

# Policy trees and threshold-based adaptation of water resources systems under climate change

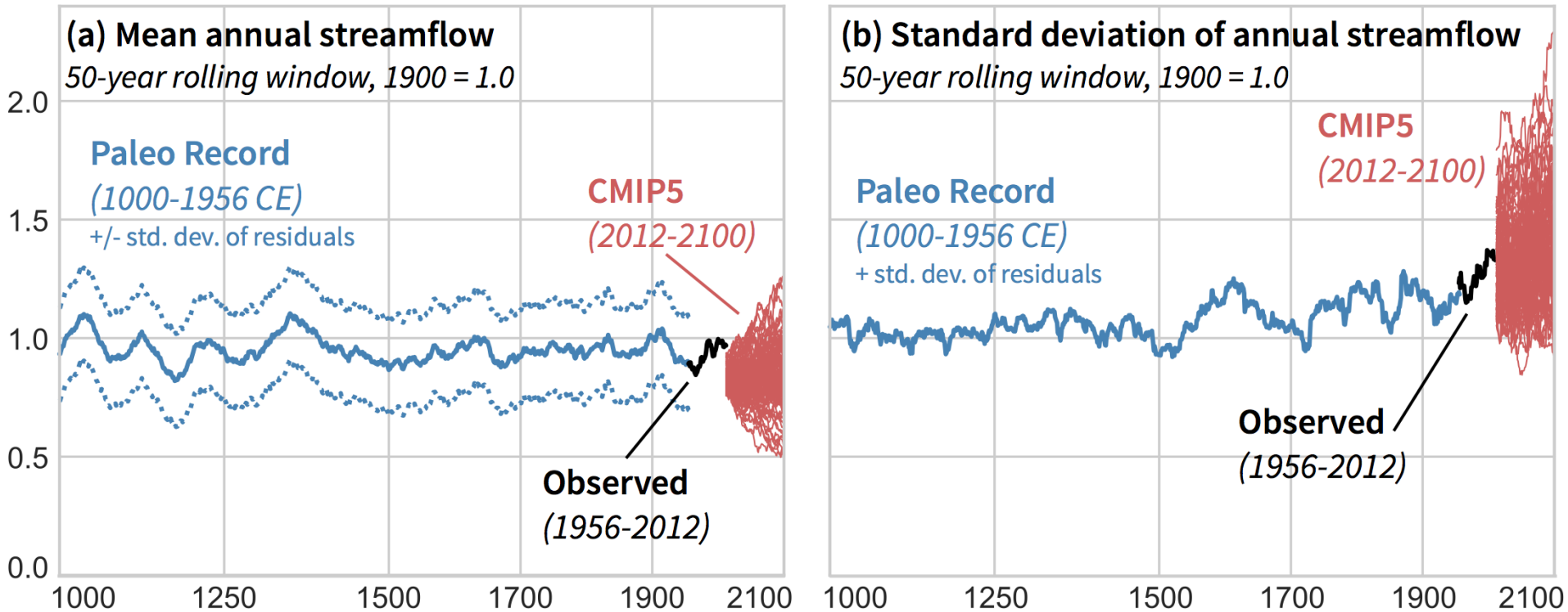
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Contributions from:

Beth Robinson (UC Davis), Matteo Giuliani (Politecnico di Milano), Julie Quinn (University of Virginia), and Jan Kwakkel (TU Delft)

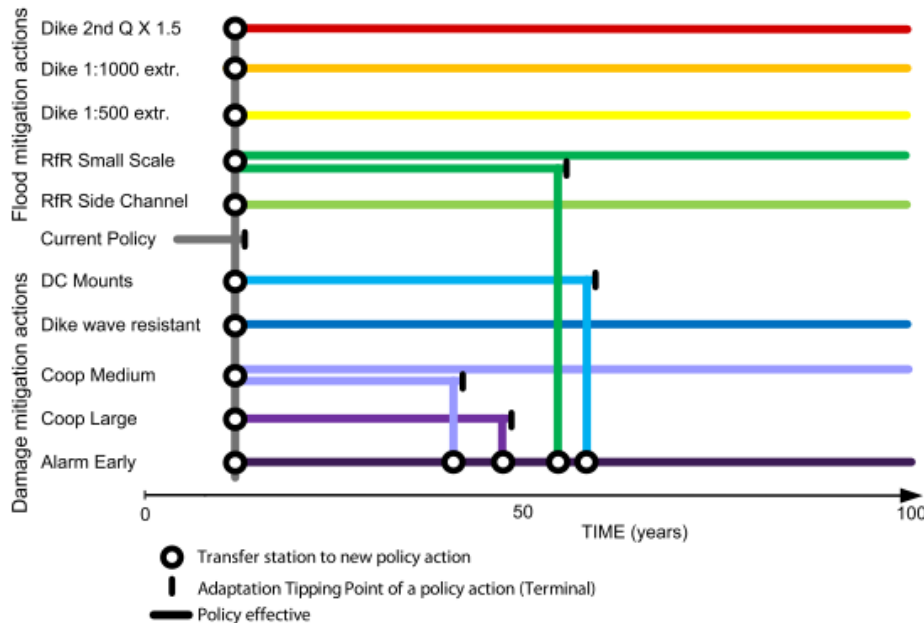
# Uncertainty in water resources systems



Example: Sacramento River at Bend Bridge (California)

[Data sources: TreeFlow, USBR CMIP5 simulations]

# Policies that respond to observed conditions: climate adaptation as a control problem



[Kwakkel et al. 2015]

