and tail

(a)

0.4 -

Probability density 7.0 8.0 8.0 8.0 8.0

0.1

0.0

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- 05

- Regulated (reservoirs) simulation vs discharge: general overestimation (b)
- Naturalised simulation vs naturalised streamflow: no clear pattern (c)

![](_page_16_Figure_5.jpeg)

GEV shape parameter estimates with 90% confidence intervals

# Conclusions

- Promising 2-step calibration approach:
  - 1. Hydrologic parameters using naturalised streamflow
  - 2. Reservoirs parameters using discharge
- Representation of extremes to be improved

- Future work:
  - Precipitation correction
  - Develop a more realistic reservoir model

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# Thank you

#### **Extra slides**

#### Naturalised streamflow results

![](_page_22_Figure_1.jpeg)

### Naturalised streamflow results

![](_page_23_Figure_1.jpeg)

#### Naturalised streamflow results

![](_page_24_Figure_1.jpeg)

### **Calibrated vs uncalibrated reservoirs**

![](_page_25_Figure_1.jpeg)