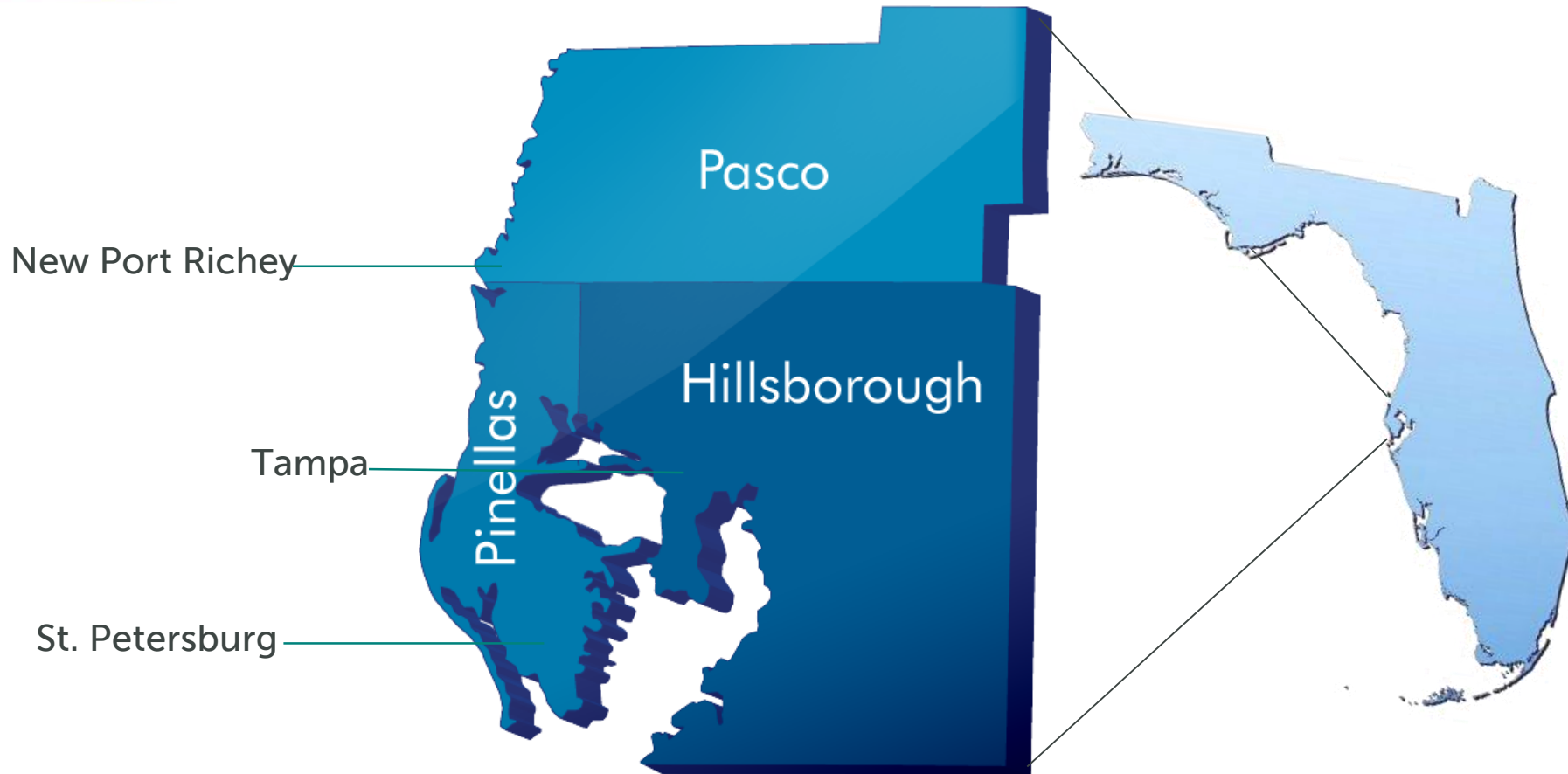


# Multi-Objective Evolutionary Algorithm Model for Water Supply Sources Allocation

Tirusew Asefa, Ph.D., P.E., D.WRE, F.ASCE  
Planning and Decision Support Manager