Discrete actions,  
continuous costs

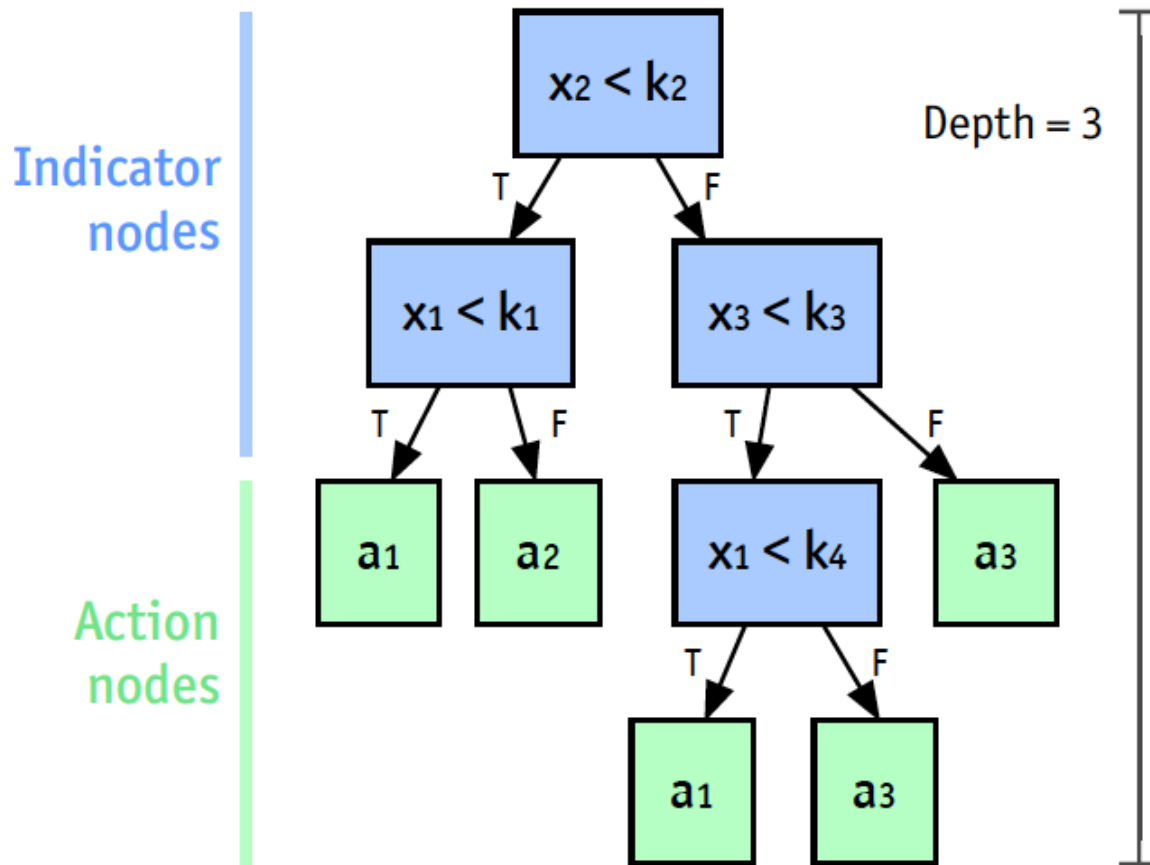
$$\min_{\pi} \mathbb{E}_{\mathbf{e}_t} \left[ \sum_{t=0}^N J(\mathbf{x}_t, a_t) + J_{N+1}(\mathbf{x}_{N+1}) \right]$$

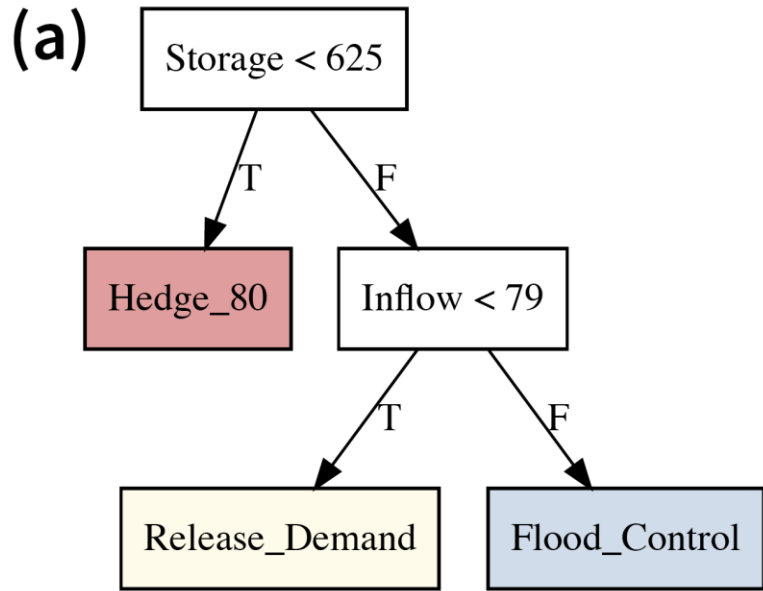
subject to:  $\mathbf{x}_{t+1} = f(\mathbf{x}_t, a_t, \mathbf{e}_{t+1}), a_t = \pi(\tau_t)$

- The policy is a function mapping observations to actions

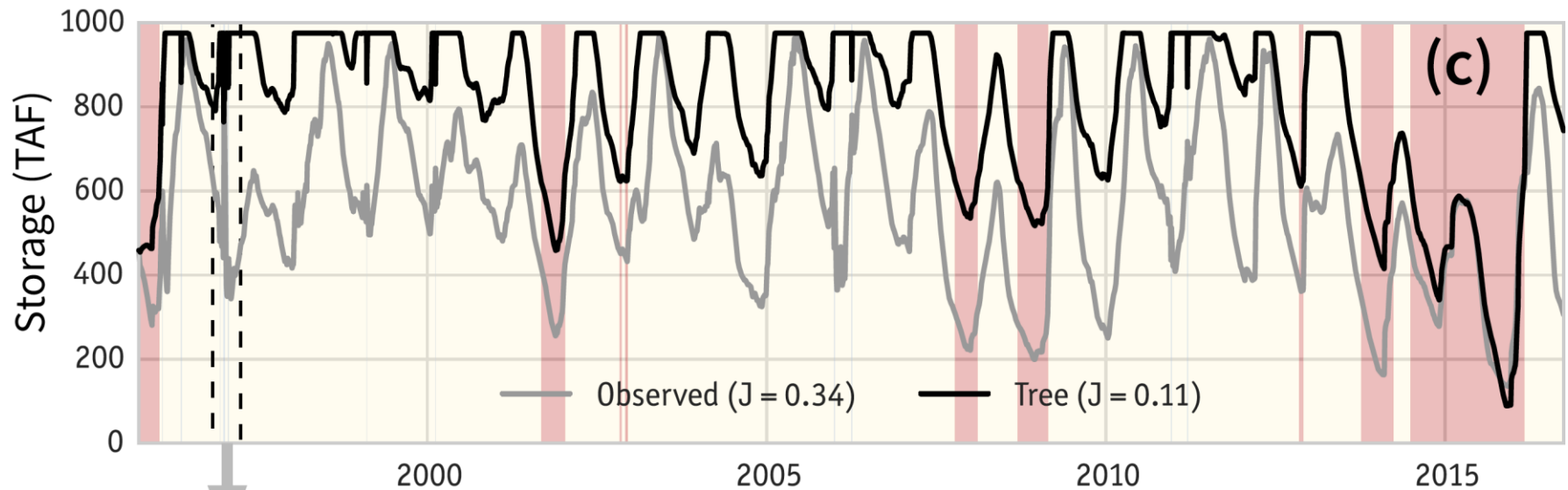
# How to structure a policy?

