

CALIFORNIA DEPARTMENT OF WATER RESOURCES

Decision Scaling: How

CWEMF ANNUAL MEETING

MAY 14TH, 2025



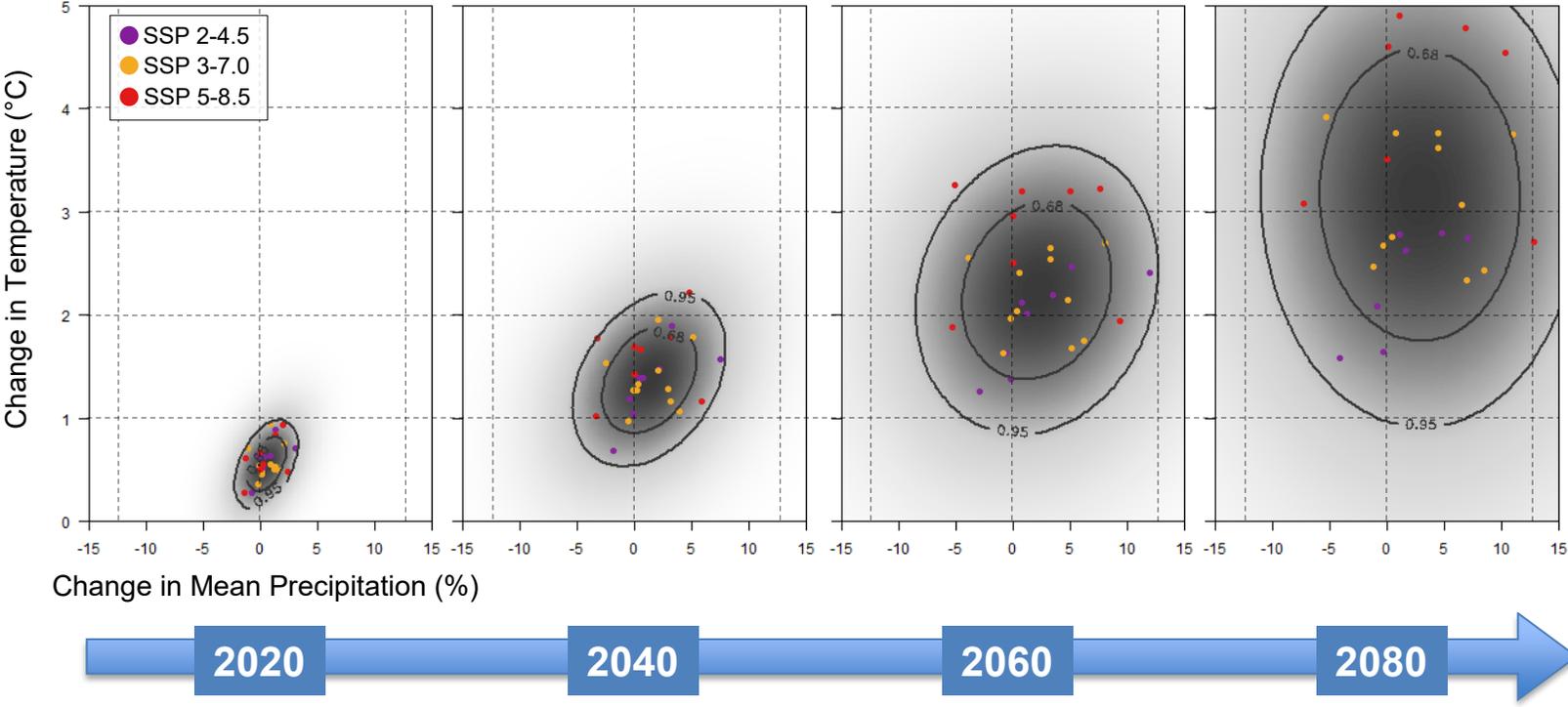
Overview

- **Climate Change Analysis**
 - GCM Projections
 - Decision Scaling
 - System-risk Informed Scenarios
- **Application 1: California Water Plan 2023**
- **Application 2: SJR Flood-MAR Watershed Studies**

Climate Change Analysis

Wrestling with uncertainty

Projected range of likely climate changes over the Delta Catchment relative to the baseline 30-yr period 1992-2021

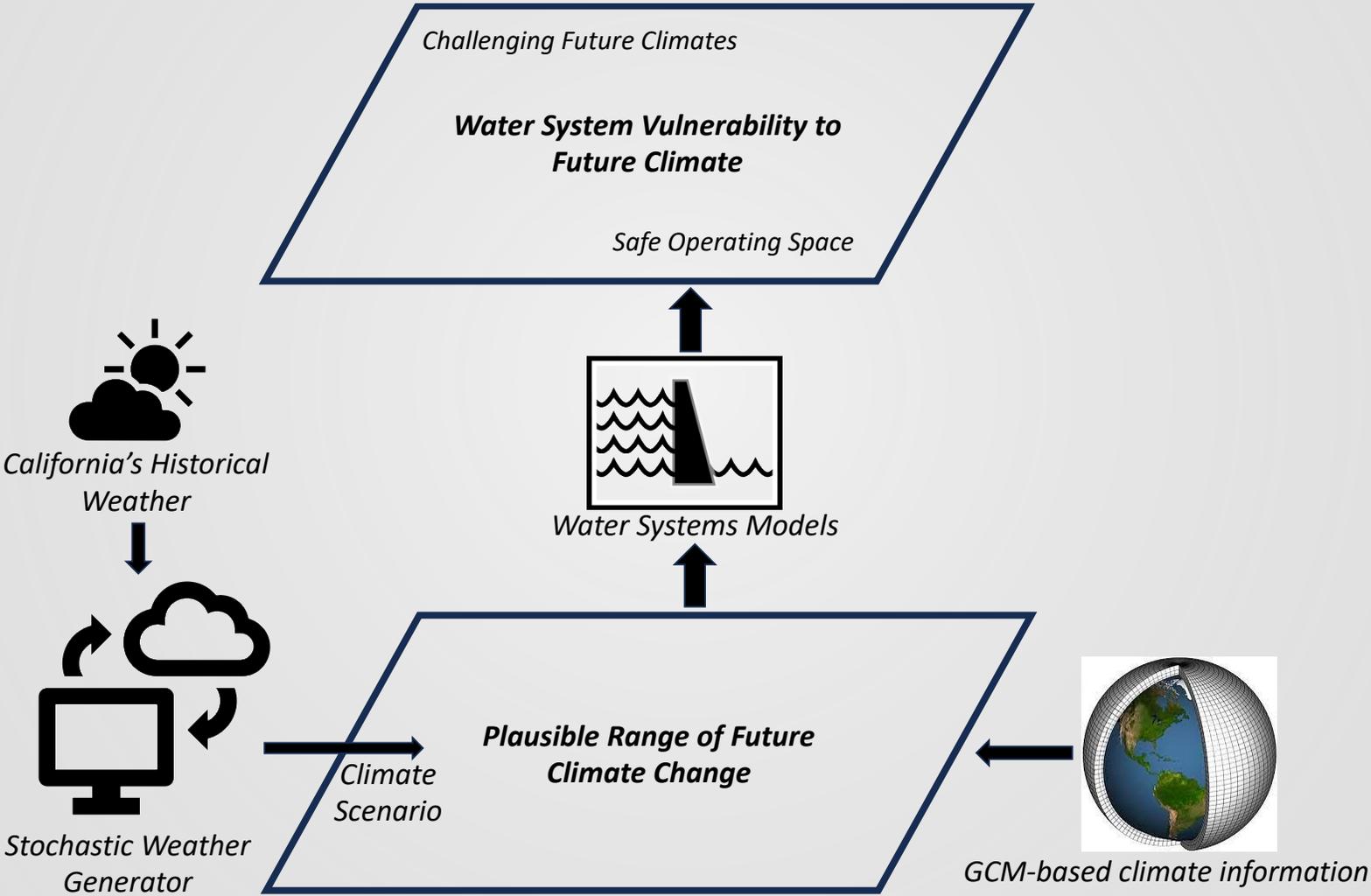


Each dot:

- One of the 27 CMIP6/LOCA-2 downscaled projections of temperature and precipitation
- Spatial averaging over watersheds draining into the Sacramento-San Joaquin Delta

Decision Scaling Analysis

A Bottom-up Framework



Climate Change Projections



Global

Regional

Filtering by UCSD/Scripps & CA Energy Commission to target high-performing models and diversity of data

Downscaling



199 projections over California
(available at Cal-Adapt)



What DWR uses

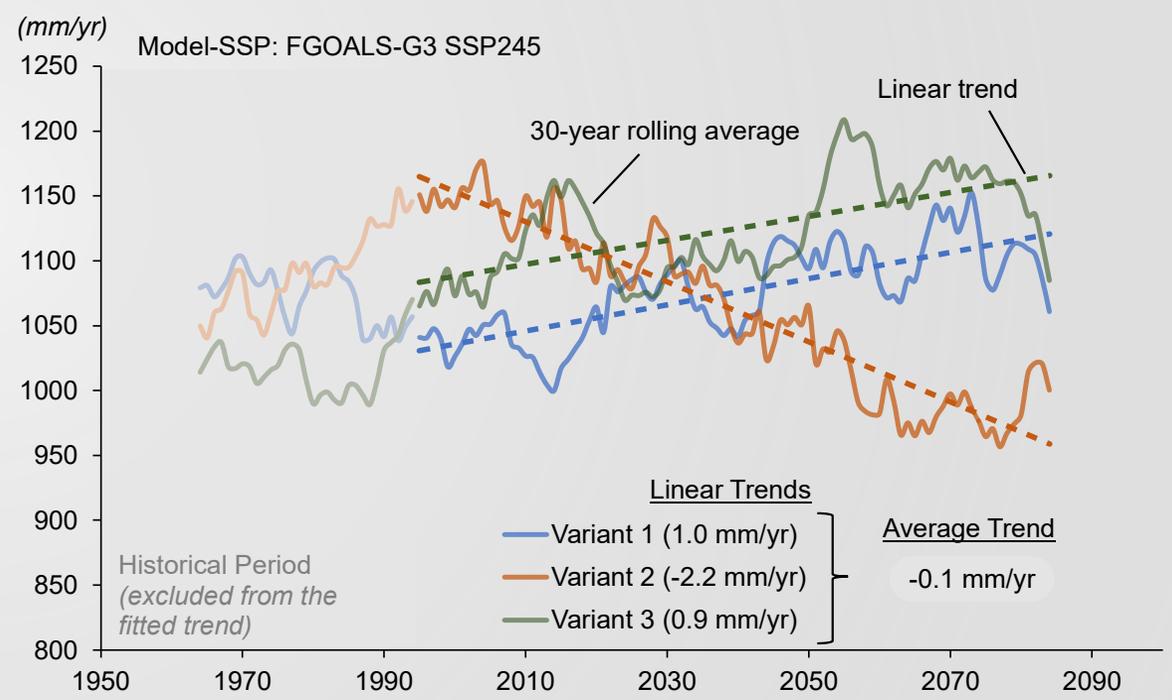
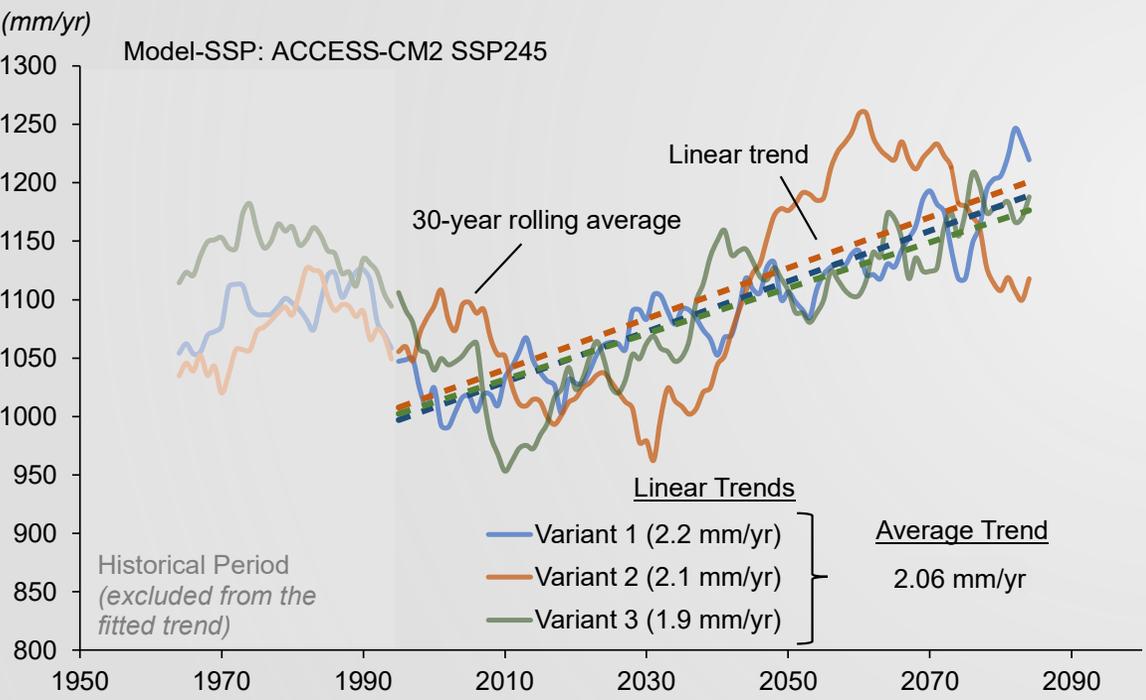
- 8x 5-85
- 11x 3-70
- 8x 2-45

27 variant-averaged projections of temperature and precipitation change

Climate Change Projections

Why variant-averaging?

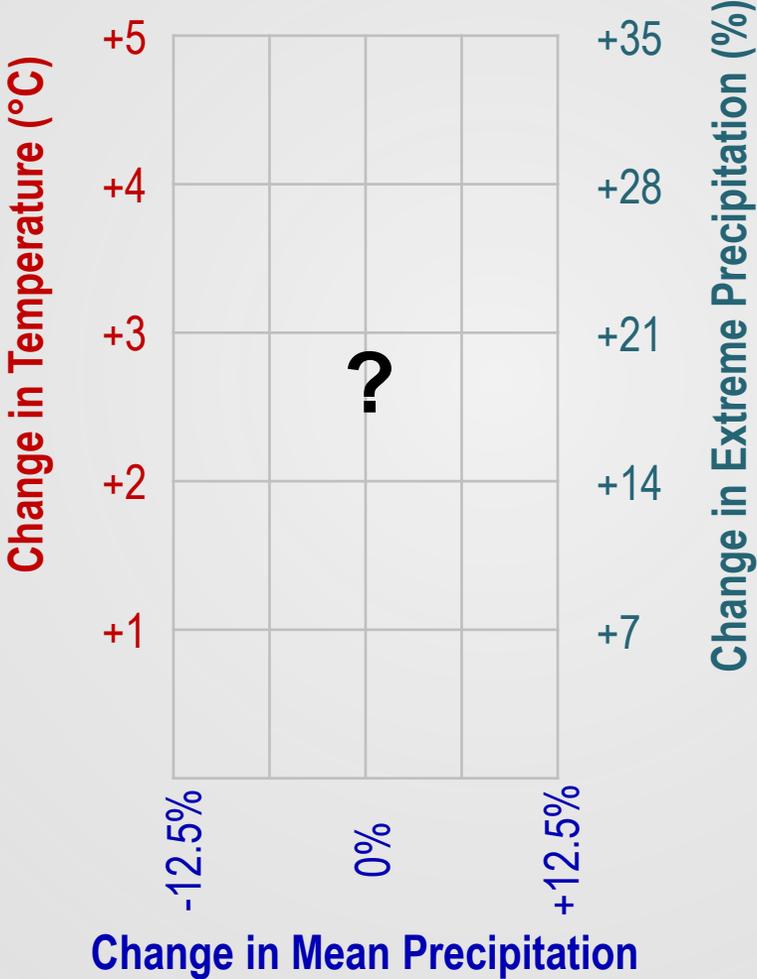
Long-term mean precipitation changes appear mostly driven by natural variability



Climate Change Analysis

Wrestling with uncertainty

Magnitude ?



Change in Mean Precipitation

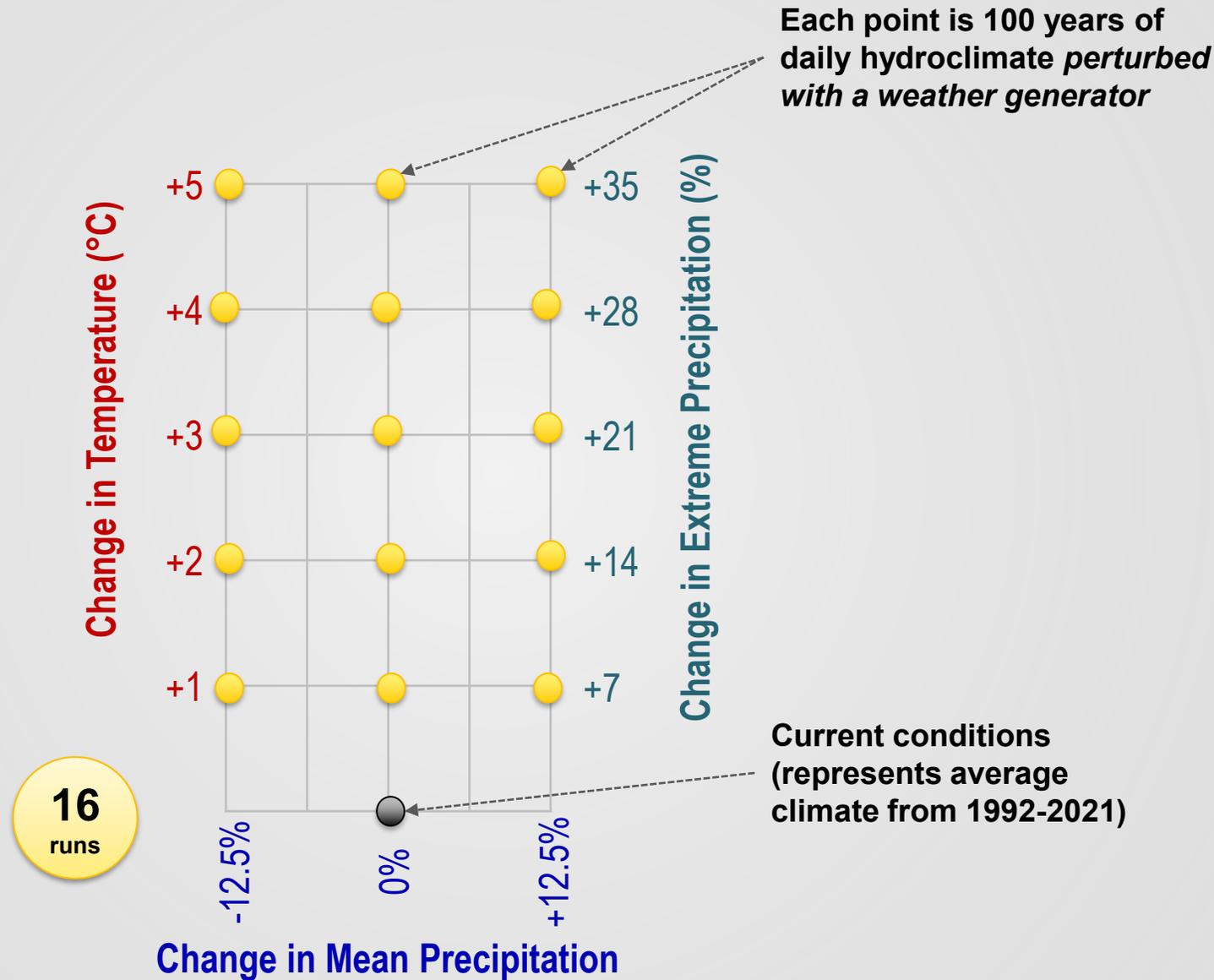


Direction / Magnitude ?

