

The background of the entire slide is a high-speed photograph of water splashing, creating a dynamic pattern of droplets and ripples in various shades of blue.

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# Exploring Impacts of Model Calibration on Regionalization within the Next Generation Water Resources Modeling Framework (NextGen)



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## Motivation

Regional/continental hydrologic modeling depends on **proper regionalization** to identify model and parameters for uncalibrated areas, which in turn depends on **robust calibration**

## Methodology

Run and evaluate different **calibration/regionalization experiments within NextGen** (in terms of model formulation, objective function, calibration/regionalization algorithm etc.)

## Message to take home

NextGen's **heterogeneous** process representation offers flexibility in **model and parameter identification at catchment level** to achieve best results

# The Next Generation Water Resources Modeling Framework (NextGen)

- ❑ Features a **model-agnostic** modular modeling environment with a plug-and-play interface
- ❑ Incorporates **multiple modules** for each hydrologic process
- ❑ Supports **heterogeneous** process representation
- ❑ Facilitates collaborative **community** development



Multiple model  
calibration &  
regionalization  
experiments

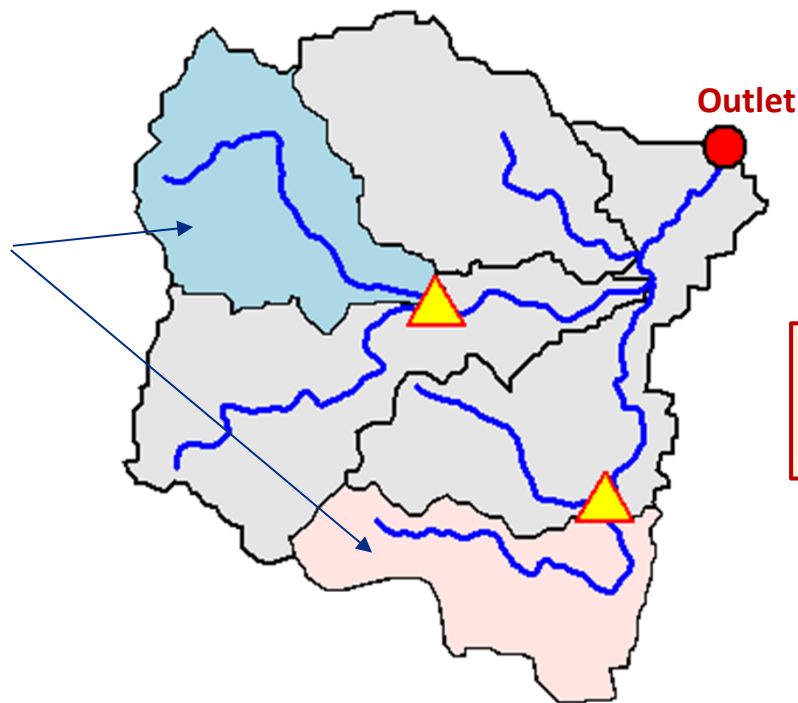




## Model/Parameter Identification with NextGen



2 calibratable  
catchments  
2 USGS gages

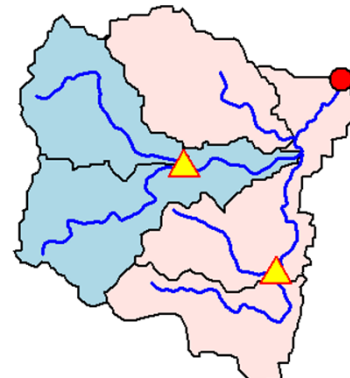
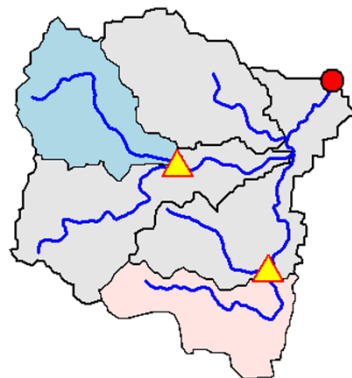


Question: How to identify appropriate model and parameters for each catchment (in order to simulate flow at the outlet)?

