

EGU General Assembly 2022

Flood Forecasting With LSTM Networks: Enhancing Input Data With Statistical Precipitation Information

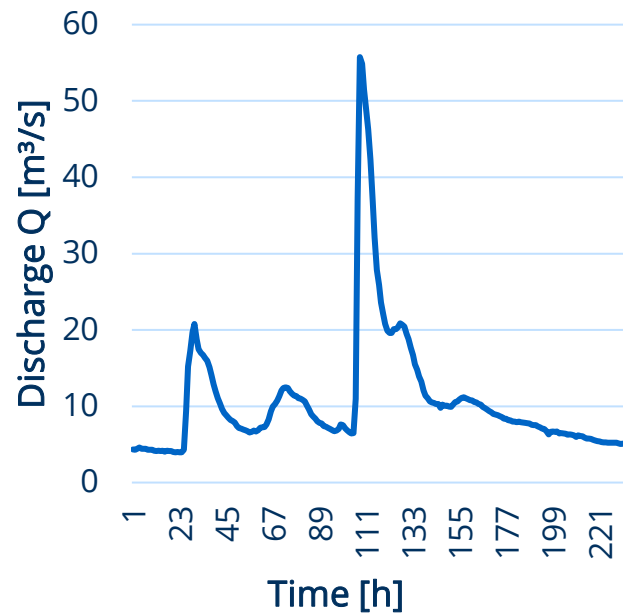
Tanja Morgenstern
Dr. Jens Grundmann
Prof. Niels Schütze

Vienna 25.05.2022

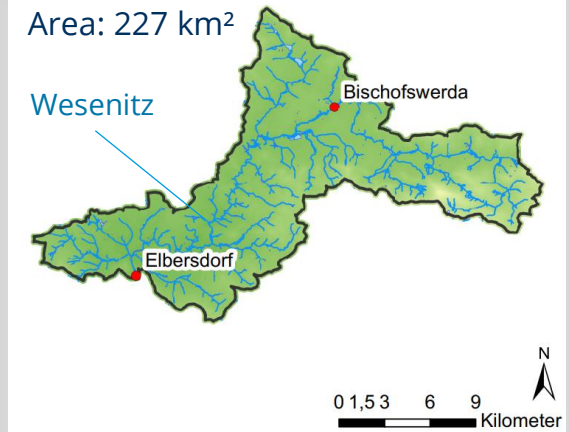
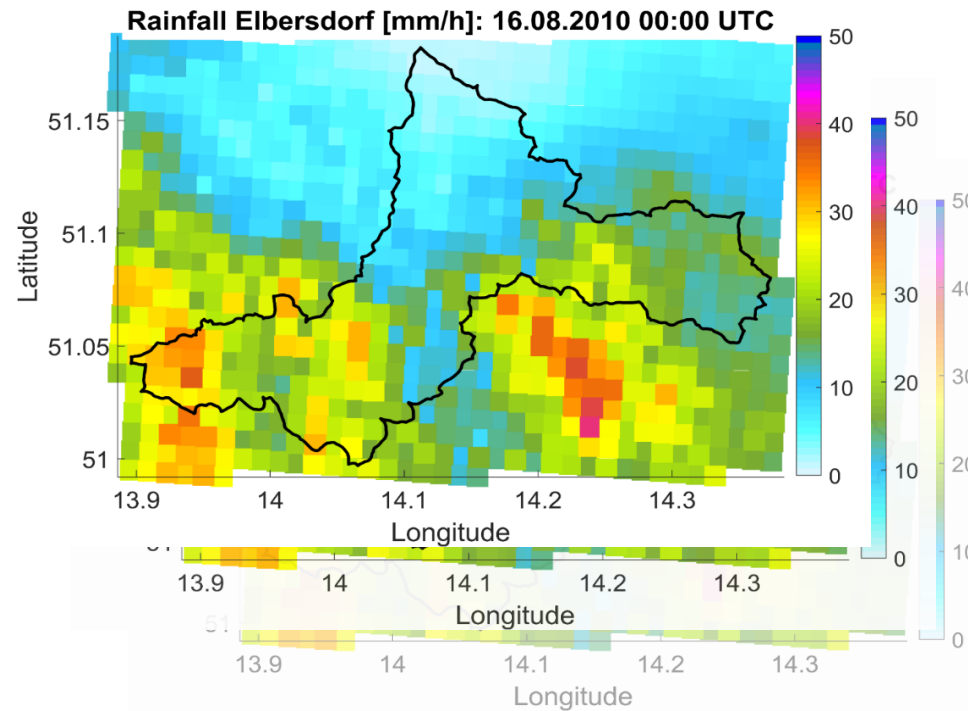
Data

