



Exploring the Application of Ensemble Prediction Methods Across Regional Forecasting Domains

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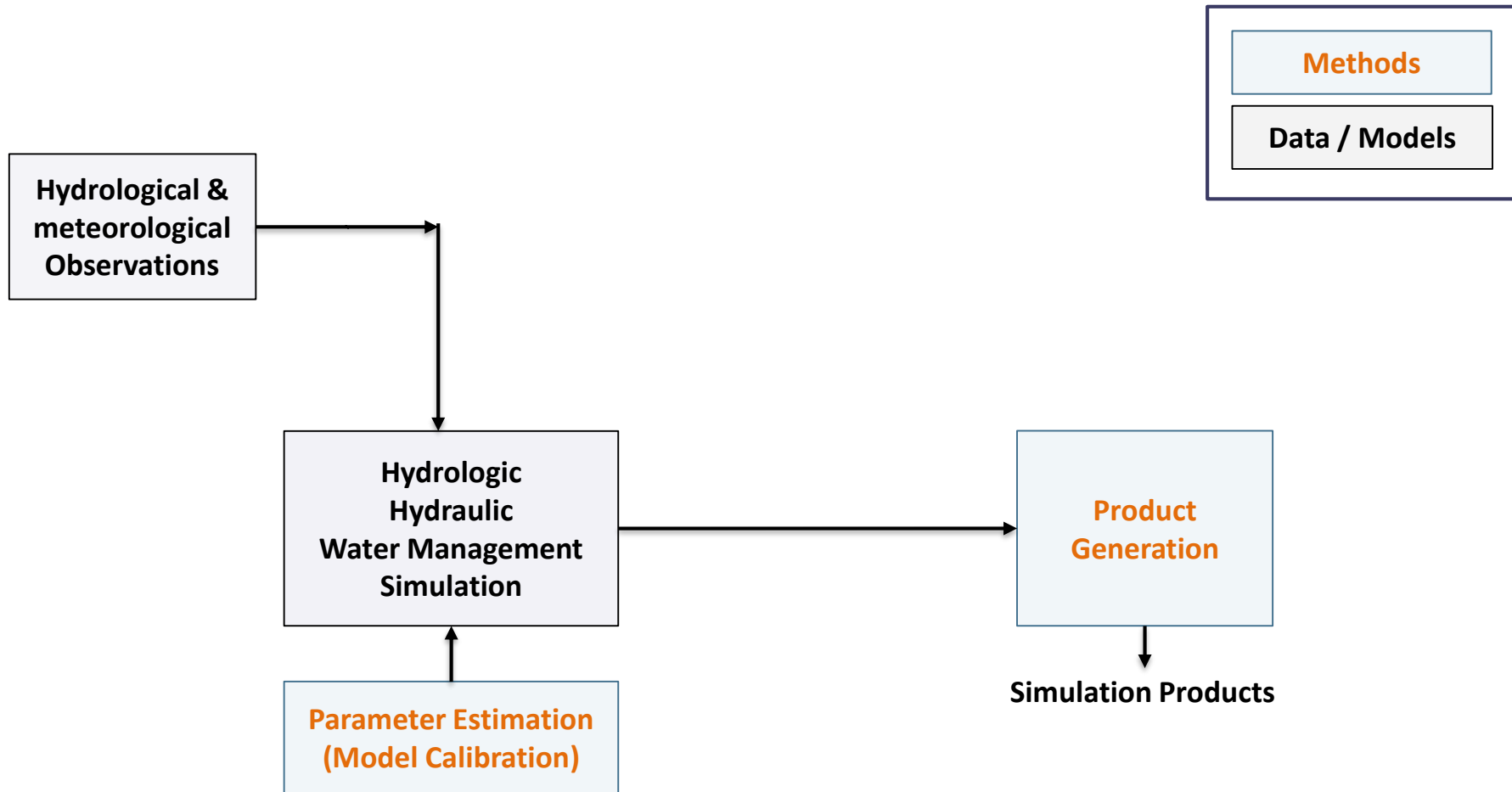
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HEPEX

U. Melbourne, Australia – February 2018

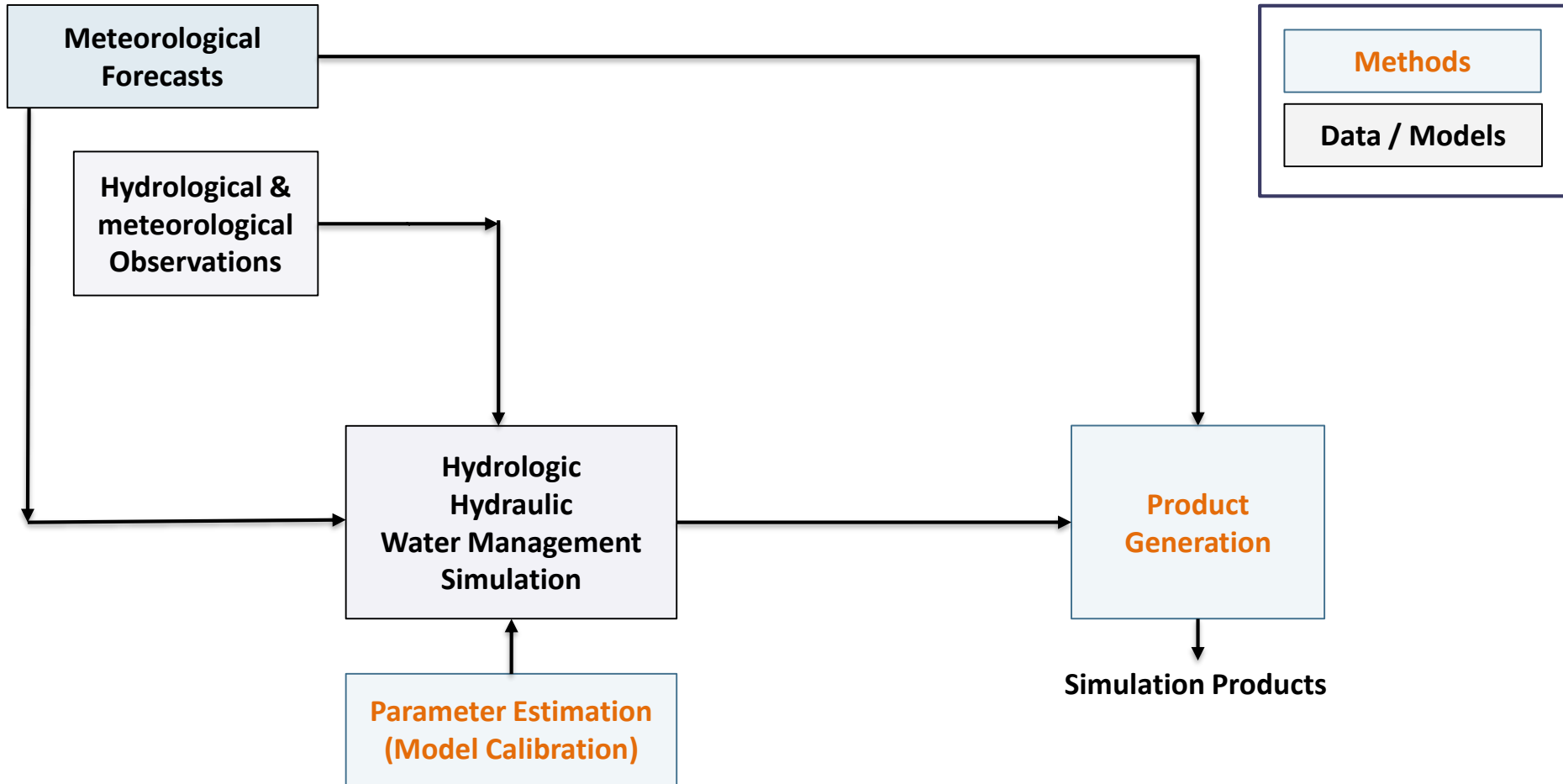


We have broad community experience in modeling



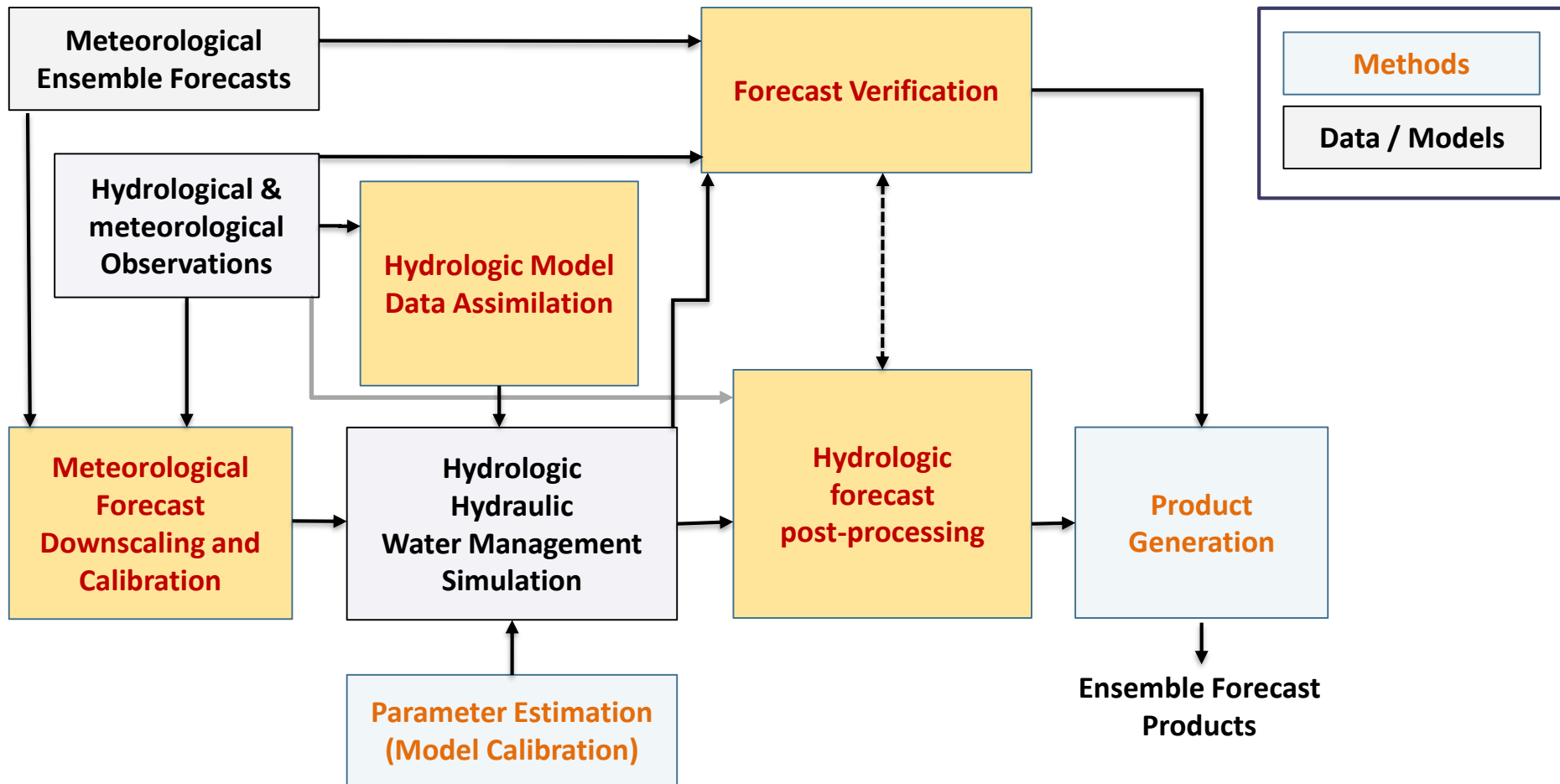
- Uncoupled systems – derive forcing meteorology from external sources to run hydrologic model

