Interactive Visualization of **Ensemble Data Assimilation Forecasts for Hydrology Models**

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1. Ensemble Data Assimilation

- A sequential approach used to estimate the dynamical state of a system with its uncertainties
- > Combines multiple sources of information, each having its own uncertainties, including:
 - Model predictions with model bias uncertainty,
 - real-world observations with measurement uncertainty
- to get an improved estimate of the system state.

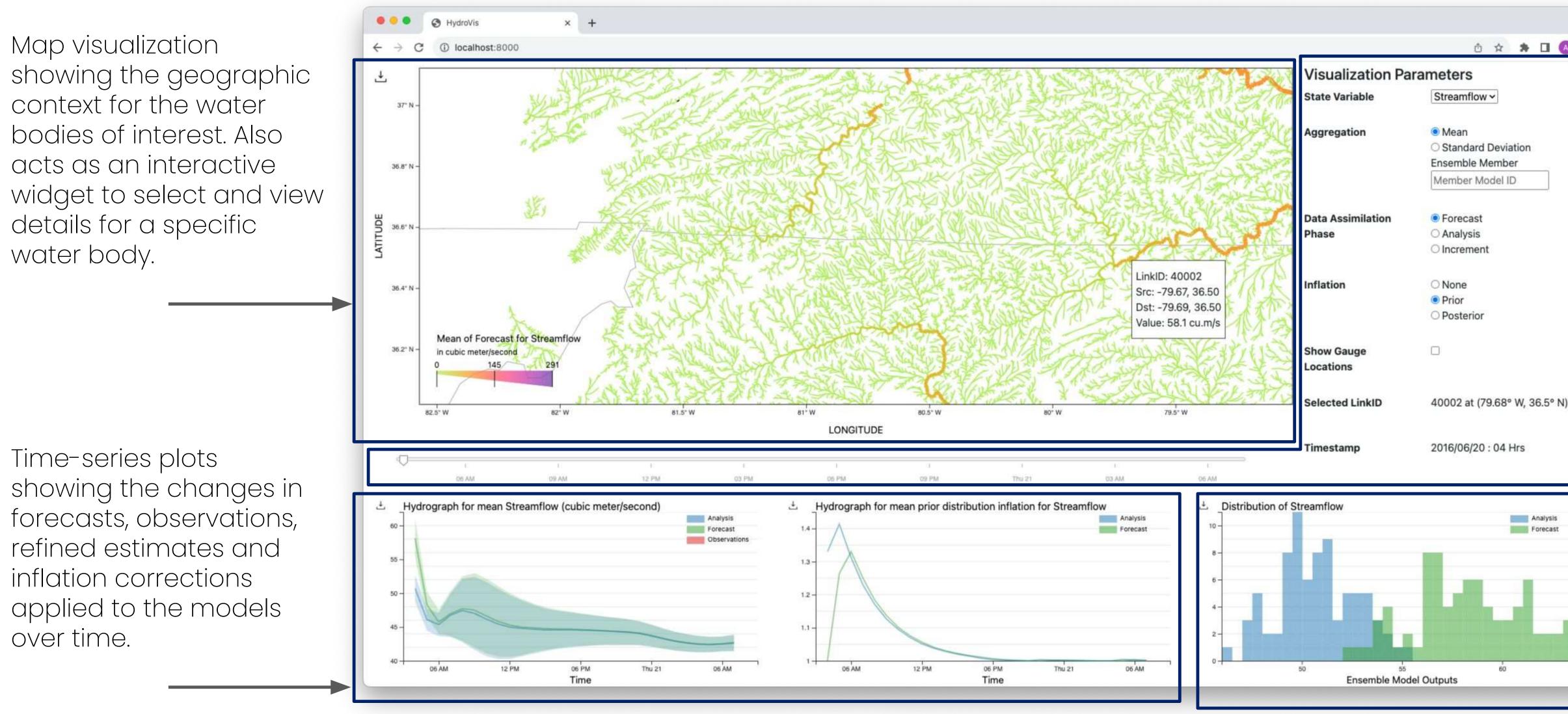
Used for forecasting weather, floods, and even spread of diseases like covid.

ensemble data assimilation, with the associated uncertainties.