The Basis

Hourly discharge observation (representative of catchment state)

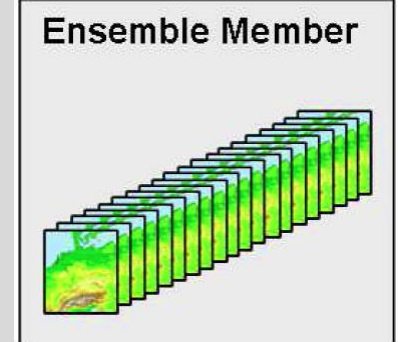


Hourly precipitation observation from RADOLAN RW product (driving force)



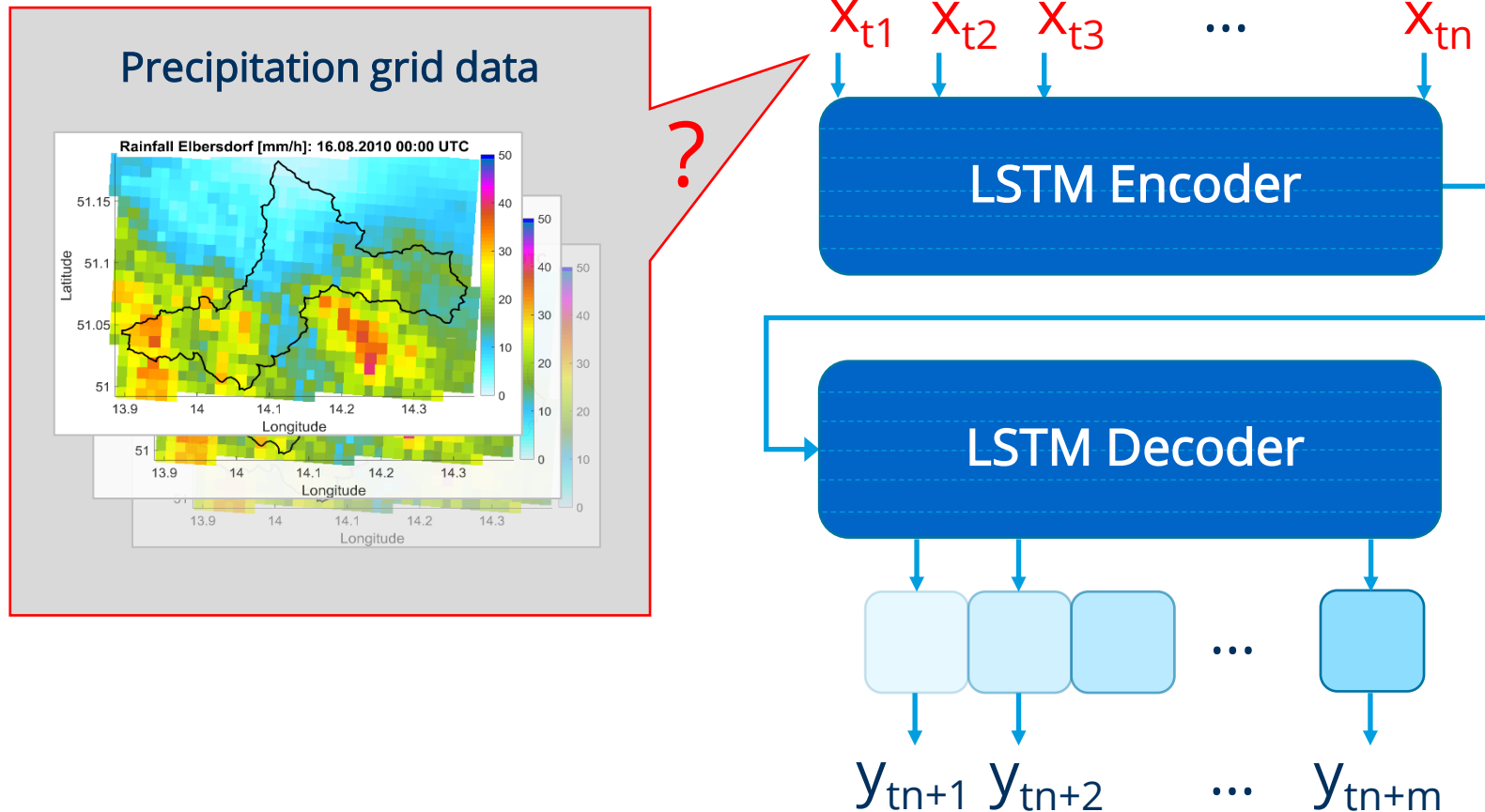
Later used as well:
meteorological
forecast ensembles