Decision Scaling Approach

Stress testing

- Instead of “picking” one future or selecting a handful of models etc. ...
- ...**simulate the whole range, incrementally**
- Provides a stress (or sensitivity) test, independent of GCM model assumptions

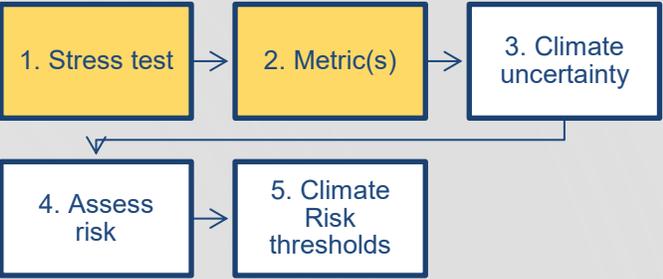
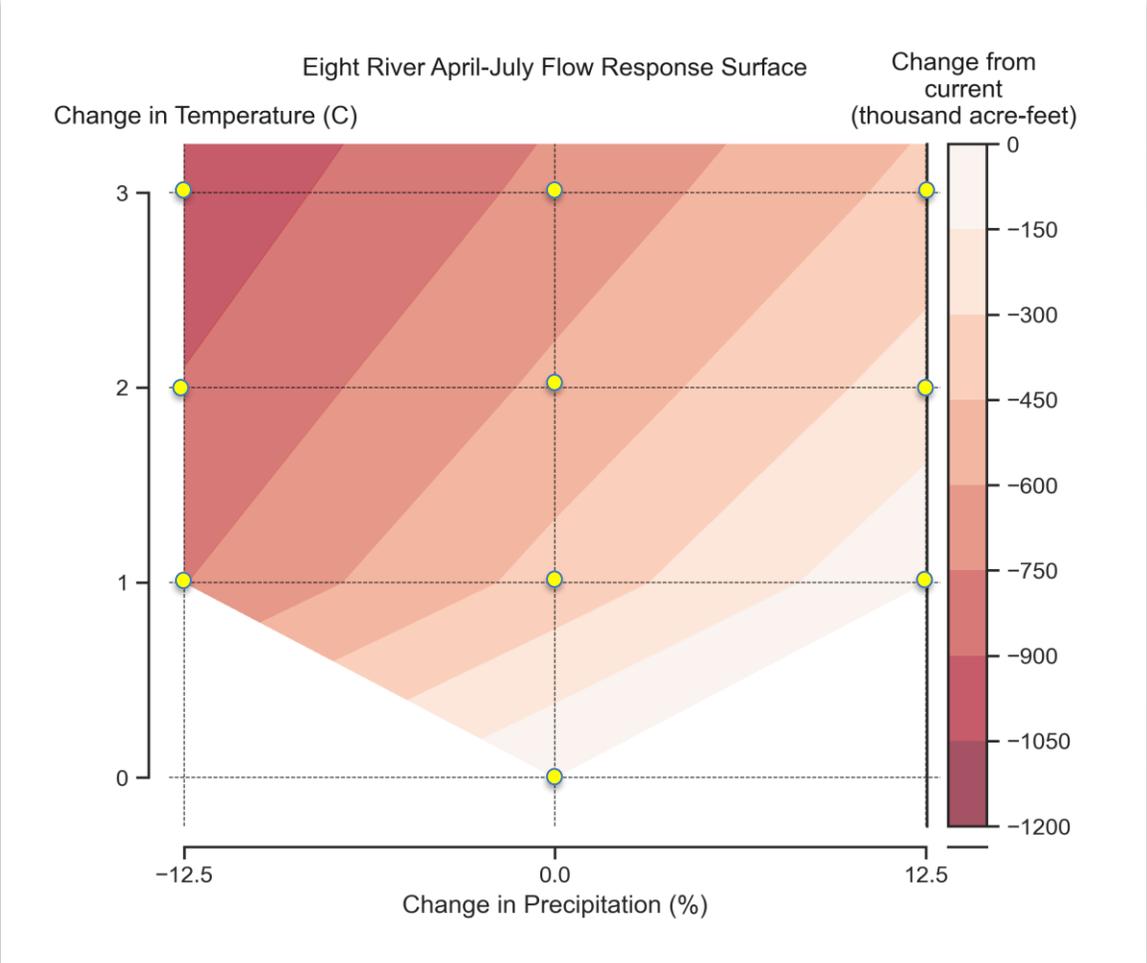


Decision Scaling Process Steps

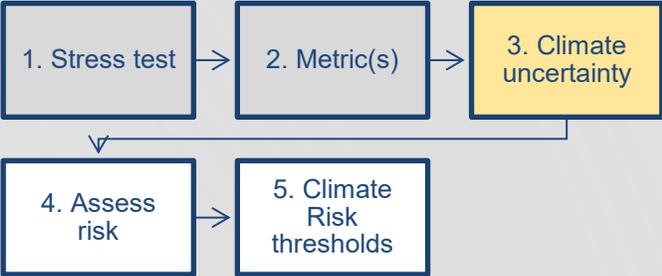


Building the Response Surface

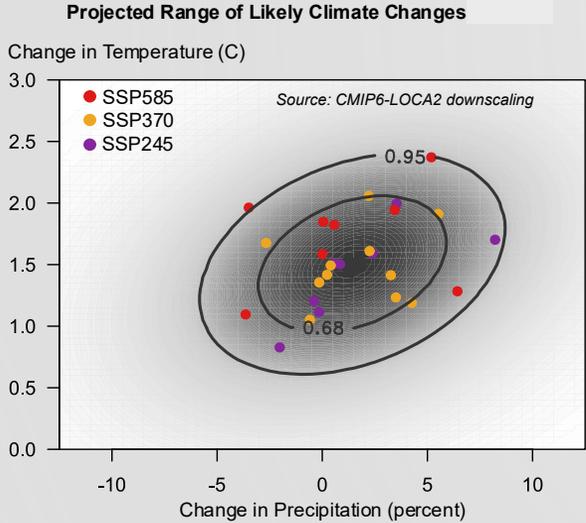
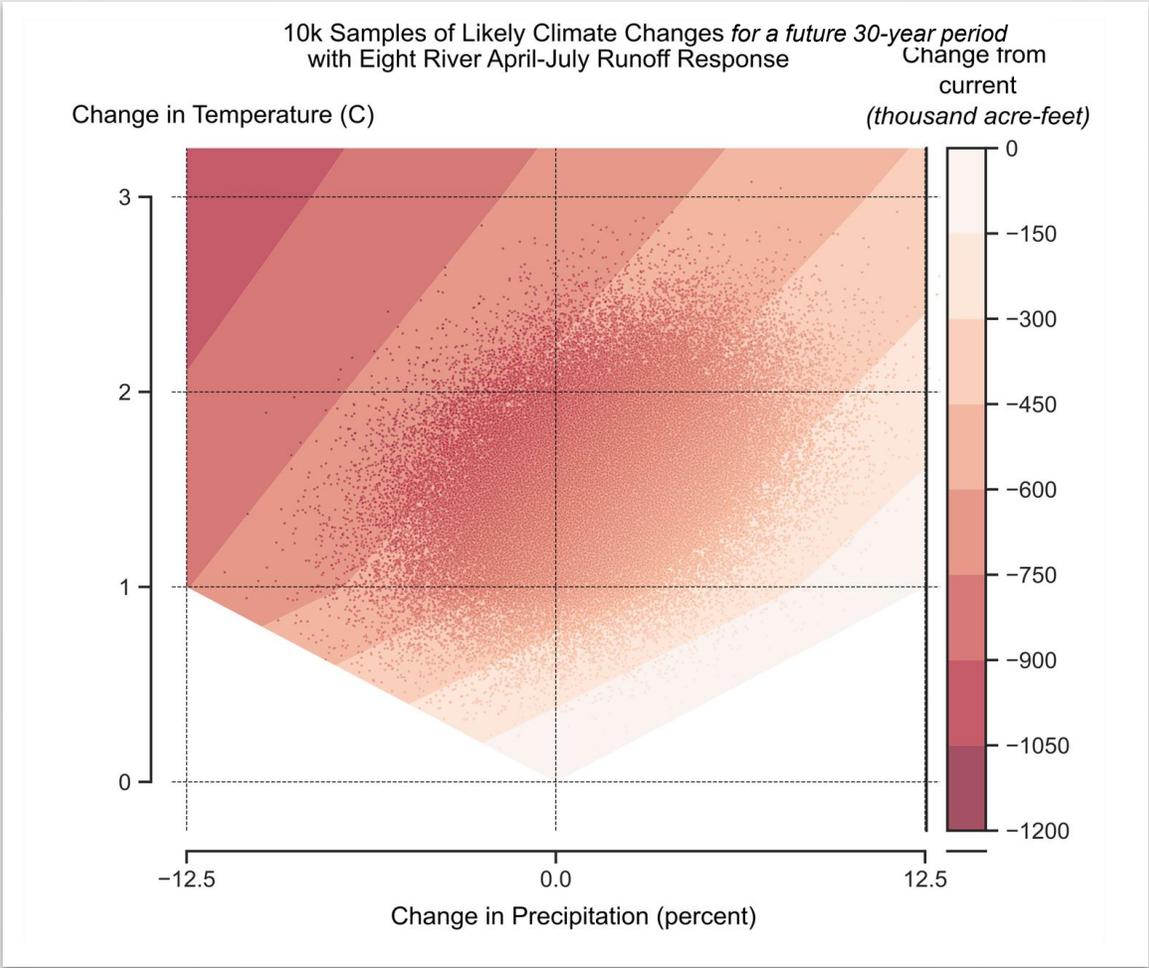
- 1. Simulate models for each combination of temperature and precipitation
- 2. Calculate indicators/metrics of performance based on the simulation time-series
- 3. Interpolate to create the system response surface



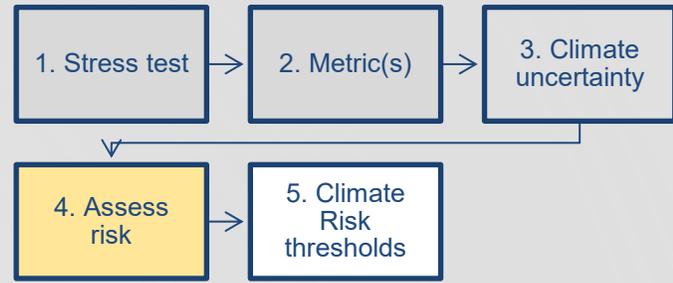
Calculating Climate Change Risk



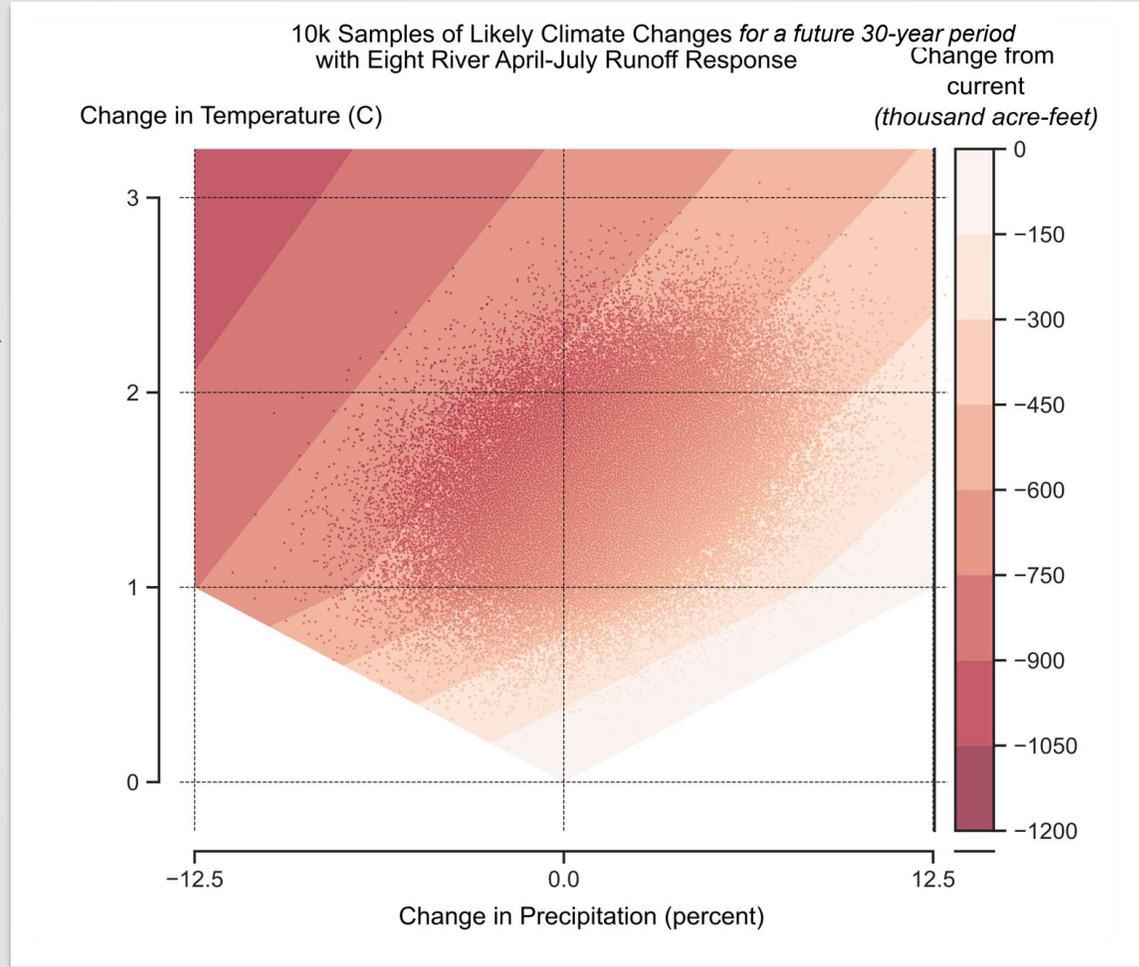
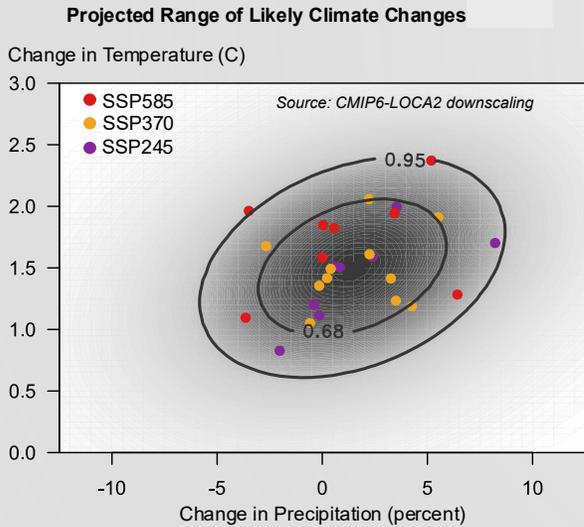
1. Sample the PDF of changes in temperature and precipitation projected by GCMs



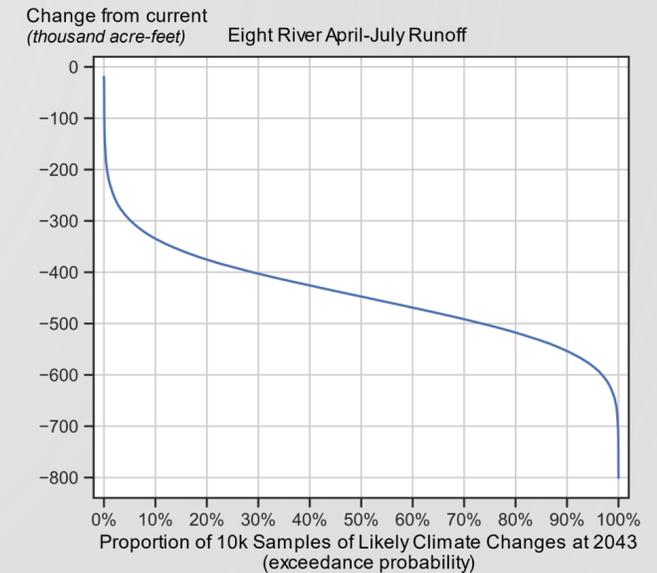
Calculating Climate Change Risk



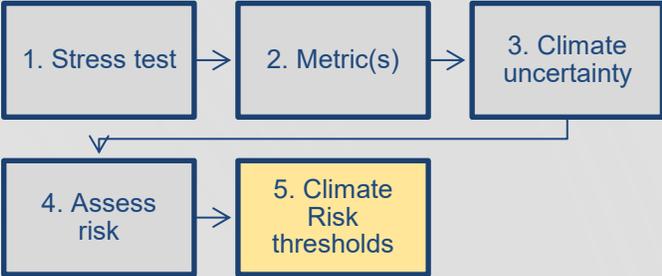
1. Sample the PDF of changes in temperature and precipitation projected by GCMs



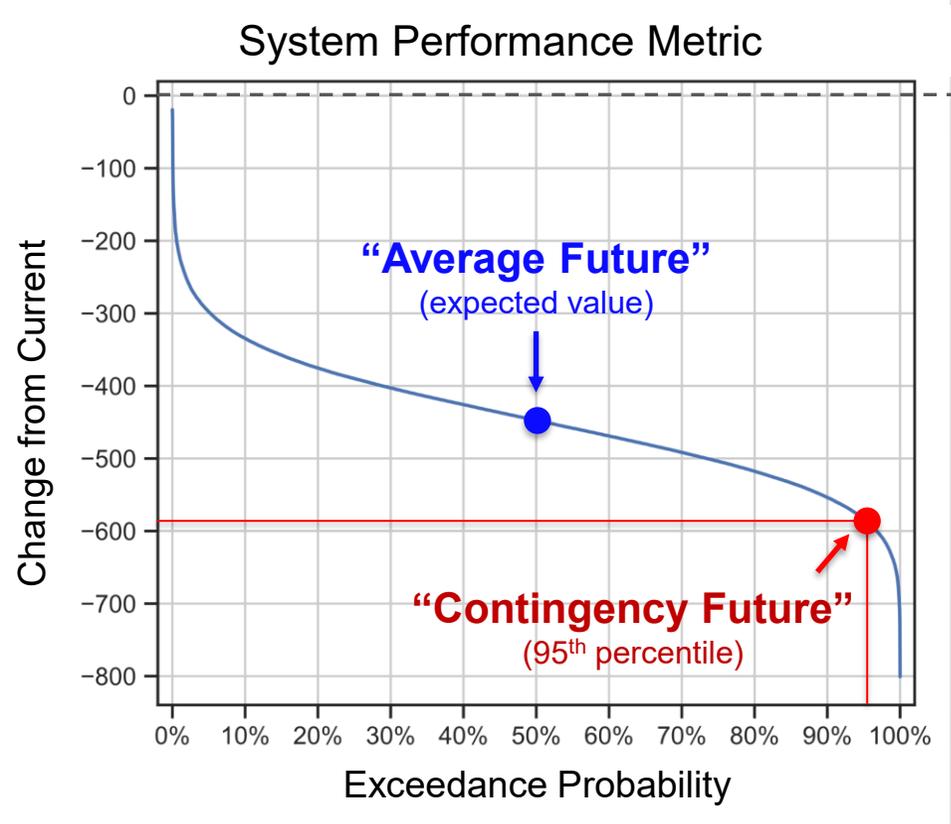
2. CDF of system performance conditioned on climate PDF



Climate Change Risk Reporting



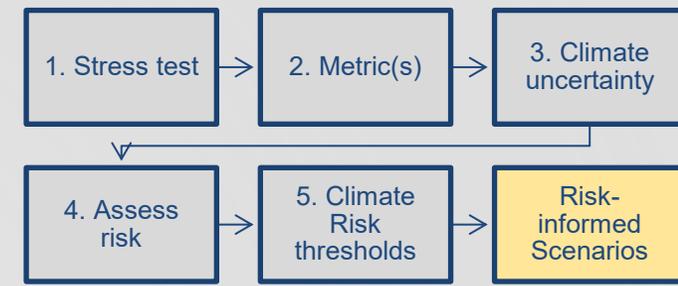
- What to expect
- If worse, by how much (5% chance)?
- Likelihood of being worse (by any amount)



Performance threshold
(set at current performance)

Probability of Declining Performance

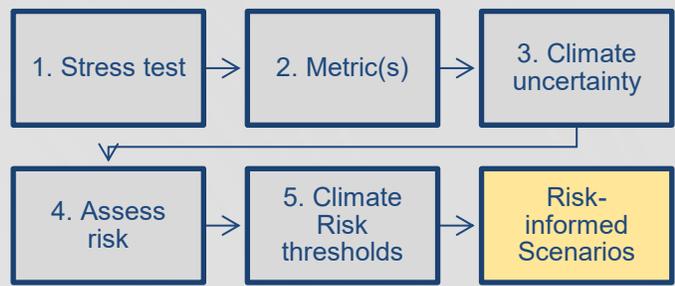
System Risk-Informed Scenarios



Q: Great. But what if I have a really, really, complex (and awesome) system model and multiple other complex system models that depend on that complex model's output ???

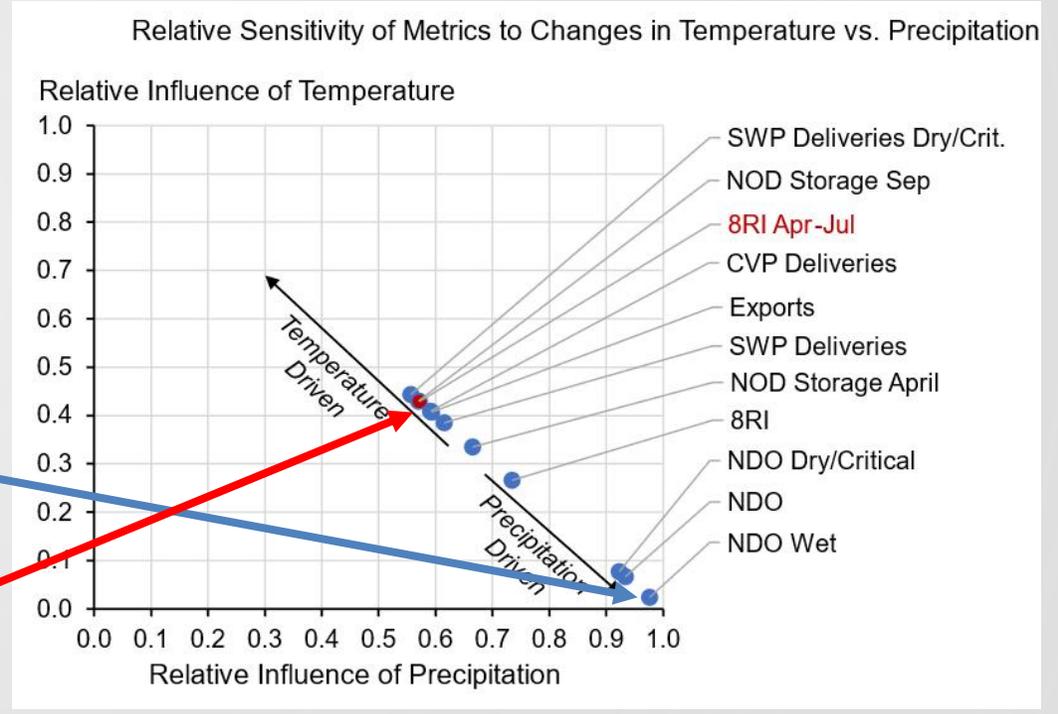
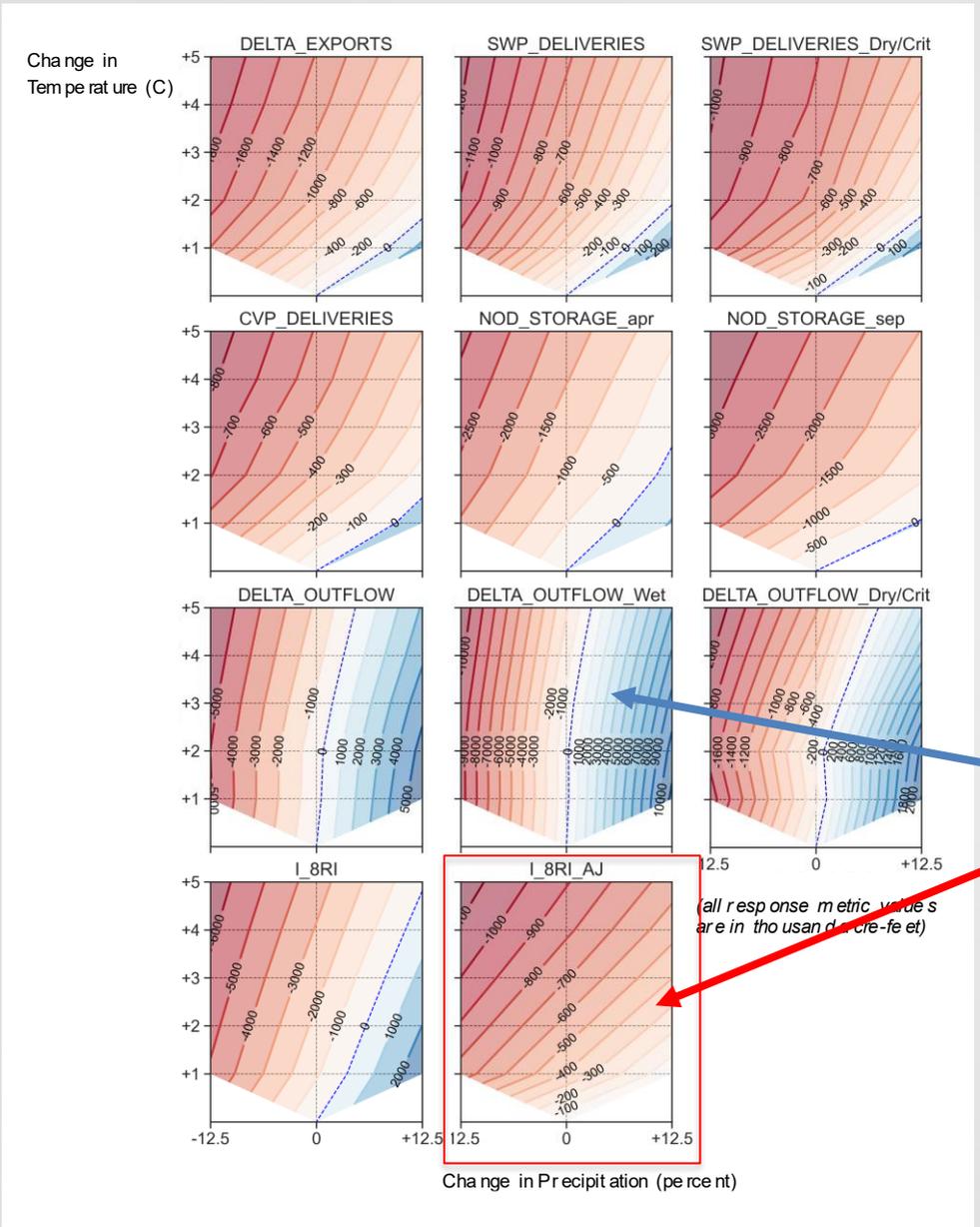
A: Risk-informed scenarios, a.k.a. “Level of Concern” (LOC)

