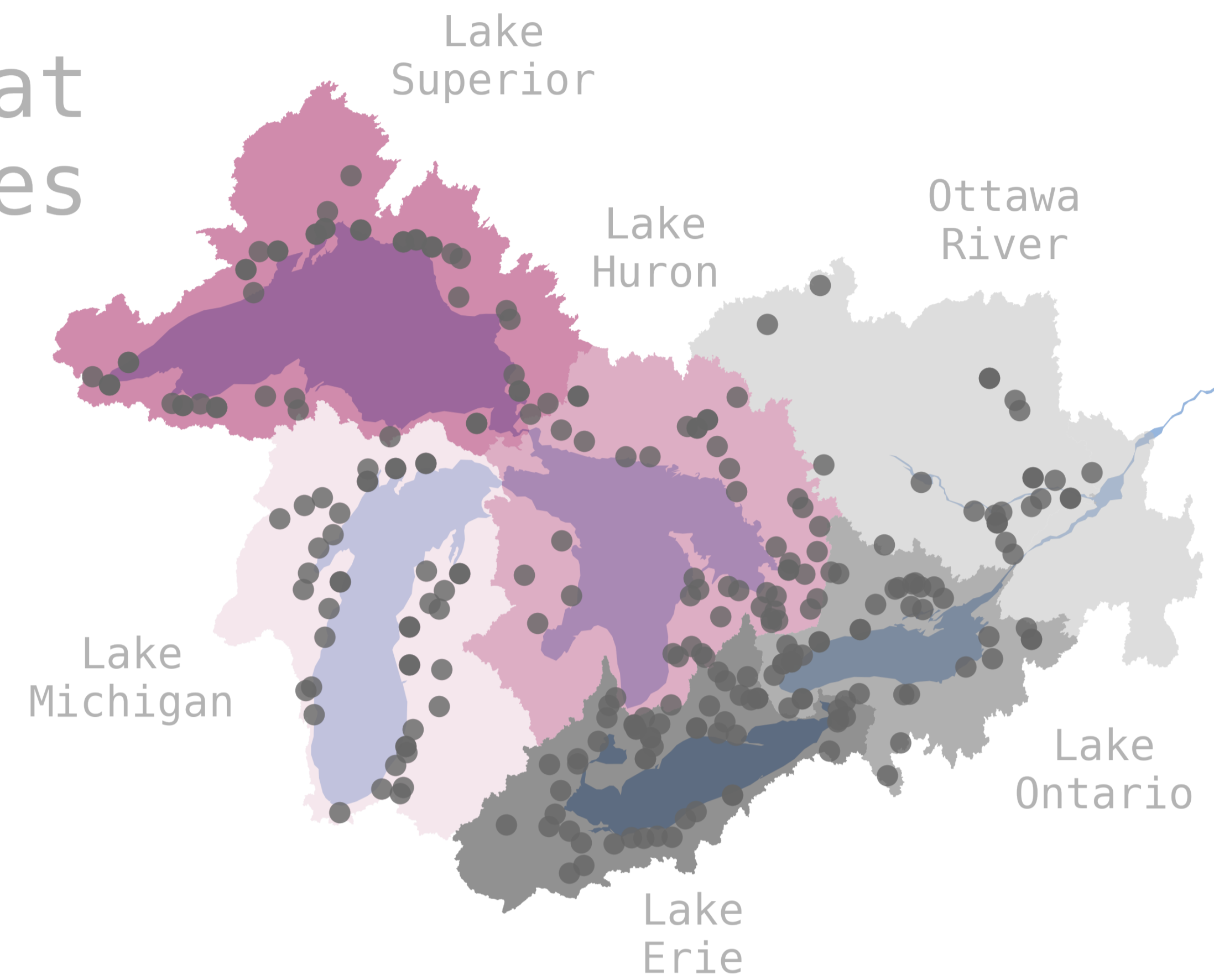


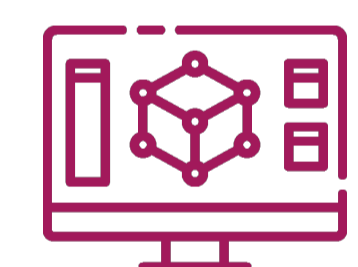
# Great Lakes Runoff Intercomparison Project