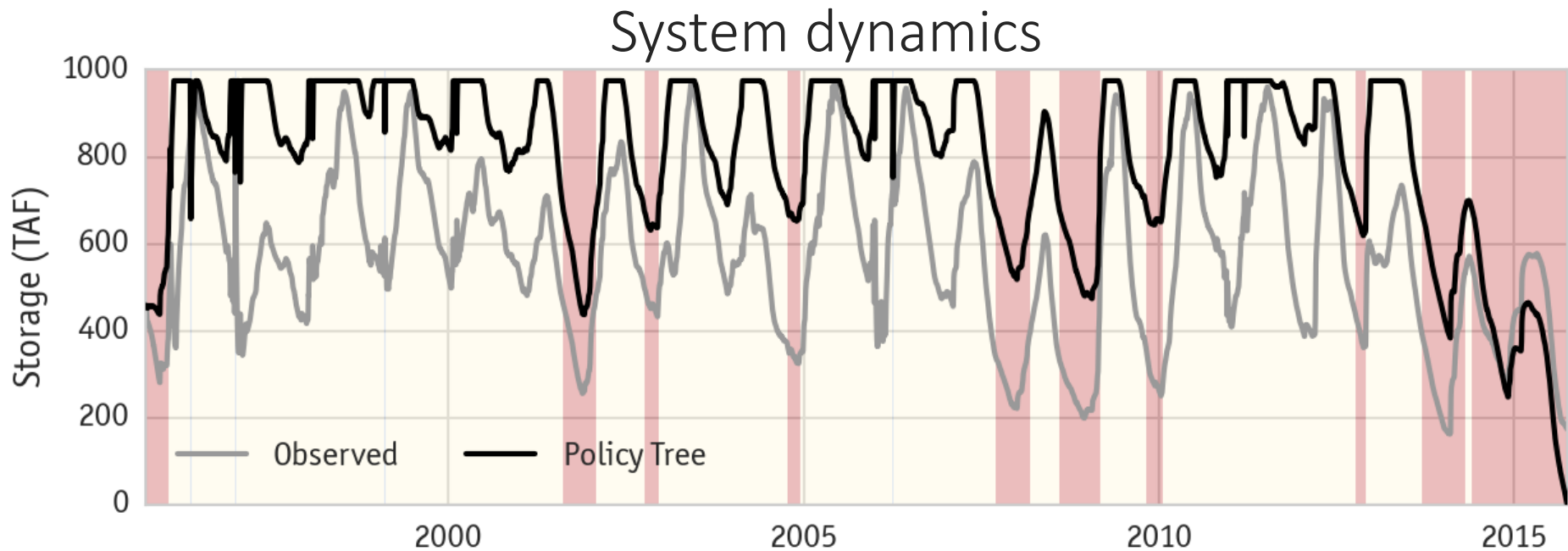
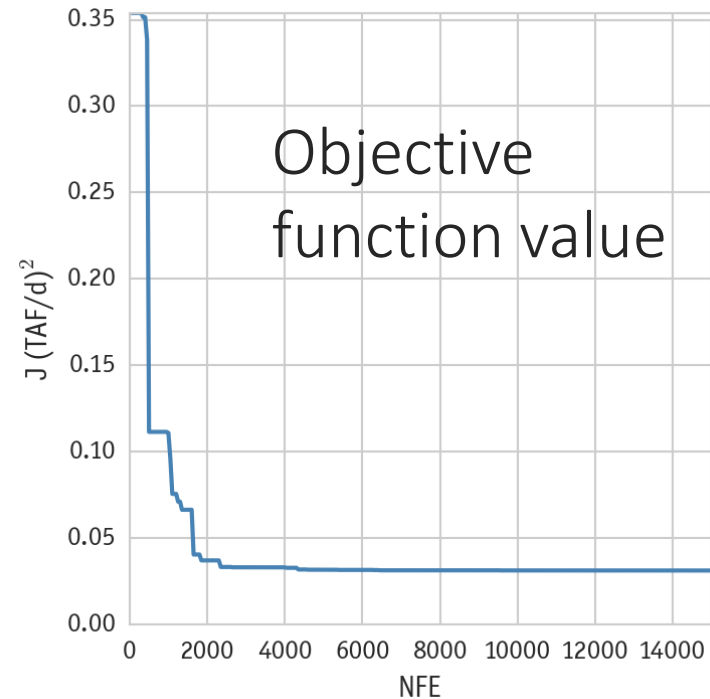
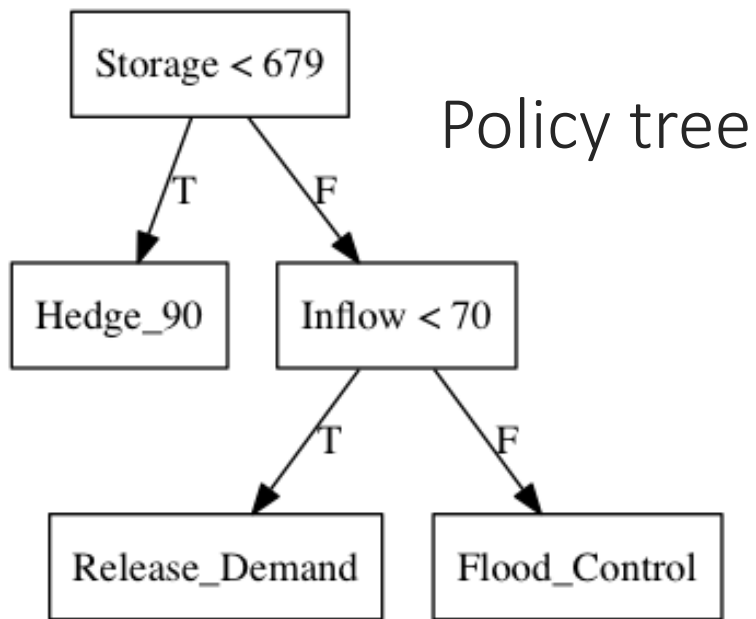
One idea: a tree mapping observations ( $\mathbf{x}$ ) to actions ( $\mathbf{a}$ ) based on the values of thresholds ( $\mathbf{k}$ )