System Risk-Informed Scenarios

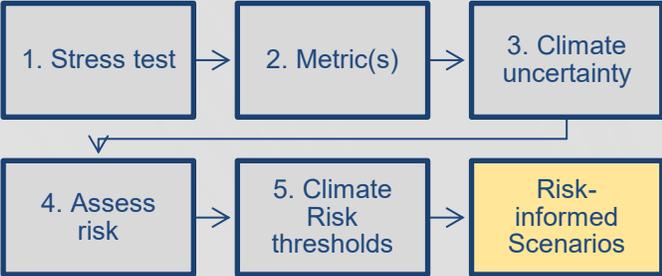


Part 1: Use your screening model to:

1. Select a key metric that reflects systemwide risk

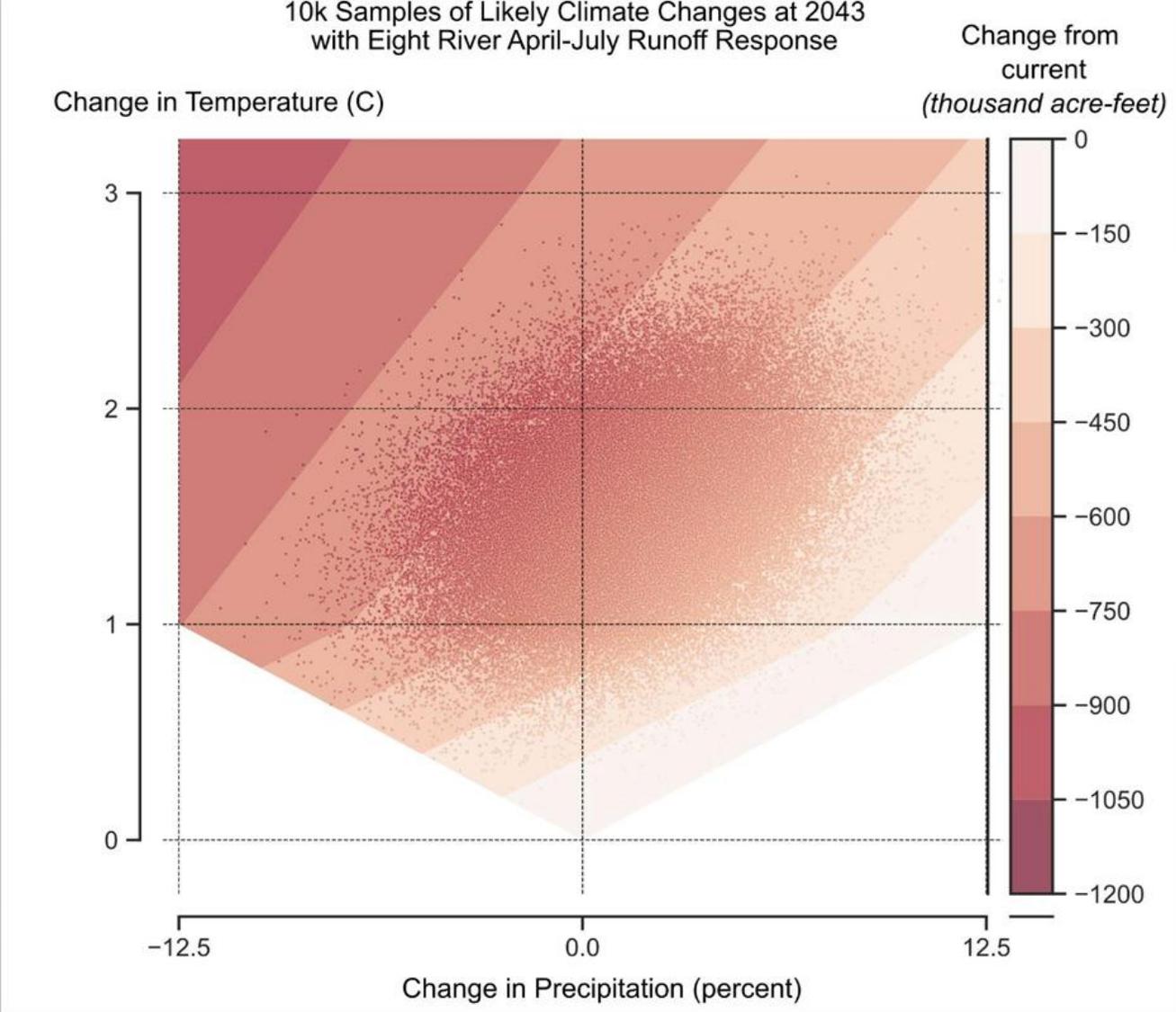


System Risk-Informed Scenarios

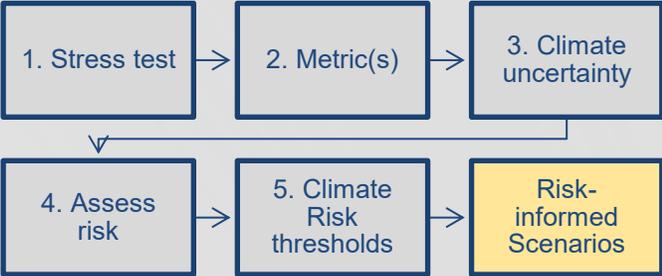


Part 1: Use your screening model to:

- 1. Select a key metric that reflects systemwide risk
- 2. Response surface with climate PDF

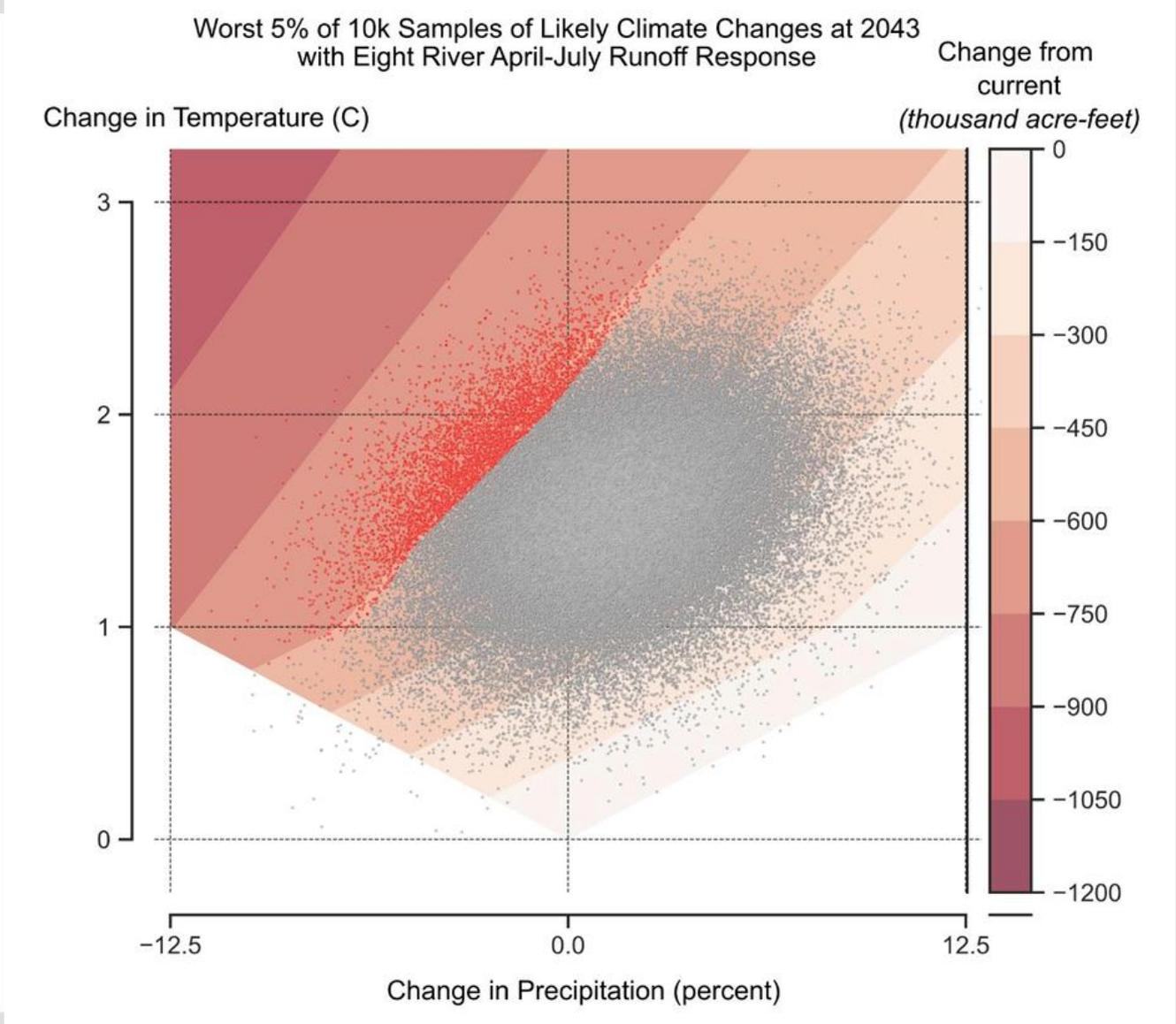


System Risk-Informed Scenarios

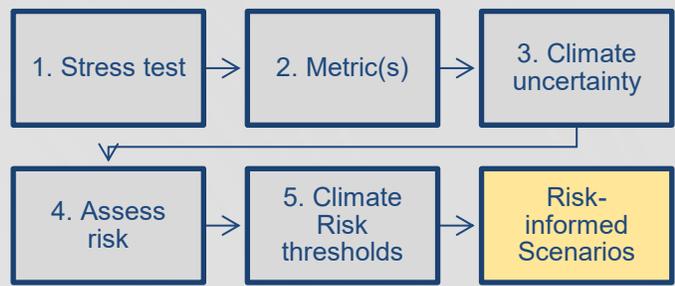


Part 1: Use your screening model to:

- 1. Select a key metric that reflects systemwide risk
- 2. Response surface with climate PDF
- 3. Isolate sampled points along the X percentile “Level of Concern”

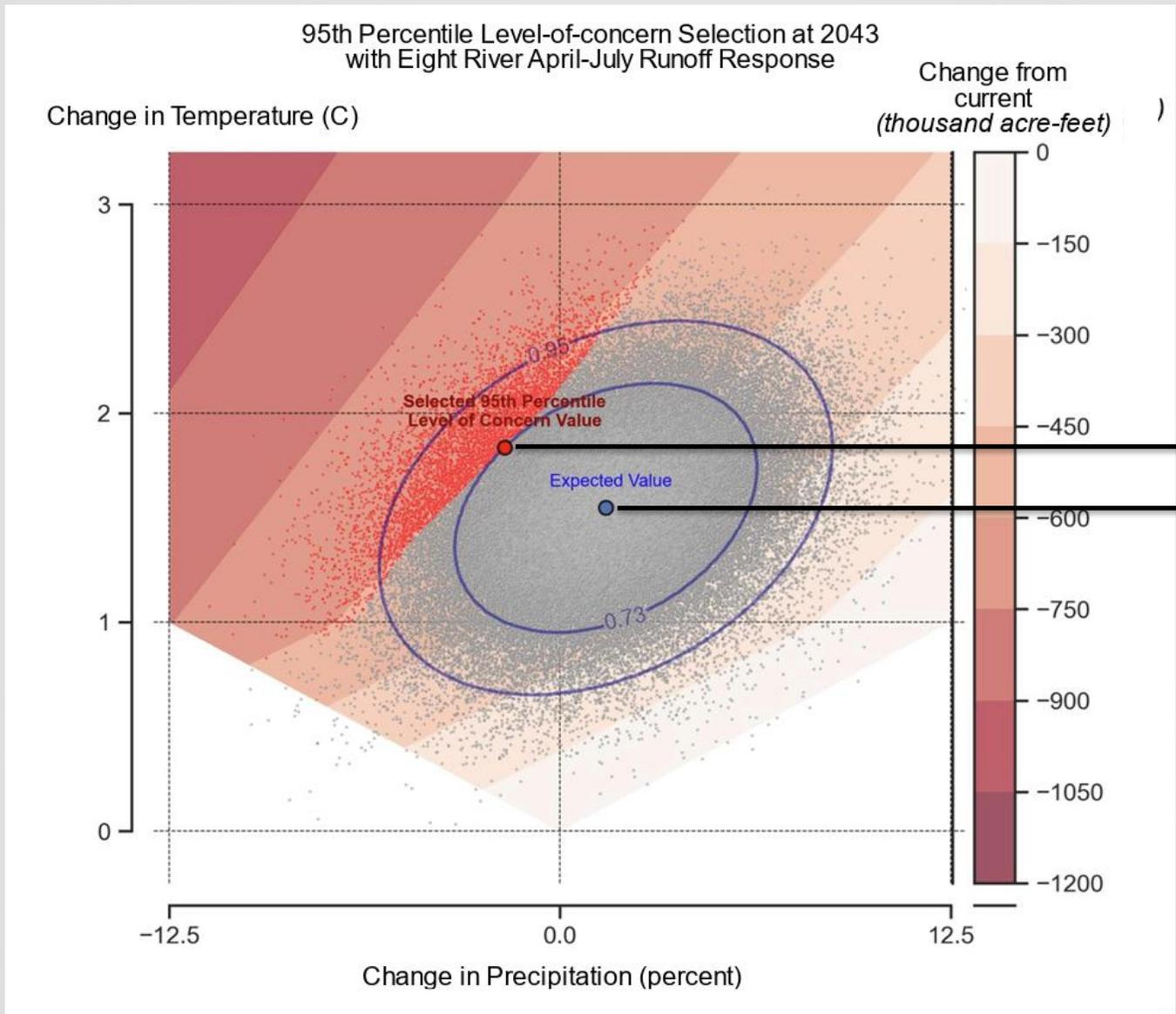


System Risk-Informed Scenarios



Part 1: Use your screening model to:

1. Select a key metric that reflects systemwide risk
2. Response surface with climate PDF
3. Isolate sampled points along the X percentile "Level of Concern"
4. Select the point closest to the center of the climate PDF density



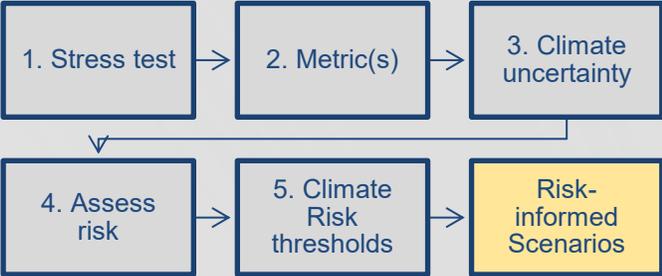
System-Risk Informed T/P Scenarios:

95th LOC: -1.8% Pr, +1.8°C

50th LOC: +1.5% Pr, +1.5°C

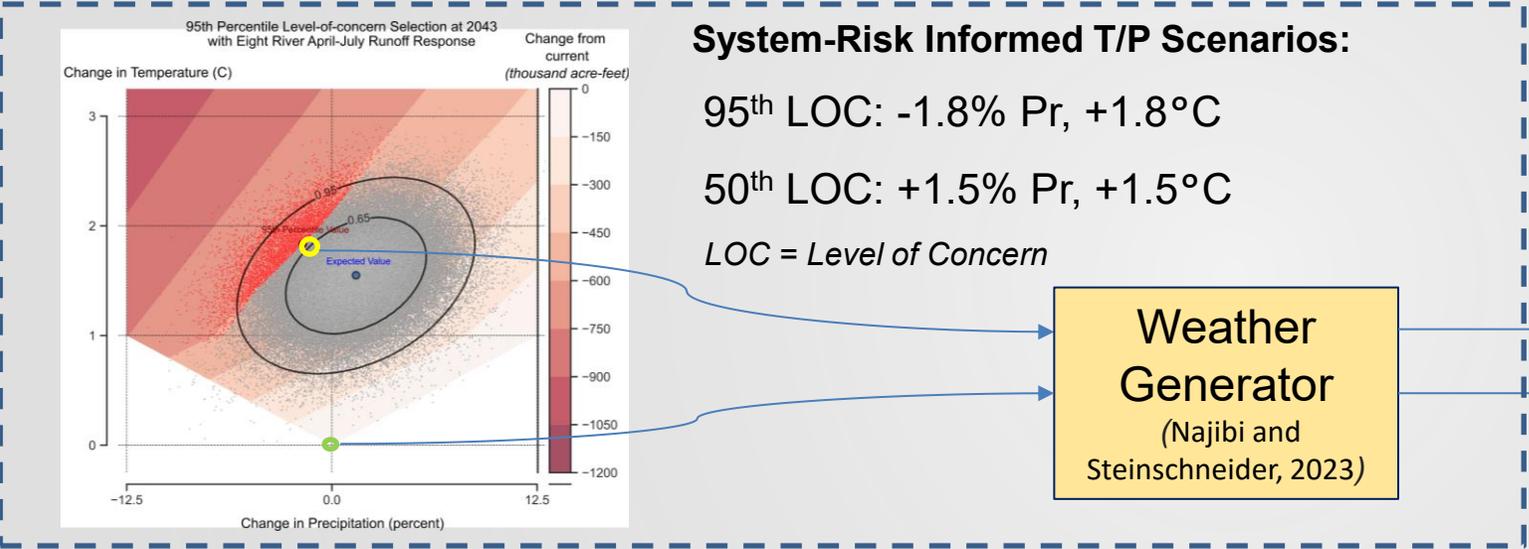
LOC = Level of Concern

System Risk-Informed Scenarios

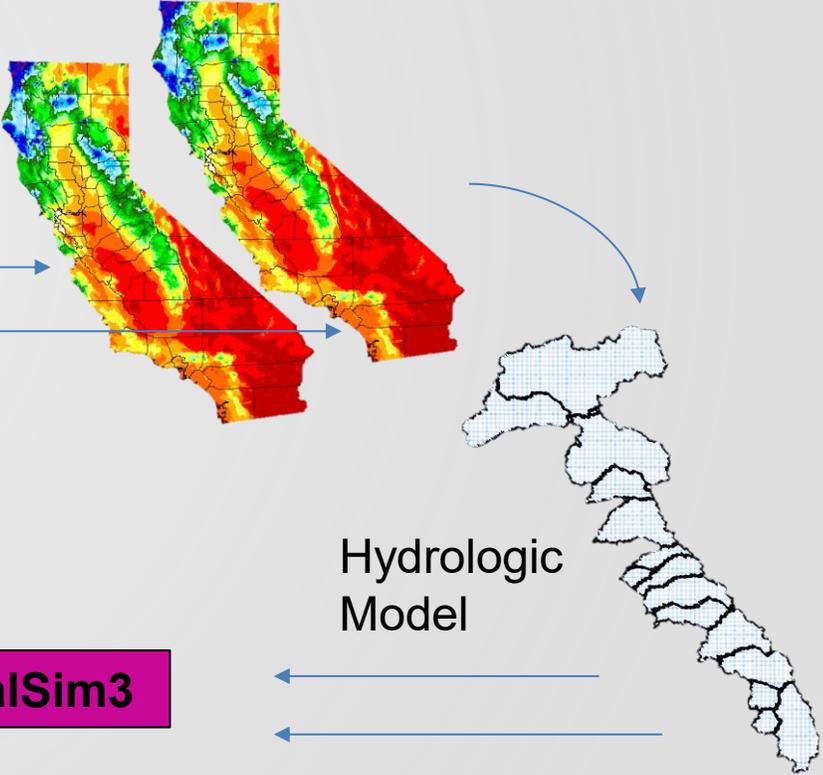


Part 2:

Generate the risk-based T/P scenarios with Weather Generator and input them to the detailed, complex model:



Weather Generator
(Najibi and Steinschneider, 2023)



