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MACHINE LEARNING AIDED FORECASTING OF SUB-SEASONAL DROUGHT INDICATORS IN THE EUROPEAN ALPS

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1 Introduction

• There is a need for decision makers to have early warnings of

3 Preliminsteen/lowRetSults

Initial conditions



- drought as many sectors in Central Europe depend heavily on its water resources, e.g. agriculture, hydropower production and transportation.
- "Hybrid forecasts" combining traditional models and machine learning techniques have shown potential for improved skills and they are recommended for sub-seasonal to seasonal forecasts.
- "Weather pattern approach" has shown promising results for meteorological forecasts, but there has been less effort on applying such approach directly to hydrological forecast.

Research Questions:

- 1. Can machine learning model be used to perform hydrological routing and generate streamflow forecasts at monthly timescale?
- 2. Can large scale weather patterns help improve the predictability of local drought indicators?
- 3. Can low flow forecasts be combined with forest fire algorithms to explore the potential of early prediction of forest fire?

Keywords: Drought, Machine Learning, Weather Regimes, Forest Fire, Hybrid Forecasting

2 Work Packages

Fig. 2. Discharge forecast from 22.11.2018 to 23.12.2018 at Lake Brienz outlet. The classes of drought emergency, drought warning, normal and high correspond respectively to 1-10, 33-10, 66-33 and 99-66 percentiles of climatology from 1991-2000. The solid black line represents the multi-model mean with uncertainty derived from the 51 ensemble members. Upper left panel: with PREVAH outputs only. Upper right panel: with added observed initial conditions. Bottom left: with added weather regime indices. Bottom right: with added Julian days as hydropower proxy. Model runs are with pre-processed forecasts.

Initial conditions

- WP1: Predicting low flows and lake levels in Swiss rivers and lakes using machine learning on sub-seasonal timescales.
 - Data: PREVAH^[1] monthly runoff forecast data, BAFU streamflow and lake level observation data, weather regime (WR) forecast data provided by Karlsruhe Institute of Technology (KIT).
 - Study Period: March 2018 September 2020
 - ML Algorithms: Multivariate Adaptive Regression Spline (MARS), Gaussian Process (GP), Quantile Random Forest (QRF), Gradient Boosting Machine (GBM)
- WP2: Predicting low flows and lake levels in large rivers and lakes in the European Alps on sub-seasonal scales.
 - Data: Alpine Drought Observatory (ADO) observation data, European Flood Awareness System (EFAS) forecast data, KIT WR forecast data
- WP3: Linking hydrological drought to forest fires in the Swiss Alps
 - Data: PREVAH^[1] monthly runoff and soil moisture forecasts, KIT WR forecast data, historical fire data from database "Swissfire"^[2].





Fig. 3. NSE of lake level forecast for Lake Brienz with 32-day lead time. Upper left panel: with PREVAH outputs only. Upper right panel: with added observed initial conditions. Bottom left: with added weather regime indices. Bottom right: with added Julian days as hydropower proxy. Model runs are with raw, pre-processed or reference PREVAH simulations.

4 Risk Assessment

- Sufficient data quantity needed for machine learning model training
- Interpretability of machine learning outputs to understand the driving forces of hydrological drought events.

5 References

- 1. Viviroli, D., Zappa, M., Gurtz, J., and Weingartner, R. (2009). An introduction to the hydrological modelling system PREVAH and its pre- and post-processing-tools. Environmental Modelling and Software, 24(10):1209–1222.
- 2. Pezzatti, G., Reinhard, M. and Conedera, M. (2010), 'Swissfire: die neue schweizerische Waldbranddatenbank', *Schweizerische Zeitschrift für Forstwesen* **161**, 465–469.