2. HydroVis

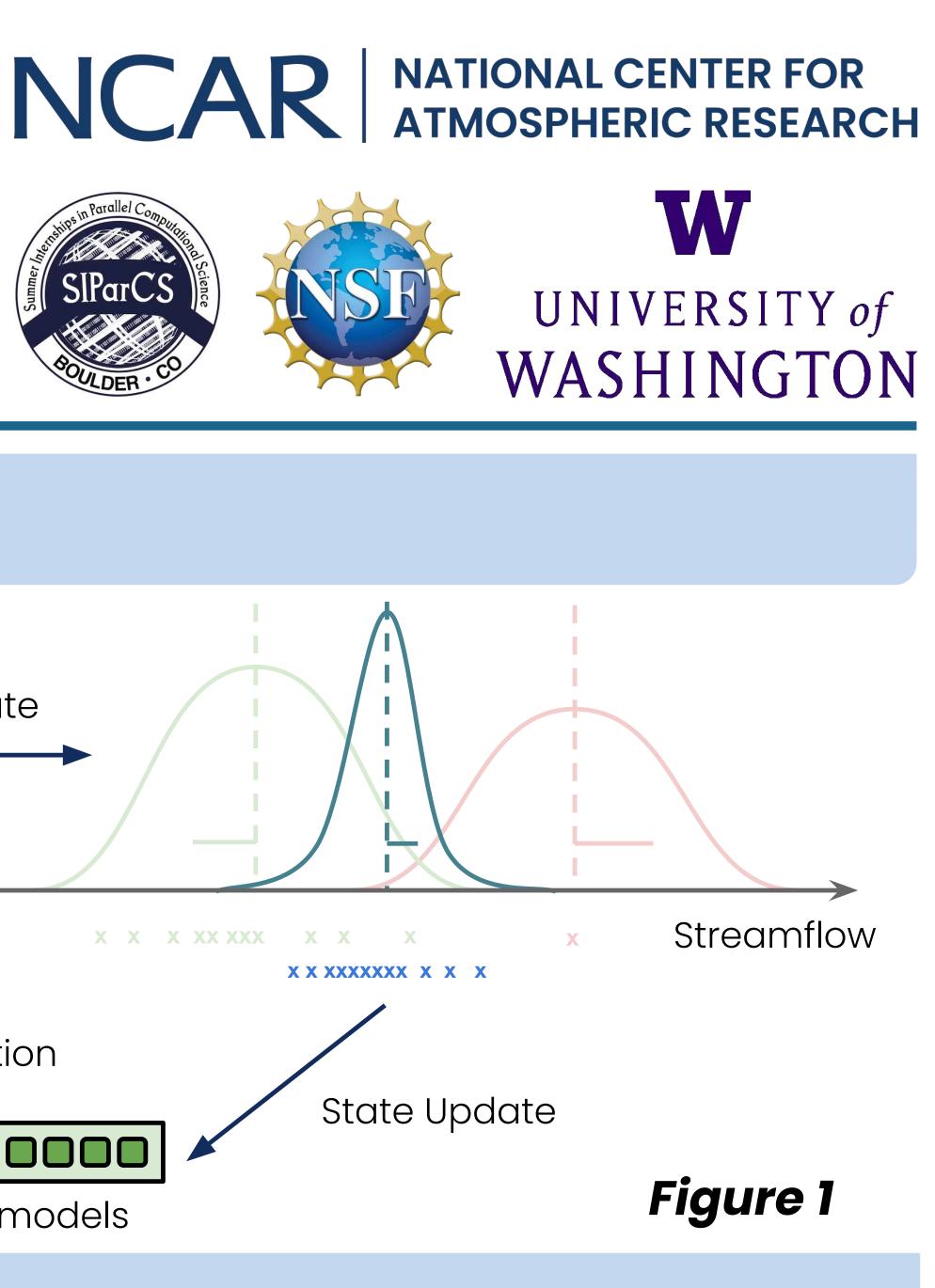
Problem: Understanding and making reliable decisions based on ensemble forecasts for Earth systems can be challenging due to the high dimension of earth system models, the various sources of uncertainties, and the massive volume of observation data. This becomes critical to plan precautionary measures, during extreme events like hurricanes.