Example: short-term control rules for reservoir operations in California  
 [Herman and Giuliani, 2018]







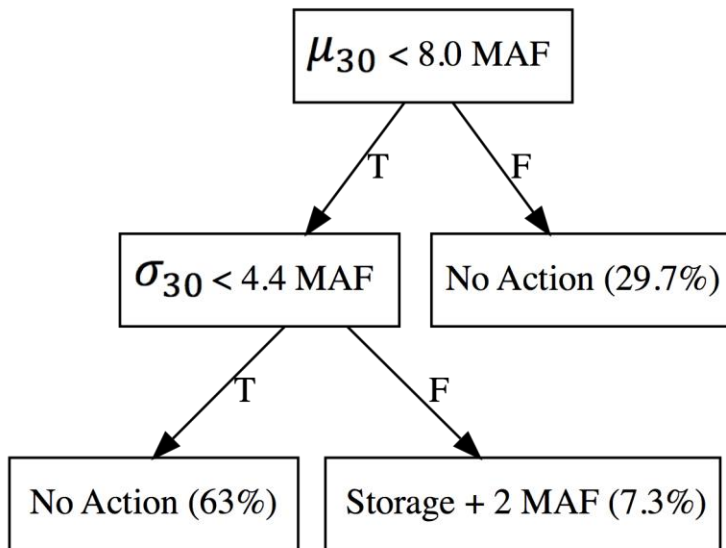
# Can trees be used to represent long-term adaptation policies? An illustrative case study

- Sacramento River CMIP5 projection ensemble, annual timestep 2000-2100
- One reservoir with fixed annual water demand
- *Indicators:* 30-year mean and std. deviation of reservoir inflow ( $\mu_{30}, \sigma_{30}$ )
- *Actions:* Do nothing, or increase reservoir storage by a set amount  $\Delta S$
- *Objective function:* Minimize NPC (water shortage plus construction costs)

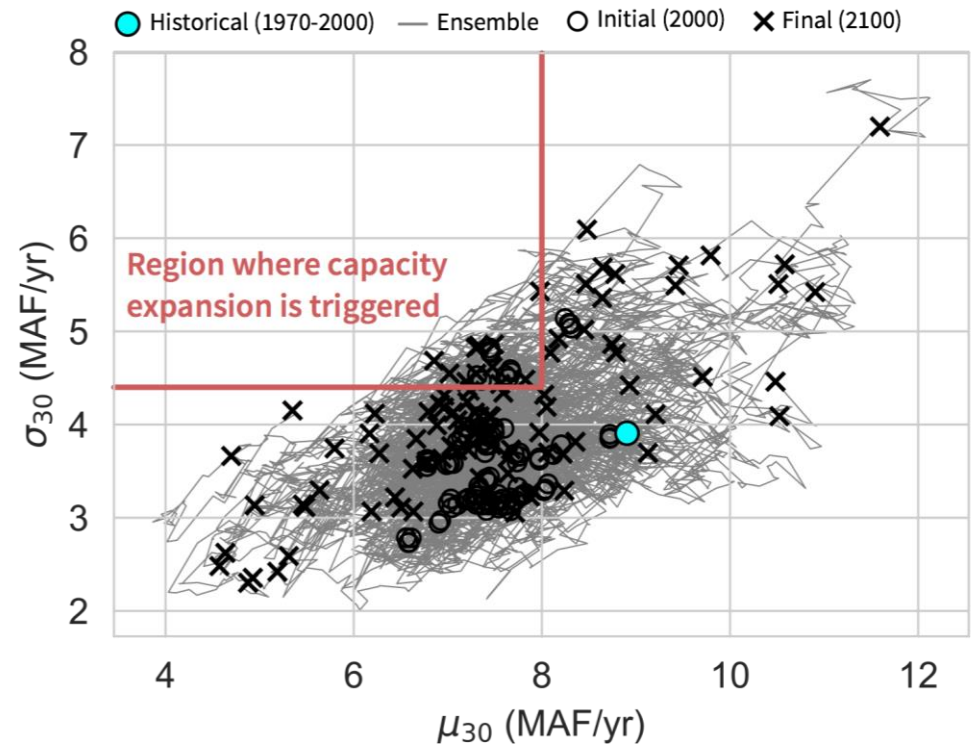


# Illustrative case study: policy result optimized to the ensemble mean NPC

(a) Policy tree with minimum NPC  
(50,000 NFE, 30 random seeds)



(c) Indicator variable space



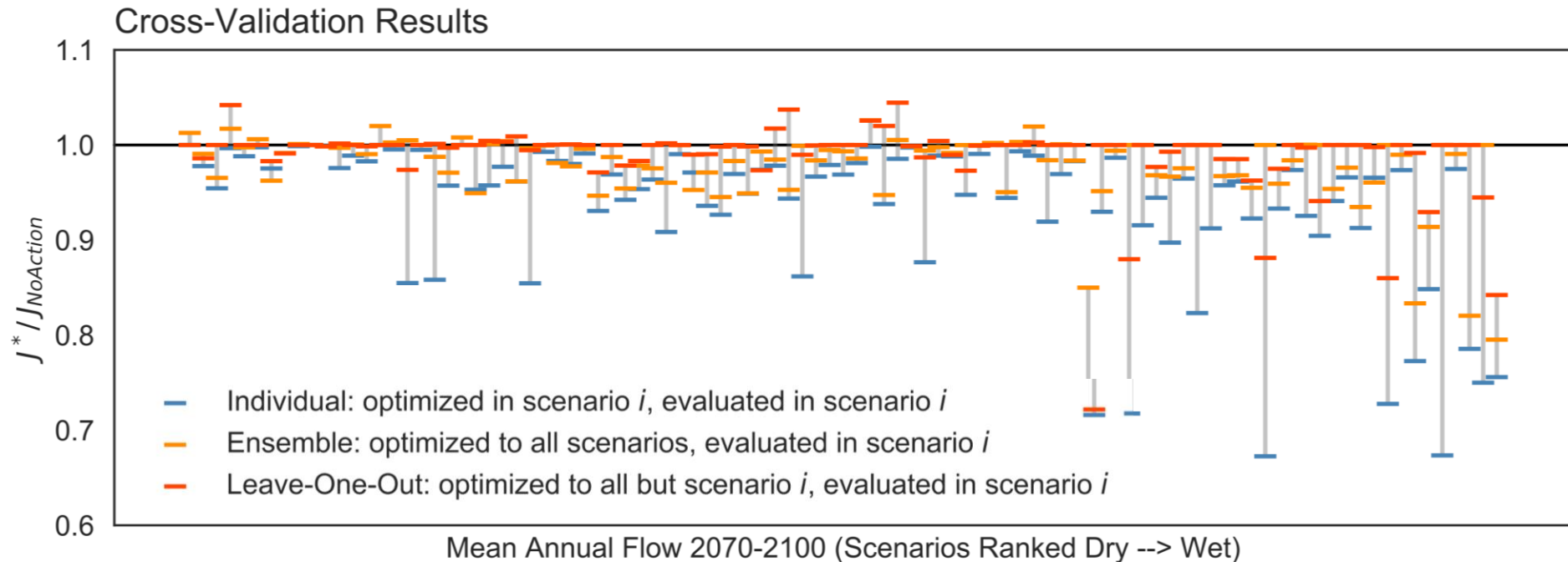
- It works! But, raises some other questions...