Generating forecasts seems a small leap



- Obtain similar forcings from NWP to run through same model, creating forecasts ... easy!

Generating *skillful* forecasts is more complex



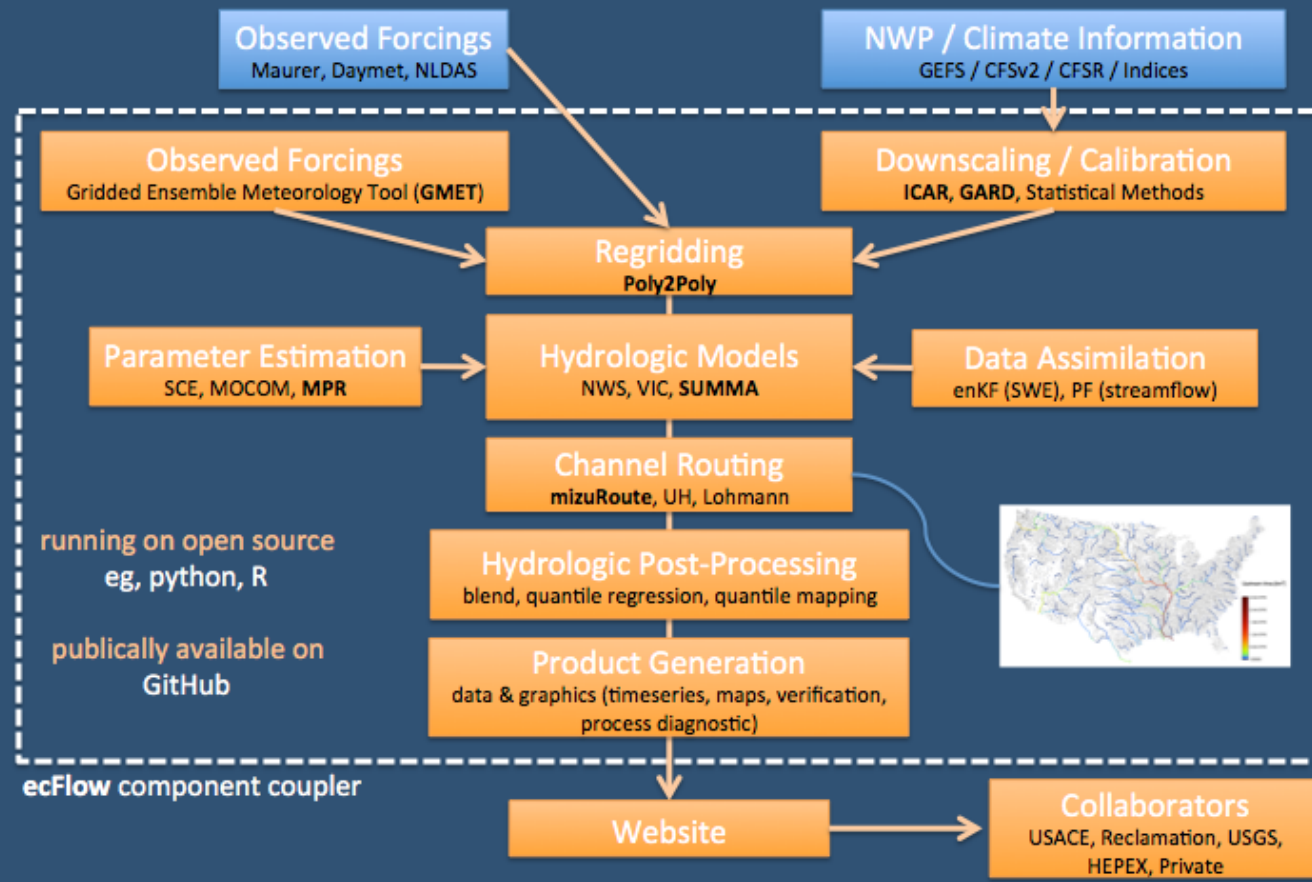
- Interdisciplinary
 - geosciences, statistics, computing and software engr., social science
- Not taught as a coherent program in any discipline
- HEPEX focus

SHARP Testbed forecasting system

We have developed an operational system to assess methods for real-time short and seasonal range forecasts

System for Hydromet Analysis Research and Prediction (SHARP)

an agile effort supporting ensembles, hindcasting, benchmarking, and development



workflow web monitor

SHARP System Status Report

Updated: Tue Dec 13 15:13:57 UTC 2016

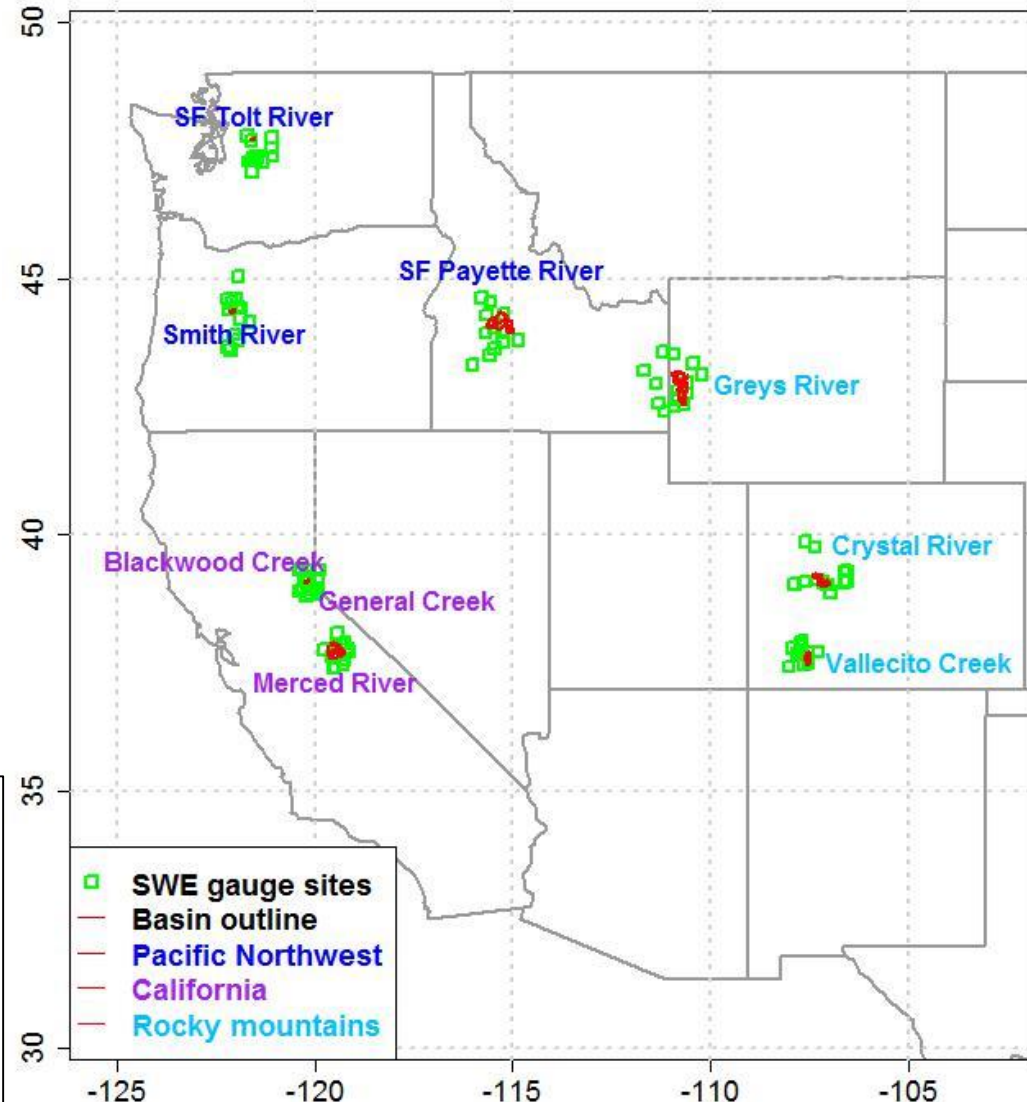
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|---------------------------|-----------|-----------|--------|
| get_ghcnd | 14:00:00 | 14:16:35 | |
| get_nwcc | 14:00:01 | 14:05:17 | |
| get_gefs | pending | pending | |
| get_cfsr | 14:00:01 | 14:02:13 | |
| get_flow | 14:00:01 | 14:01:03 | |
| reformat_ghcnd | 14:16:36 | 14:33:58 | |
| reformat_nwcc | 14:33:59 | 14:34:12 | |
| QC_stn_data | 14:34:12 | 14:38:30 | |
| fill_stn_data_pass1 | 14:38:31 | 15:13:56 | |
| fill_stn_data_pass2 | 15:13:56 | pending | |
| fill_stn_data_pass3 | pending | pending | |
| fill_stn_data_pass4 | pending | pending | |
| gen_ens | pending | pending | |
| grid2poly | pending | pending | |
| make_nws_forc | pending | pending | |
| run_nws_spinup | pending | pending | |
| downscale_gefs_fcst | pending | pending | |
| downscale_gefs_fcst_regr | pending | pending | |
| reformat_gard_output | pending | pending | |
| reformat_gard_output_regr | pending | pending | |
| met_forecast_grid2poly | pending | pending | |
| make_nws_met_forecast | pending | pending | |
| run_nws_gefs_fcst | pending | pending | |
| plot_stn_data_map | pending | pending | |
| plot_mr_fcst | pending | pending | |