CalSim3 Climate Scenario
Runs/Operational Adjustments
and Refinements

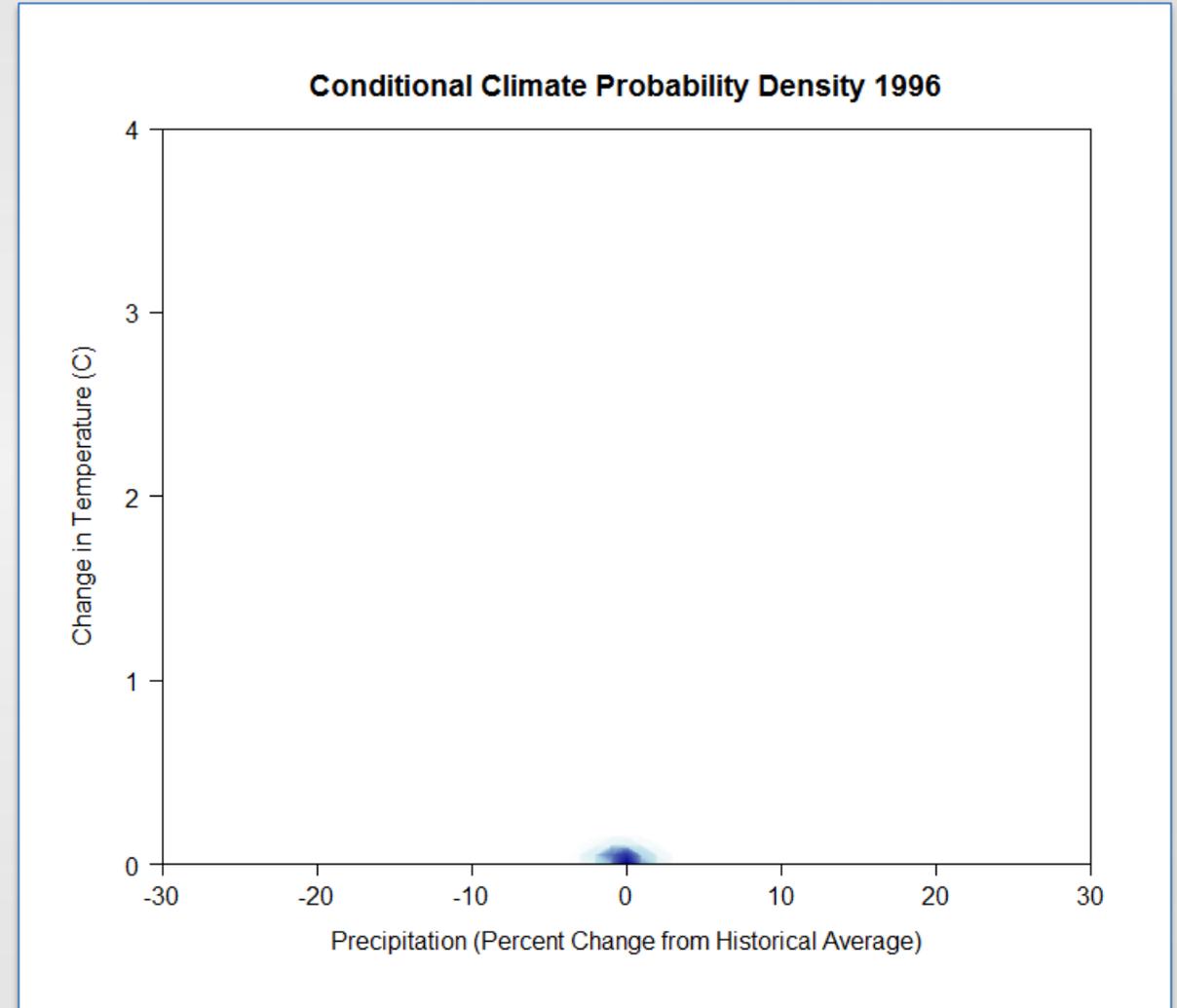
Adjust inputs to CalSim3

Application 1: Water Plan Update 2023



Decision Scaling

- A stress test under the *full range of relevant and credible* changes
- Estimate risk conditional on climate projection-based evidence



How to apply decision Scaling

1. Select an appropriate model
2. Identify a representative hydrology
3. Simulate conditions and extract model results
4. Plot model results for metrics on response surface
5. Apply GCM projections to response surface to develop probability of occurrence
6. Sum probabilities of occurrence to develop CDFs for risk-based assessments
7. Identify a risk threshold and assess adaptations to meet that threshold

Example models for decision scaling analysis

1. WEAP

Simulates physically based interactions through a simplified model allowing for very low run times

2. CalLite

Simplified version of CalSim very effective for anything depending on SWP or CVP deliveries

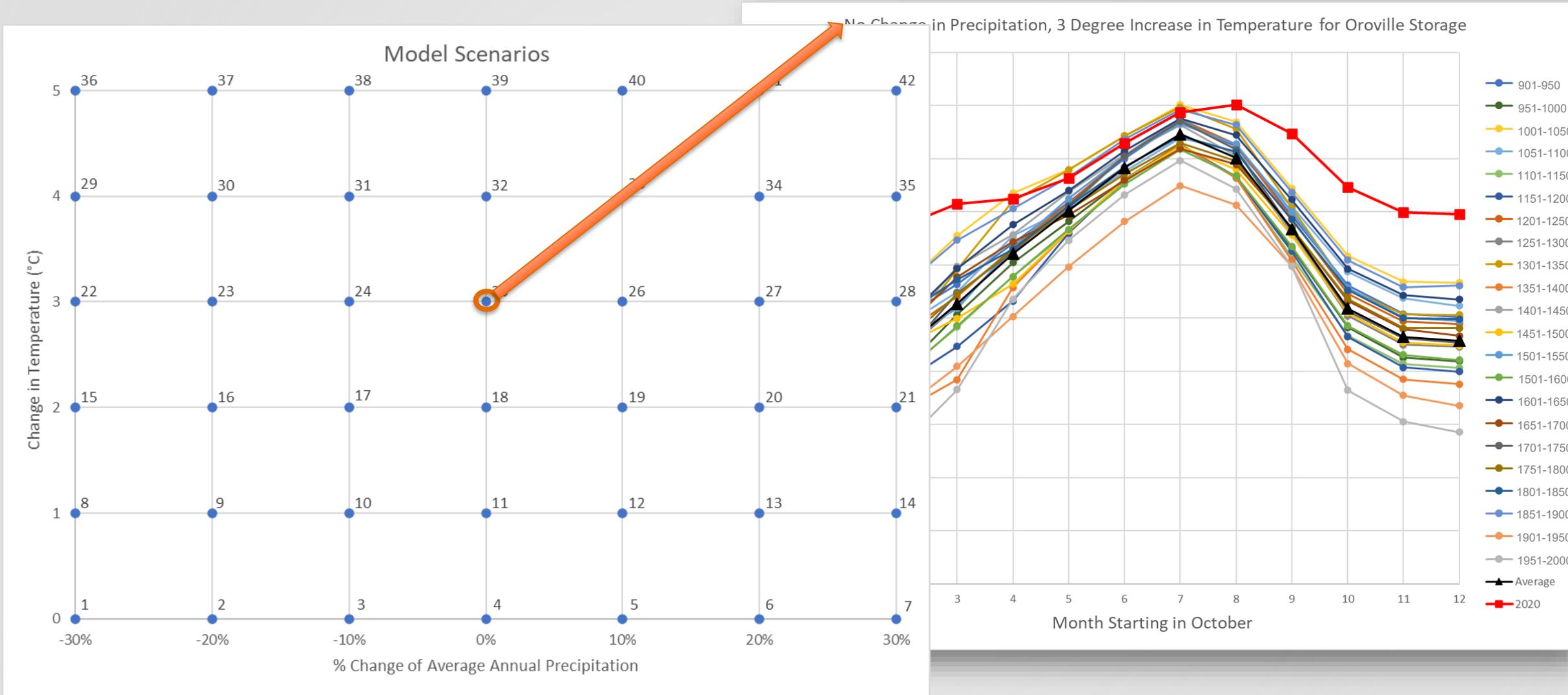
3. Other models with shorter run times

Theoretically, any model with enough cloud computing could run fast enough to do decision scaling, just be aware of when cost of cloud computing exceeds benefit of the more detailed modeling

Identify a representative hydrology

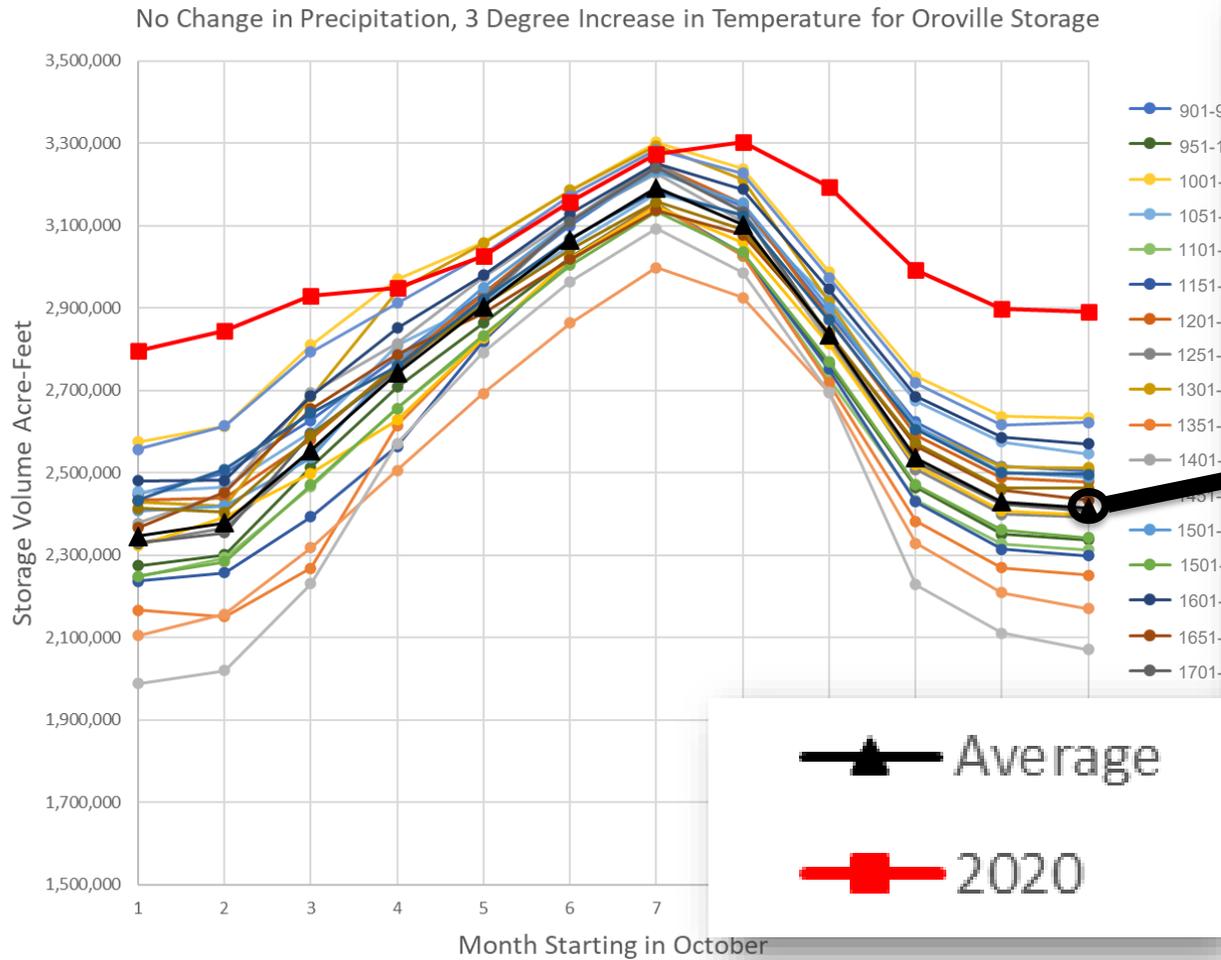
1. Gridded Weather Generator Perturbations of Historical Detrended and Stochastically Generated Temperature and Precipitation for the State of CA and HUC8s
2. Paleo-Dendrochronological "Tree-Ring" Hydroclimatic Reconstructions for Northern and Southern California River Basins
3. Other datasets as available that capture possible future conditions

Simulate conditions and extract model results

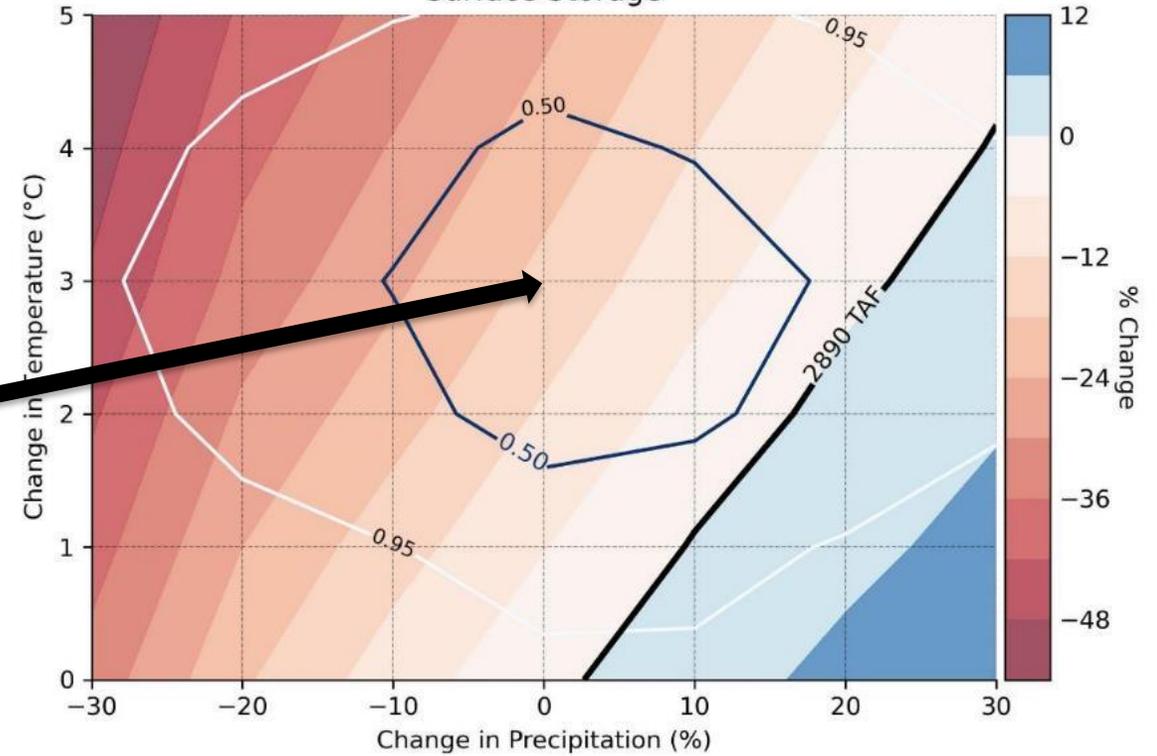


Each run represents 50 years where the only change is the long-term climate variability. Average results are then developed from the runs. Model was reset every 50-years to mitigate any effects of long term draw down.

Plot model results for metrics on response surface



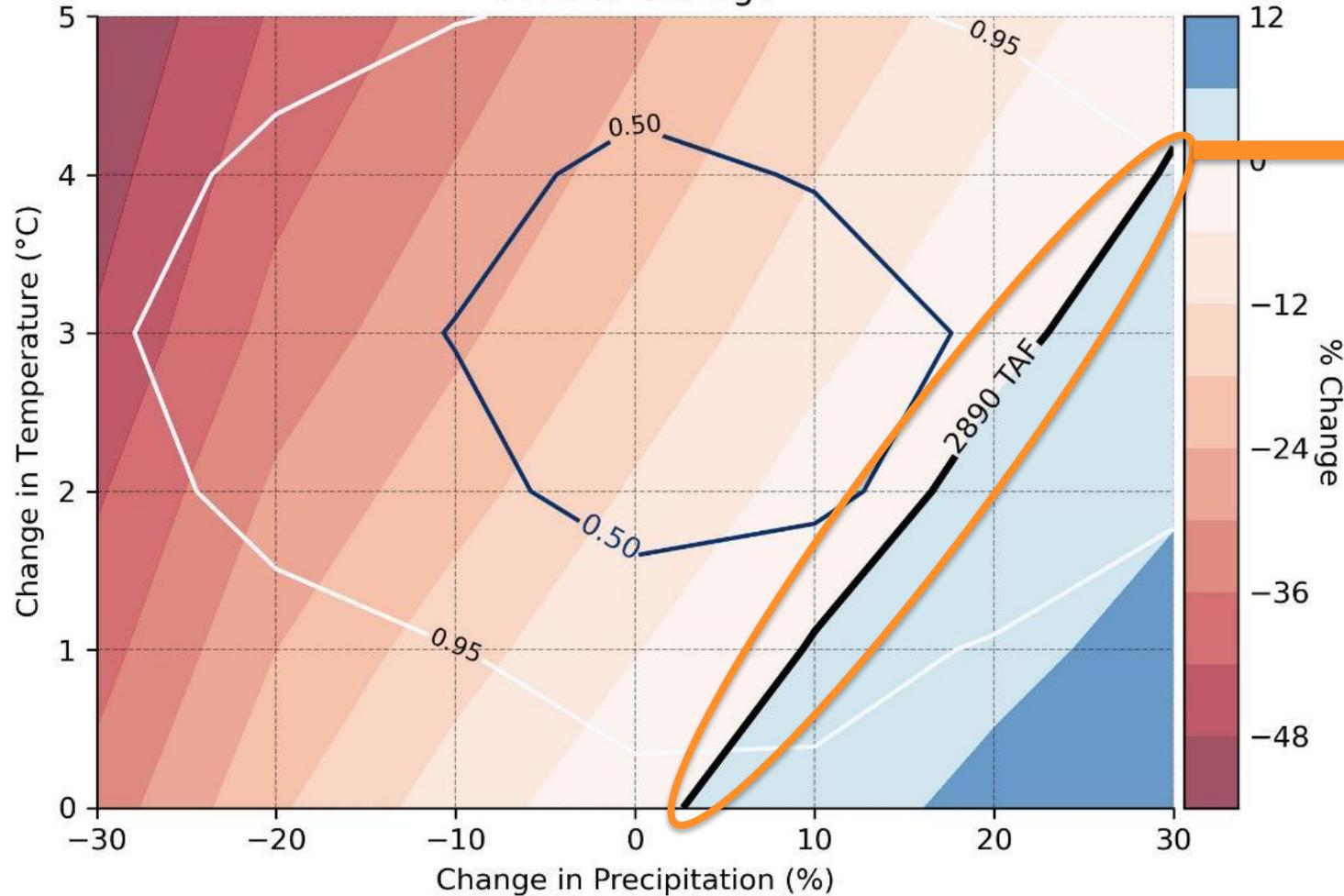
Percent Change in Oroville Reservoir | End of Water Year (September)
Surface Storage



**APPLY GCM PROJECTIONS TO
RESPONSE SURFACE TO
DEVELOP PROBABILITY OF
OCCURRENCE**

Metric 1 Surface Storage – Oroville

Percent Change in Oroville Reservoir | End of Water Year (September)
Surface Storage

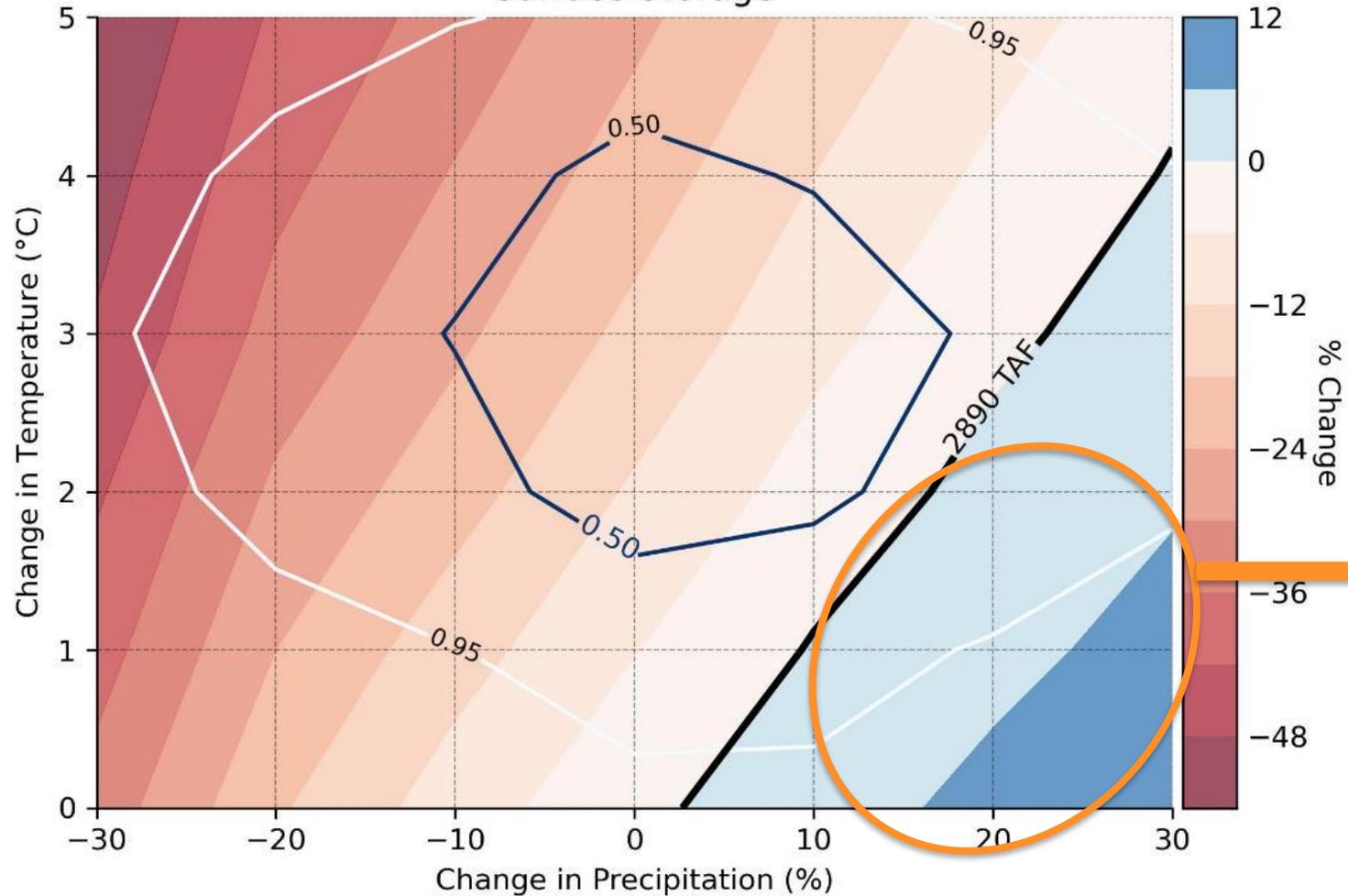


The black line represents the combination of climate conditions that would result in performance similar to existing conditions. For this study, existing conditions is the target performance threshold.