# Model/Parameter Identification with NextGen – cont.

## STEP1 : Calibration

In gaged catchments, identify a model and optimize parameters so model simulates observations well



## STEP2: Regionalization

From the calibrated catchments, identify the best donor of model/parameters for each uncalibrated catchment (receiver) based on physical & spatial similarities

**Calibration**  
Configurable  
Components

Model

Parameters

Objective function

Optimization algorithm

Physical attributes

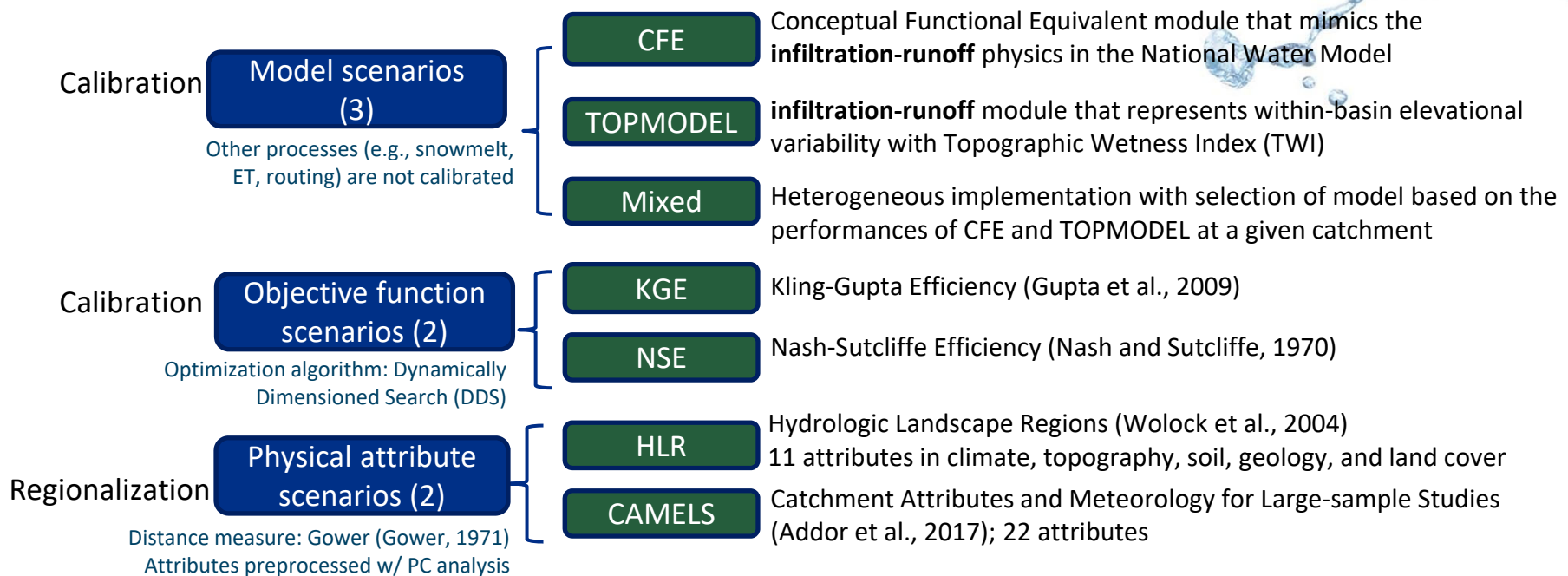
Distance measure

Donor-receiver pairing  
algorithm

**Regionalization**  
Configurable  
components



# Calibration & Regionalization Scenarios



Total number of scenarios:  $3 \times 2 \times 2 = 12$

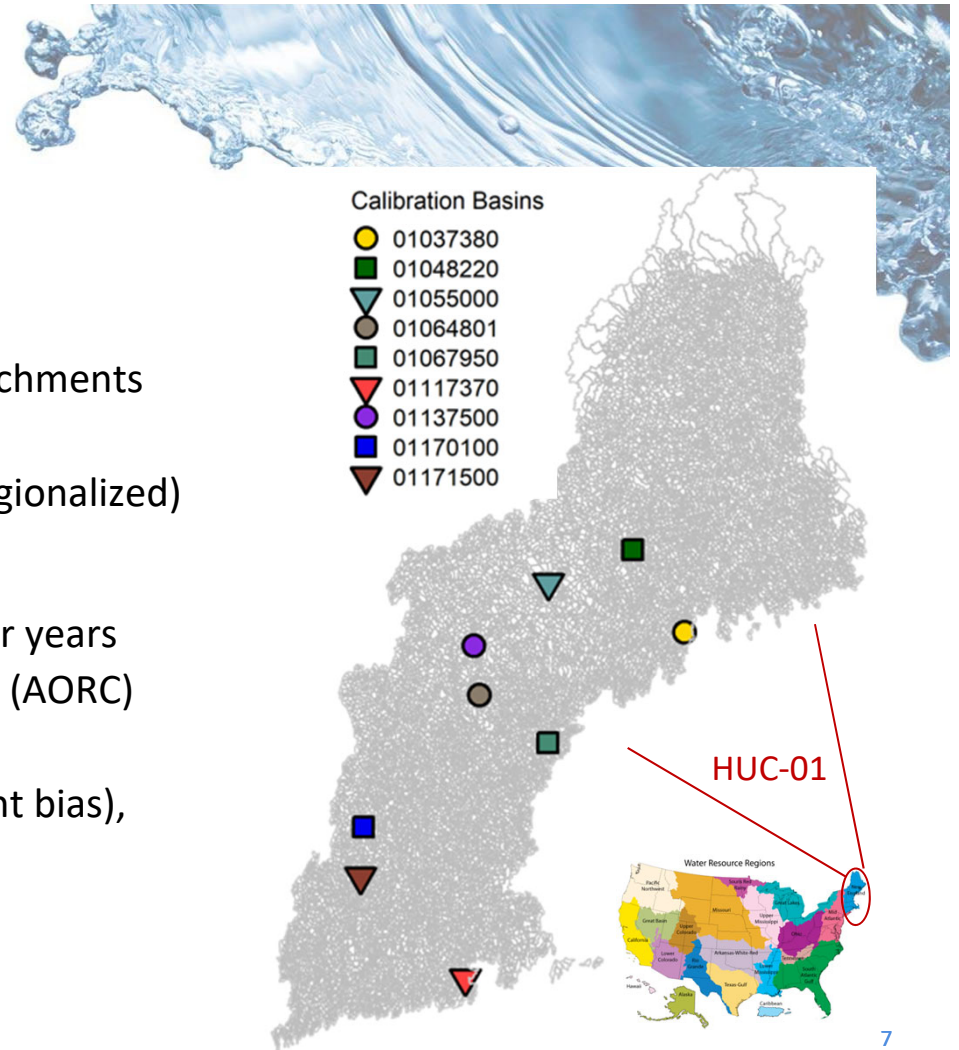
# Experimental Setup

## Study area

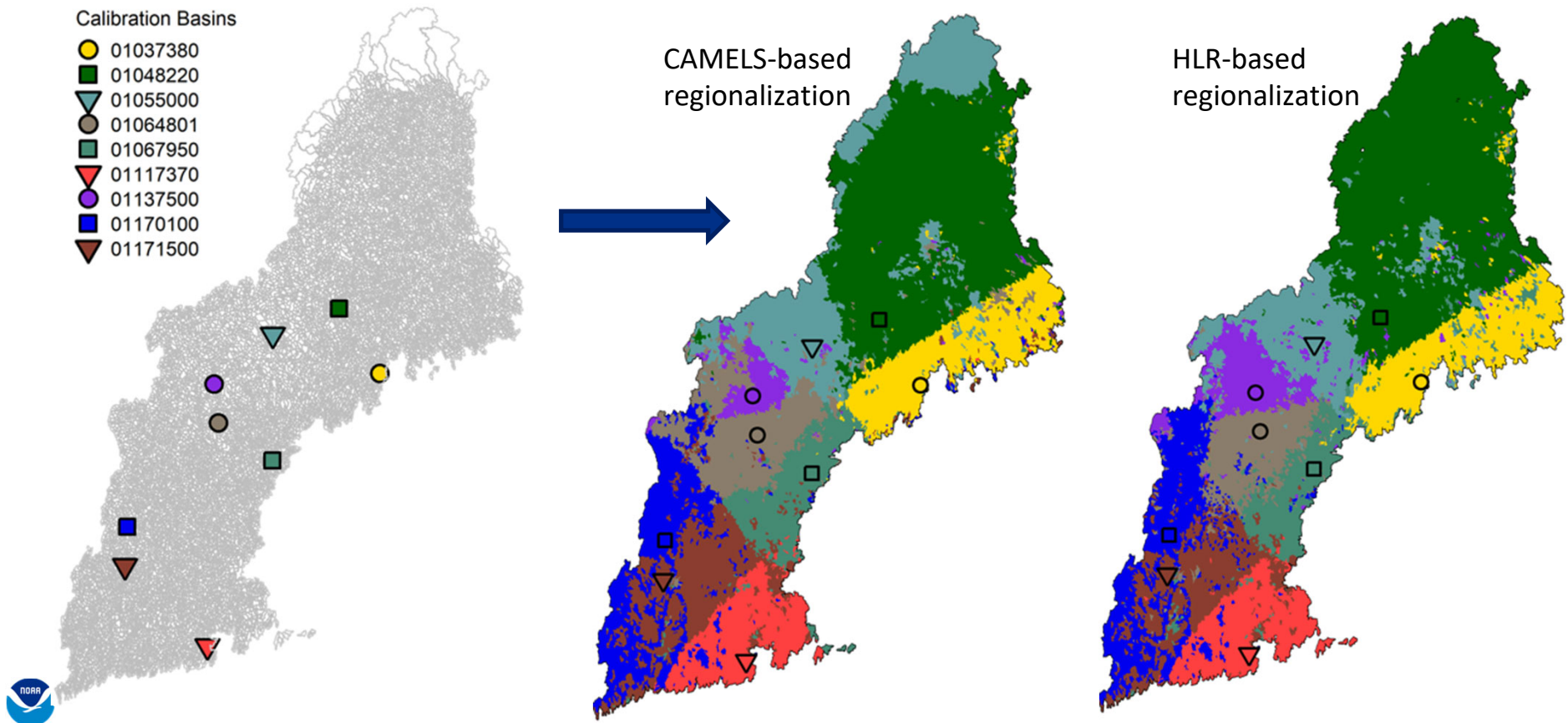
- New England region (HUC-01); 14,632 catchments
- 9 calibration basins (122 catchments)
- 239 validation basins (uncalibrated but regionalized)

