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What is the role of noise and uncertainty in neural network modelling? Almost 30 years of research and applications to karst aquifer modelling and flash-floods forecasting

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What is the role of noise and uncertainty in neural network modelling?

Almost 30 years of research and applications to karst aquifer modelling and flash-floods forecasting

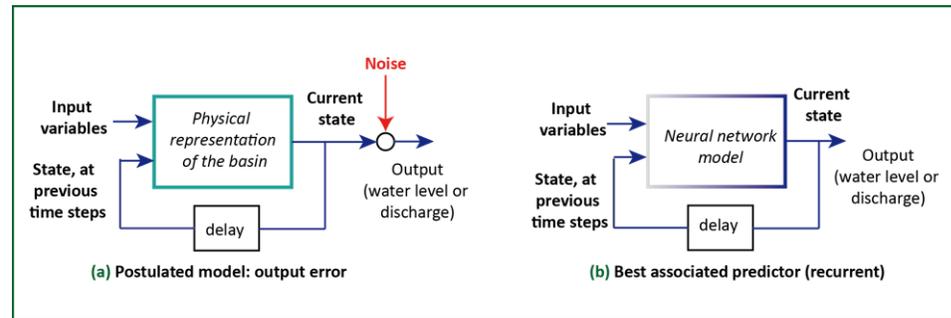
Neural networks to model watersheds or karst basins

Multilayer perceptron are universal approximators:

- ▶ The main challenges in data-driven modeling are the inaccuracy, the variability, and the statistical significance of the data,
 - Important to consider the relevant hypothesis relative to the noise impact [Nerrand et al, 1993, Taver et al, 2014]
- ▶ Two hypothesis: “output error” and “state error”

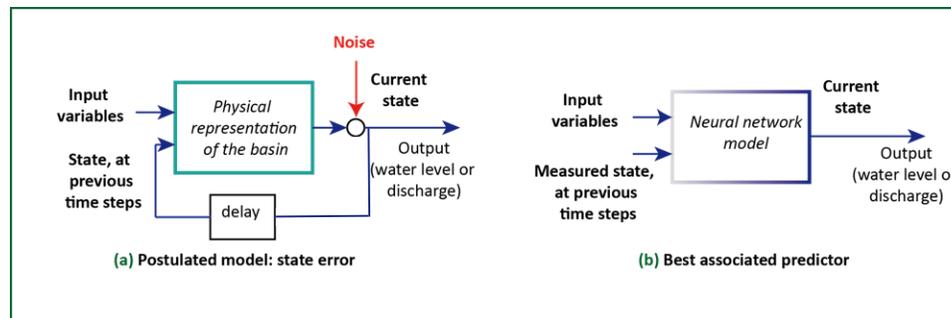
Output error

The hypothesis of output error supposes that uncertainties are essentially due to output measurement errors (for example rating curve). The best predictor to choose is a **recurrent model**. As the output is corrupted by noise, it is better to inform the model about its state using the estimated output at previous time steps.

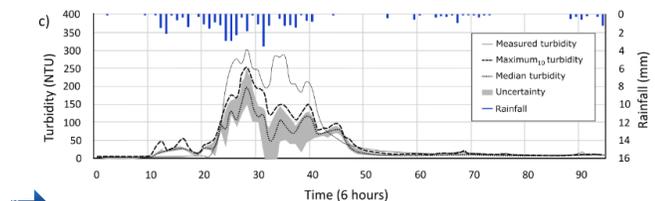
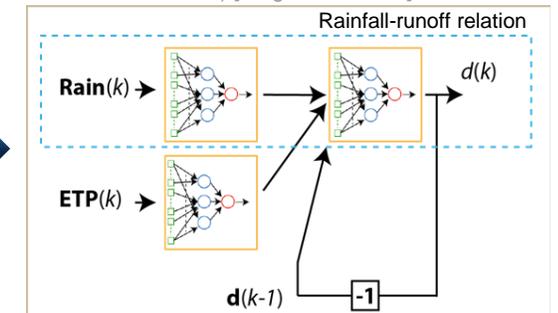
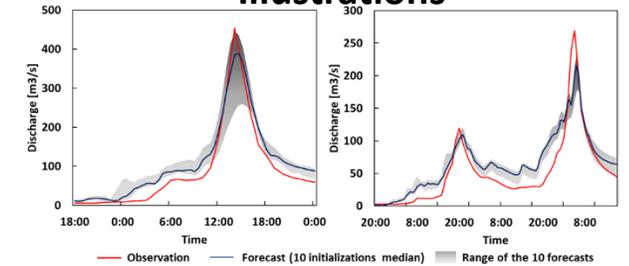


State error

The state of the postulated model is corrupted by noise or uncertainties that can originate from the model (parameters, function) or from rainfalls (forcing data). As the state estimated by the model is corrupted, it is thus logical to use as inputs the observed state of the basin (discharge or water-level) within a **feedforward model** (non recurrent).



Illustrations



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