# Questions/challenges with optimized policies

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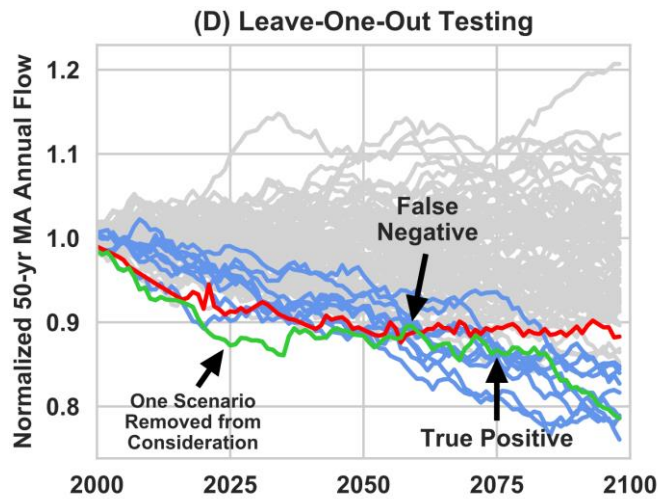
- ① Cross-validation: how well does the policy perform in scenarios it hasn't seen before?
- ② Irreversible actions: how well can we classify vulnerable scenarios in advance?
- ③ Input variable selection: what other indicators would be informative?

# (1) Cross-validation (example)



- In this case, leave-one-out costs are equal or greater in most scenarios than the “do nothing” policy

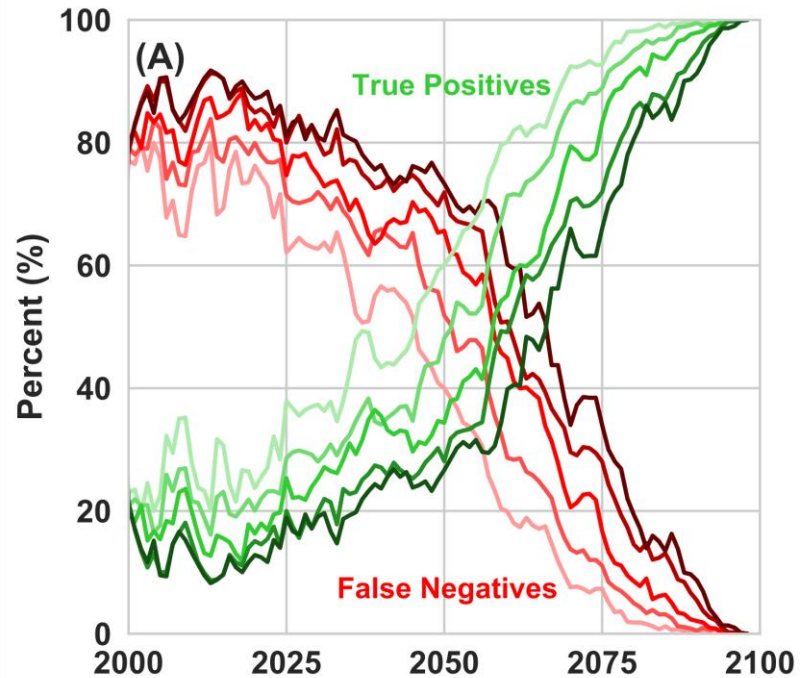
## (2) Vulnerability classifications and irreversible infrastructure decisions *[Robinson and Herman, in review]*



Vulnerable

Predicted —————

Not Vulnerable



- Predict a “not vulnerable” scenario when it should be: will incur costs by waiting too long to adapt
- Predict a “vulnerable” scenario when it should not be: will over-invest in adaptation measures

### (3) Input variable selection: what other long-term observations could be informative?

- **Variables:** temperature, precipitation, land use
- **Timescales and quantiles:** e.g. daily 99% flow (floods), or annual average flow (water supply)
- **Aggregation windows:** 5-yr, 10-yr, 30-yr (tradeoff between adapting quickly vs. correctly)
- **Lead time:** could CMIP5 scenarios serve as a long-term “forecast” input to the policy?

General strategy: run optimization with many of these (ideally not too correlated) and see how costs improve

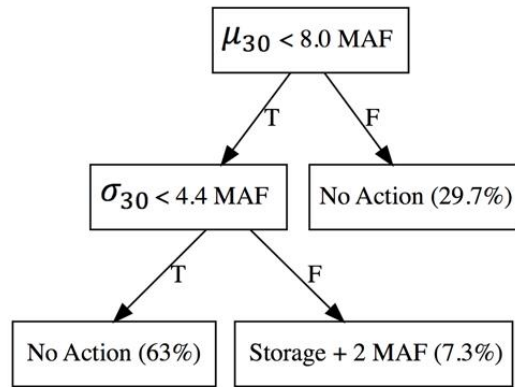
# Key points

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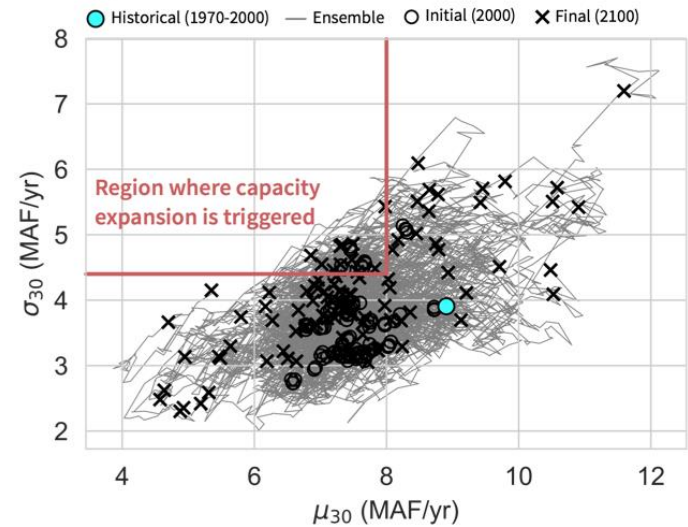
- ① We can optimize threshold-based adaptation policies structured as binary trees
- ② All such policies (tree or not) would benefit from cross-validation against other scenarios
- ③ Open-source tool available; many interesting and challenging research questions remain.