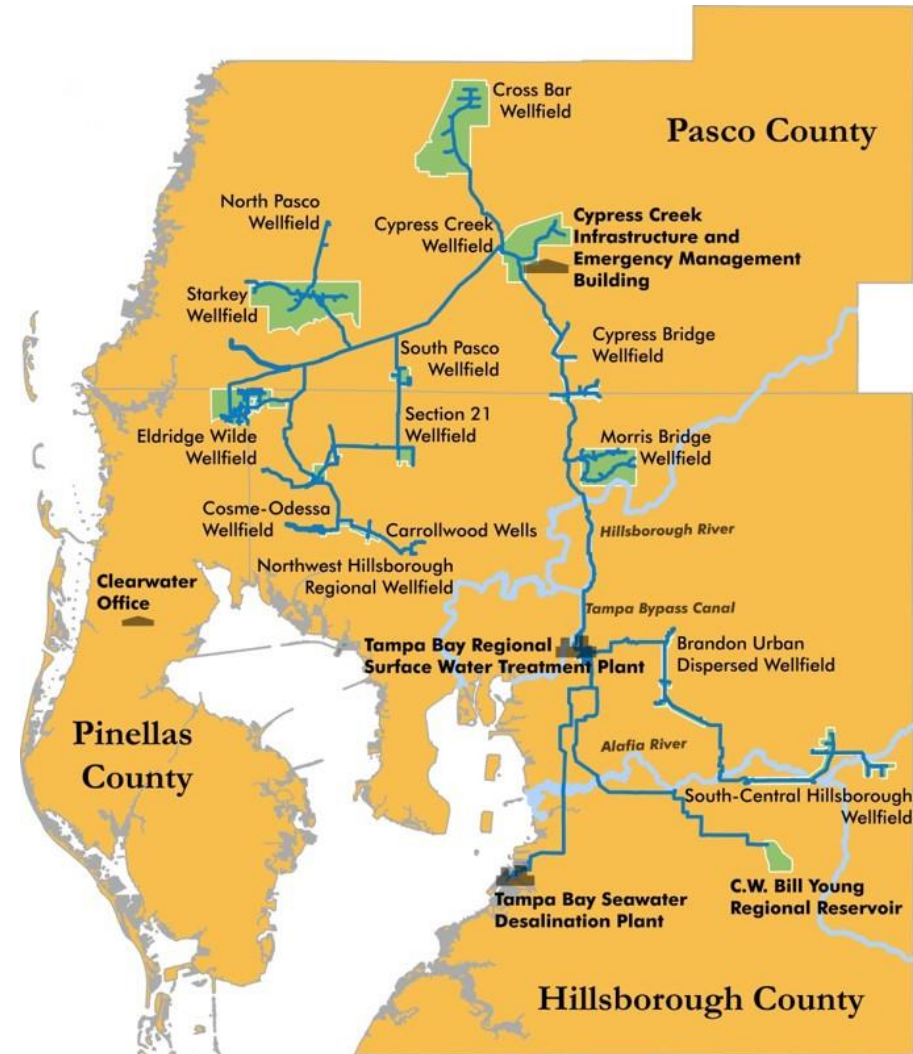


# Tampa Bay Water's Member Governments



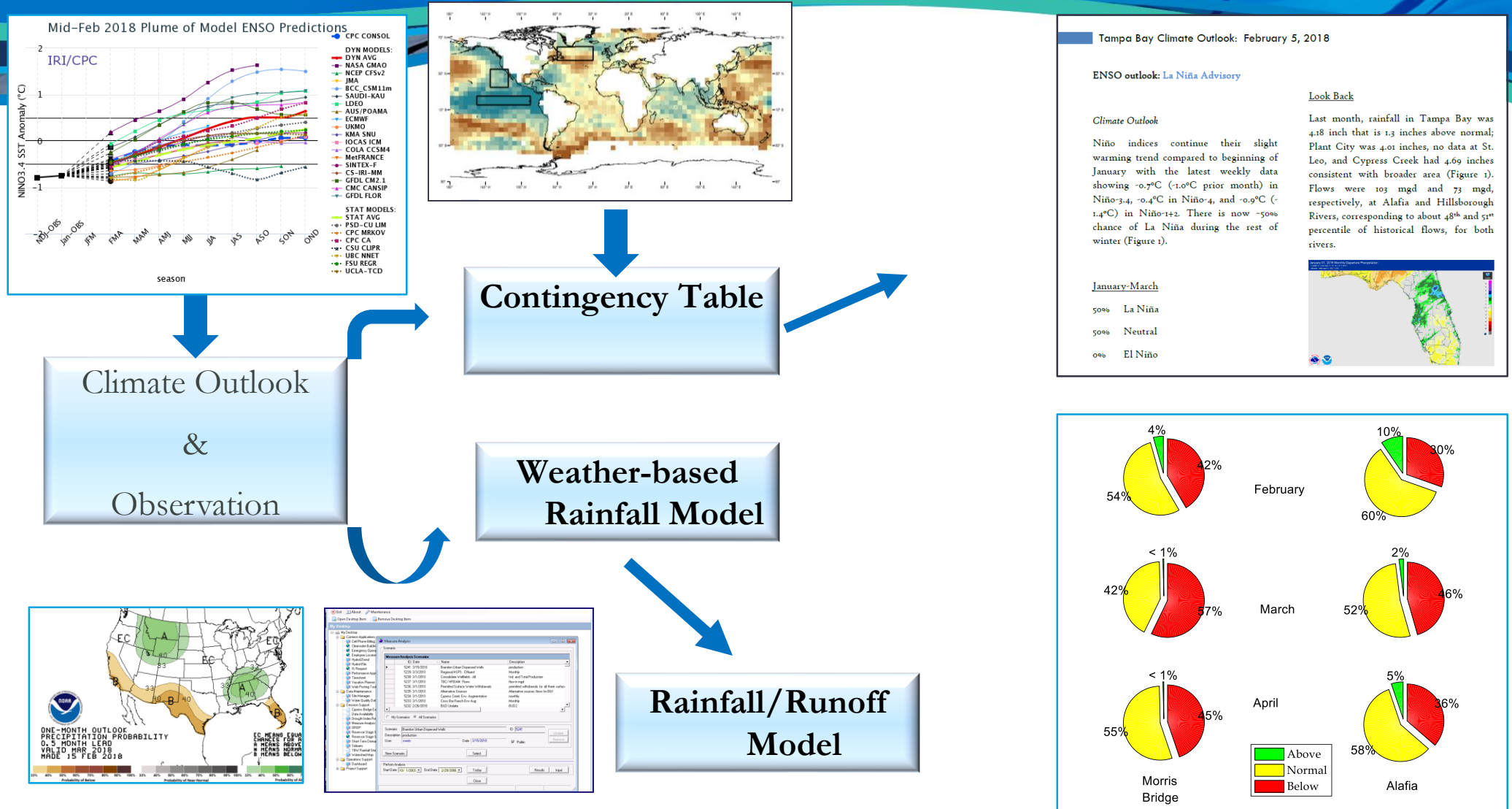
# Tampa Bay Water's Supply System

- Integrated drought-resistant supply system
  - 13 wellfields
  - 8 groundwater treatment facilities
  - Surface Water Treatment Plant
  - Desalination Treatment Plant
  - 9 pump stations
  - 270 miles of transmission mains