Hydrologic DA for seasonal prediction

- SWE measurements (snotel) can be used objectively to update hydrologic model states and improve forecasts
- Using Ensemble Kalman Filter (EnKF)
- Hindcast-based study
- Huang et al, 2016 (HESS)

Position of 9 case basins and SWE gauge sites

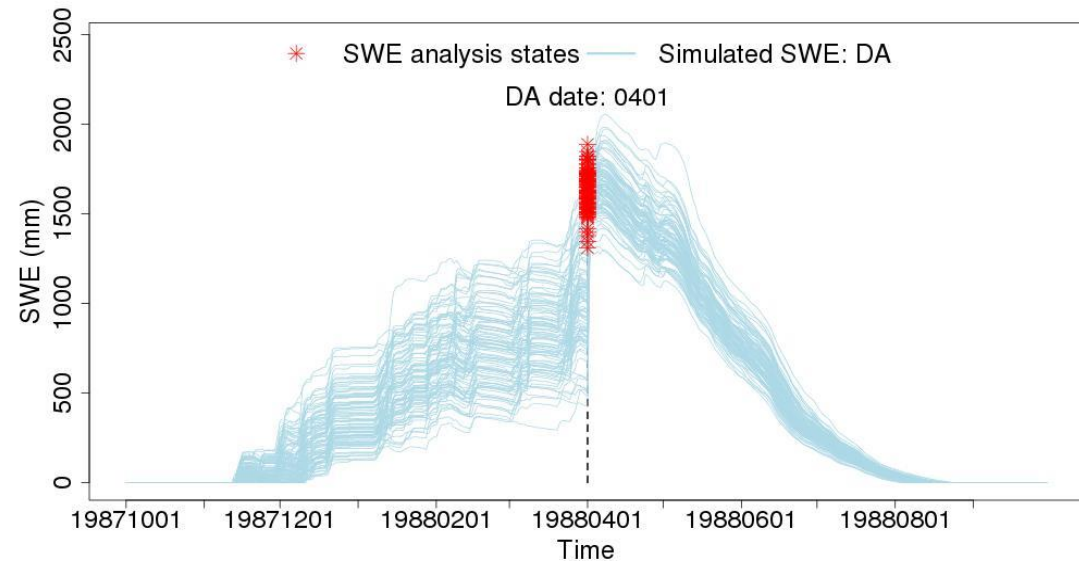
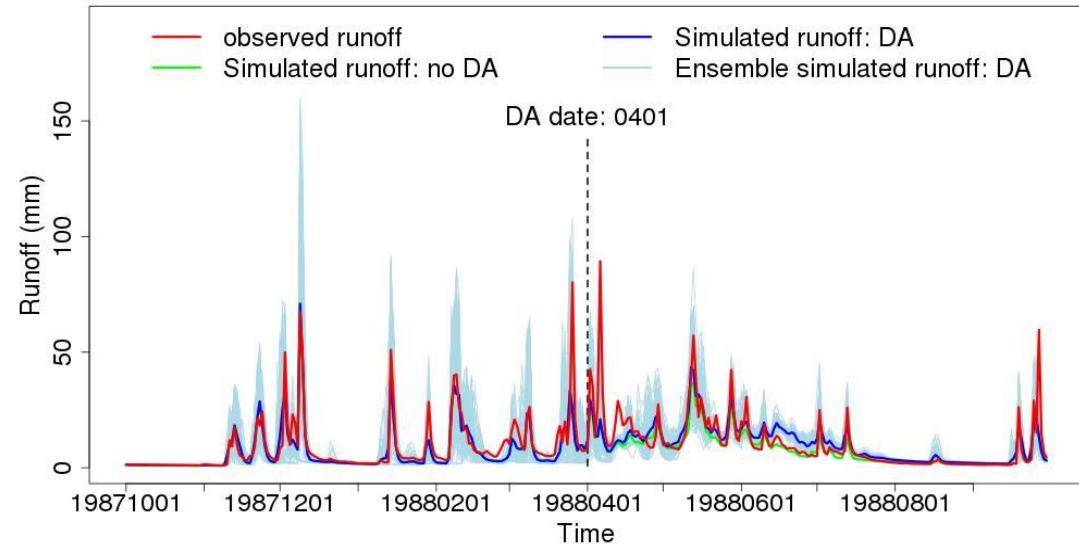


Hydrologic DA for seasonal prediction

Example

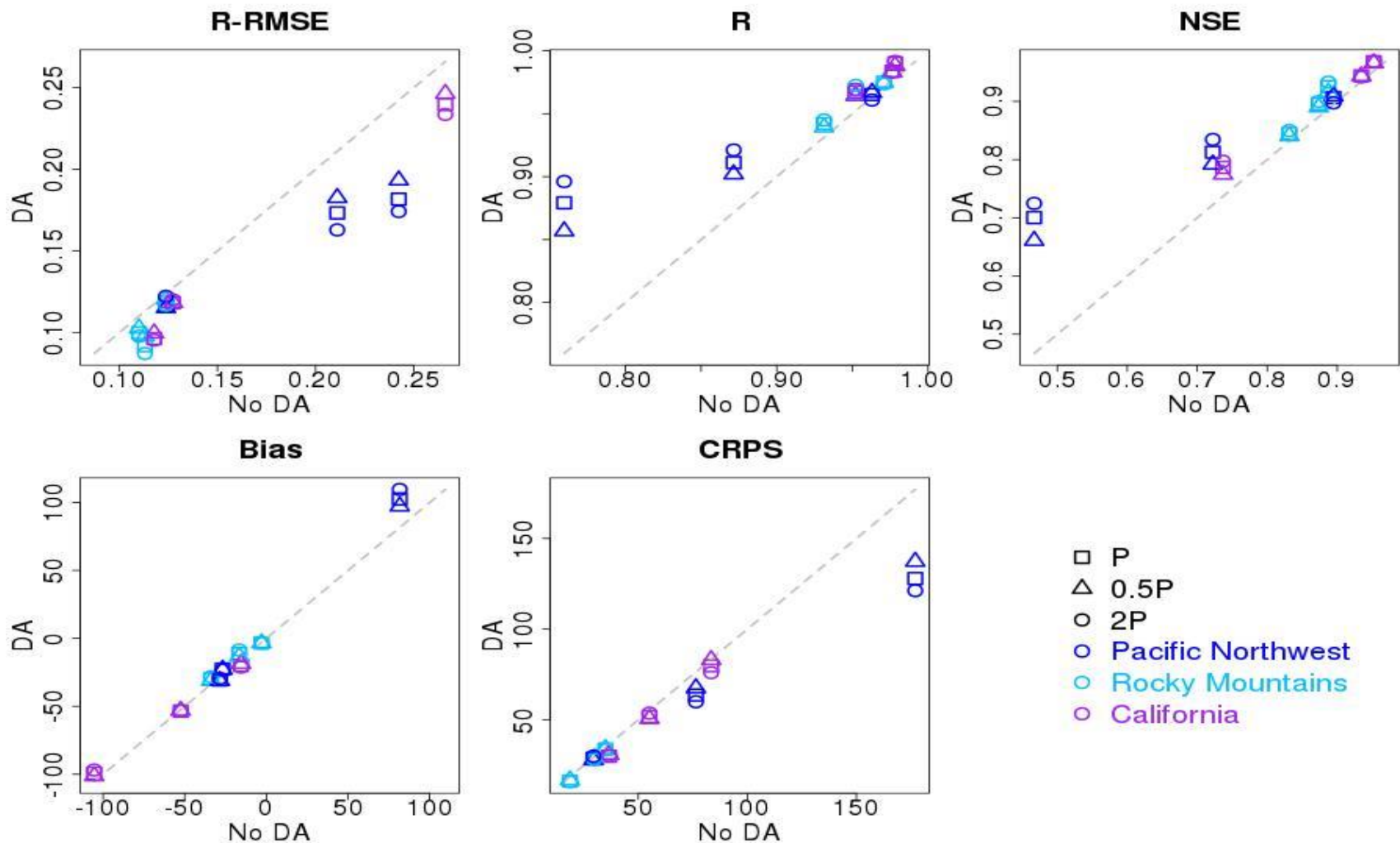
- Use an ensemble method to estimate initial conditions
- Update those conditions with SWE observations
- Make ensemble predictions from mean model states
- Assess forecast skill after assimilation