Color bars represent lines of consistent performance.

Metric 1 Surface Storage – Oroville

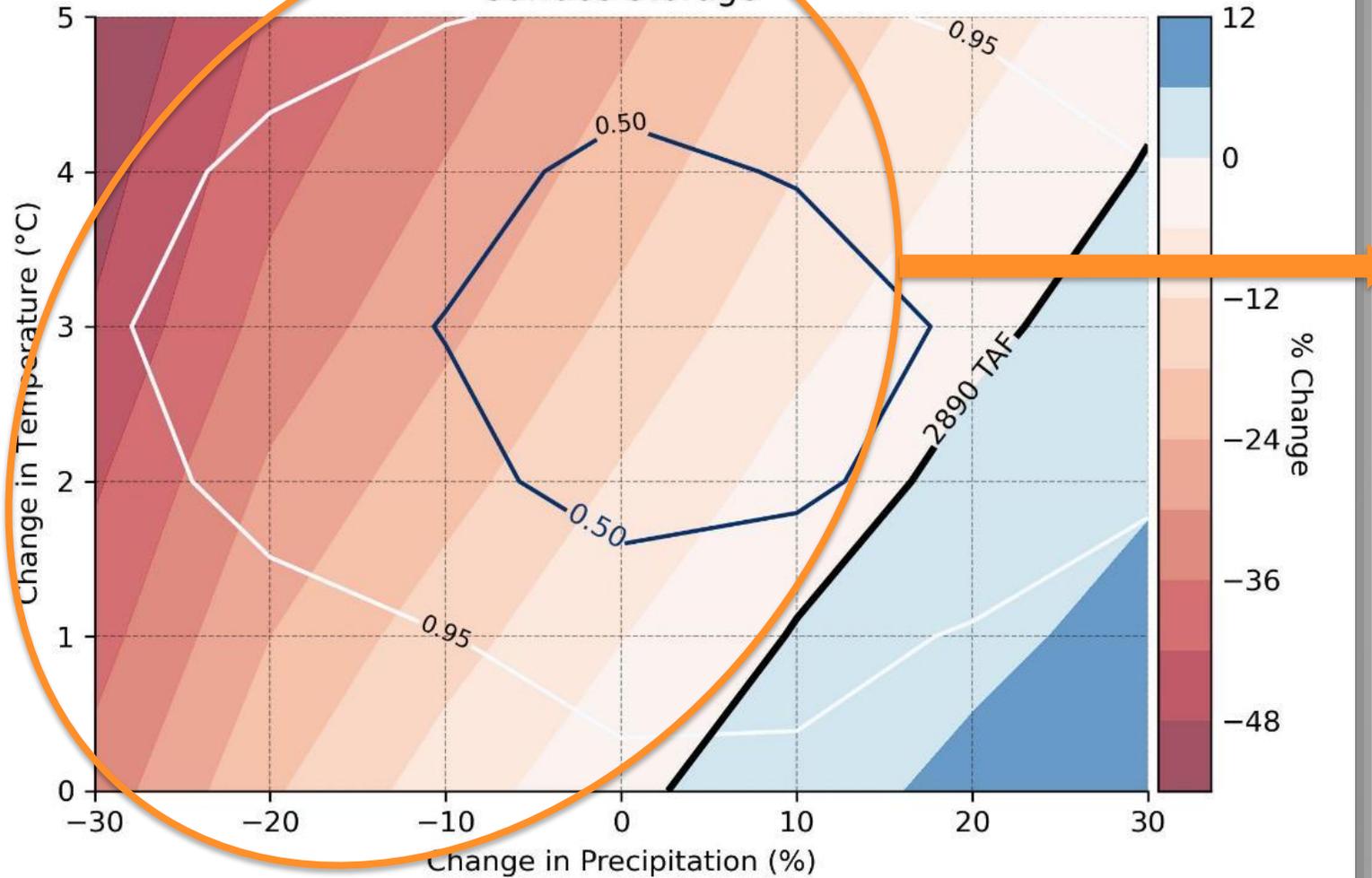
Percent Change in Oroville Reservoir | End of Water Year (September)
Surface Storage



Blue areas indicate improved system performance

Metric 1 Surface Storage – Oroville

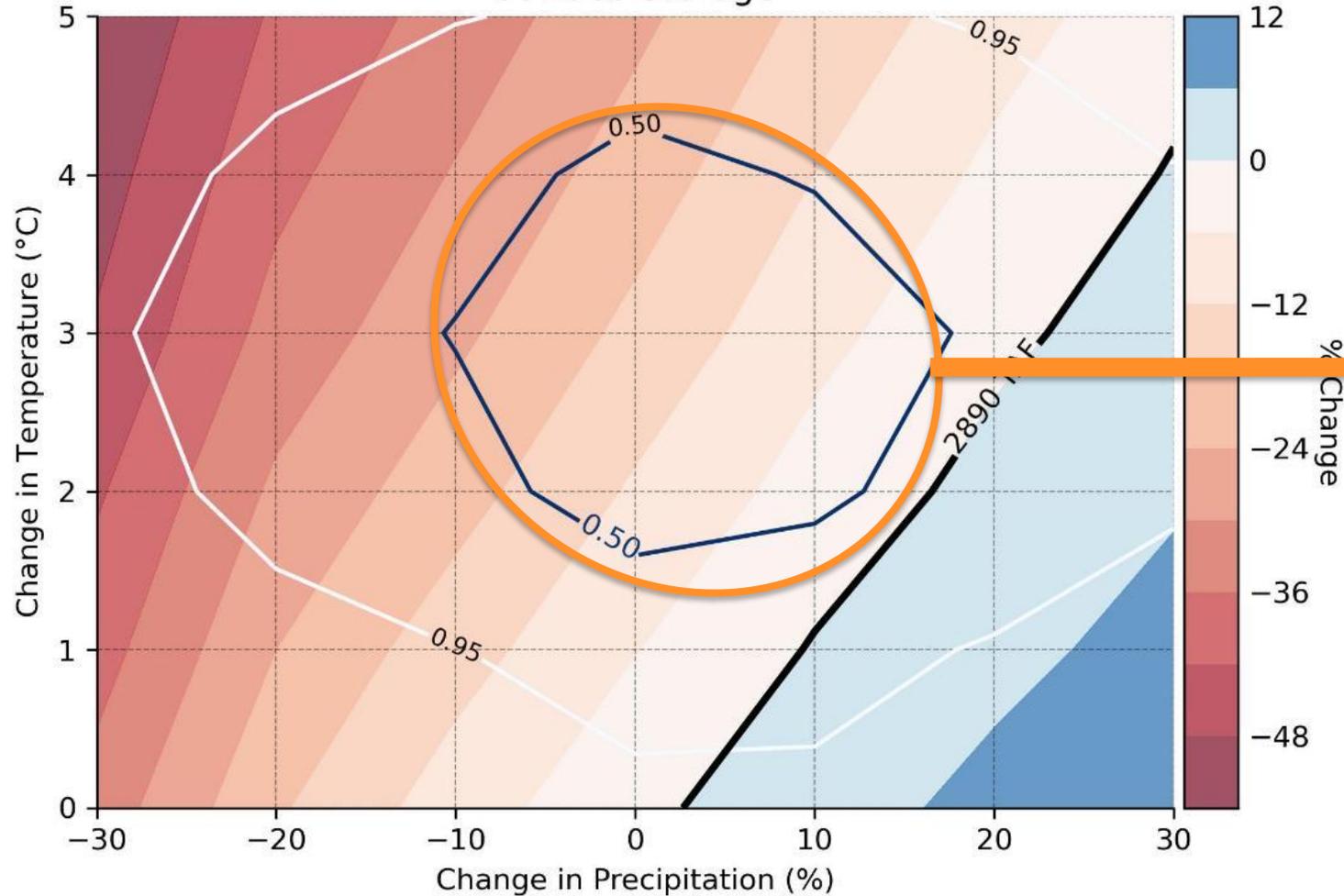
Percent Change in Oroville Reservoir | End of Water Year (September)
Surface Storage



Red areas indicate areas of decreased performance.

Metric 1 Surface Storage – Oroville

Percent Change in Oroville Reservoir | End of Water Year (September)
Surface Storage



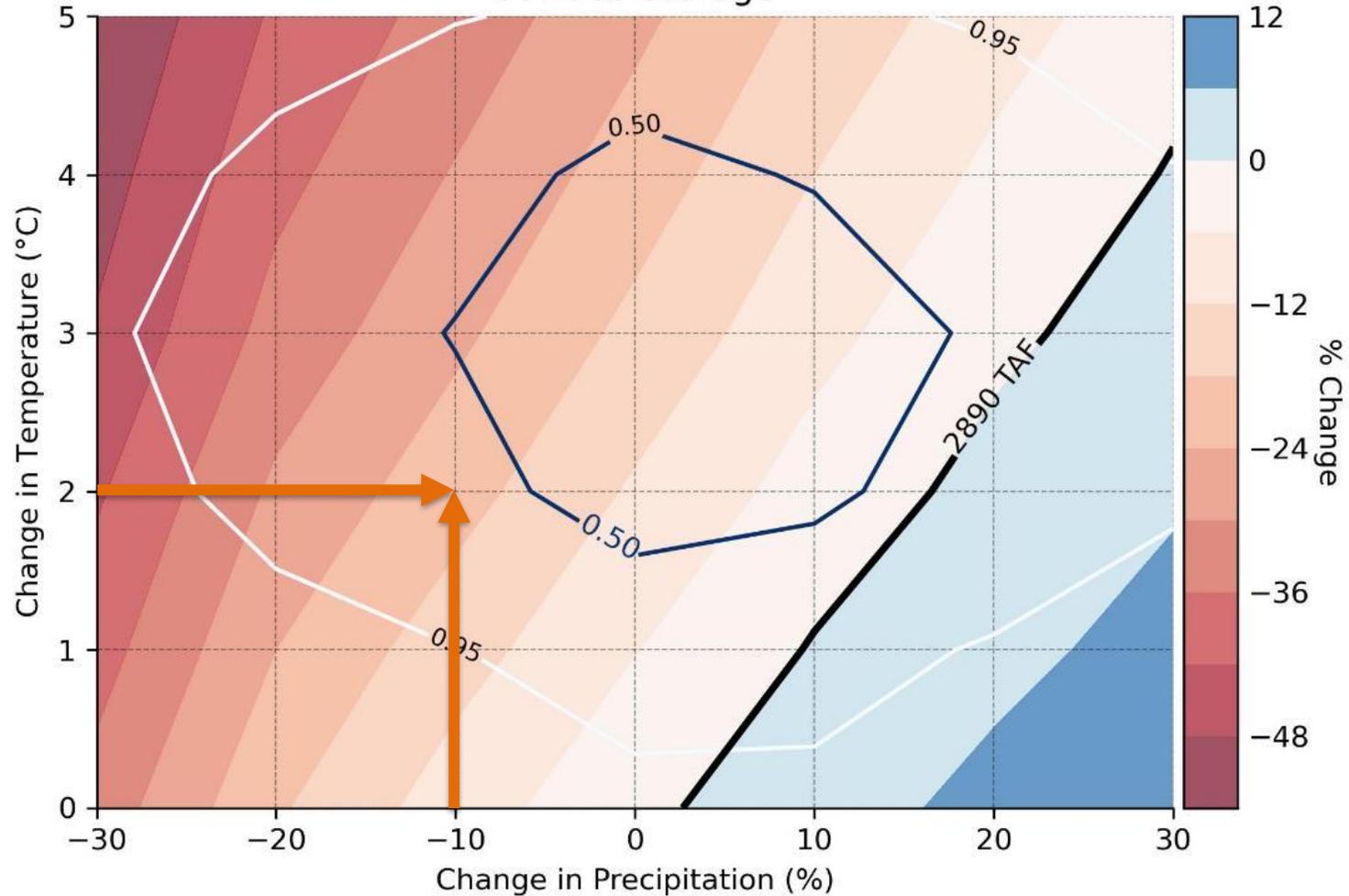
The contours represent the probability of occurrence based on GCM projections.

The center of the circle is the central tendency or expected value.

Areas further out are plausible but less and less likely the further from the center.

Metric 1 Surface Storage – Oroville

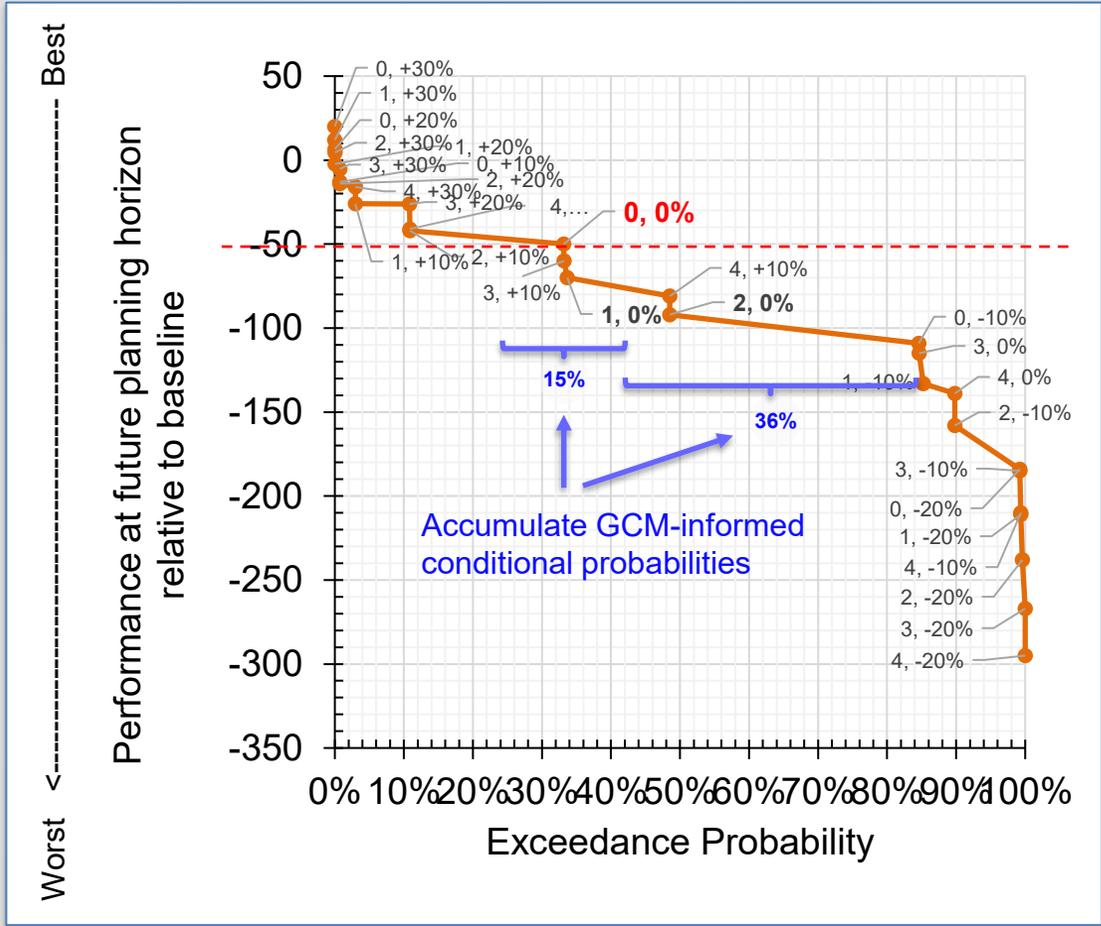
Percent Change in Oroville Reservoir | End of Water Year (September)
Surface Storage



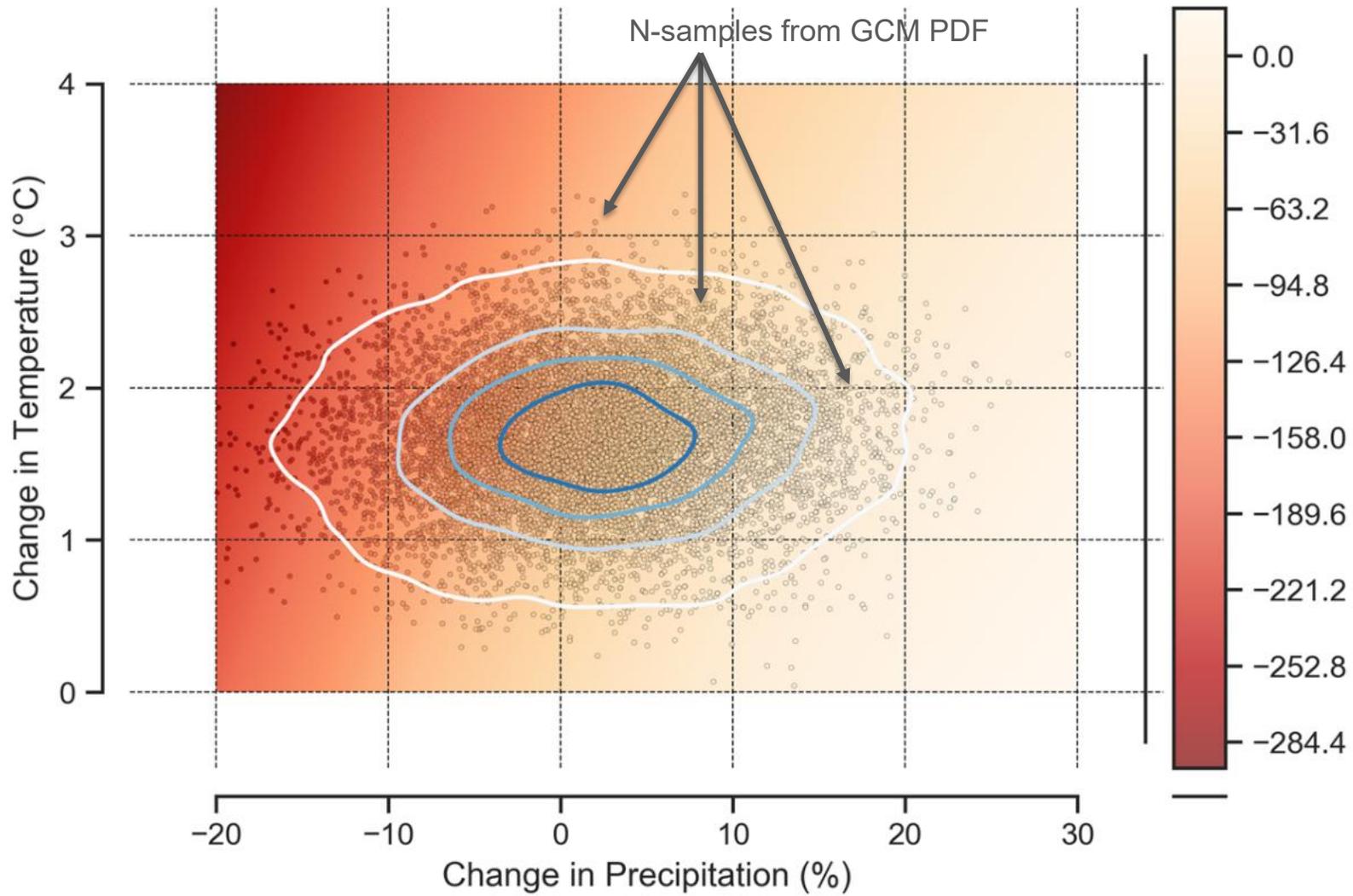
Response surfaces can be used to look at any combination of temperature and precipitation you are interested in regardless of probability of occurrence.

Response surfaces themselves are based on system parameters and are independent of the GCM probabilities.