DWD: COSMO/Icon-D2-EPS



Methods

The Deep Learning Network



Hidden units: 35

Optimizer: Adam

Loss function: RMSE

Trained Networks per variation: 20

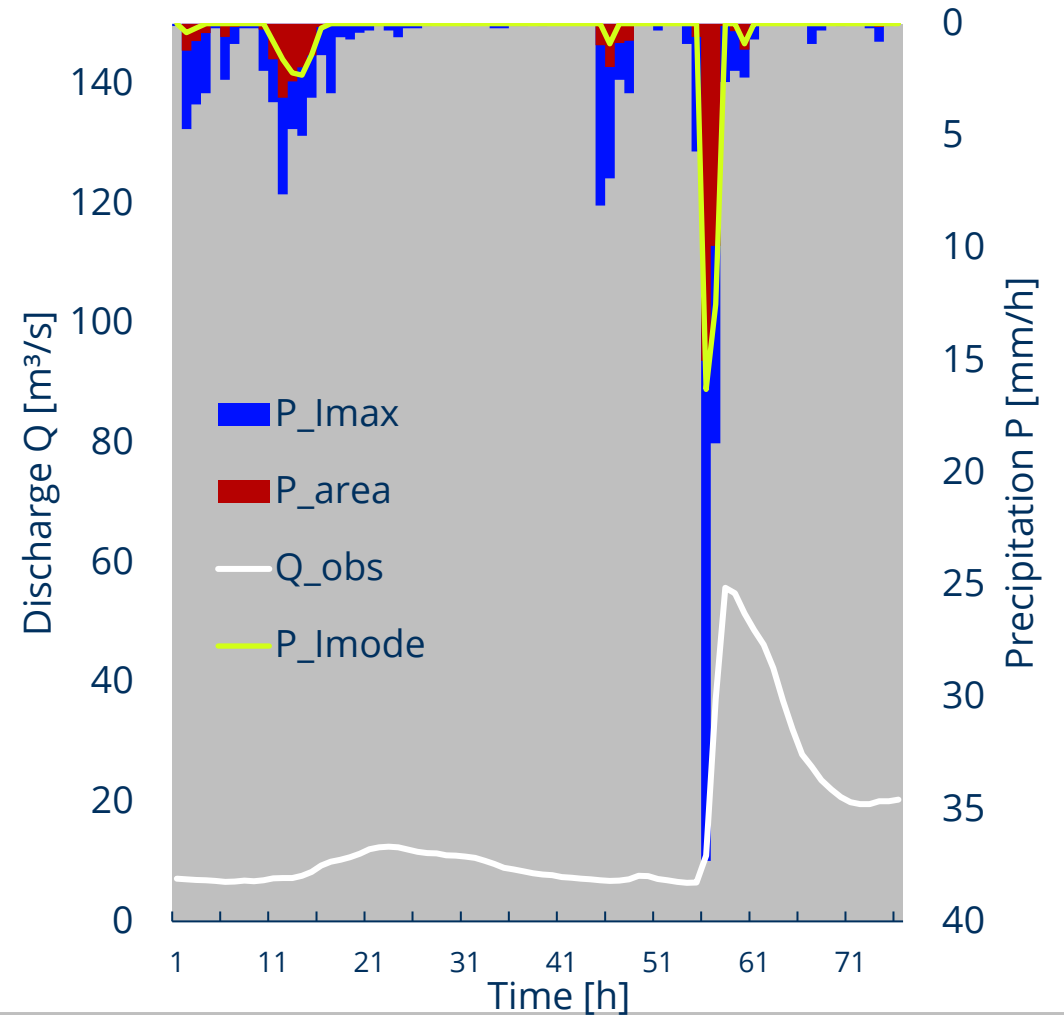
Event based training:
8 train events
(containing largest event)
4 validation events
8 test events