Region: 17 Basin ID: 12147600 Name: SF Tolt River

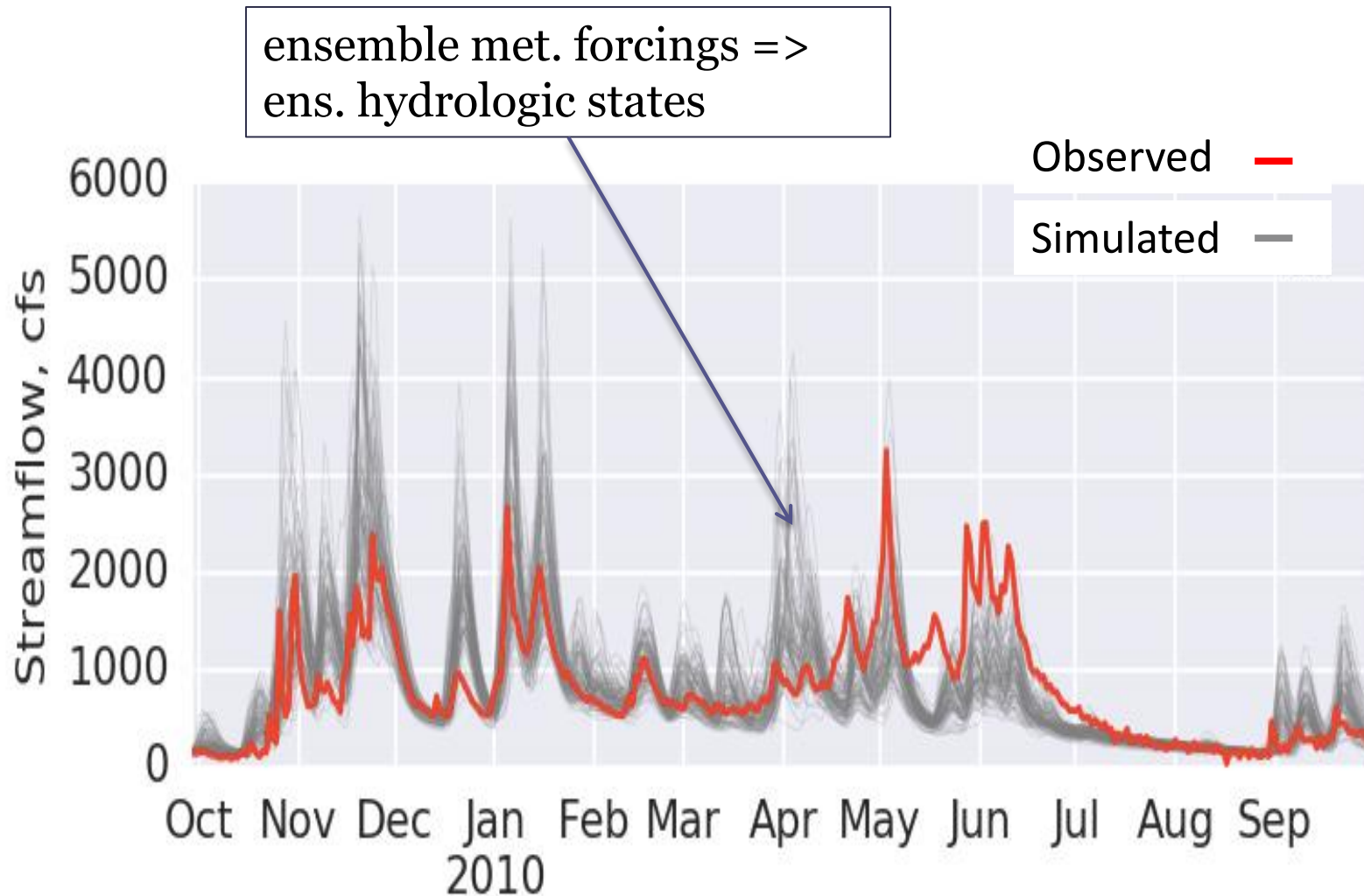


Hydrologic DA for seasonal prediction

- Evaluation metrics generally show improvements for April-July ESP mean streamflow forecast for the nine case basins.



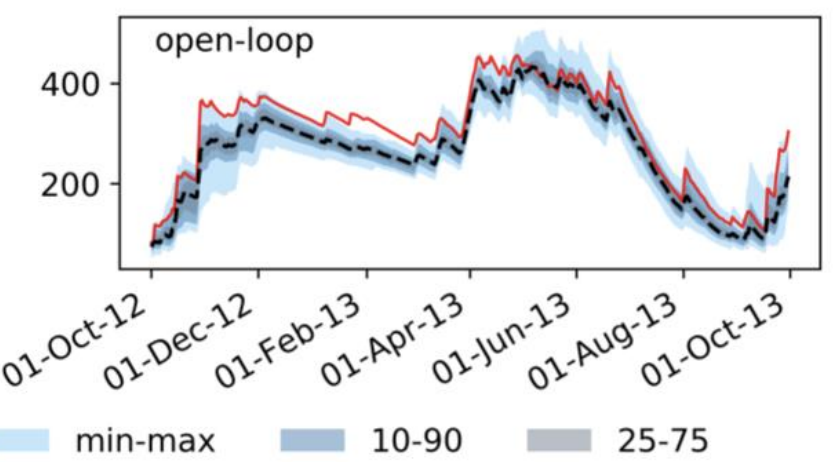
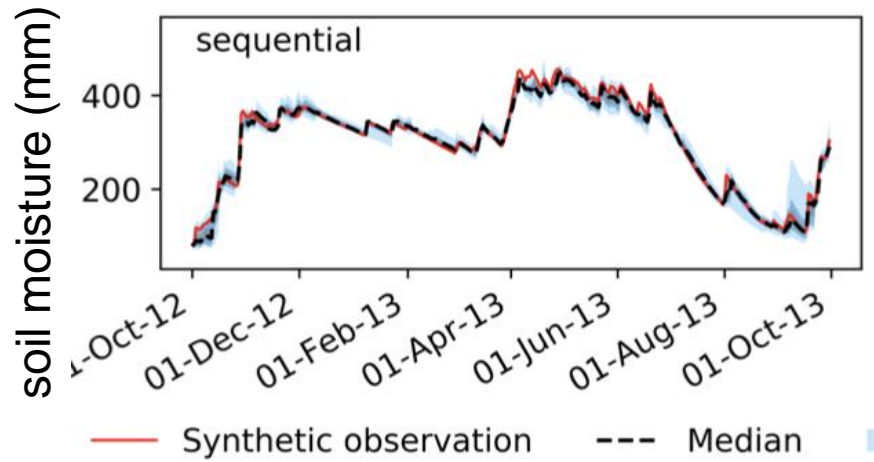
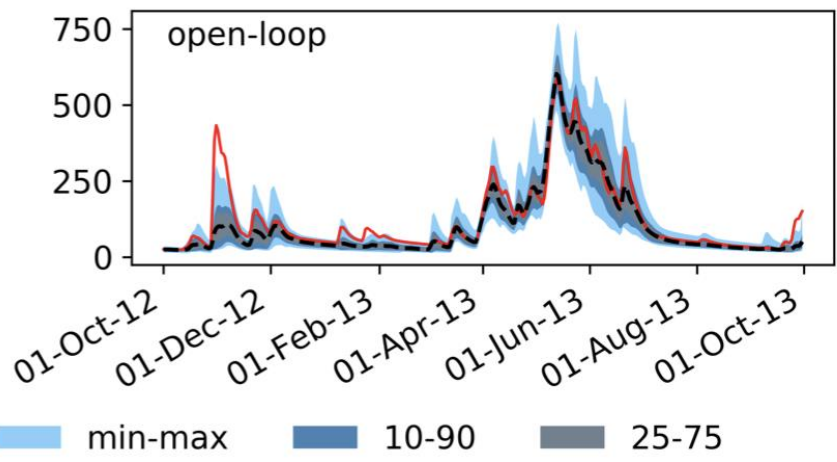
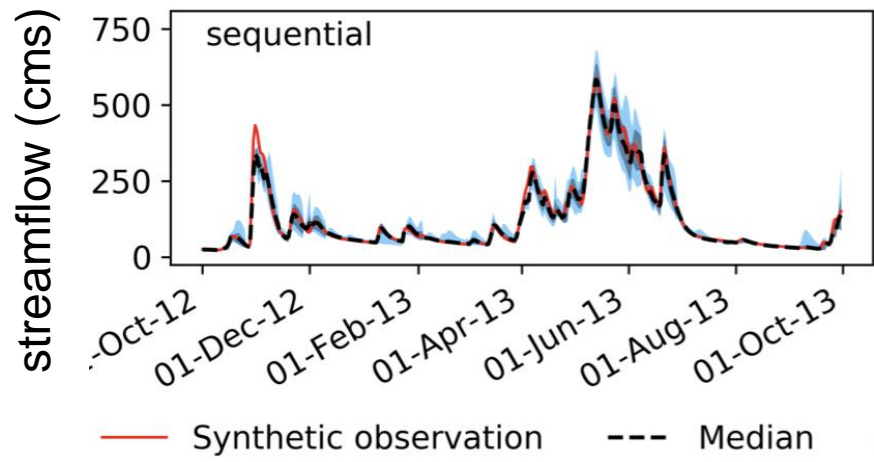
Hydrologic DA for short-range prediction



- Generate hydrologic states representing uncertainties to support data assimilation (and initialize forecasts)

Hydrologic DA for short-range prediction

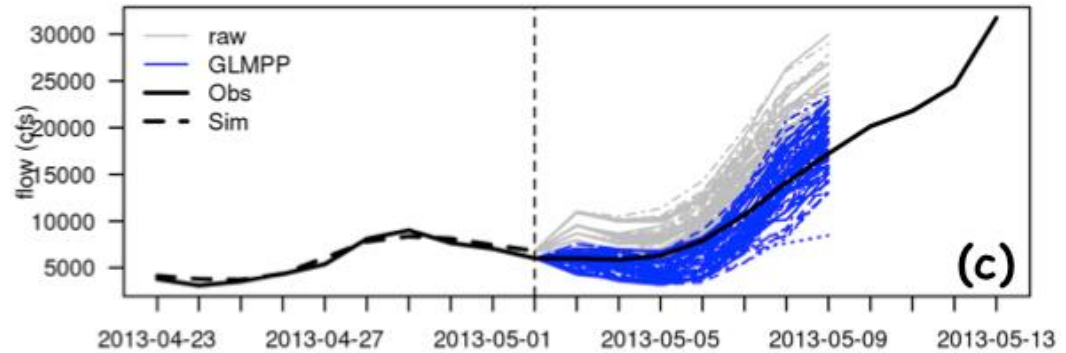
Particle Filter DA (Sequential Importance Resampling) improves both flow and SM