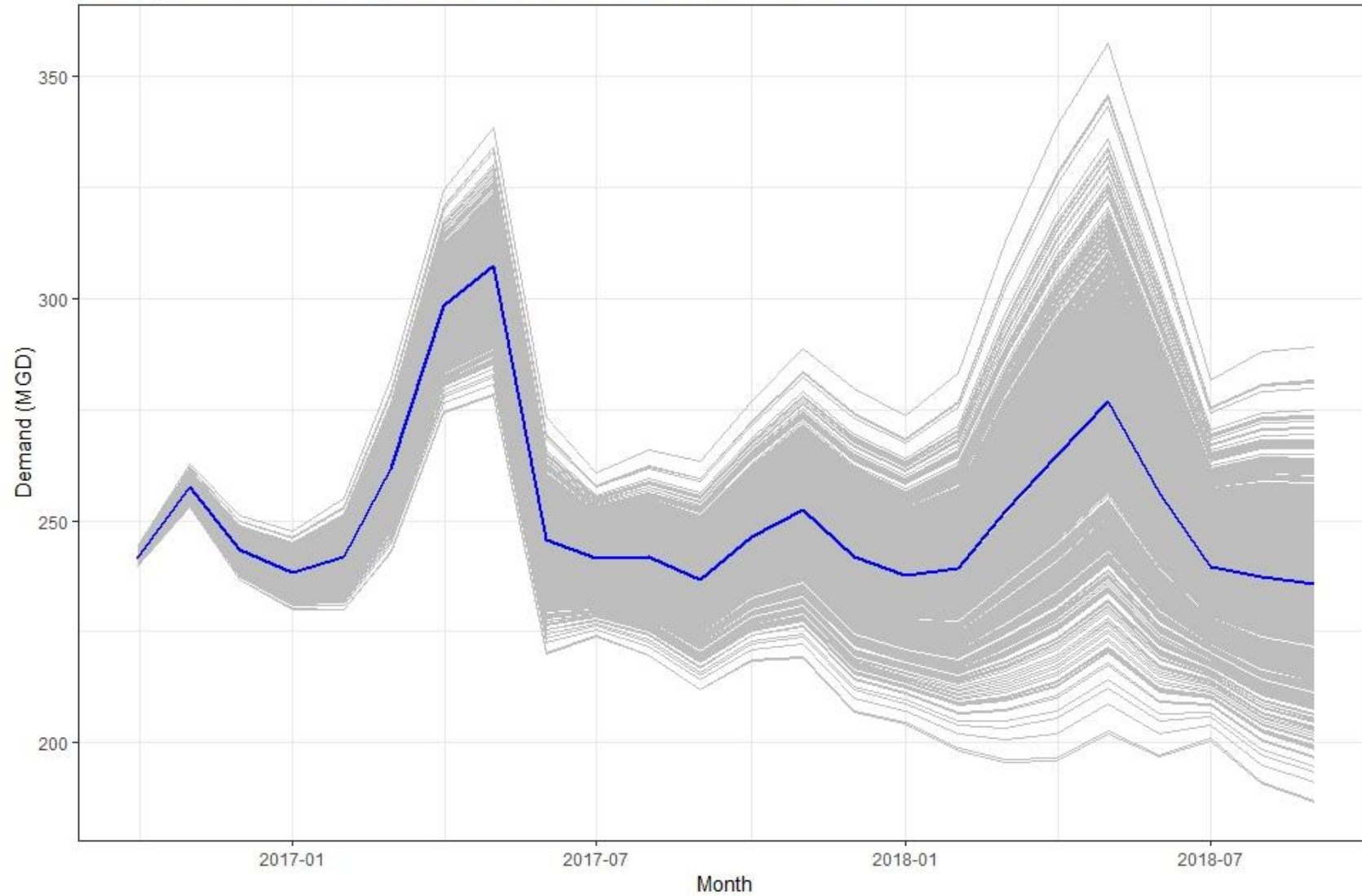


# Seasonal flow forecasting system



# Seasonal demand forecast

Monthly Probabilistic Demand Forecast for Water Years 2017 and 2018



# Four Objectives defined

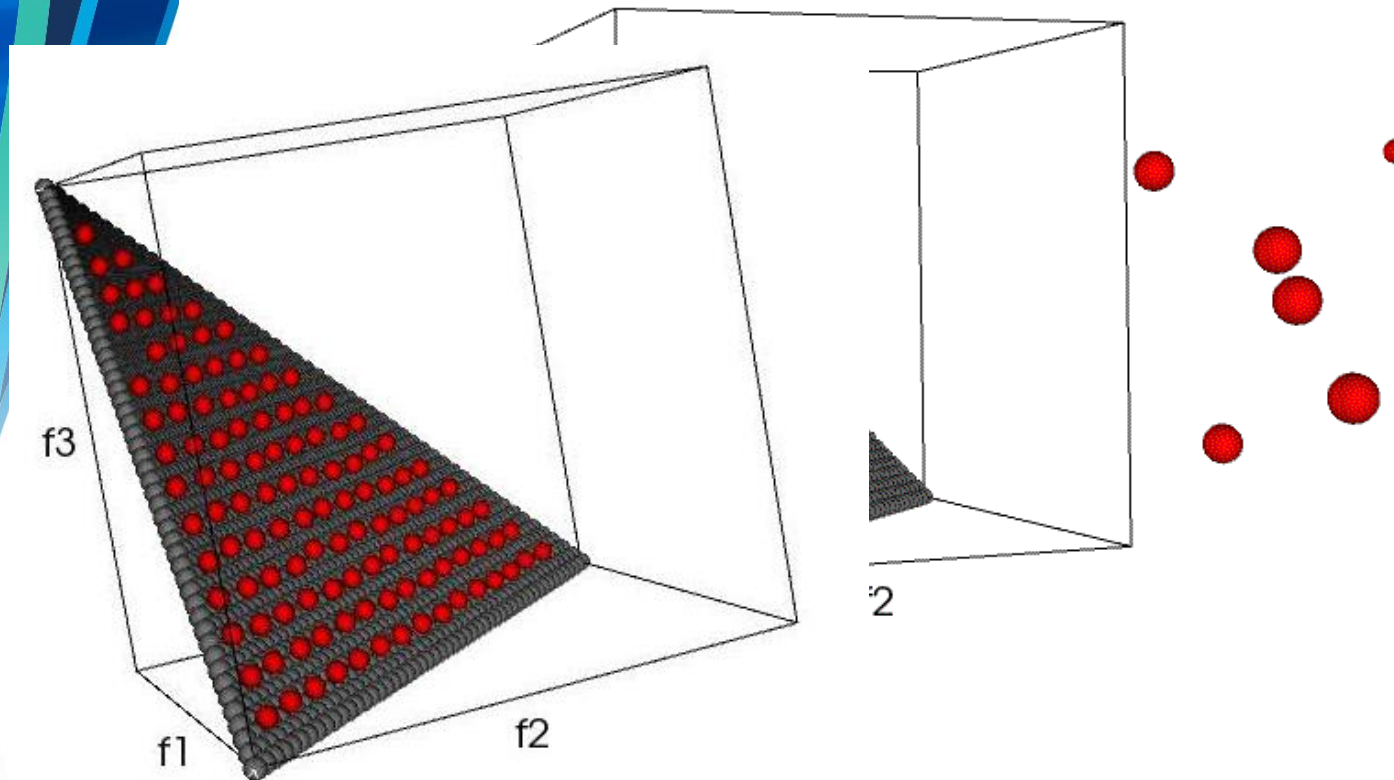
## Minimize:

- deviation from annual budget (in mgd)
- total cost of water production (in relative monetary values)
- Under-Utilization of groundwater use
- Over-Utilization of groundwater use

## Examples of Operating Constraints:

- Keep reservoir storage full at the end of water year
- surface water treatment operation efficiency