Moving-window-like samples in
batches (events):
Window length $n = 60$ h

Data

Statistical precipitation characteristics

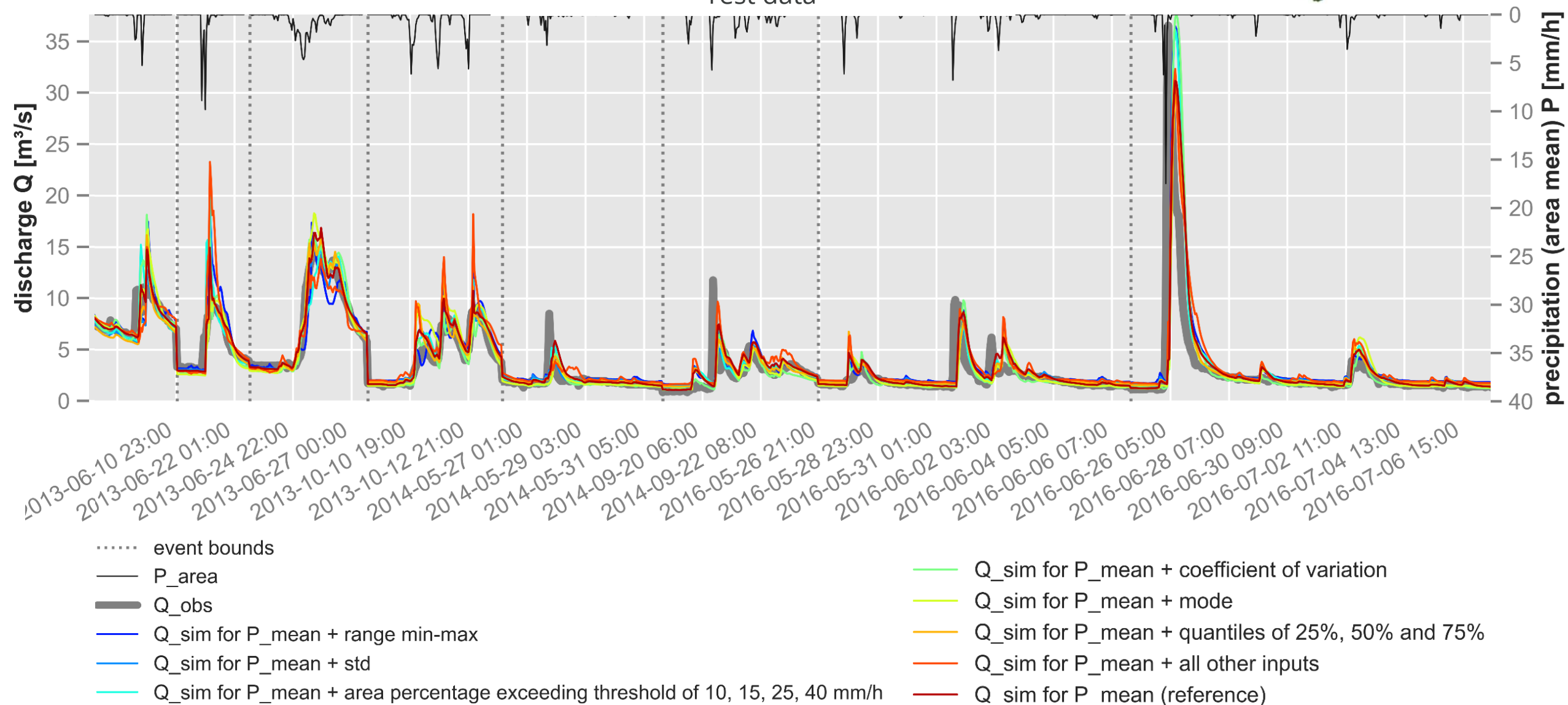
	Reference: discharge + area averaged precipitation
+	Precipitation intensity minimum & maximum
+	Standard deviation over area
+	Area percentage exceeding threshold of 10, 15, 25, 40 mm/h
+	Coefficient of variation over area
+	Mode value over area
+	Precipitation quantiles of 25%, 50% and 75% over area
+	All precipitation information