# Thanks!

(a) Policy tree with minimum NPC  
(50,000 NFE, 30 random seeds)



(c) Indicator variable space



## References

Robinson and Herman, Testing threshold-based identification of future water supply vulnerabilities in the Western U.S., in review.

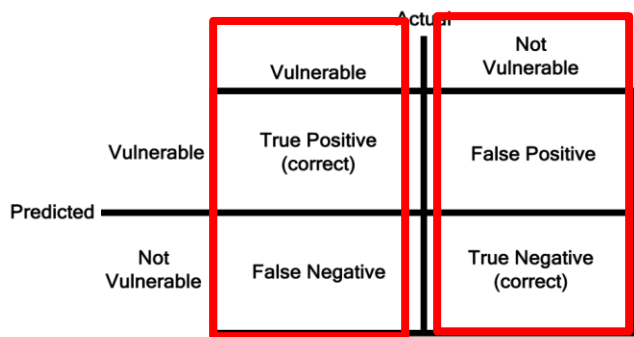
Herman and Giuliani (2018) Policy tree optimization for adaptive management of water resources systems. Environmental Modelling & Software, 99.

Zeff et al. (2016) Cooperative drought adaptation: Integrating infrastructure development, conservation, and water transfers into adaptive policy pathways. WRR.

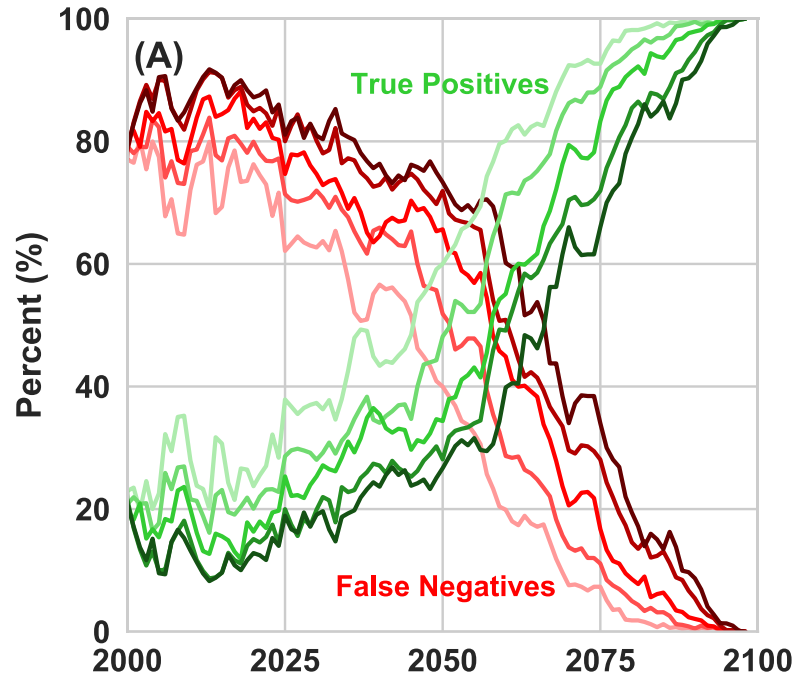
Herman et al. (2015) How should robustness be defined for water systems planning under change? JWRPM.