## Great Lakes

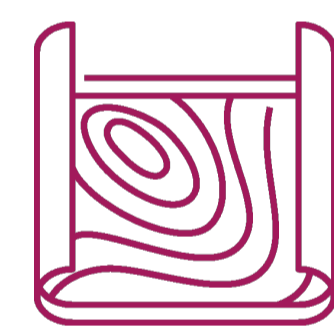


Mai, J., H. Shen, B.A. Tolson, É. Gaborit, R. Arsenault, J.R. Craig, V. Fortin, L.M. Fry, M. Gauch, D. Klotz, F. Kratzert, N. O'Brien, D.G. Princz, S. Rasiya Koya, T. Roy, F. Seglenieks, N.K. Shrestha, A.G.T. Temgoua, V. Vionnet, J.W. Waddell

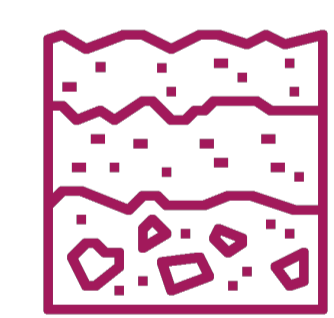
### What we did and used



Models: LSTM, LBRM, HYMOD2, GR4J, HMETS, Blended-lp, Blended-sd, VIC, SWAT, WATFLOOD, MESH-CLASS, MESH-SVS, GEM-Hydro



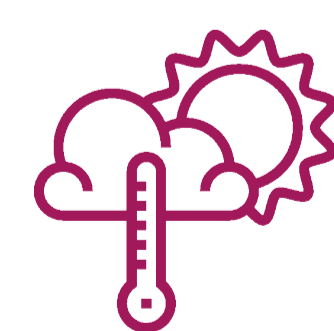
HydroSHEDS 90 m (3") digital elevation model (DEM)



Global Soil Dataset for Earth System Models (GSDE)



North American Land Change Monitoring System (NALCMS)



Regional Deterministic Reanalysis System (RDRS) v2



Common routing product and routing approach



Common calibration period (2001-2010) and evaluation period (2011-2017)