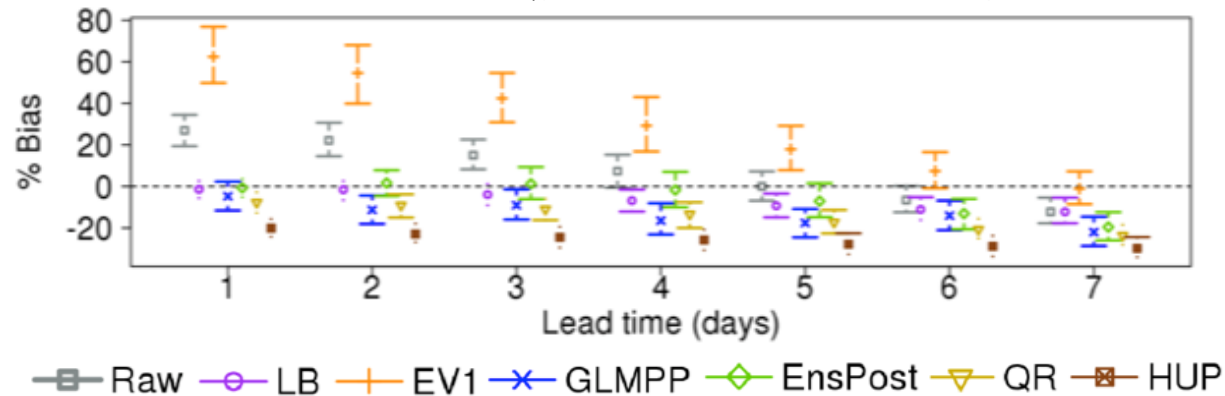
Post-processing is also critical

- reduces residual uncertainty after other parts of the process

| Method | Name |
|---------|---|
| LB | Linear blending |
| EV1 | Error in variable Model Output Statistics with one variable |
| GLMPP | Generalized Linear Model Post-Processor |
| EnsPost | Ensemble Post-Processor |
| QR | Quantile Regression |
| HUP | Hydrologic Uncertainty Processor |



hindcast based skill evaluation



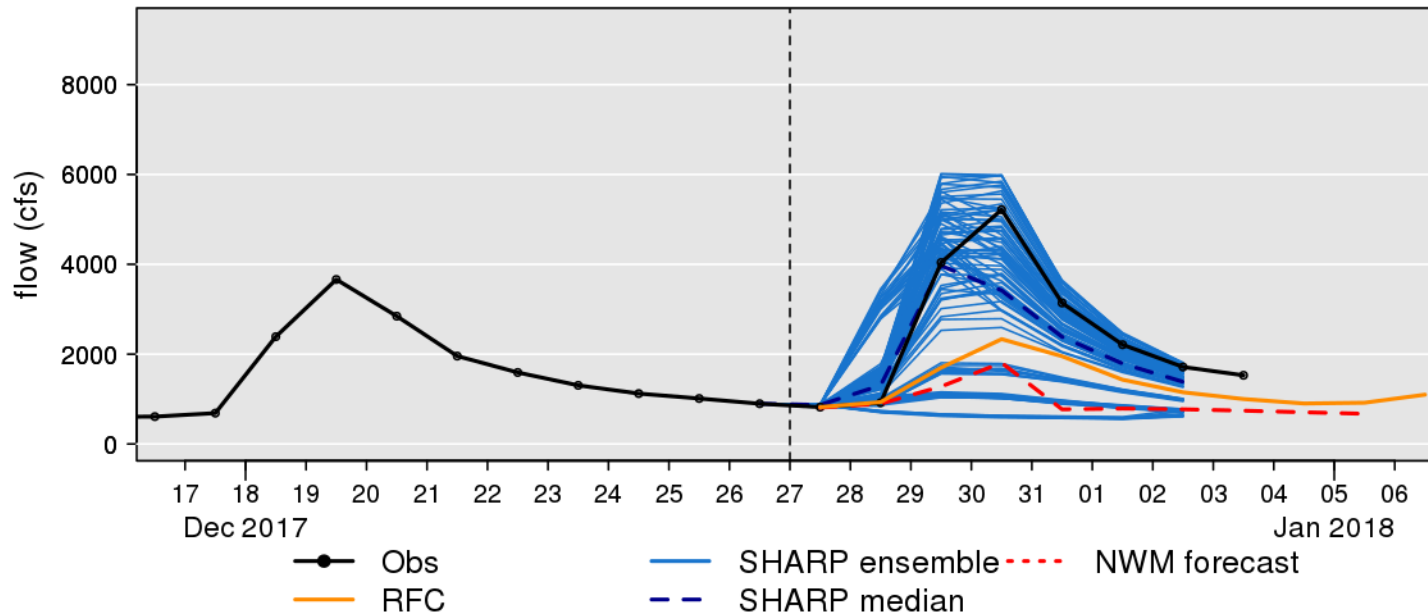
contribution by Pablo Mendoza

Require consistent hindcasts to implement

Hydrologic DA for short-range prediction

- Operational demonstration shows hydrologic DA is workable in the pilot watersheds
- Hydrologic DA can replace some of the manual adjustment (MODs) now made by human forecasters

Streamflow forecasts (daily mean) for Howard Hanson Reservoir Inflow WA (HHDW1)
 Initialized on Dec 27 2017



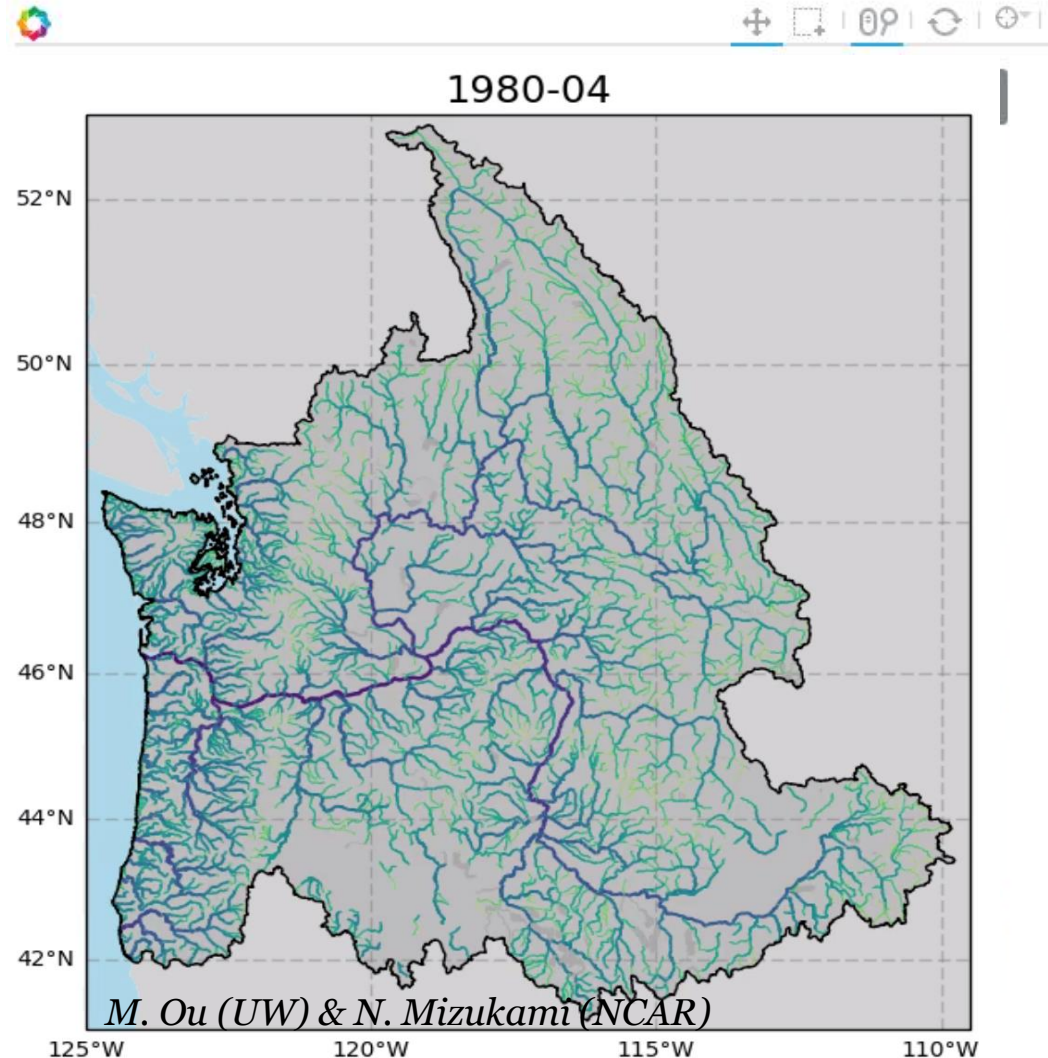
- Clark, E.A., A.W. Wood, and B. Nijssen, 2017: Assessing ensemble particle filters for the estimation of model states for streamflow forecasting, *Water Resources Research* (in review).
- Clark, E. A., A. W. Wood, B. Nijssen, and M. P. Clark. Implications of streamflow data assimilation via particle filter on streamflow forecasts in basins with seasonal snow. *Hydrology and Earth System Sciences*, in preparation.