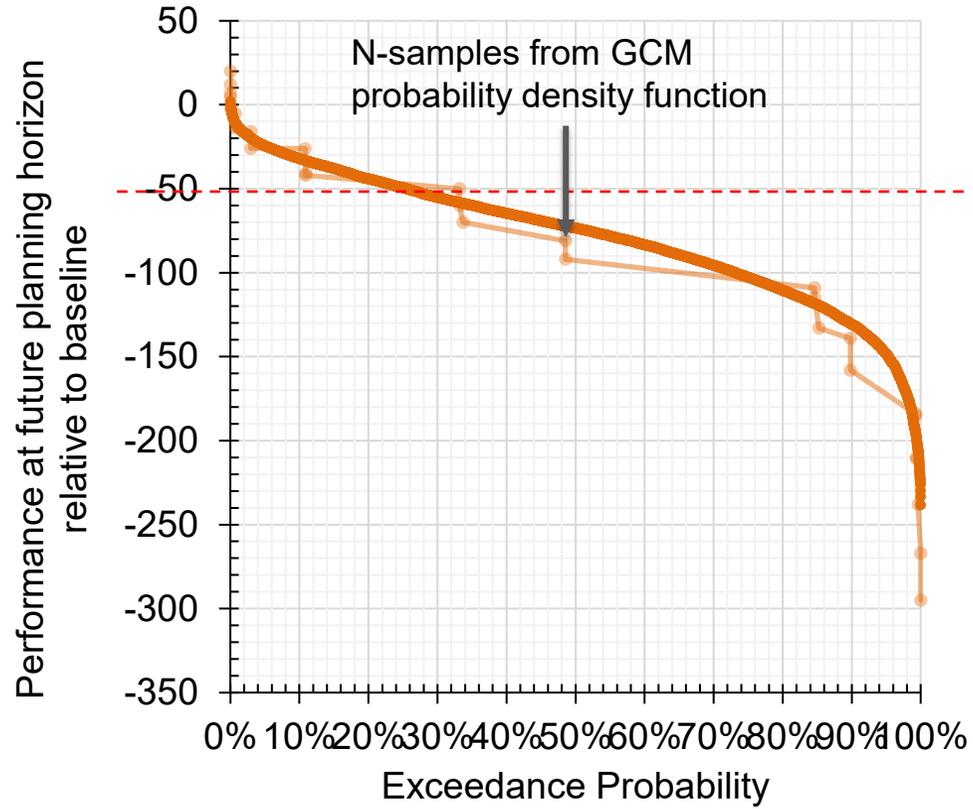
**SUM PROBABILITIES OF
OCCURRENCE TO DEVELOP
CDFs FOR RISK-BASED
ASSESSMENTS**



Some important system metric



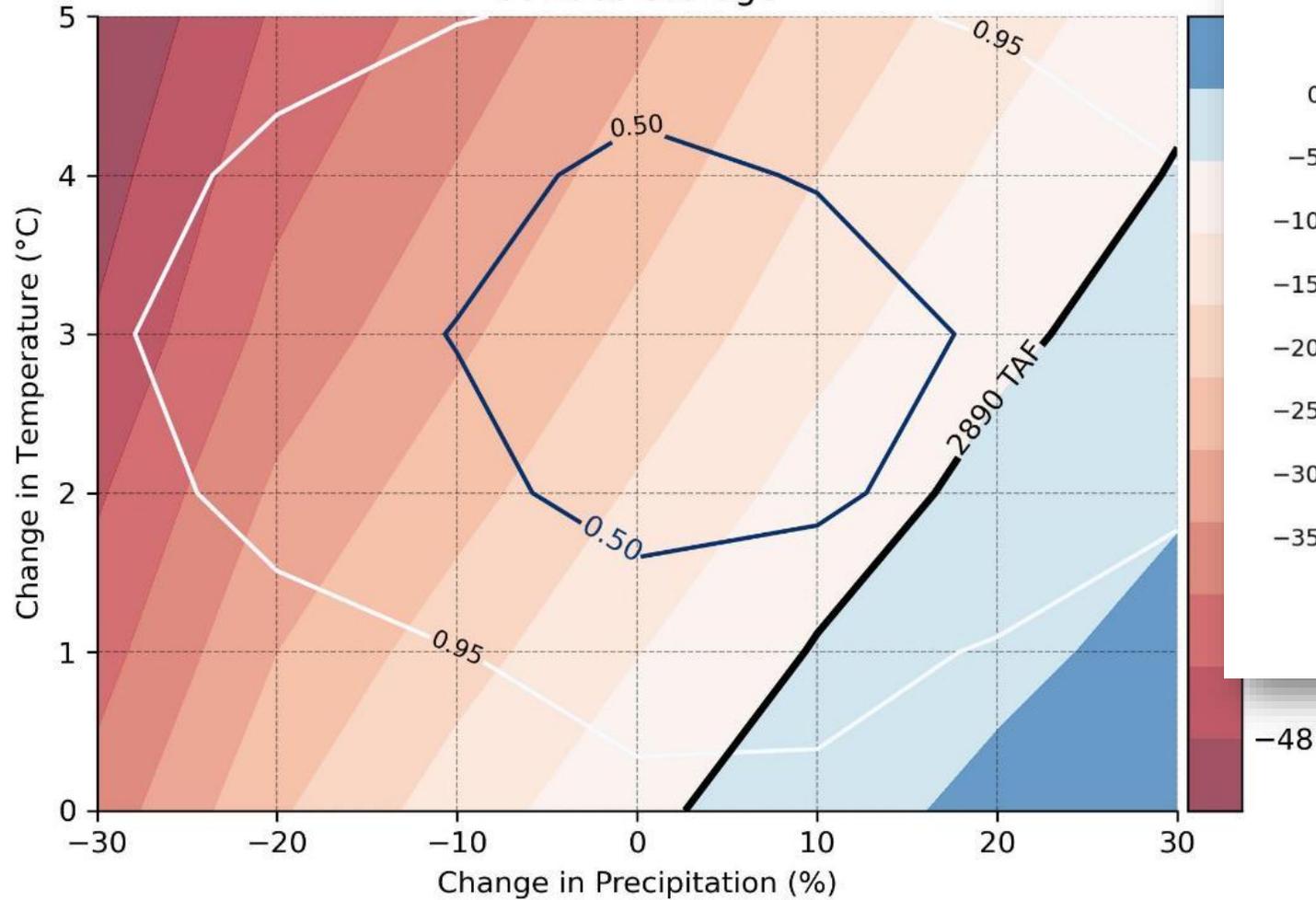
Worst <----- Best



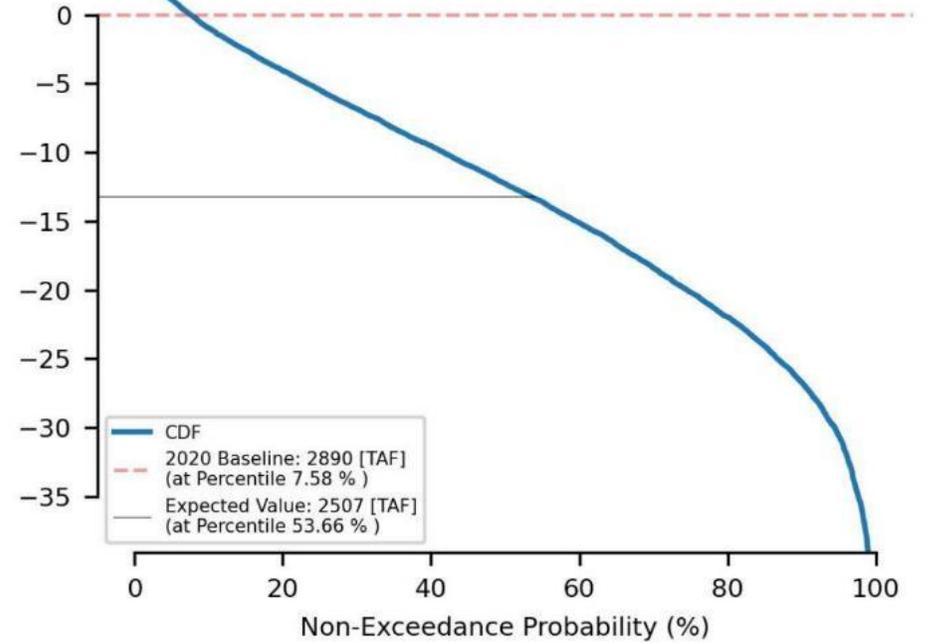
ECDF (rank/n) ----->

Metric 1 Surface Storage – Oroville

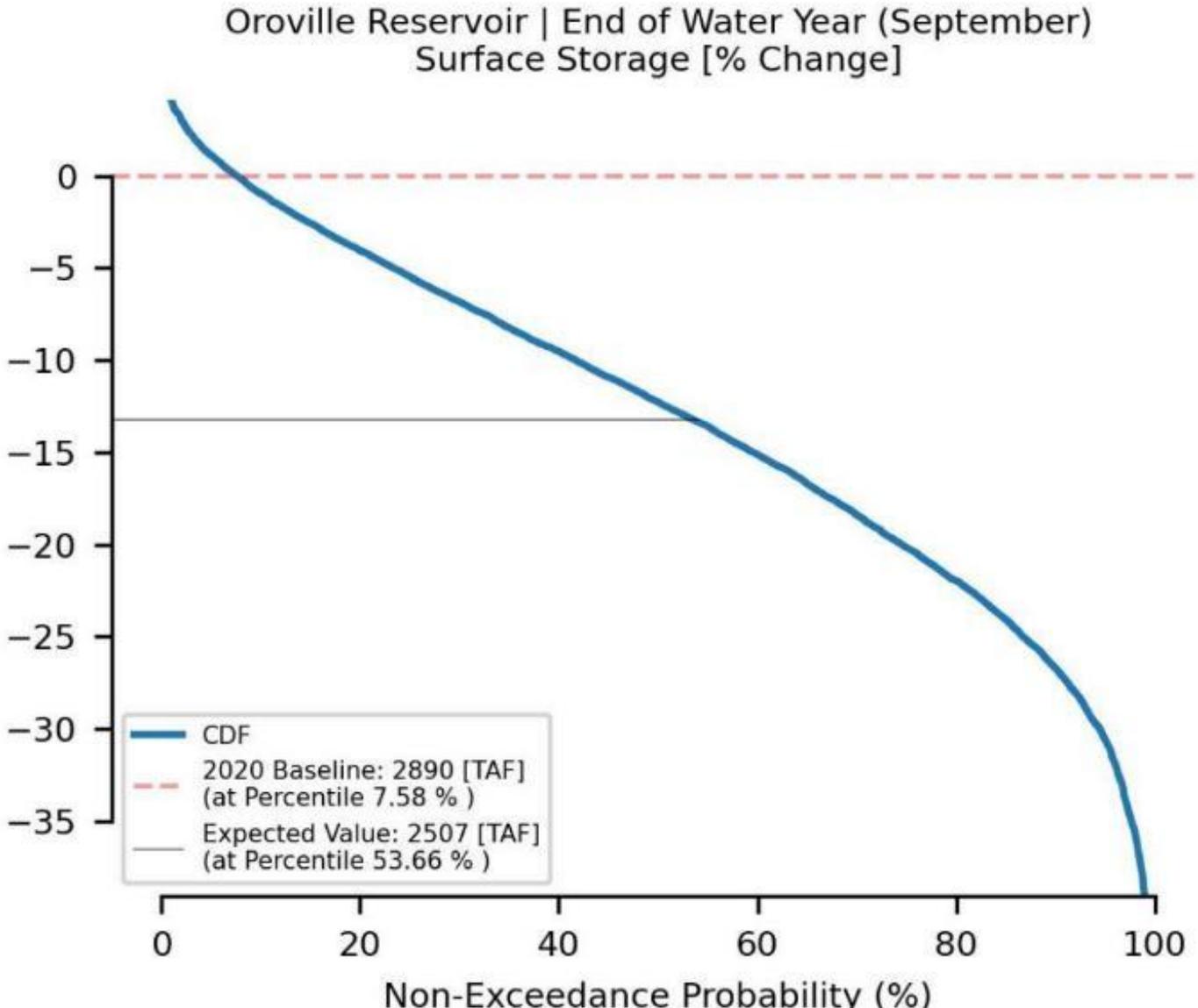
Percent Change in Oroville Reservoir | End of Water Year (September)
Surface Storage



Oroville Reservoir | End of Water Year (September)
Surface Storage [% Change]



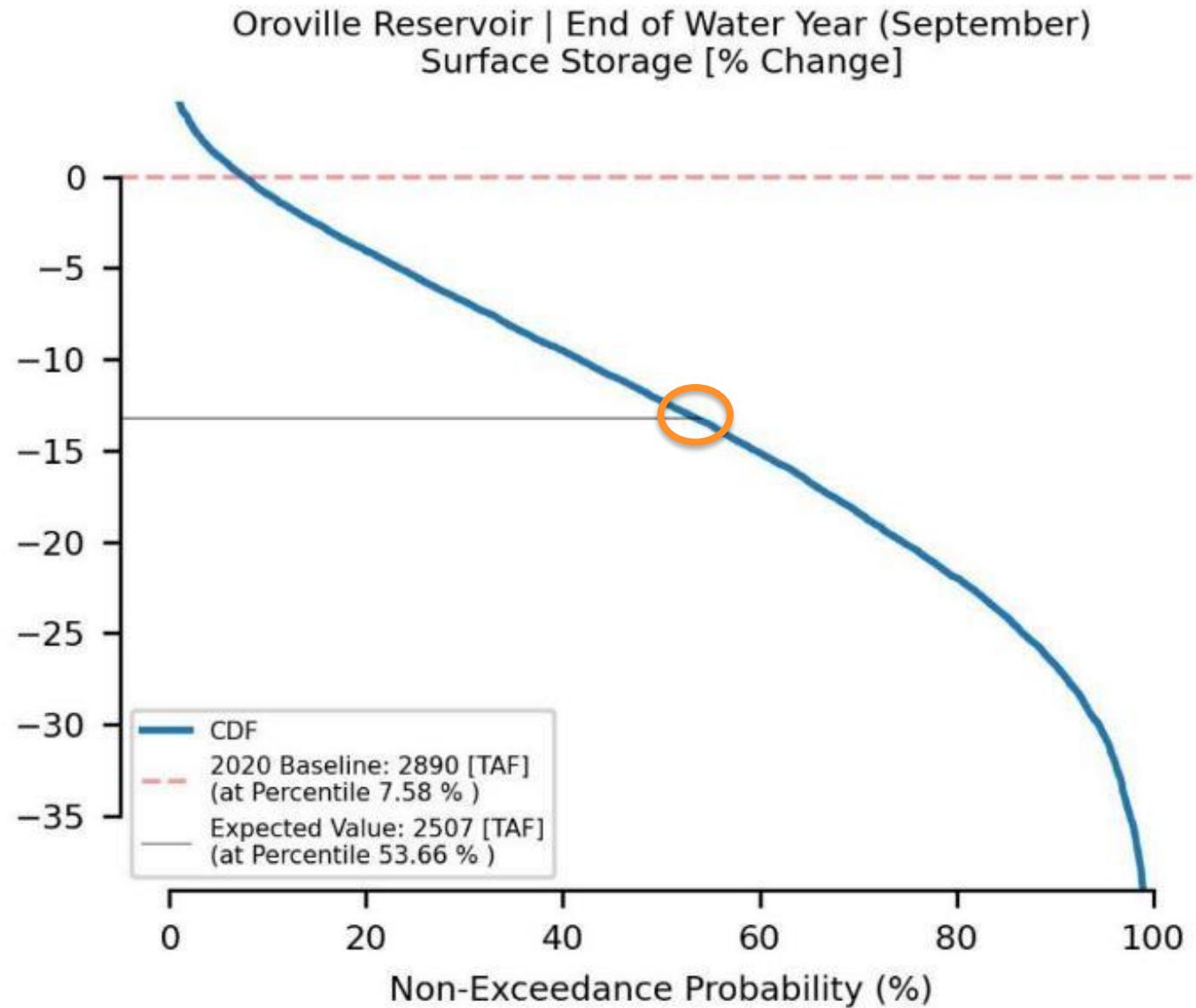
Metric 1 Surface Storage – Oroville



CDFs provide a probabilistic framework for the outcomes.



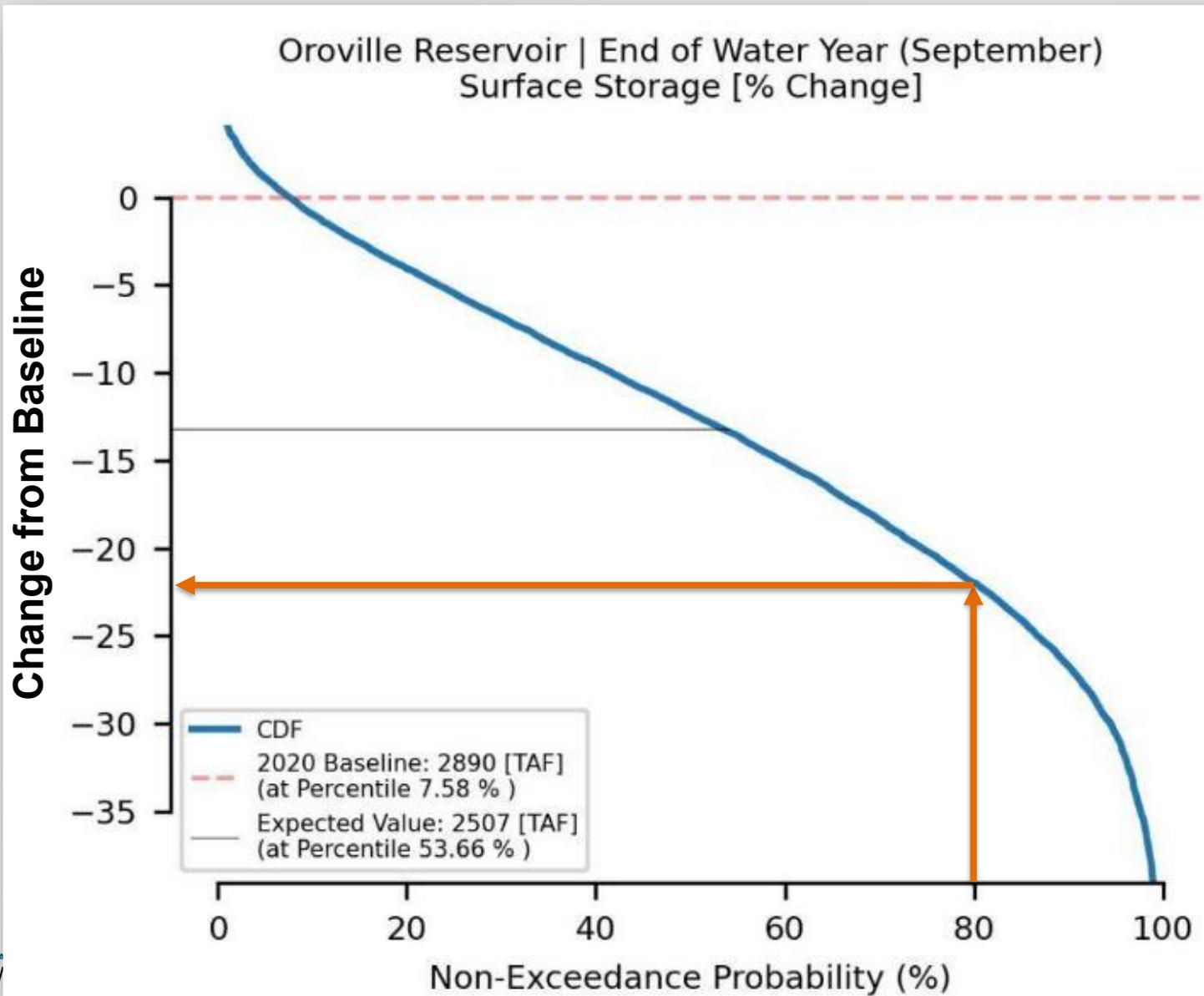
Metric 1 Surface Storage – Oroville



CDFs provide a probabilistic framework for the outcomes. These let you identify most likely conditions ...

**IDENTIFY A RISK THRESHOLD
AND ASSESS ADAPTATIONS TO
MEET THAT THRESHOLD**

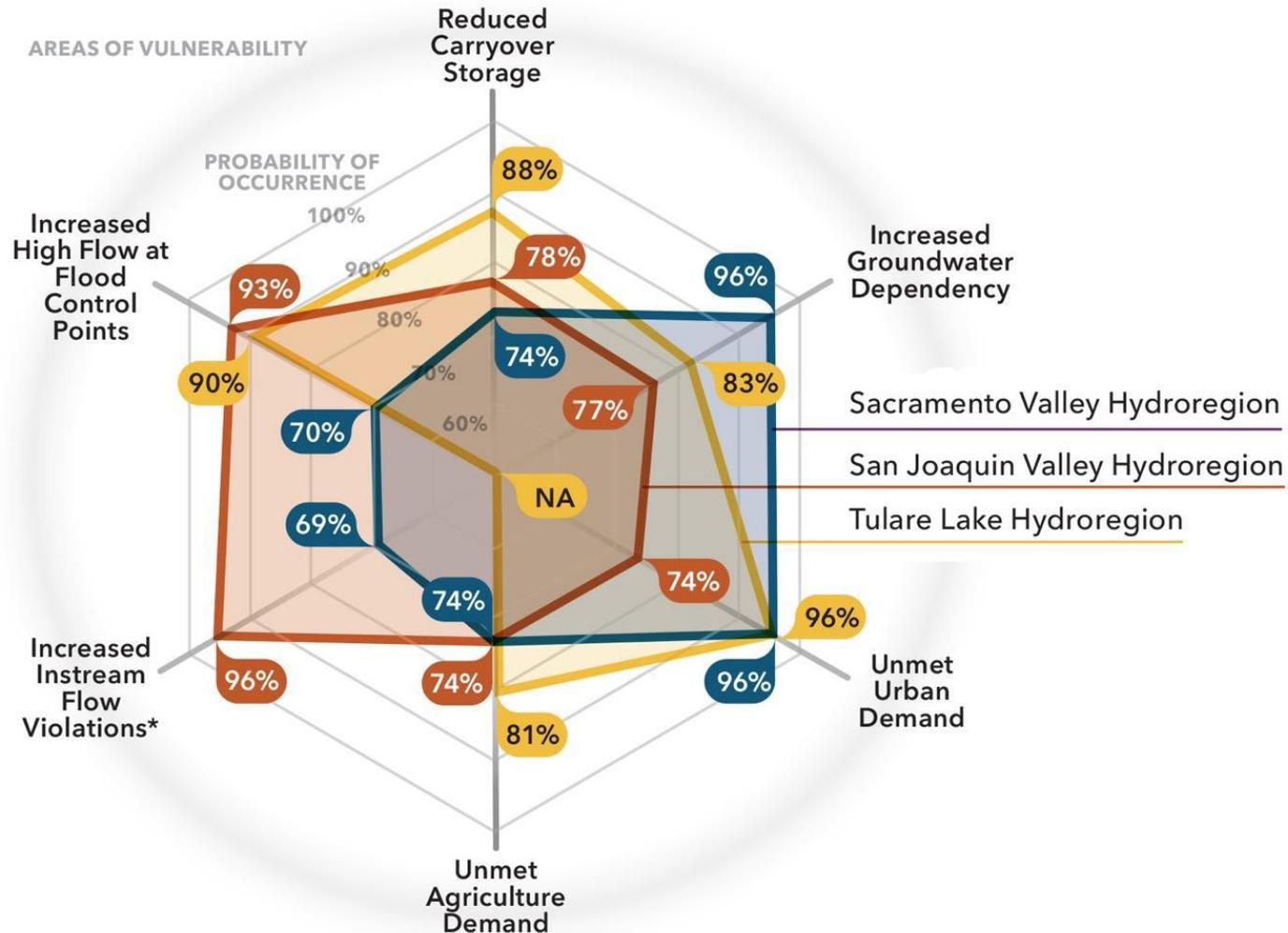
Metric 1 Surface Storage – Oroville



CDFs provide a probabilistic framework for the outcomes. These let you identify most likely conditions but they also let you set your own risk thresholds.