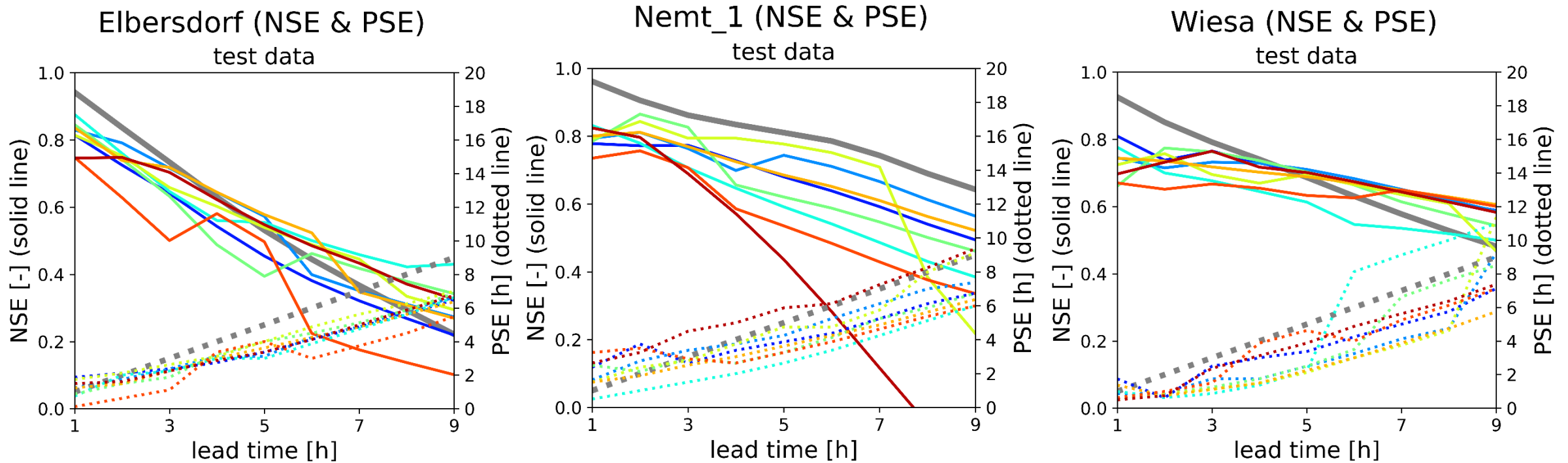
First results

Elbersdorf (lead time 4 h)