Next - Prediction over large domains

- There is a need to re-connect hydrologic prediction science & uncertainty methods to large-domain prediction efforts

Scaling Challenges

- regional model calibration
- spatial coherence in downscaling and post-processing
- Spatial propagation of obs info in data assimilation
- understanding appropriate complexity of modeling
 - scale
 - physics
 - tradeoffs

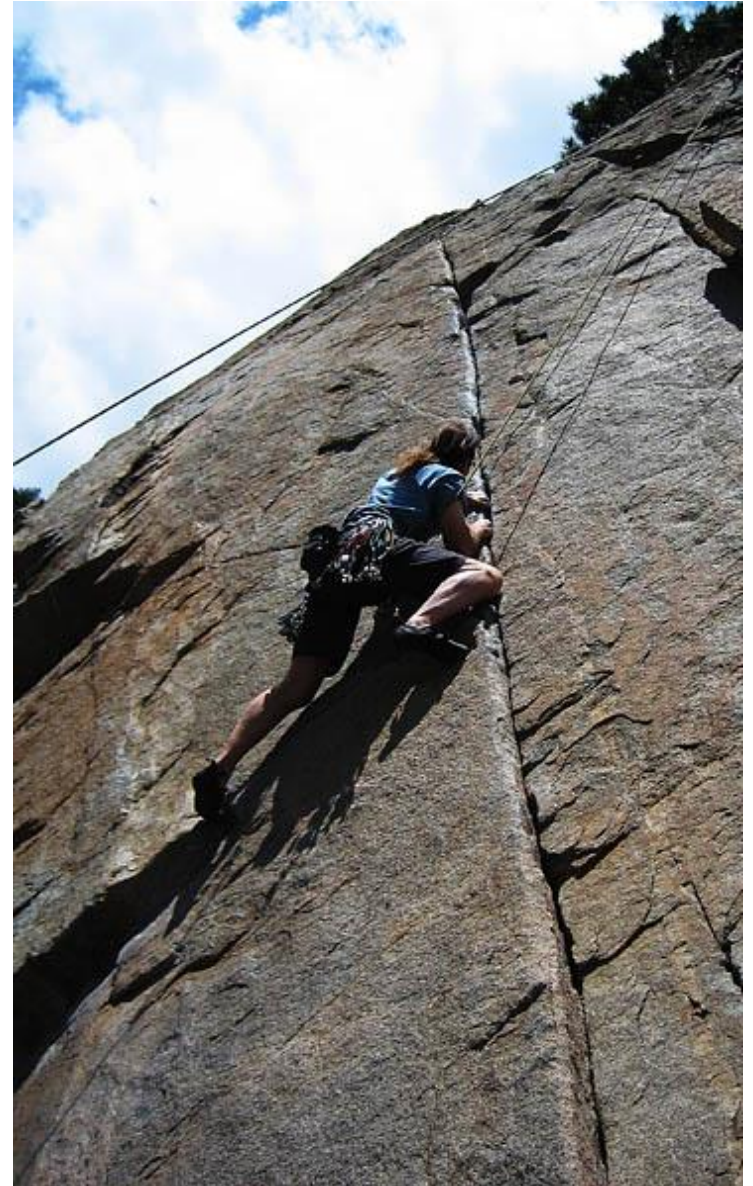


In-the-Loop

- Traditional In-the-Loop forecast paradigm

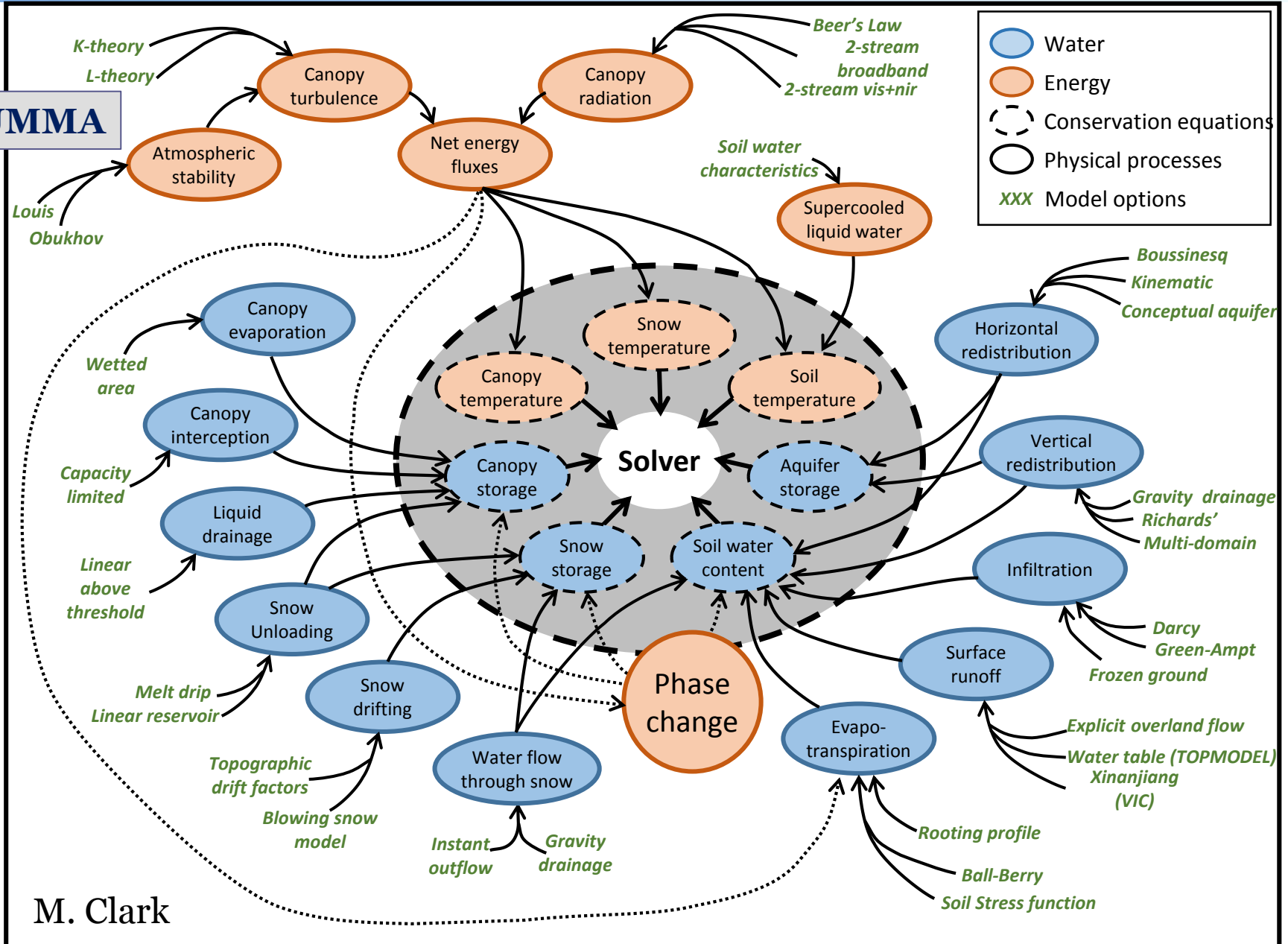
Over-the-Loop

- Medium-resolution / Ensemble / Uncoupled
- High-resolution / Deterministic / Uncoupled
- Intermediate-resolution / Ensemble / *Coupled*