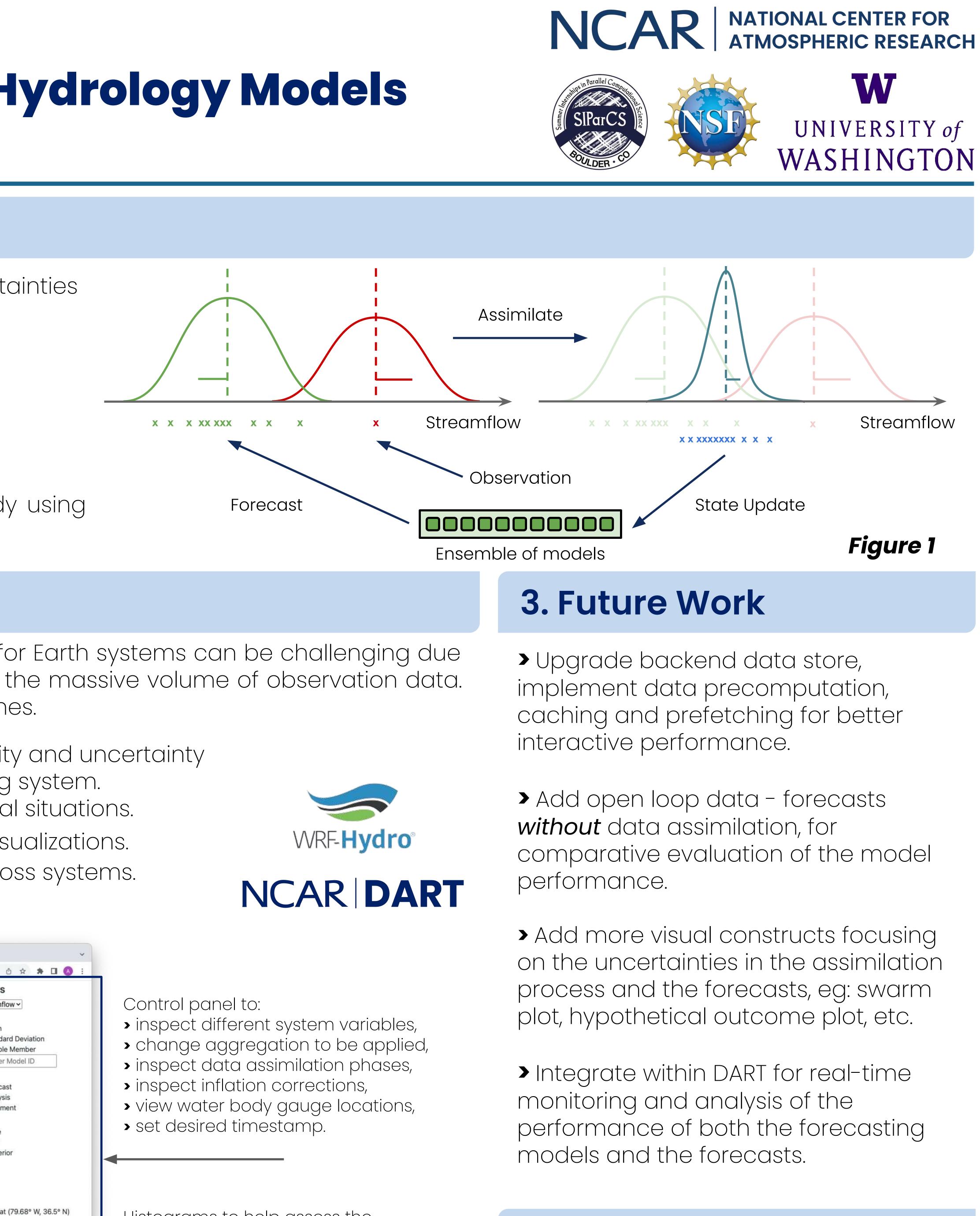
Solution: HydroVis - interactive dashboard for easy analysis and assessment of the quality and uncertainty of ensemble data assimilation forecasts, for the WRF-Hydro hydrological flood forecasting system. This would in turn help in making reliable ensemble forecasts based decisions, in life critical situations. > Developed as a web application with Python Flask for the web server and D3.js for the visualizations. ► Uses open-source libraries with limited software dependencies for better portability across systems. ▶ Interfaced with the Data Assimilation Research Testbed (DART) developed at NCAR.



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- Fig. 1 shows a schematic representation for forecasting the streamflow in a water body using





Histograms to help assess the uncertainty of the estimated streamflow at locations of interest. This panel also provides insights on the Analysis nature of the underlying streamflow Forecast probability distribution: Gaussian, near or non-Gaussian.

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Figure 2

4. Acknowledgements

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