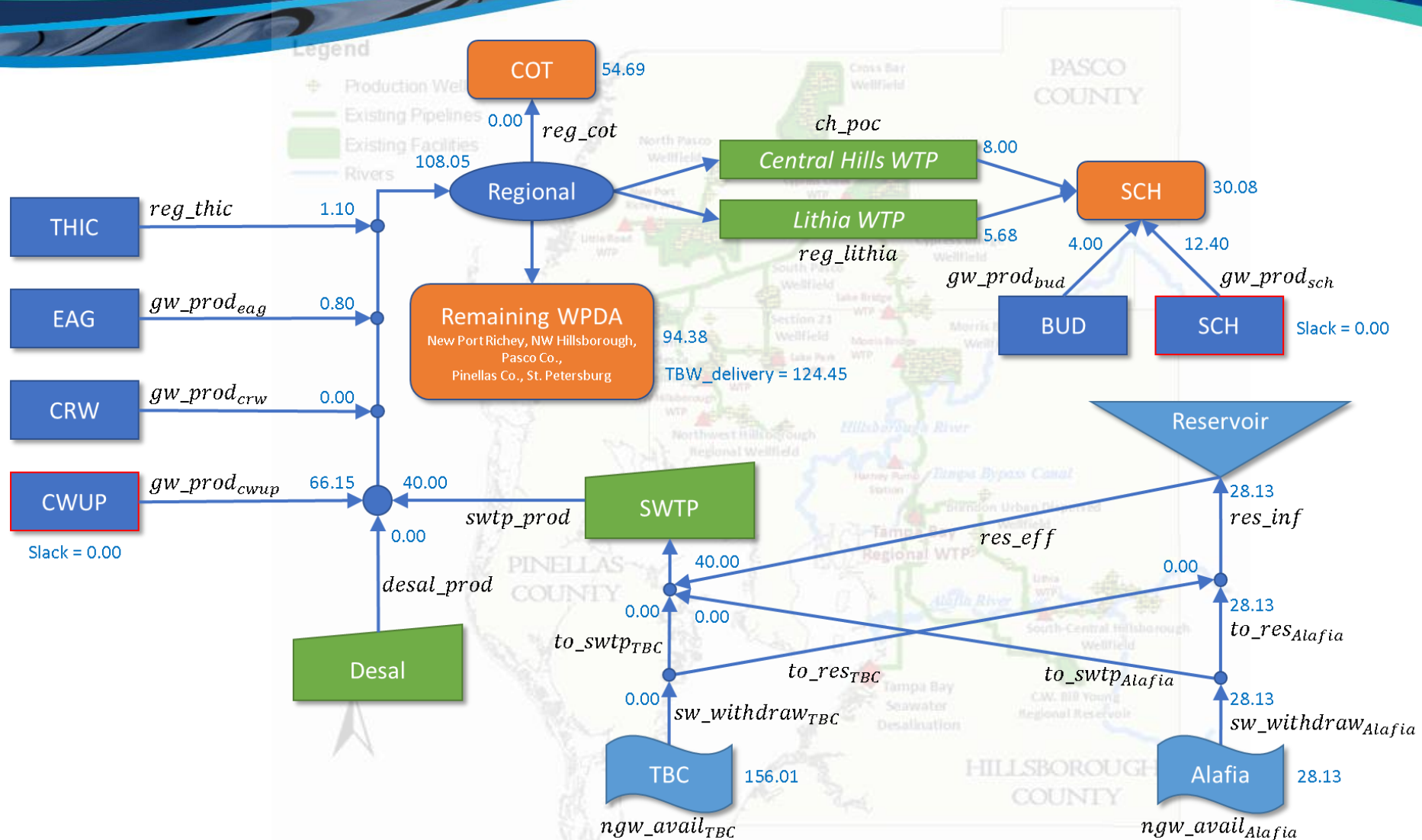
### Three-objective Test Problem



- Heuristic method: flexibility for stochastic problems with unknown gradients
- Search balances convergence and diversity
- Borg MOEA: efficient, reliable performance broad range of applications