Test data



First results



NSE ... Nash Sutcliffe Efficiency

PSE ... Phase shift error

— Baseline out of Q_obs

— P_mean + range min-max

— P_mean + std

— P_mean + area percentage exceeding threshold of 10, 15, 25, 40 mm,

— P_mean + coefficient of variation

— P_mean + mode

— P_mean + quantiles of 25%, 50% and 75%

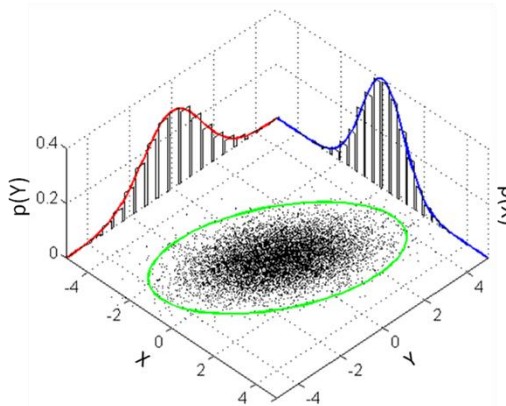
— P_mean + all other inputs

— P_mean (reference)

Prospective investigations

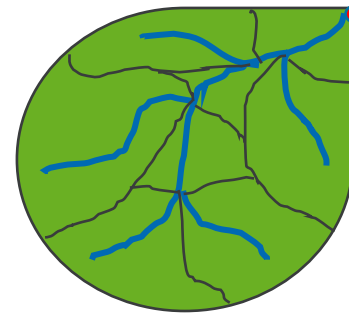
Input represents whole catchment

- correlation between best suitable statistical quantity and precipitation patterns?
- Precipitation characteristics containing **spatial** information (e.g. distance peak – gauging station, bivariate gaussian distribution as approximated spatial precipitation distribution)



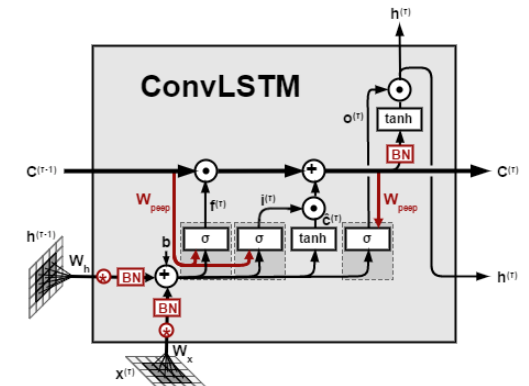
Input represents catchment sections

- symmetrical grid aggregation
- isochrones
- subcatchments



Input represents catchment grid

- LSTM networks with internal processing adapted to 3D (spatio-temporal) input (e.g. ConvLSTM)



<https://upload.wikimedia.org/wikipedia/commons/8/8e/MultivariateNormal.png>

https://miro.medium.com/max/942/1*u8neecA4w6b_F1NgnyPP0Q.png

Thu 26th May – Session HS1.2.2:



Thank you for your attention!

Contact: tanja.morgenstern@tu-dresden.de

Abstract:

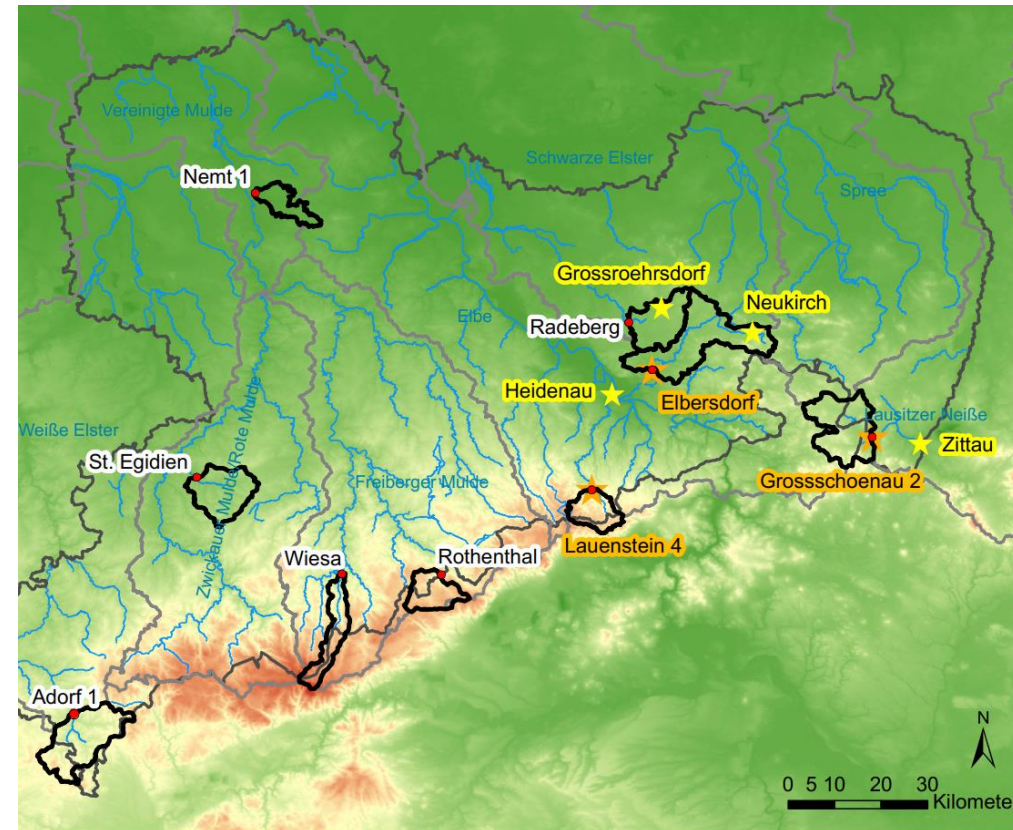
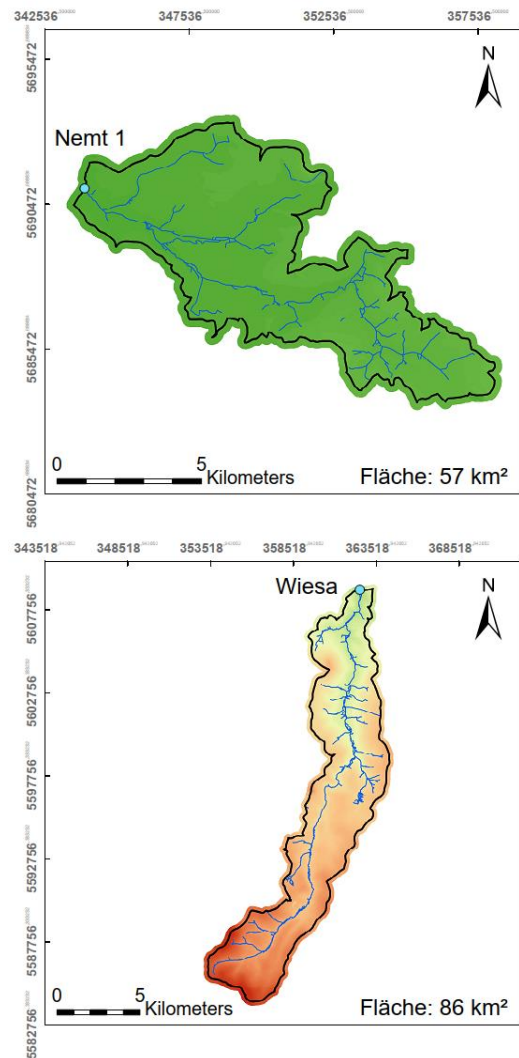


Towards automatic real-time
water level estimation using
surveillance cameras
Xabier Blanch
17:03 – 17:10



Determination of continuous
discharge time series based on
the optical Particle Tracking
Velocity (PTV)
André Kutscher
17:31 – 17:38

The other catchments



Legende

- Pegel
- ★ Messstellen Typ A
- ★ Messstellen Typ B
- Haupteinzugsgebiete
- Einzugsgebietsgrenze
- Landesgrenze
- Fließgewässer Übersicht
- Geländehöhe (DGM5)**
- 1242 m
- 0 m

Datengrundlage:
 DGM5: © GeoBasis-DE / BKG
 Pegel, EZG & Gewässer:
 Landesamt für Umwelt,
 Landwirtschaft und Geologie
 Sachsen