Require agility: process modularity

SUMMA



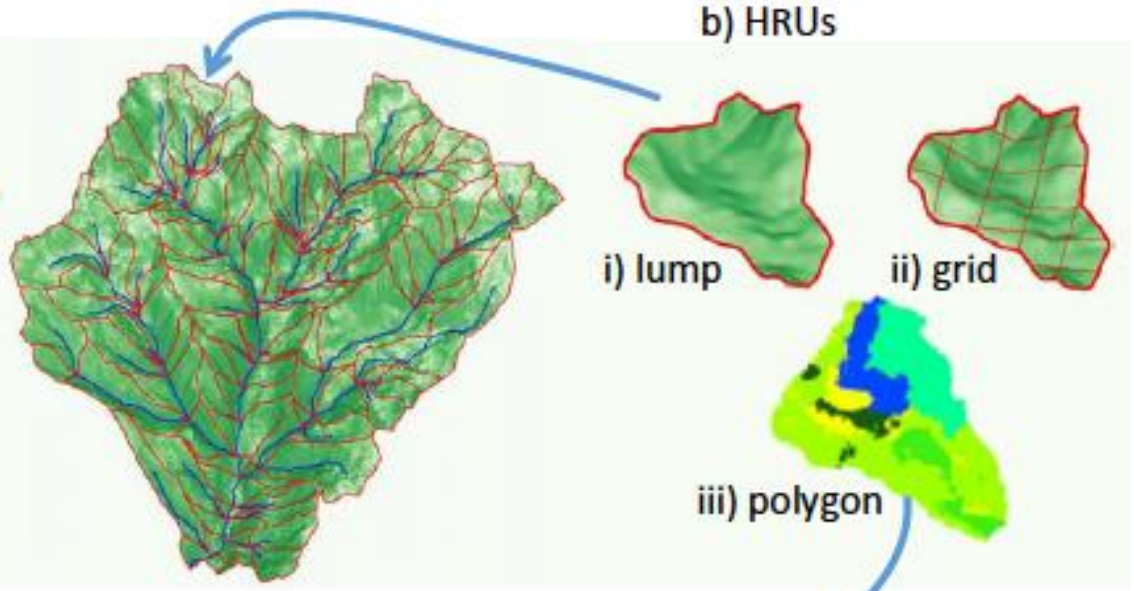
M. Clark

Require agility: spatial/coupling flexibility

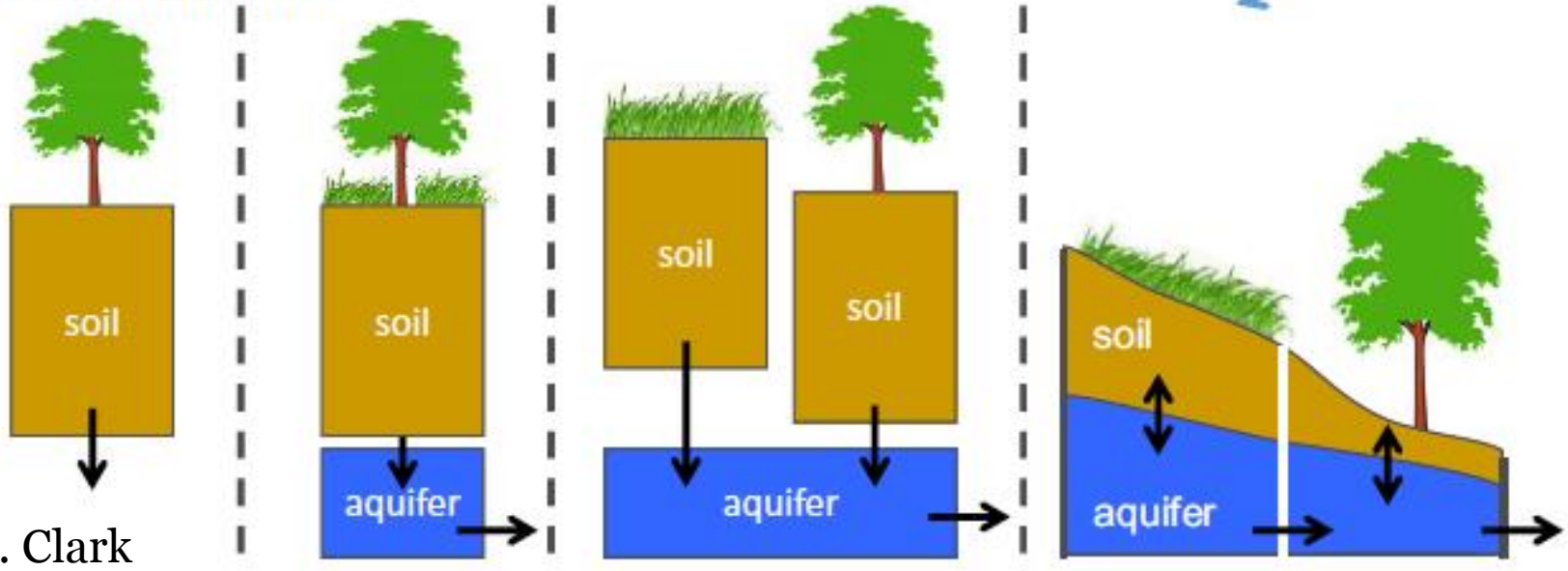
a) GRUs



b) HRUs

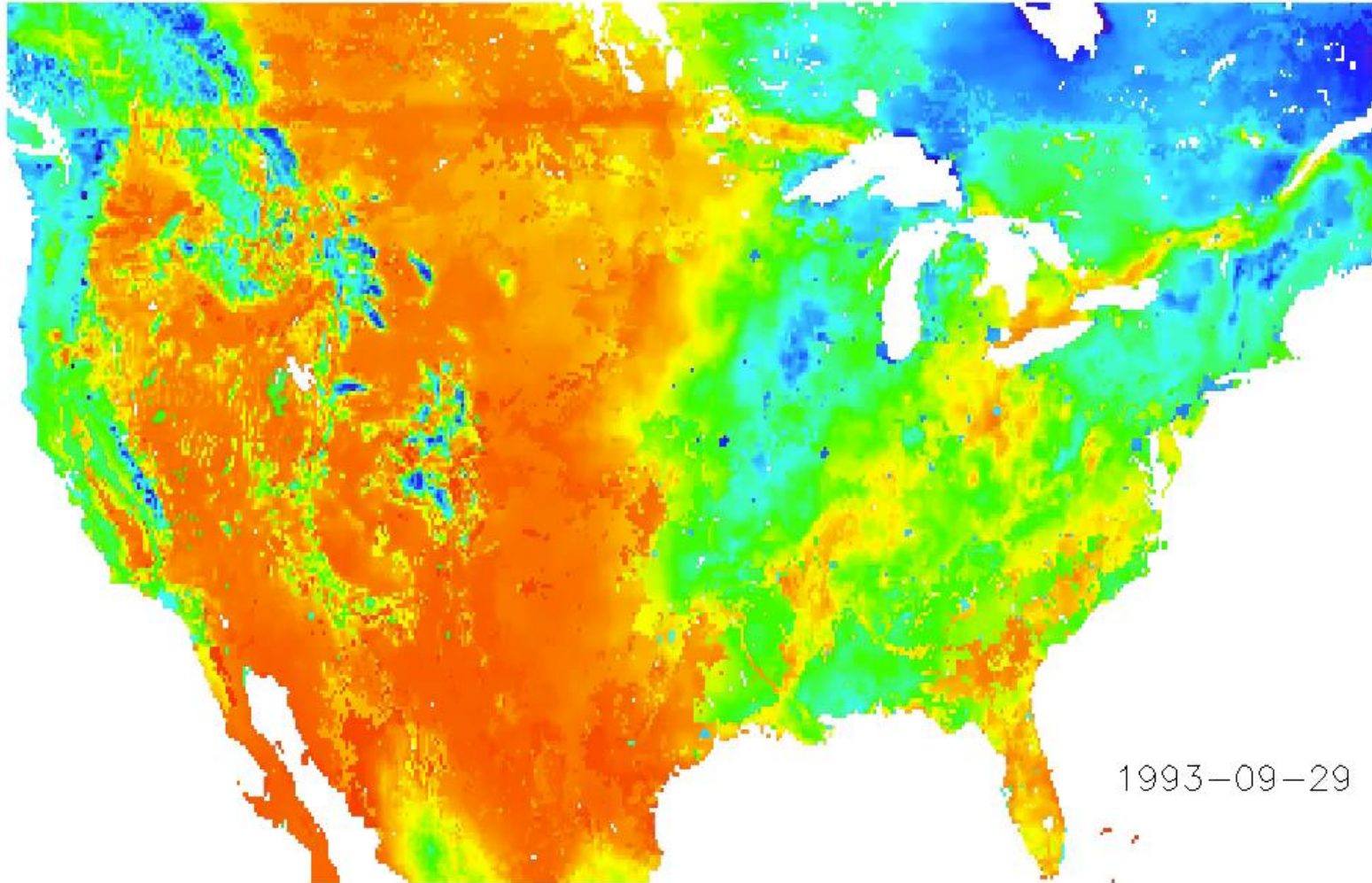


c) Column organization



SUMMA at 1/8th degree over US NLDAS domain

Soil water (mm) – 25 yr hourly simulation, 30 min on 8000 cores

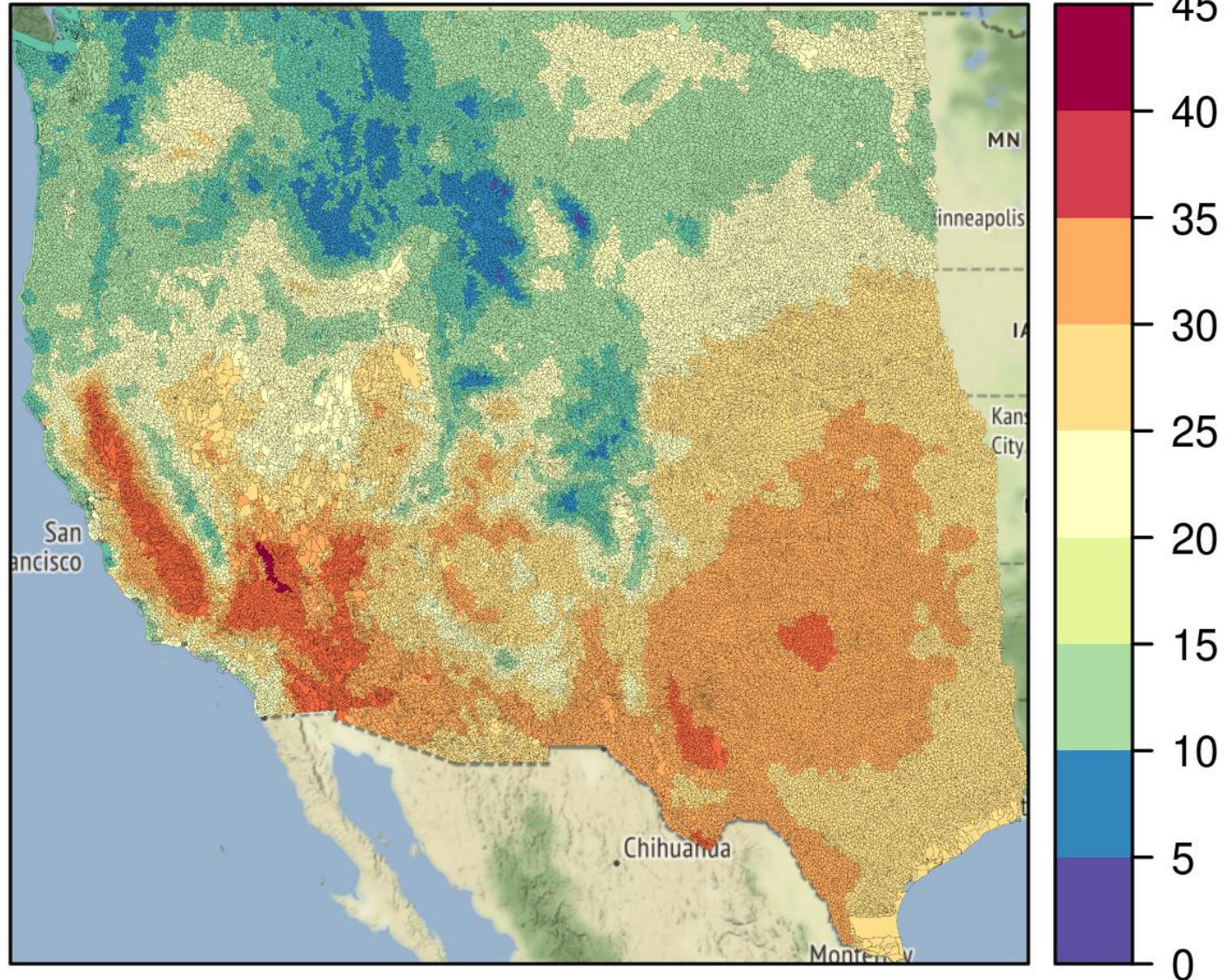


Focus on intermediate scale-modeling

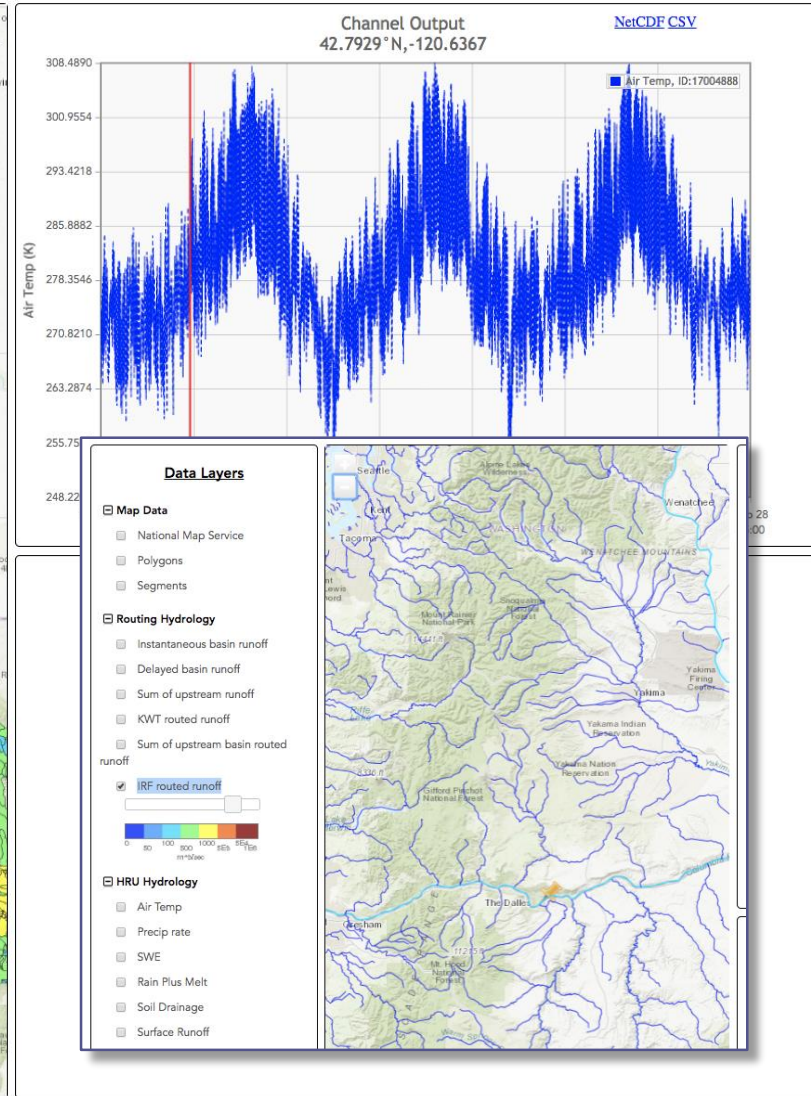
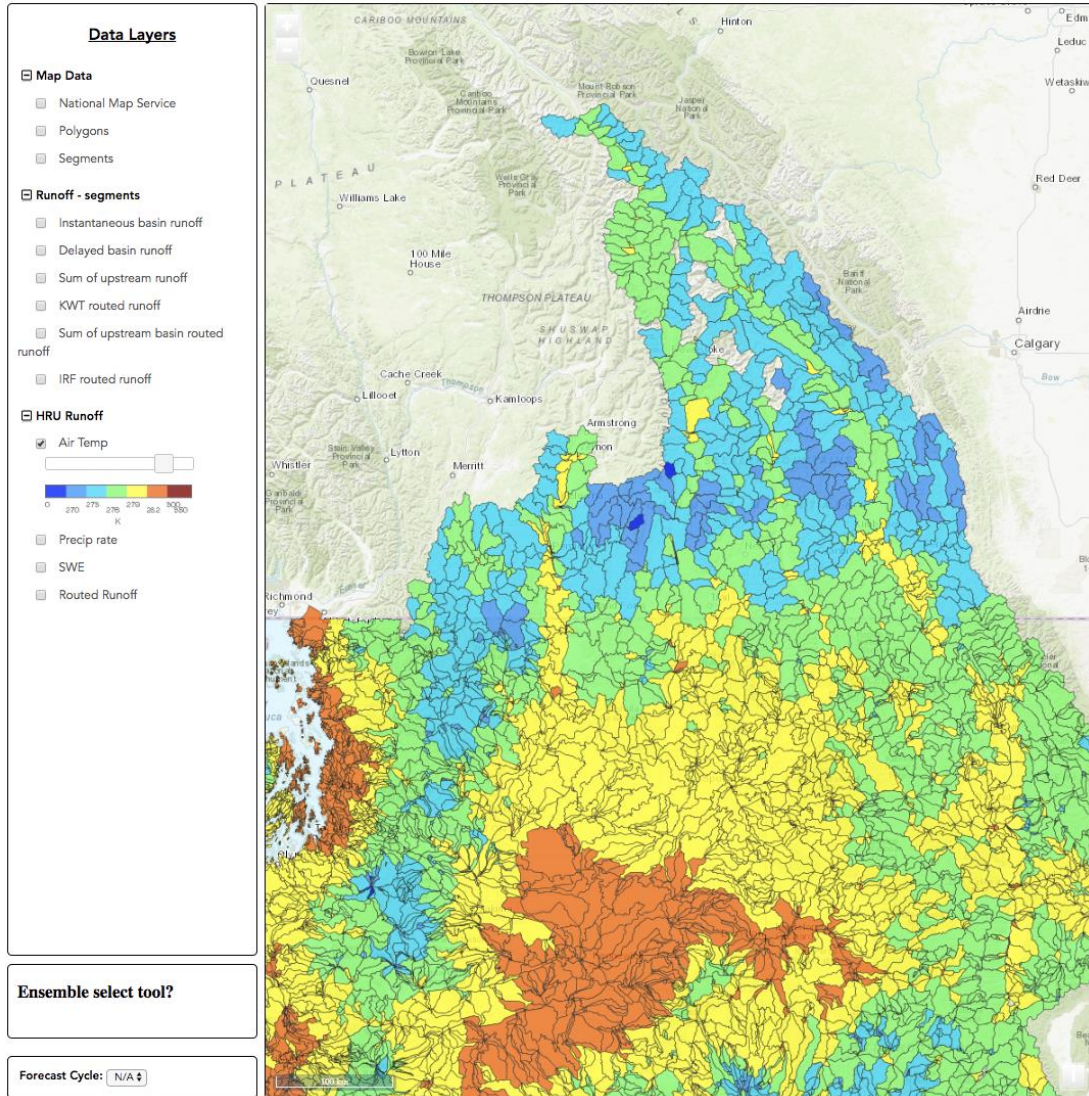
User-oriented domain

- Reclamation management area
- HUC12 scale simulations

Air Temperature (deg C) – 20140715



Focus on intermediate scale-modeling

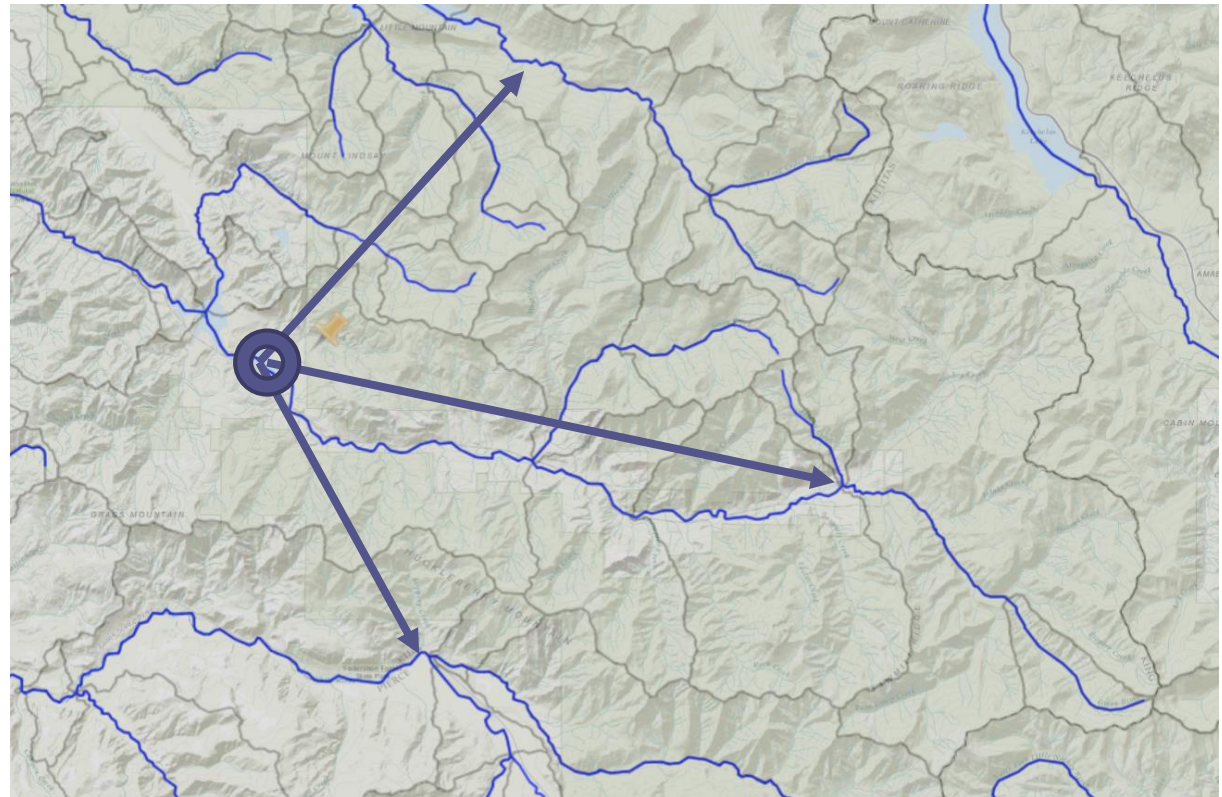


Challenges / Efforts

- Most past work on PF DA localization is not in hydrology, and not for streamflow (SWE DA localization is easier)
- We're beginning to investigate strategies for PF ensemble weighting across ungauged areas.

Next Steps

- Applying SUMMA in real-time, HUC12 scale (~100,000 watersheds)
- Ensemble inputs and outputs
- Model calibration and DA



A scenic landscape featuring a large body of water in the foreground, a forested hillside in the middle ground with a small building on top, and a range of mountains in the background under a clear sky. A dark horizontal band is overlaid across the middle of the image, containing the text "Thank You".

Thank You

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