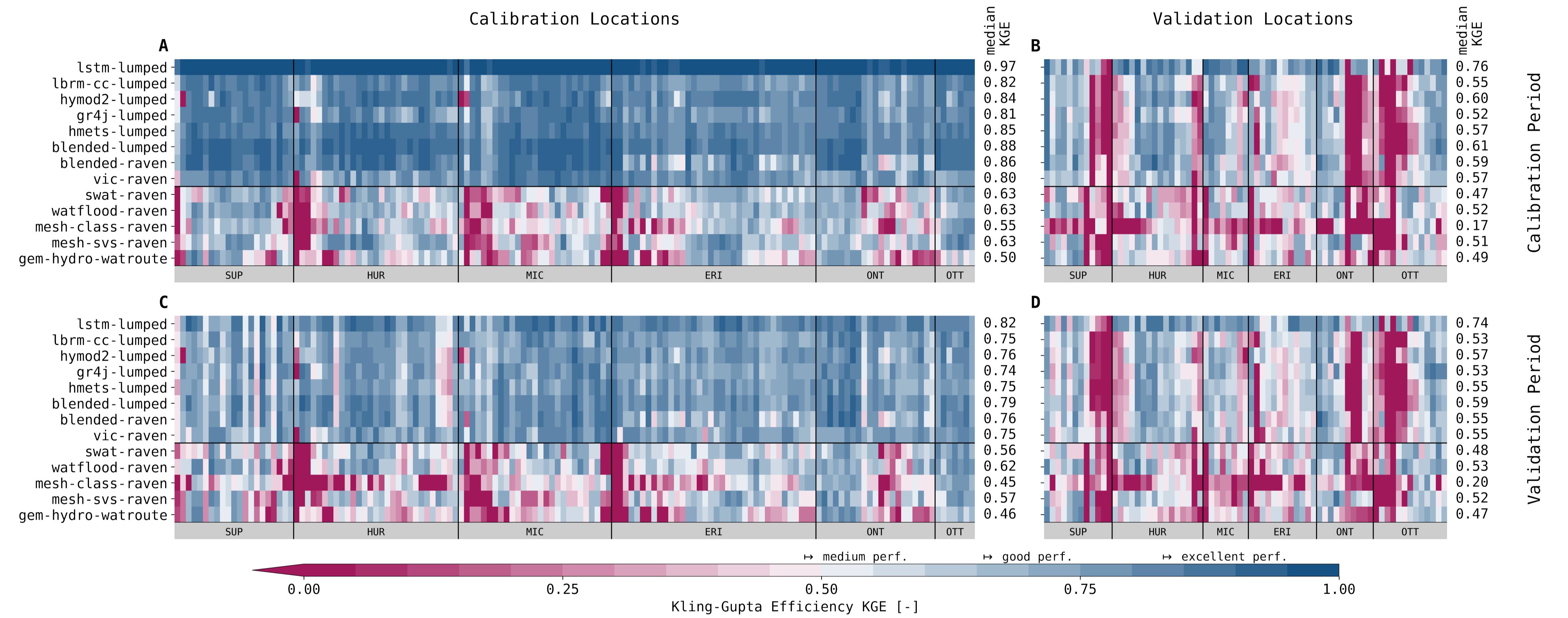


Common calibration stations (141) and evaluation stations (71)

# We found that the Machine Learning-based model outperformed any other physically-based model regarding simulation of streamflow.

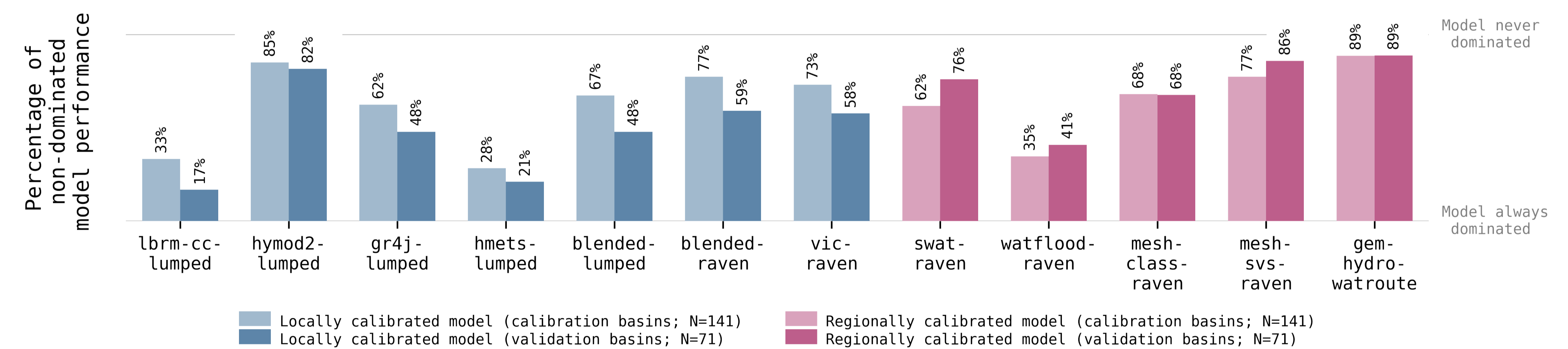
### Basin-wise comparison of streamflow

reveals superior performance of machine-learning based model in calibration (A) and all validation experiments (B-D). Locally calibrated models transfer well in time (A vs C) while spatial transfer shows potential for improvement (A vs B and C vs D). On average, regionally calibrated models are the one with least performance.



### Multi-objective multi-variable comparison of streamflow and additional variables

identifies overall well-performing locally calibrated models (i.e., HYMOD2-lumped) and regionally calibrated models (i.e., MESH-SVS-Raven and GEM-Hydro-Watroute) due to varying reasons. The machine-learning-based model is not included here as it is not set up to simulate the additional variables such as AET, SSM, and SWE.



# This highly standardized model comparison led to a plenitude of insights on model performance regarding streamflow and beyond.