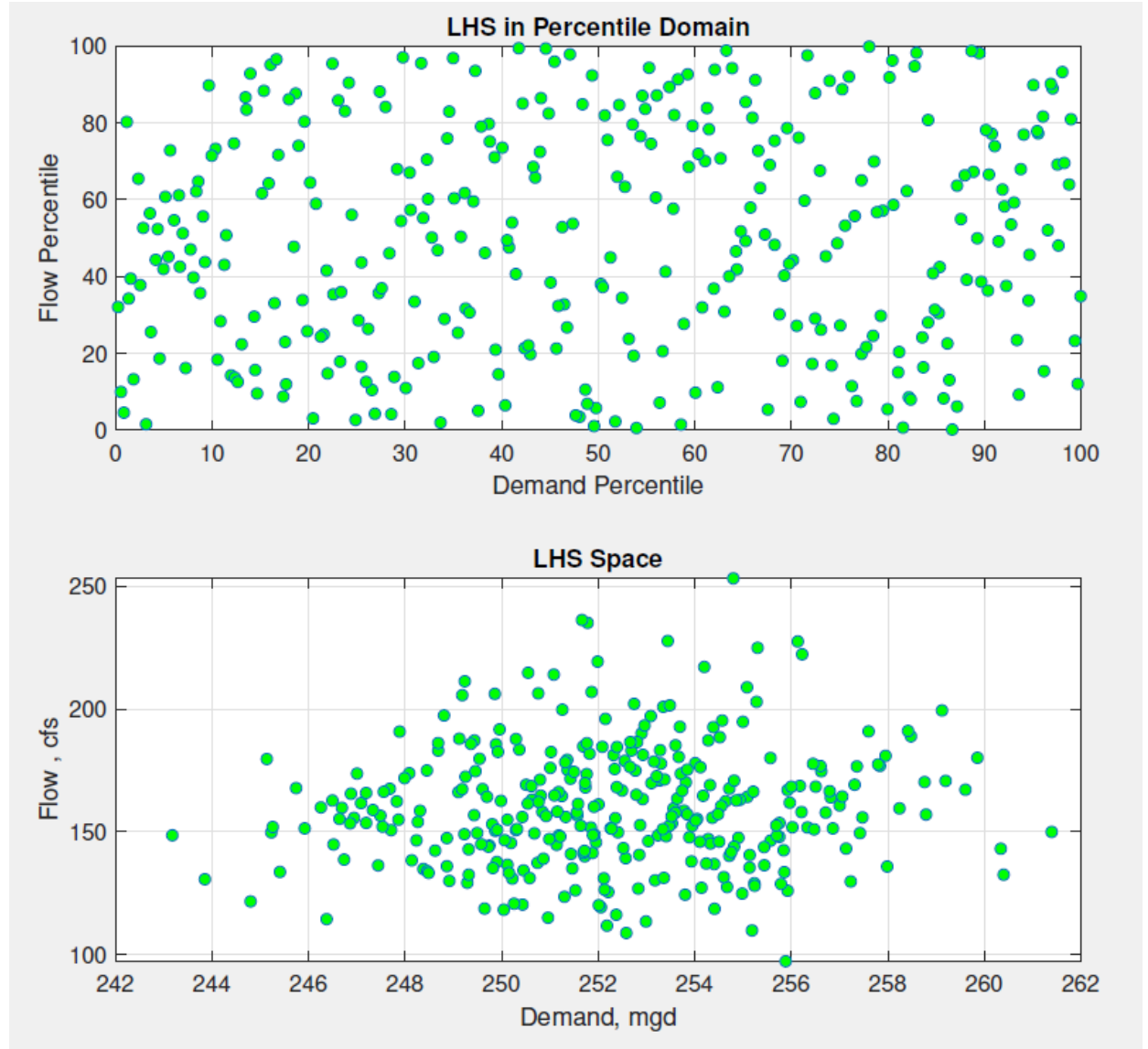
# Seasonal Resource Allocation Model





# Addressing uncertainty in flow and demand forecasts

334 realizations of flow and demand for each MOEA run.

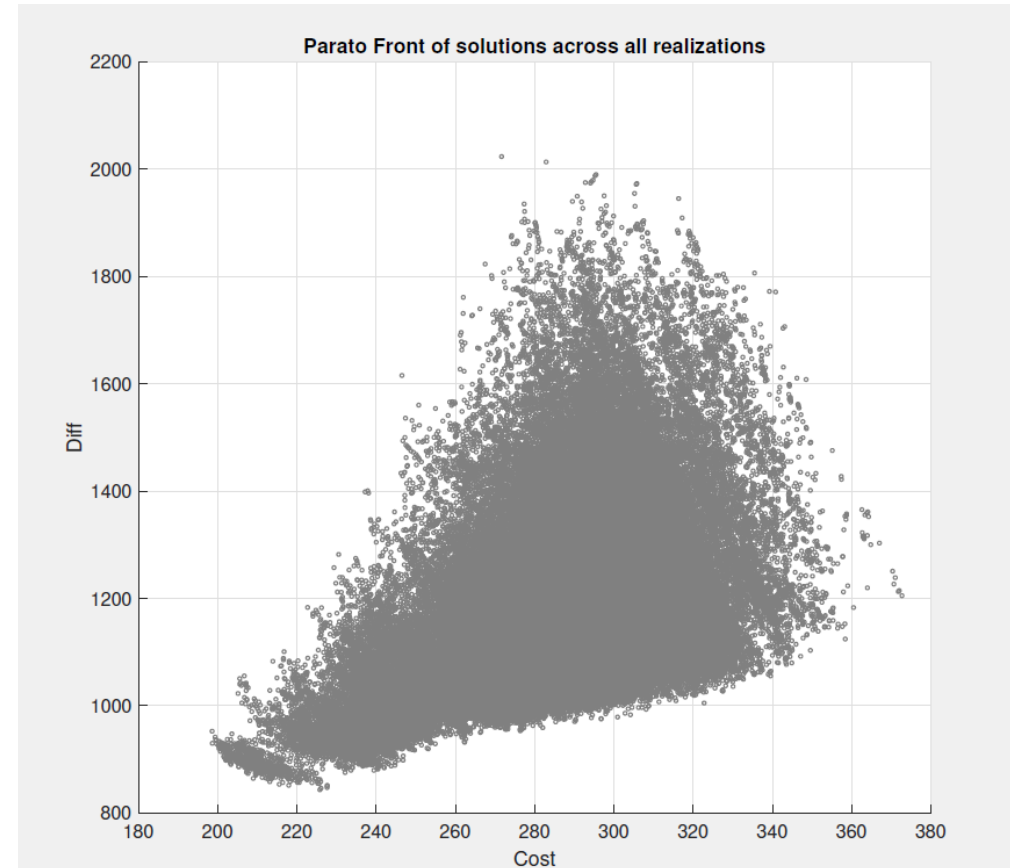


# An illustrative example using retrospective analysis for October 2016

A set of solutions, corresponding to a Parato front, are obtained for each realization.

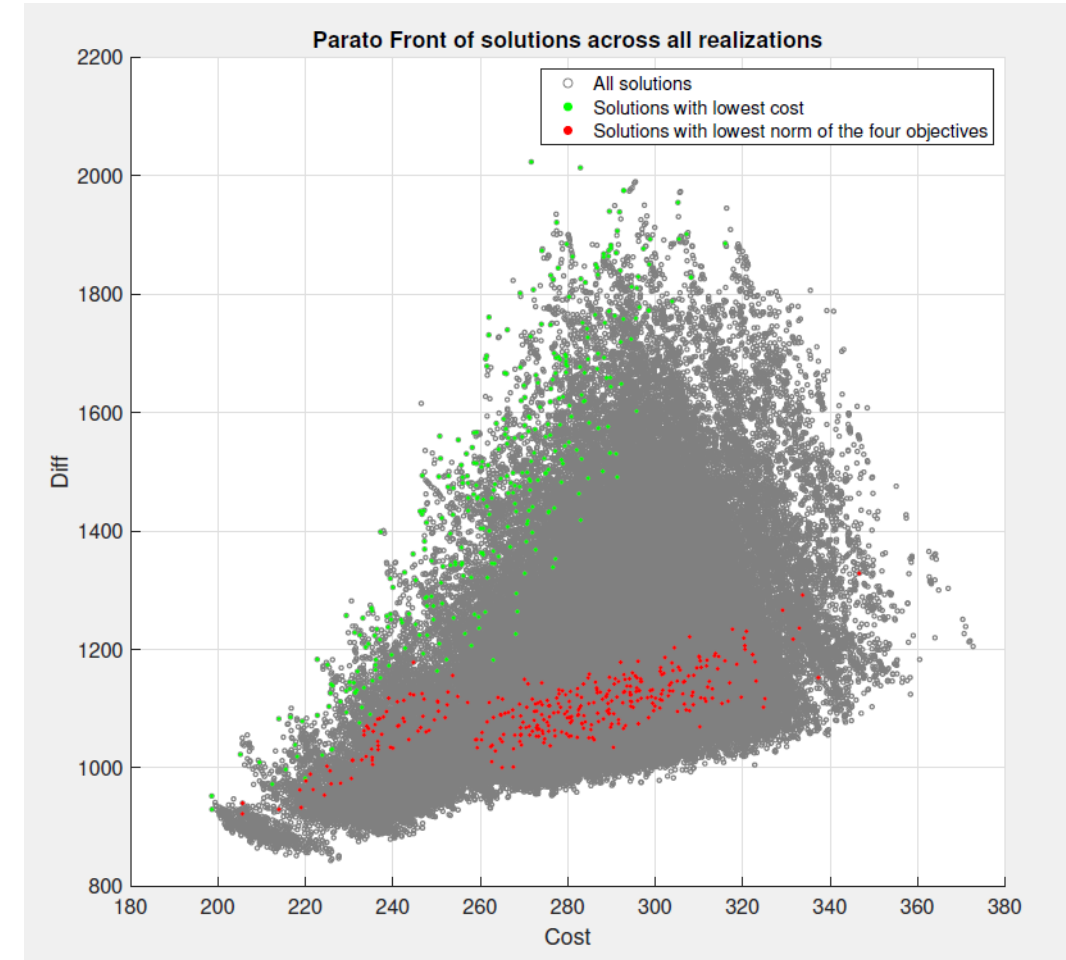
The figure shows Parato front for three realizations with different combinations of flow and demand.