## Evaluation

- Period: 10/01/2013 – 09/30/2016, 3 water years
- Forcing: Analysis of Record for Calibration (AORC)
- Validation data: USGS hourly streamflow
- Metrics: KGE, NSE, PBIAS (absolute percent bias), CORR (Pearson correlation)



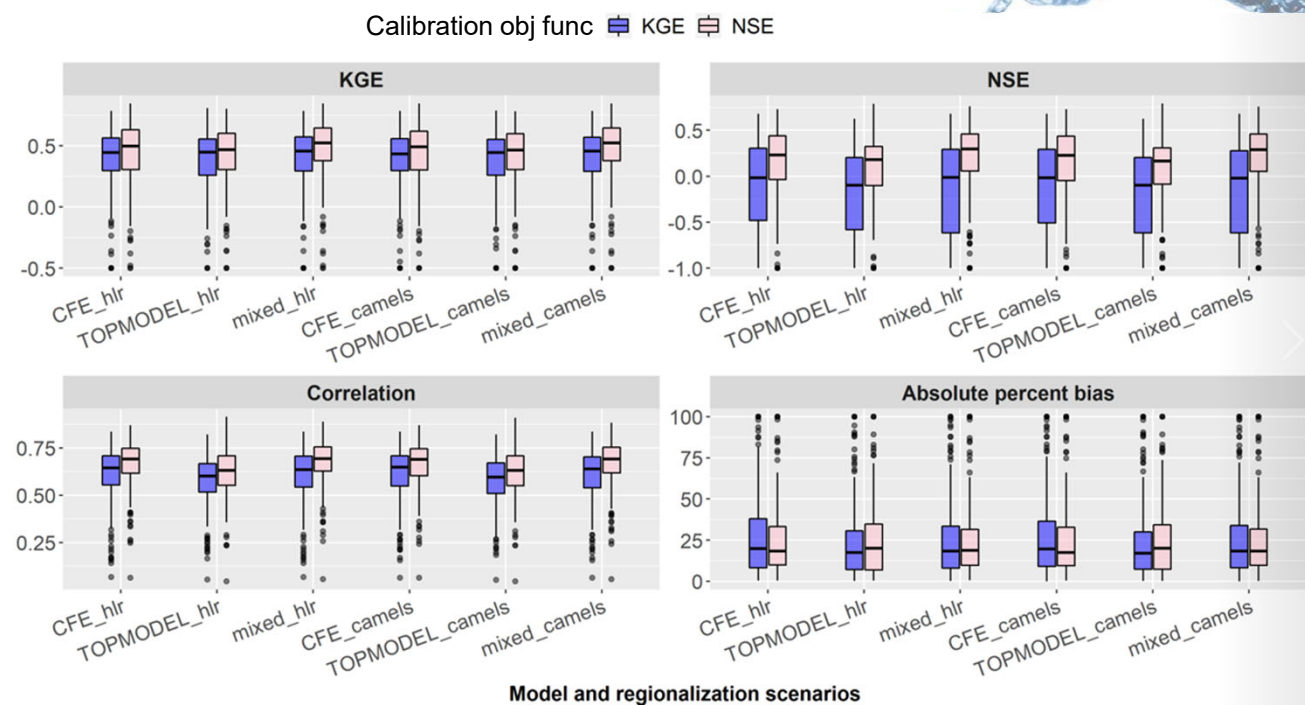
## Donor-Receiver Pairing from Regionalization (based on **Physical & Spatial Similarities**)