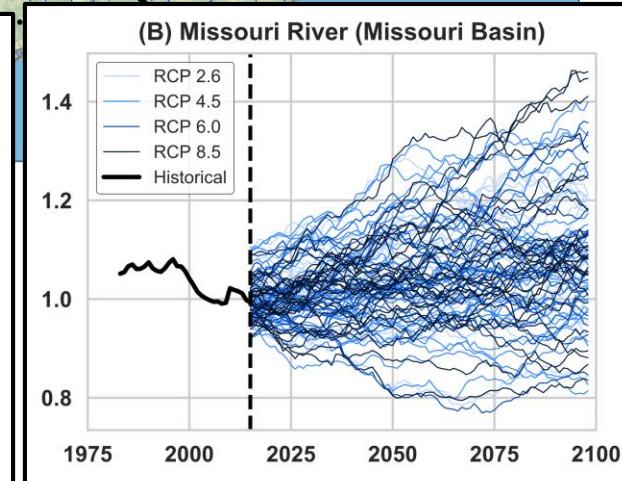
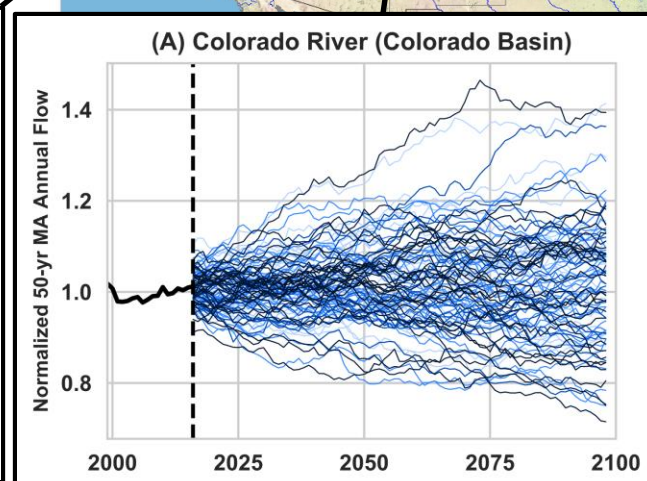
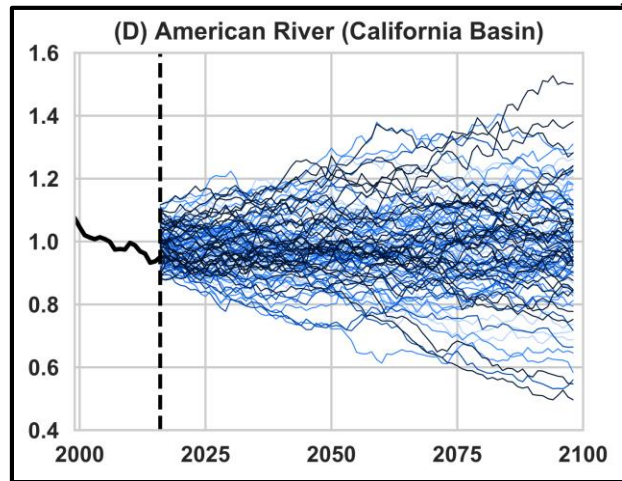
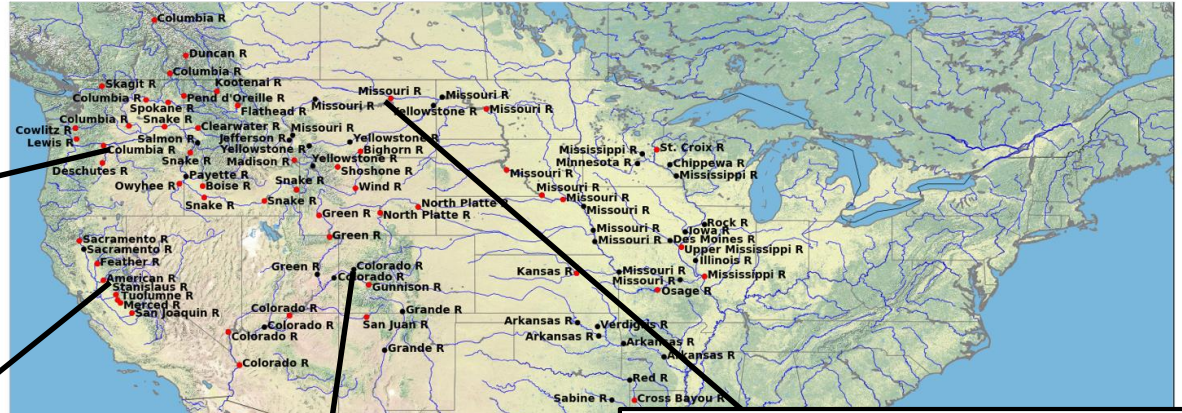
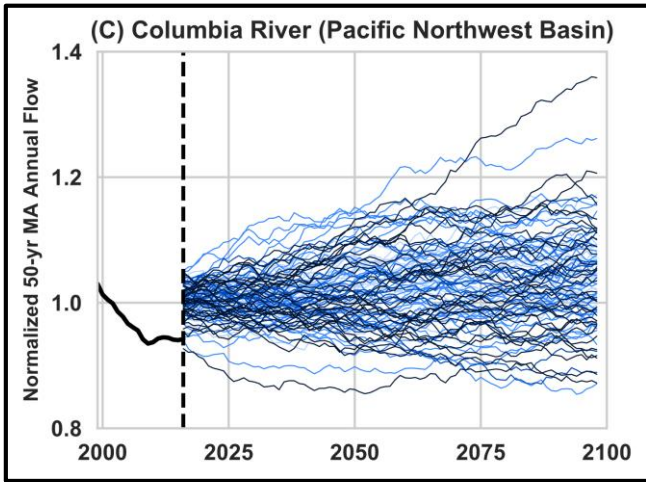


# Classification error rates decrease over time



- FP% generally low; FN% remains high until much later
- Model agreement controls FP-FN tradeoff

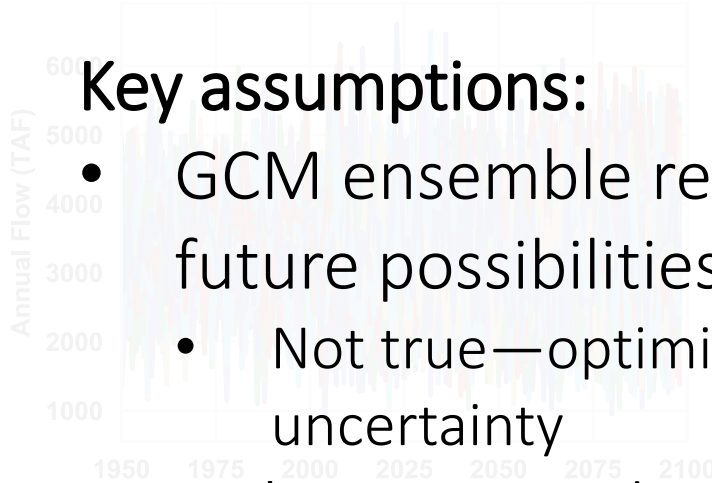
# GCM reservoir inflow projections through 2100 (Brekke et al., 2014)



50-year moving average (1950-2000 = 1.0)