Figure 2-8 The Central Valley Is Likely to be Increasingly Vulnerable in 2070 Based on Six Water Metrics

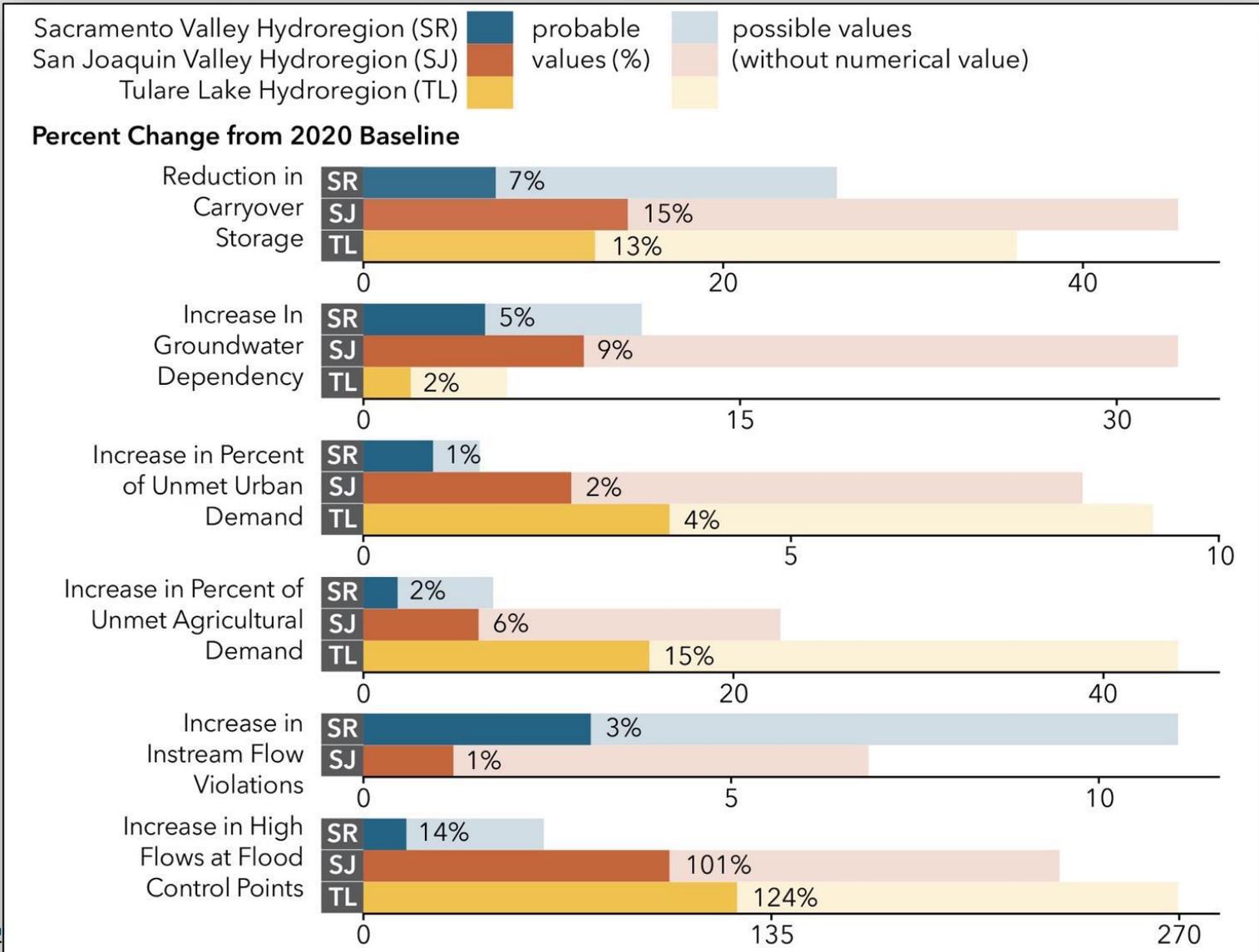
Probability of Increased Vulnerability of Conditions by 2070



Results of decision scaling analysis as depicted in the California Water Plan Update 2023.

This figure provides the likelihood of future conditions being worse than current conditions in 2070 for each metric.

Figure 2-9 Bar Plot Shows Large Magnitude Changes for Six Water Metrics by 2070



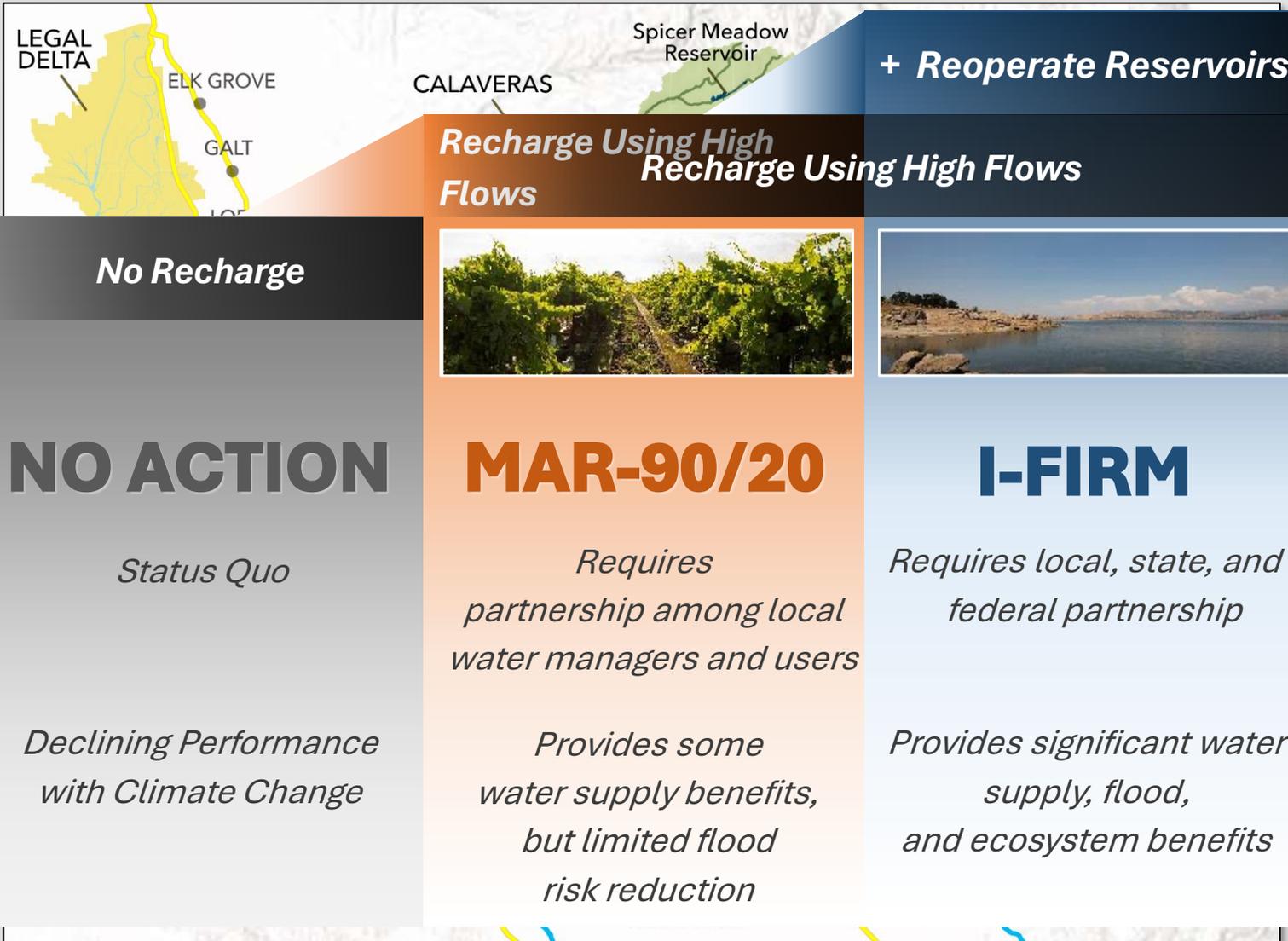
Results of decision scaling analysis as depicted in the California Water Plan Update 2023.

This figure provides the expected magnitudes of change as well as the 95th percentile value for extreme possible changes in the metrics.

Application 2:
San Joaquin Basin Flood-MAR
Watershed Studies



San Joaquin Watershed Studies Overview



- Five watersheds on the eastern side of the San Joaquin Basin
- 16 climate conditions
- 100-year hydrologic sequence
- Three recharge actions
 1. **No Action** – status quo, maintain baseline operations
 2. **MAR 90/20** – opportunistically recharge high flows
 3. **I-FIRM** – integrate recharge and forecast-informed reservoir operations

Study Objectives

**ASSESS AND QUANTIFY
MULTI-SECTOR VULNERABILITY
TO CLIMATE CHANGE**

**ASSESS AND QUANTIFY
MULTI-SECTOR FLOOD-MAR
ADAPTATION PERFORMANCE**

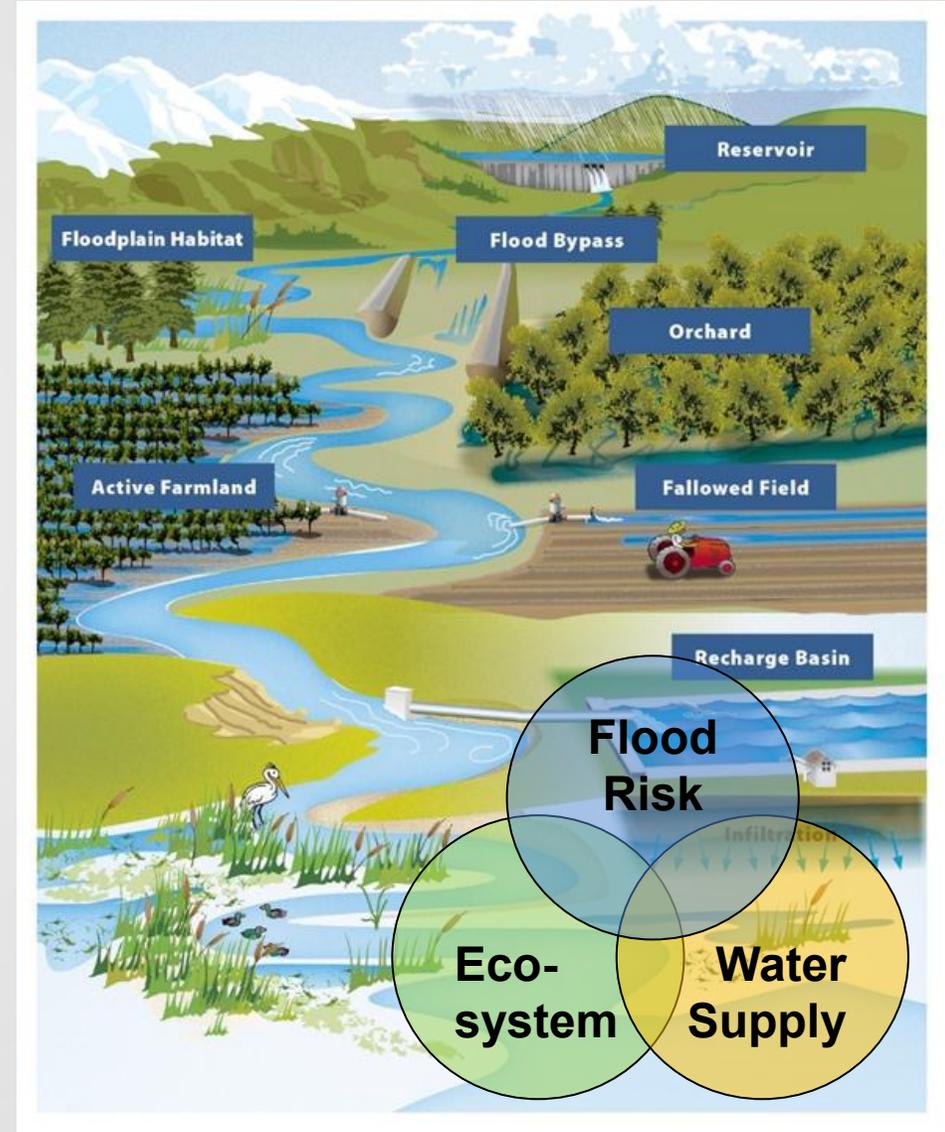
**WATERSHED
STUDIES**

**REPORT RESULTS FROM THE
LOCAL-WATERSHED-
BASINWIDE SCALE**

**FOUNDATION TO SUPPORT
COORDINATION, ALIGNMENT
AND IMPLEMENTATION**

Metric Selection Process

- Track systemwide performance
- Three major water management sectors
- Reliably measured using the available toolset resolution



Performance Metrics

182 Metrics

- Watershed Conditions – 13
- Flood Risk – 18
- Groundwater – 32
- Surface Water – 29
- Ecosystem – 67
- Flood-MAR – 23

174 Reporting Units

- GW Sub-basin – 6
- SW Watershed – 5
- District – 38
- Reservoir – 8
- Control Point – 17
- Stream Reach (GW) – 8
- ResSim Diversion – 59
- WAFR Source – 33



1632
Reported
Metrics



3
Adaptation
Strategies



16
Climate
Conditions



100-year
Continuous
Timeseries



**7.8
Million**
data points at
annual timestep

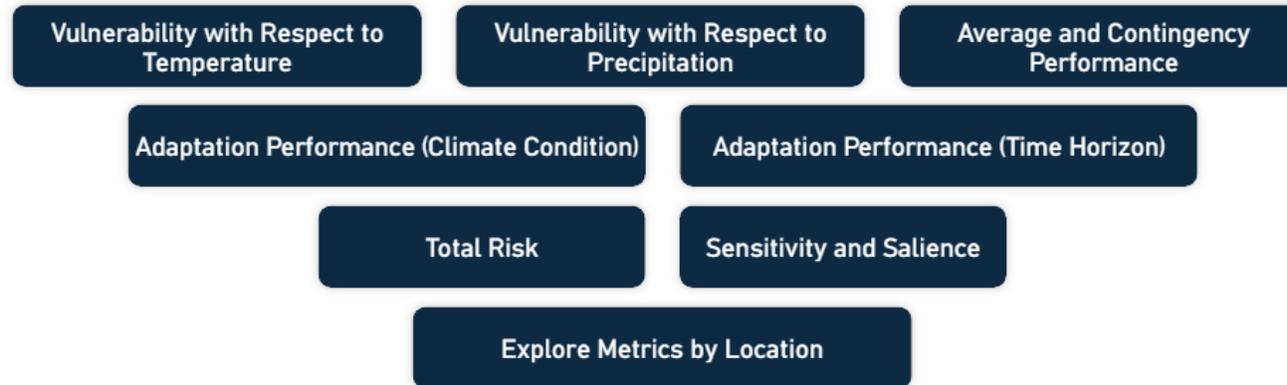
Performance to Planning



CALIFORNIA DEPARTMENT OF
WATER RESOURCES

METRICS DASHBOARD SAN JOAQUIN BASIN WATERSHED STUDIES

Start by selecting a page below.



Disclaimer:

The information presented on this dashboard is intended for **planning-level purposes only** and is provided as an **extension of the results documented in the Watershed Study Reports** available on the California Department of Water Resources' (DWR) website at <https://water.ca.gov/programs/all-programs/flood-mar>. The results displayed reflect the application of various models and analytical approaches based on a series of **specific study assumptions** regarding hydrology, infrastructure, operations, and other input parameters.

Dashboard Design Objectives

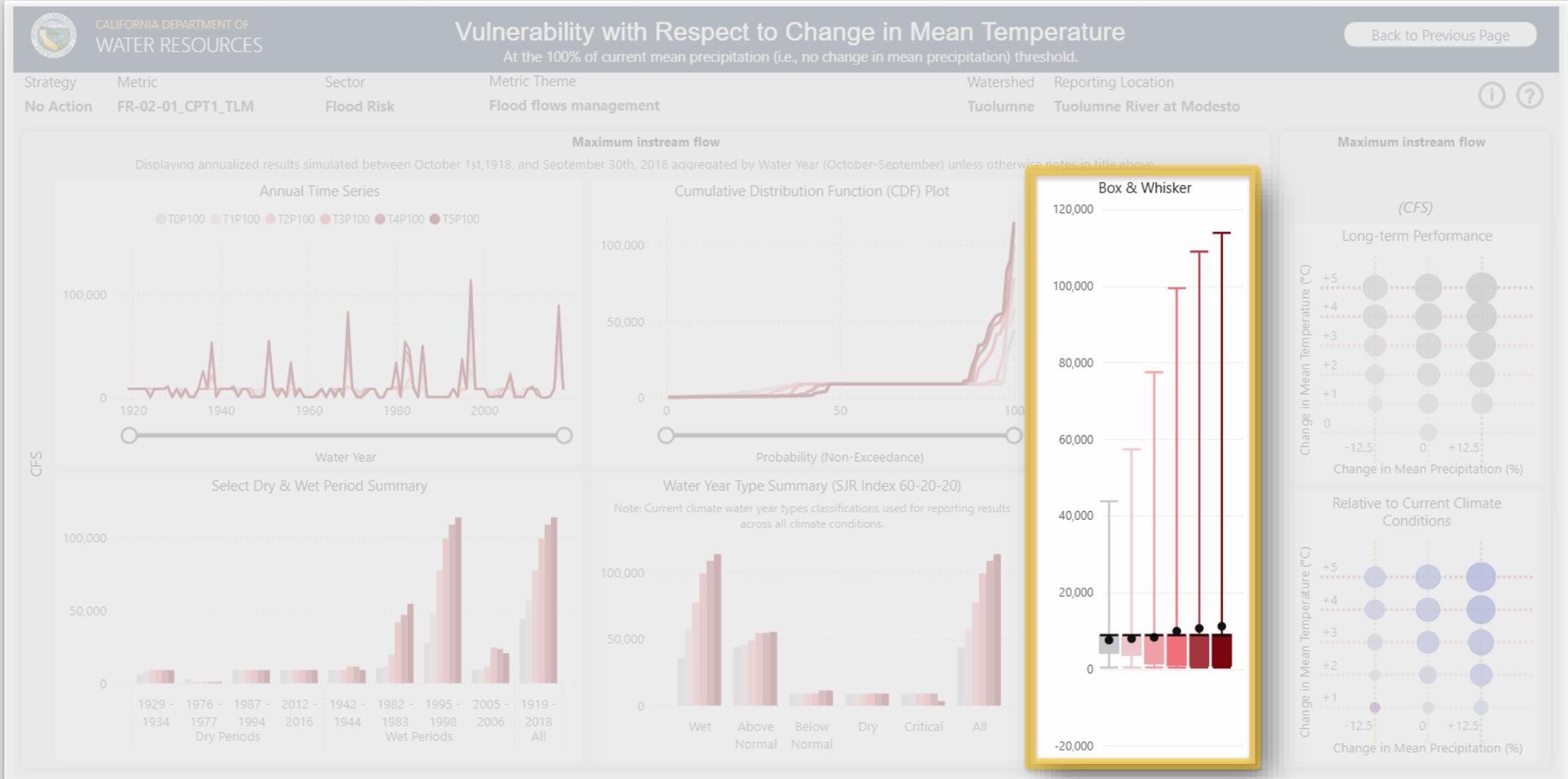
1. Understand performance trends
2. Evaluate projected future performance
3. Assess sensitivity
4. Demonstrate adaptation potential

Performance Trends

Vulnerability with Respect to Temperature

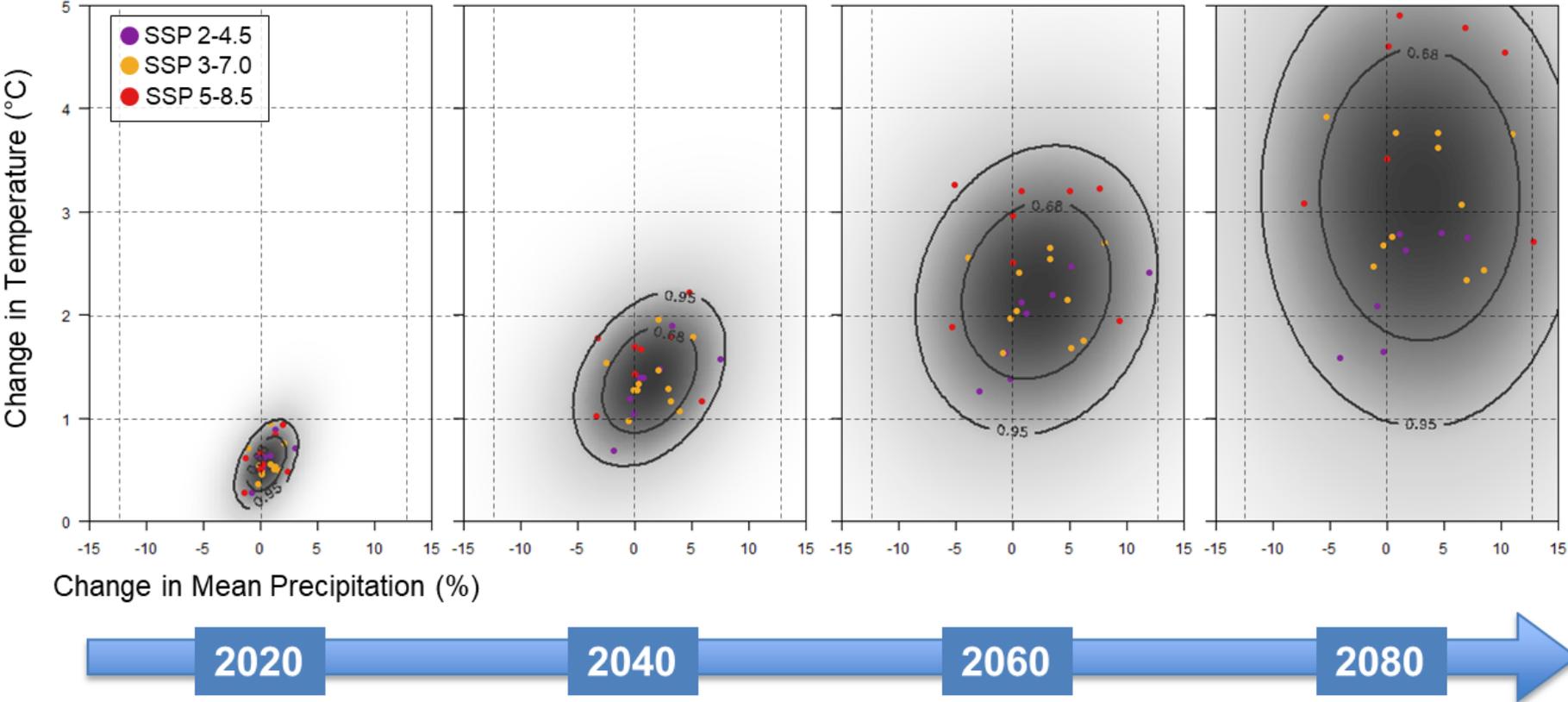
Vulnerability with Respect to Precipitation

Q. How is the peak flow in the Tuolumne River changing with temperature warming?



Projected Future Performance

Projected range of likely climate changes over the Delta Catchment relative to the baseline 30-yr period 1992-2021