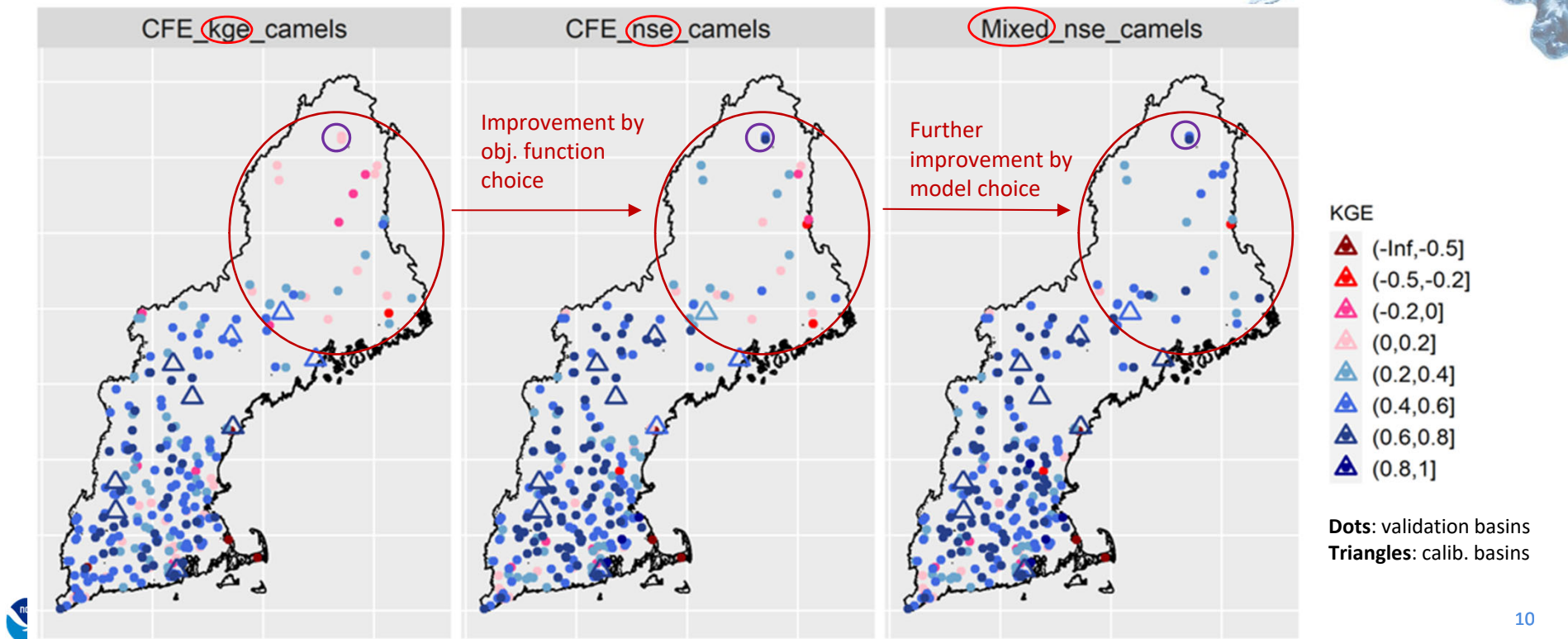
# Streamflow Results at Validation Basins: Comparison by **Objective Function**

- ❑ NSE-based calibration outperforms KGE-based calibration at validation basins
- ❑ The mixed (heterogeneous) model generally outperforms uniform implementations of CFE and TOPMODEL
- ❑ The two regionalization scenarios (HLR vs CAMELS) perform very similarly in all cases