All solutions from 334 realizations can be mapped to the same space for further analysis



# Criteria to select one solution from each Prato Front

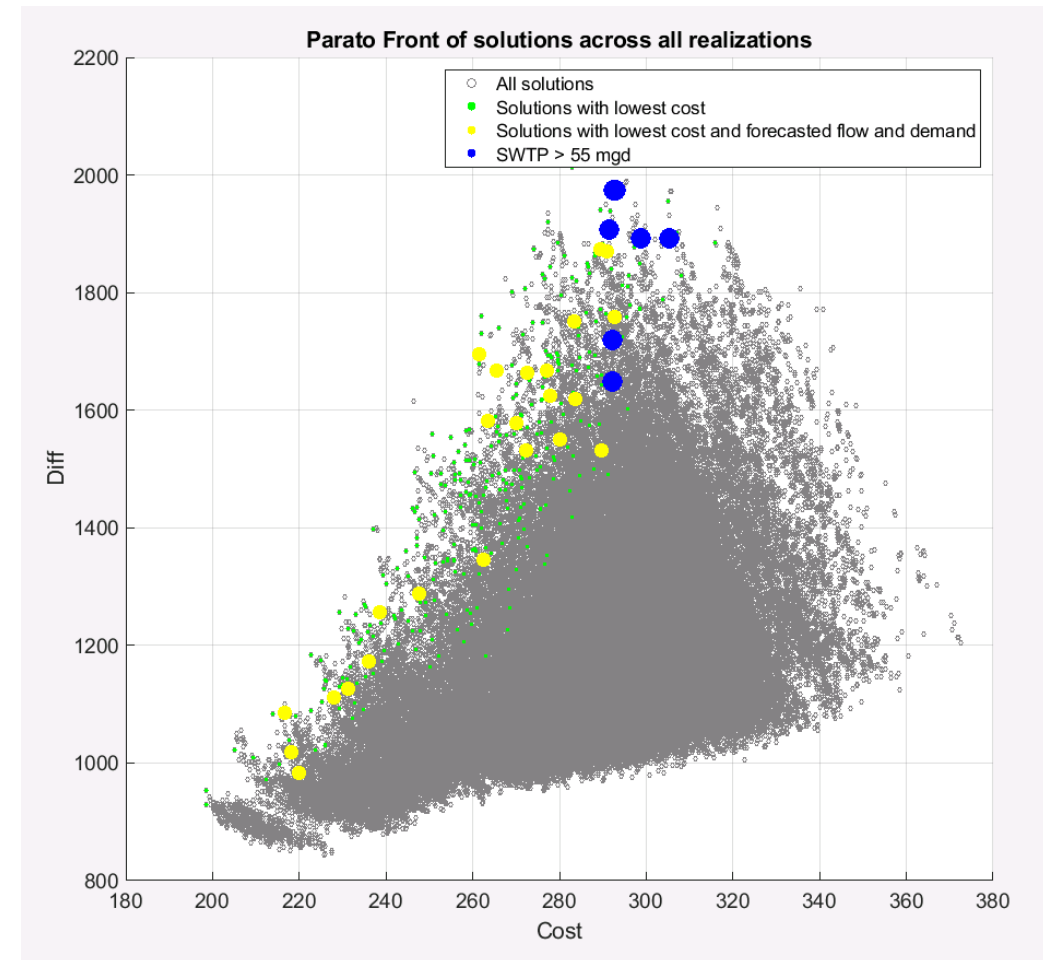
- The following two criteria were tested to select one solution for each input realization
  - ❑ Minimum cost in each set of solutions
  - ❑ Minimum norm of the four objectives in each set of solutions
- Specific criteria used for this purpose can be different depending on operational considerations.





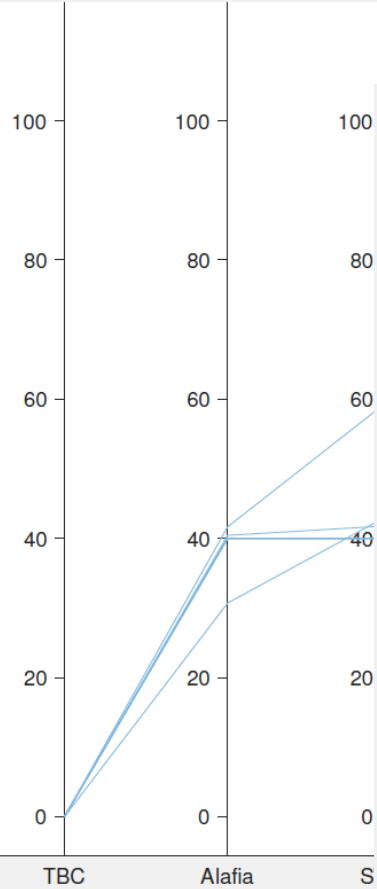
# Additional criteria to filter selected solutions

- Still, a decent number of solutions (334) to be considered for decision making at this point.
- Additional criteria can be used to further filter MOEA results.
  - ❑ First filtering: Flow and demand conditions: both are normal (between 34<sup>th</sup> and 67<sup>th</sup> percentile), shown as large yellow dots
  - ❑ Second filtering: Seasonal surface water treatment production (over Oct, Nov and Dec 2016) is greater than 55 mgd
- The number of solutions reduces from 334 to 33 after 1<sup>st</sup> filtering and reduces to 6 after 2<sup>nd</sup> filtering.