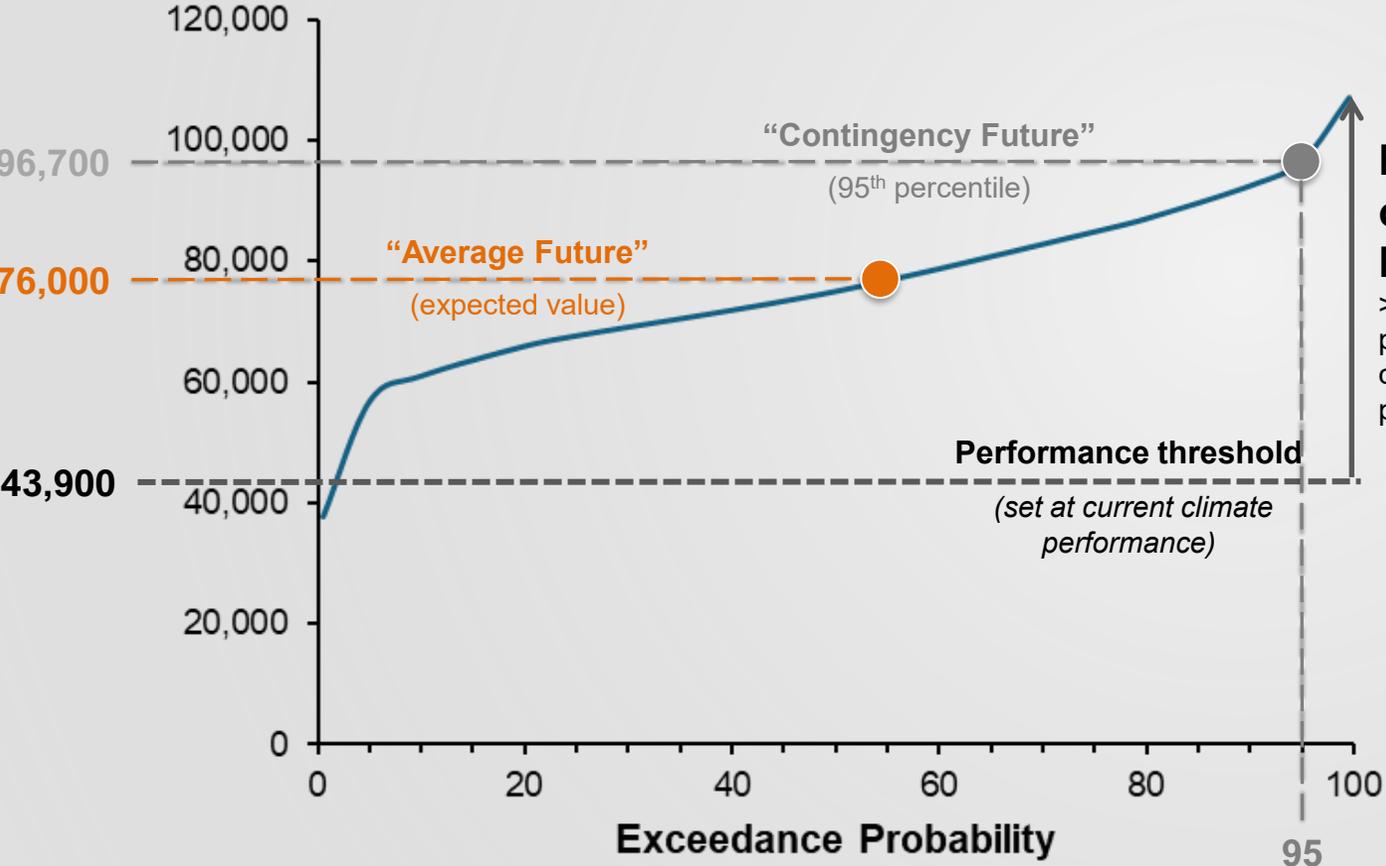
Each dot:

- One of the 27 CMIP6/LOCA-2 downscaled projections of temperature and precipitation
- Spatial averaging over watersheds draining into the Sacramento-San Joaquin Delta

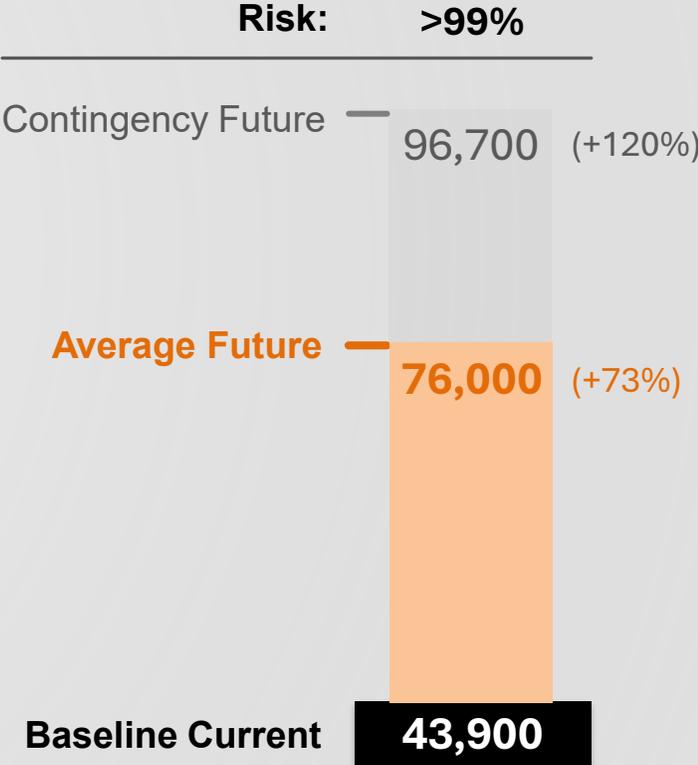
Projected Future Performance

Q. What is the likely peak flow in the Tuolumne River by 2050?

Tuolumne River Peak Flow (cfs)



Risk: "Probability of Declining Performance"
>99% probability that performance under future climate will be worse than performance threshold



Projected Future Performance

Average and Contingency Performance

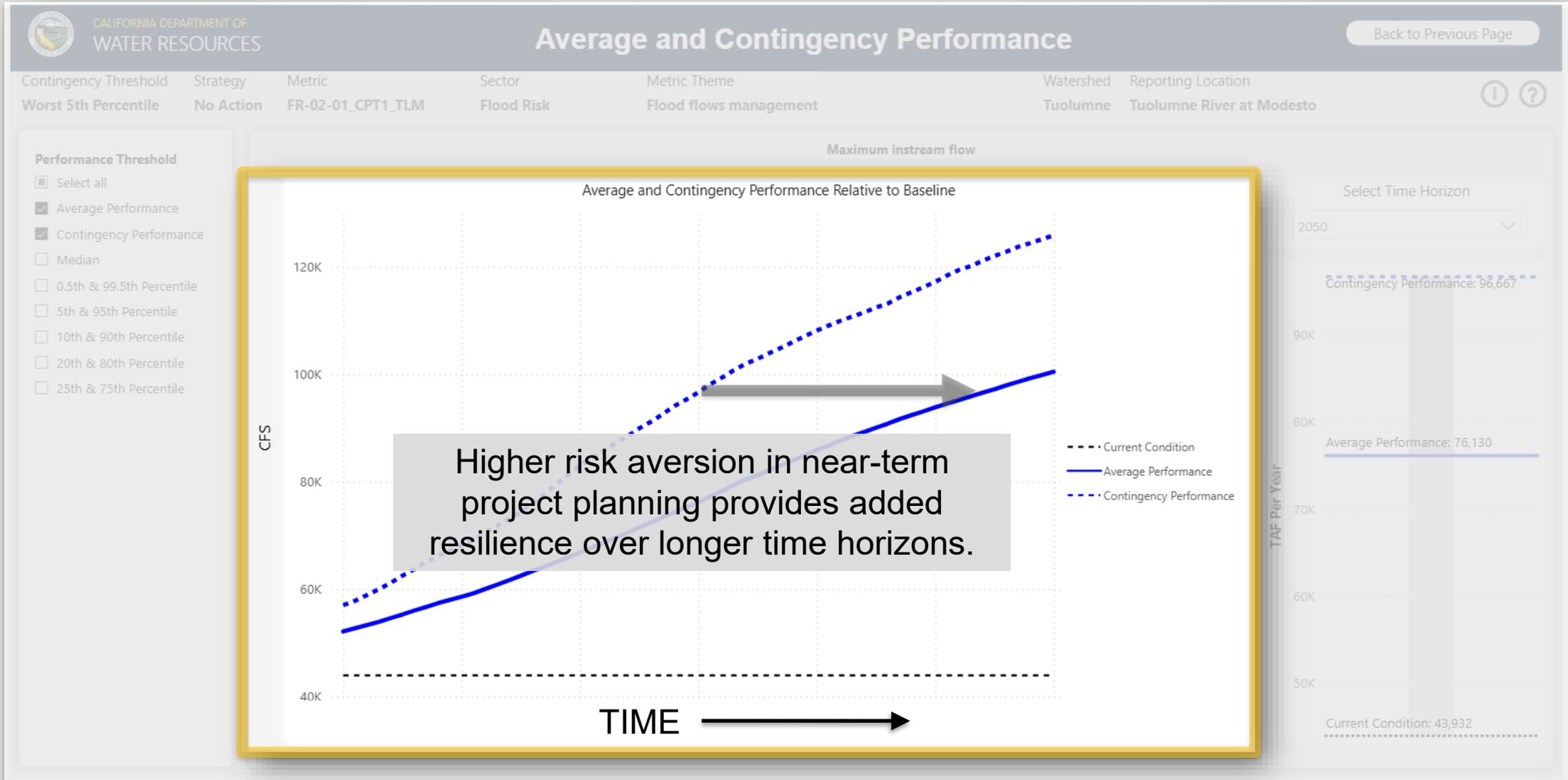
Q. What is the likely peak flow in the Tuolumne River by 2050?



Projected Future Performance

Average and Contingency Performance

Q. How does the peak flow vary for longer or shorter planning horizons?



Projected Future Performance

Average and Contingency Performance

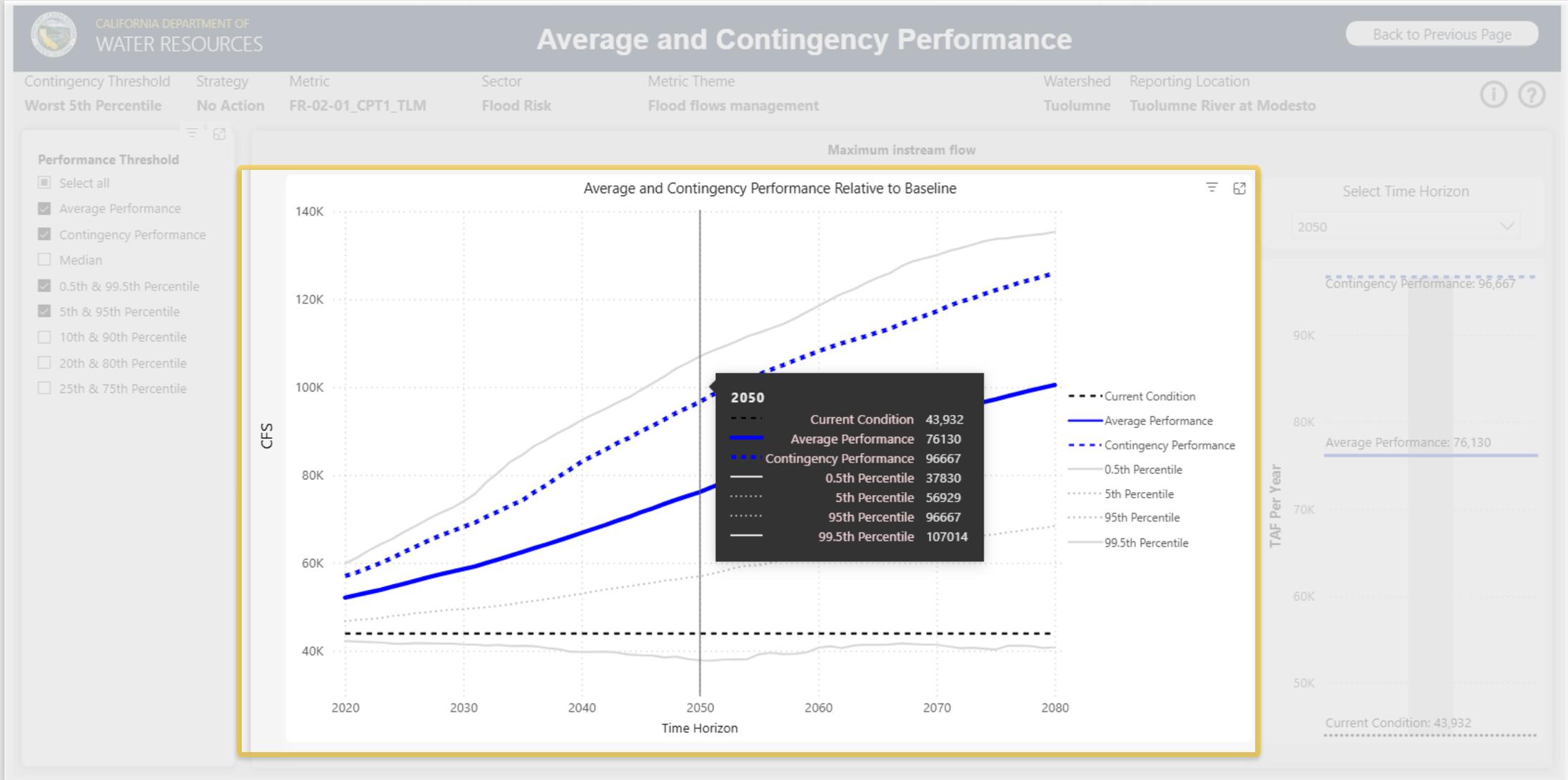
Q. What if you want to plan for a higher or lower level of concern?



Projected Future Performance

Average and Contingency Performance

Q. What if you want to plan for higher or lower levels of concern?



Sensitivity to Climate Drivers

Q. How will the peak flow change with a per-unit change in temperature or precipitation?

CALIFORNIA DEPARTMENT OF WATER RESOURCES

Home Explore Metrics by Location Vulnerability with Respect to Temperature Vulnerability with Respect to Precipitation Average and Contingency Performance Adaptation Performance (Climate...) Adaptation Performance (Time Horizon) Total Risk Sensitivity and Salience

Watershed Studies San Joaquin Basin

Select Results to View

Strategy

Sensitivity and Salience

| Metric | Watershed | Metric Description | Reporting Location | Units | Current Condition | Sensitivity to +1°C Change in Mean | Sensitivity to +1% Change in Mean | Salience of Change in Mean | Salience of Change in Mean |
|-----------------------|---------------------------|---|----------------------------|--|---|--|--|----------------------------|----------------------------|
| Metric Description | Reporting Location | Units | Current Condition (TOP100) | Sensitivity to +1°C Change in Mean Temperature | Sensitivity to +1% Change in Mean Precipitation | Salience of Change in Mean Temperature | Salience of Change in Mean Precipitation | | |
| Maximum instream flow | Tuolumne River at Modesto | CFS | 43,932 | 14,435 | 1,742 | 68% | 32% | | |
| FR-02-04_CPT1_TLM | Tuolumne | Maximum event duration during years when flow exceeds the operational threshold | Tuolumne River at Modesto | Days per Year | 20 | 8 | 2 | 23% | 72% |
| FR-02-05_CPT1_TLM | Tuolumne | Maximum annual volume above the operational threshold | Tuolumne River at Modesto | TAF per Year | 254 | 271 | 34 | 61% | 38% |
| FR-02-06_CPT1_TLM | Tuolumne | Total number of years flow exceeds the design capacity | Tuolumne River at Modesto | Years | 3 | 3 | 0 | 39% | 57% |
| FR-02-07_CPT1_TLM | Tuolumne | Maximum event duration during years when flow exceeds the design capacity | Tuolumne River at Modesto | Days per Year | 7 | 4 | 1 | 46% | 49% |
| FR-02-08_CPT1_TLM | Tuolumne | Maximum annual volume above the design capacity | Tuolumne River at Modesto | TAF per Year | 194 | 251 | 26 | 66% | 33% |

Sector: Flood Risk

Reporting Unit: All

Metric Theme: All

Sensitivity to Climate Drivers

Q. How will the peak flow change with a per-unit change in temperature or precipitation?

Q. Which factor – temperature or precipitation – has a greater influence on future performance?

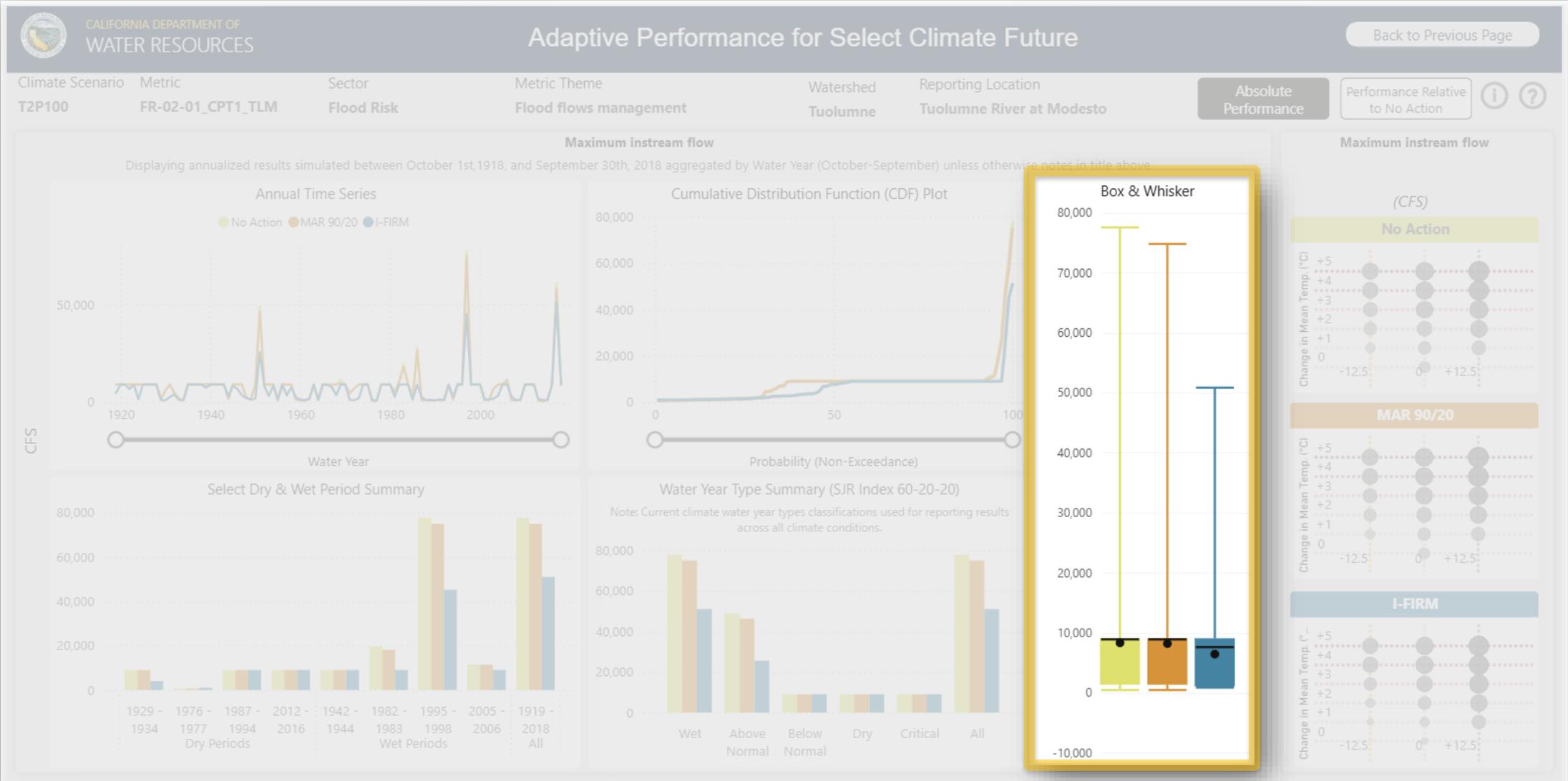
The dashboard displays the following table of metrics:

| Metric Description | Reporting Location | Units | Current Condition (TOP100) | Sensitivity to +1°C Change in Mean Temperature | Sensitivity to +1% Change in Mean Precipitation | Saliency of Change in Mean Temperature | Saliency of Change in Mean Precipitation | | |
|-----------------------|---------------------------|---|----------------------------|--|---|--|--|-----|-----|
| Maximum instream flow | Tuolumne River at Modesto | CFS | 43,932 | 14,435 | 1,742 | 68% | 32% | | |
| FR-02-04_CPT1_TLM | Tuolumne | Maximum event duration during years when flow exceeds the operational threshold | Tuolumne River at Modesto | Days per Year | 20 | 8 | 2 | 23% | 72% |
| FR-02-05_CPT1_TLM | Tuolumne | Maximum annual volume above the operational threshold | Tuolumne River at Modesto | TAF per Year | 254 | 271 | 34 | 61% | 38% |
| FR-02-06_CPT1_TLM | Tuolumne | Total number of years flow exceeds the design capacity | Tuolumne River at Modesto | Years | 3 | 3 | 0 | 39% | 57% |
| FR-02-07_CPT1_TLM | Tuolumne | Maximum event duration during years when flow exceeds the design capacity | Tuolumne River at Modesto | Days per Year | 7 | 4 | 1 | 46% | 49% |
| FR-02-08_CPT1_TLM | Tuolumne | Maximum annual volume above the design capacity | Tuolumne River at Modesto | TAF per Year | 194 | 251 | 26 | 66% | 33% |

Adaptation Potential

Adaptation Performance
(Climate Condition)

Q. How does peak flow in the Tuolumne River respond to the adaptation strategies?



Adaptation Potential

Adaptation Performance
(Time Horizon)

Total Risk

Q. How does peak flow in the Tuolumne River respond to the adaptation strategies?



Adaptation Potential

Adaptation Performance
(Time Horizon)

Total Risk

E.g., How does the peak flow in the Tuolumne River respond to the adaptation strategies?





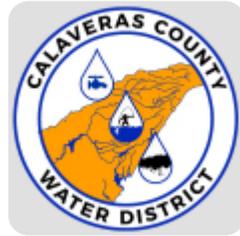
San Joaquin County
Flood Control & Water
Conservation District



WATER & POWER
Serving Central California since 1887



CENTRAL SAN JOAQUIN
WATER CONSERVATION
DISTRICT



MADERA
IRRIGATION DISTRICT