# Streamflow Results at Validation Basins: Spatial distribution

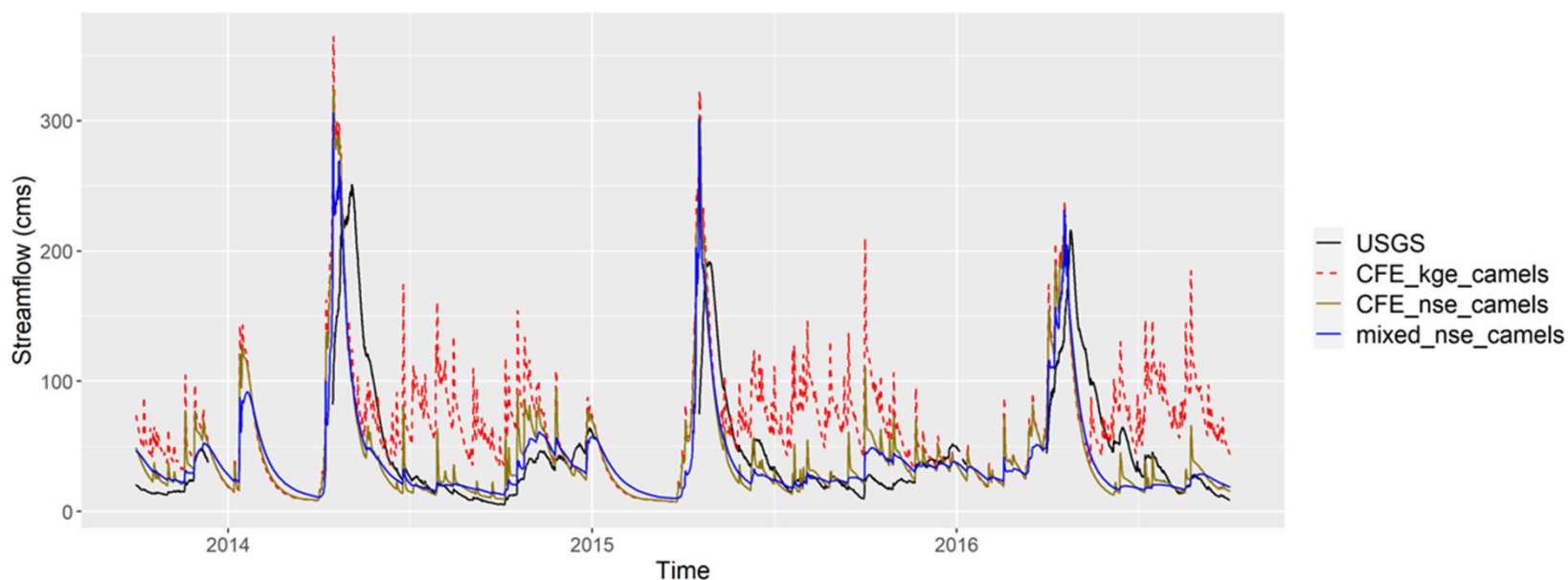
## Kling-Gupta Efficiency (KGE)



## Validation at 01013500 (Fish River near Fort Kent, Maine)

scenario	model	calib_obj_func	reg_scenario	KGE	NSE	CORR	PBIAS
CFE_kge_camels	CFE	kge	camels	0.07	-0.30	0.51	64.2
CFE_nse_camels	CFE	nse	camels	0.69	0.41	0.69	6.1
mixed_nse_camels	mixed	nse	camels	0.73	0.56	0.76	11.5

Flow Observation vs. Simulations





## Take-home Messages & Future Work

- ❑ NextGen's heterogeneous process representation offers flexibility in optimal model and parameter identification at catchment level
- ❑ Proper calibration can play a critical role in regionalization studies and hence should not be discounted
- ❑ Future work
  - ❑ Expand to other regions and contiguous US
  - ❑ Enhance NextGen calibration/regionalization capabilities
  - ❑ Explore other calibration/regionalization options





## References

- ❑ Addor, N., A.J. Newman, N. Mizukami, and M.P. Clark, 2017. The CAMELS data set: catchment attributes and meteorology for large-sample studies. *Hydrol. and Earth Syst. Sci.*, **21**, 5293-5313, doi:[10.5194/hess-21-5293-2017](https://doi.org/10.5194/hess-21-5293-2017)
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- ❑ J.E. Nash and J.V. Sutcliffe, 1970. River flow forecasting through. Part I. A conceptual models discussion of principles. *Journal of Hydrology*, **10**, 282-290.



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*Thank  
You!*

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