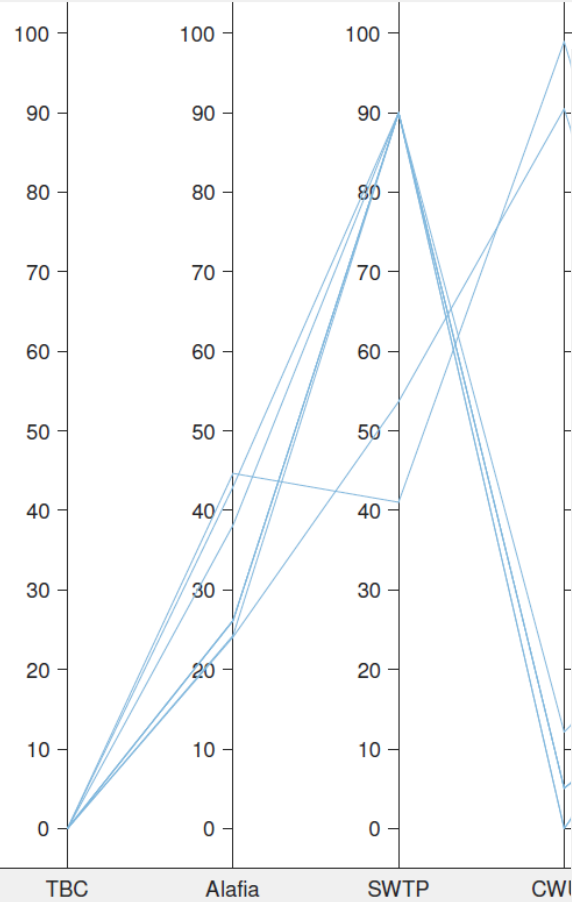


# Seasonal resource allocation

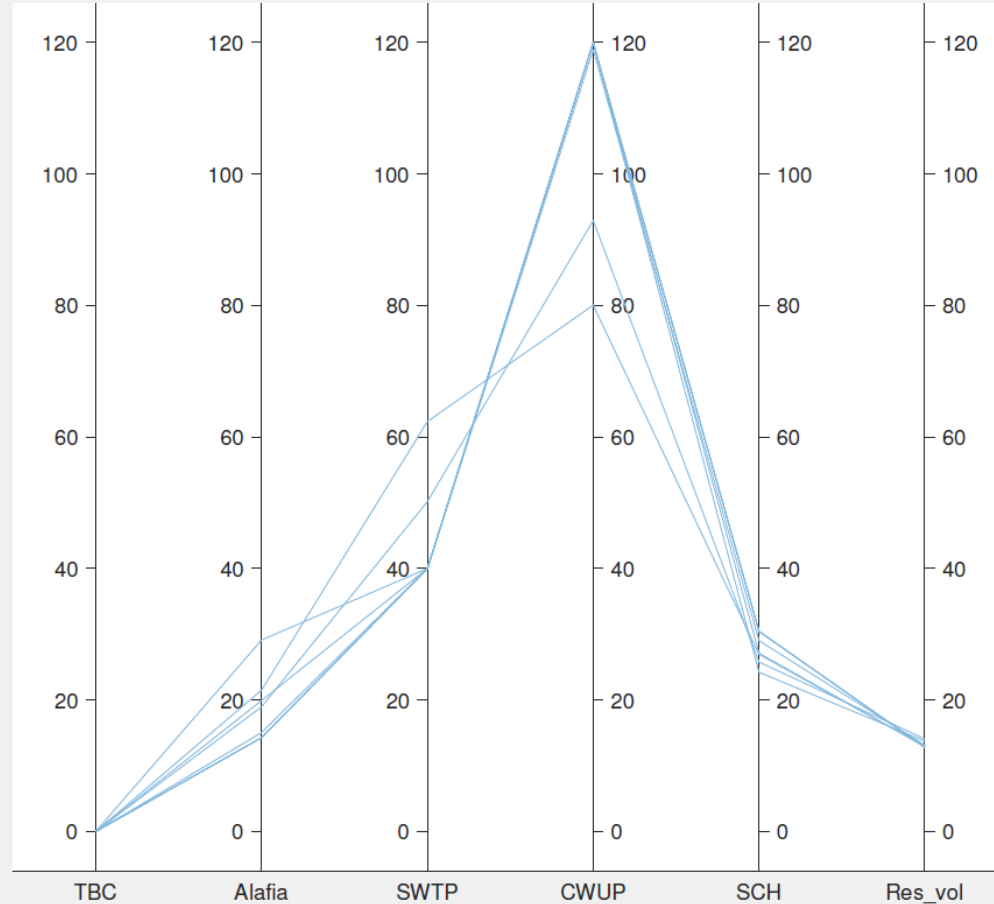
Pareto Solution for October 2016



Pareto Solution for November 2016



Pareto Solution for December 2016



# Next step in the MOEA decision tool

- Data mining to determine final decisions on allocation would based on
- Communicating uncertainty/risk
- Assessing the value of improved supply and demand forecasts (NASA project)



# Contributors

Dr. Nisai Wanakule, Lead Engineer,  
Tampa Bay Water

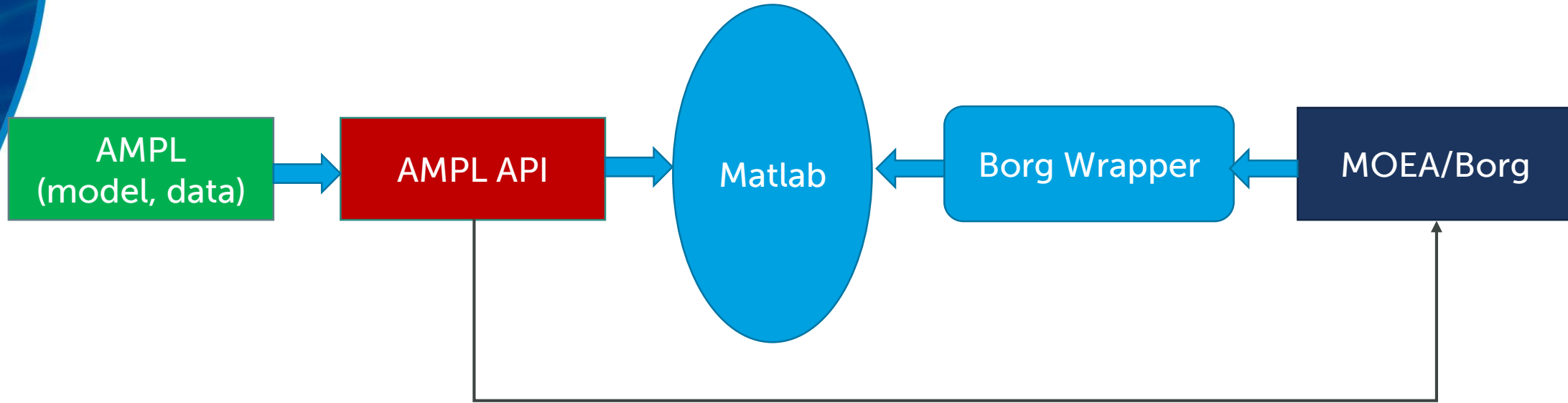
Dr. Hui Wang, Wat. Res. Sys. Engineer,  
Tampa Bay Water





Questions

# Configuration of the model: Framework



MIP fixed variables – MOEA/Borg decision variables  
(groundwater production from two major supply sources)

MIP decision variables – allocation from other sources



# Setup MIP in AMPL as Simulation

- MIP problem – 24-month source allocation to satisfy monthly demands for a given water availability (to be withdrawn from TBC and Alafia river)
- MIP objective as optimizing preferential operation via penalty functions
- MIP constraints – operating rules, facility capacity, water distribution balance, etc.
- MIP equality constraints – evaluation of multi-objectives for MOEA/Borg

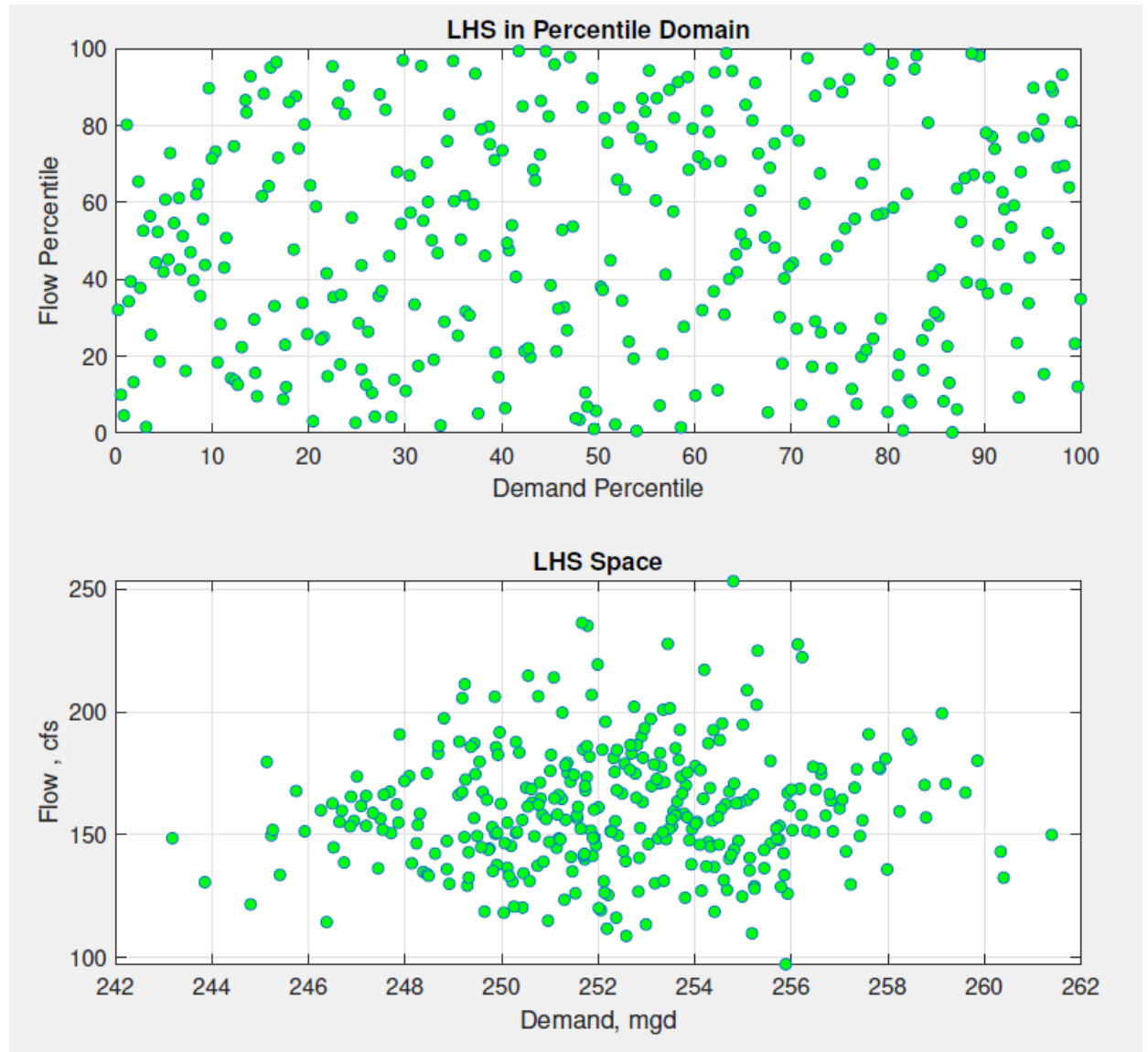
# Retrospective analysis using MOEA

- Water year 2017 with an extreme dry spring
- Retrospective forecast of demand and flow at beginning of each month
- Define data set by “RunDate” for each month
- Interim criteria to select one Pareto solution per realization

Overarching question: For each month in WY2017, how much improvement in source allocation if MOEA was used?

# Addressing uncertainty in seasonally updated flow and demand forecasts

Latin Hypercube Sampling (LHS) applied to obtain 334 realizations of flow and demand for each MOEA run.





# Computation of MOEA

- ~ One RunDate takes three hours on our cluster
- 334 realizations of flow and demand
- Matlab parallel computing toolbox
- Six virtual machines each with 20 cores and 512GB RAM
- 5,000 MOEA function evaluation (MIP) calls

How long it takes to run on a single desktop with 4 cores? (~ 90 hours)