# Experiment design: focus on decrease in $\mu_Q$

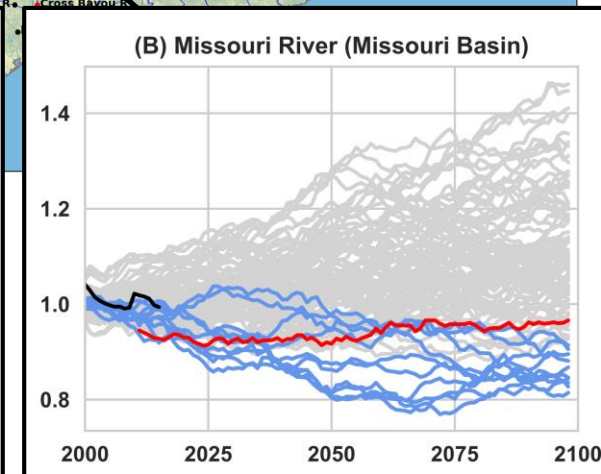
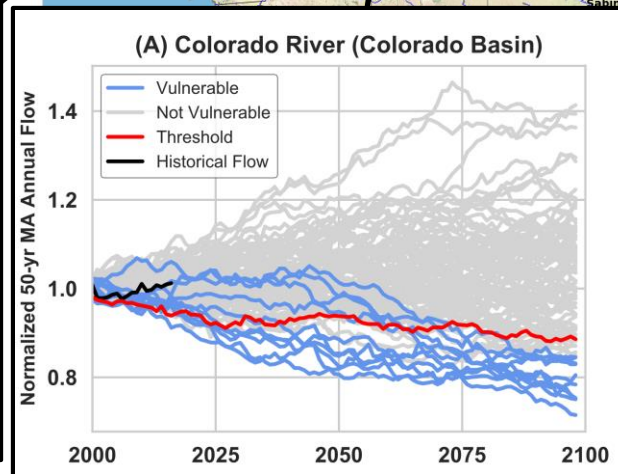
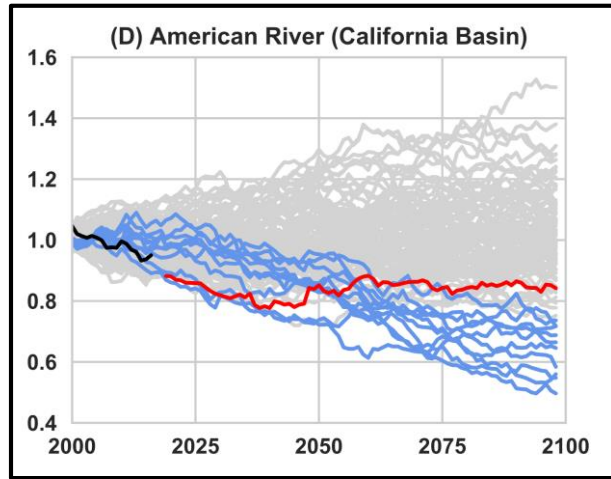
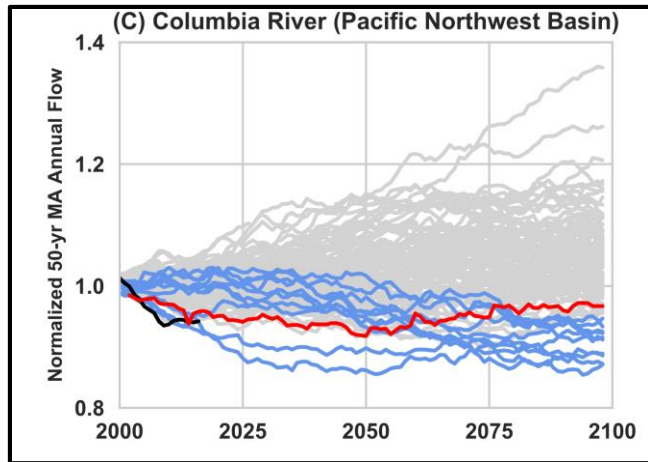
(A) GCM-based Flow Data



## Key assumptions:

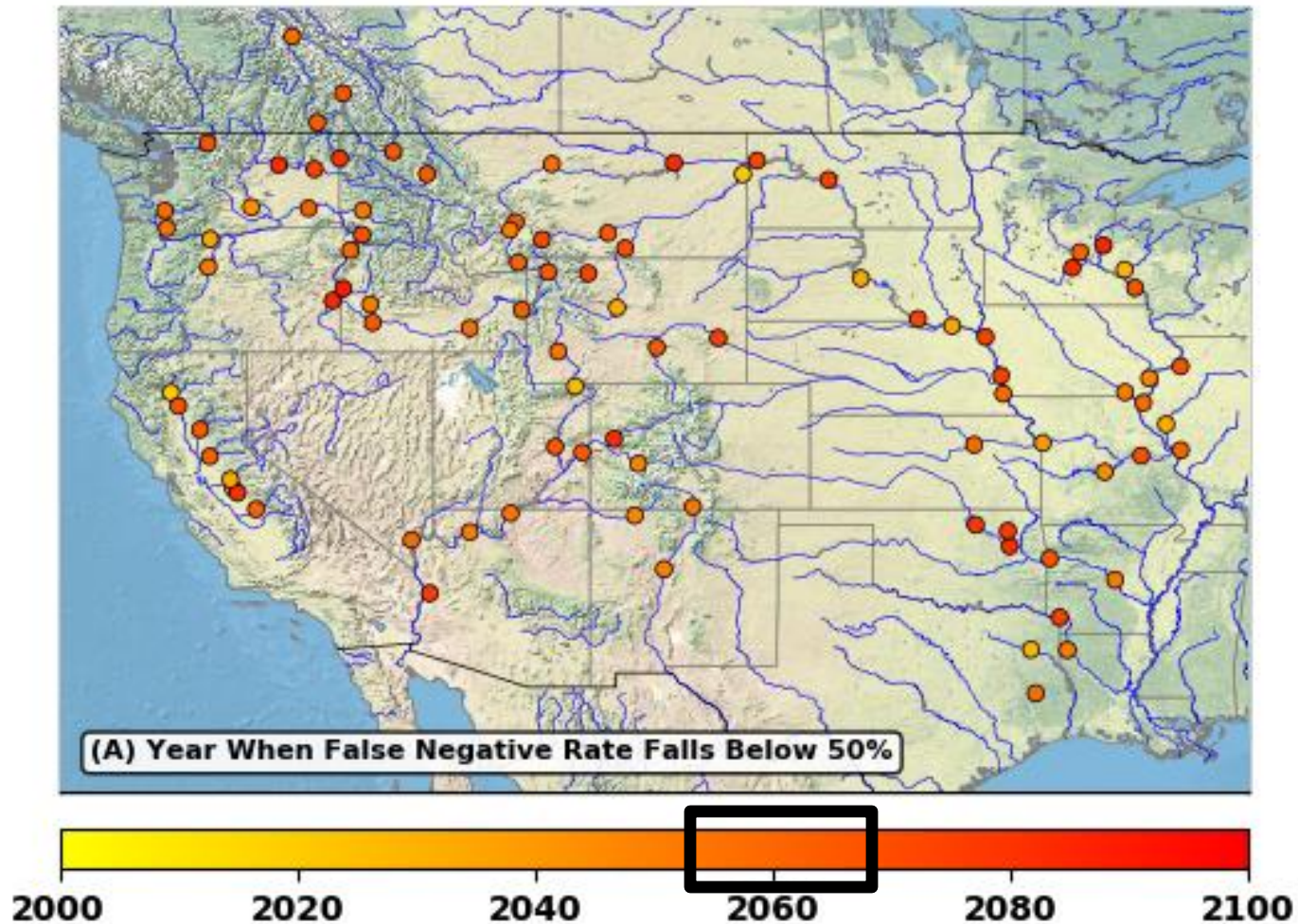
- GCM ensemble represents full range of future possibilities
  - Not true—optimistic lower-bound uncertainty
- Only concerned with average annual flow
  - Could use variance, drought frequency, flood risk, sea level rise (etc.)
- Vulnerable scenarios are the lowest 10% of average flows—illustrative case study

# Dynamic thresholds for selected rivers

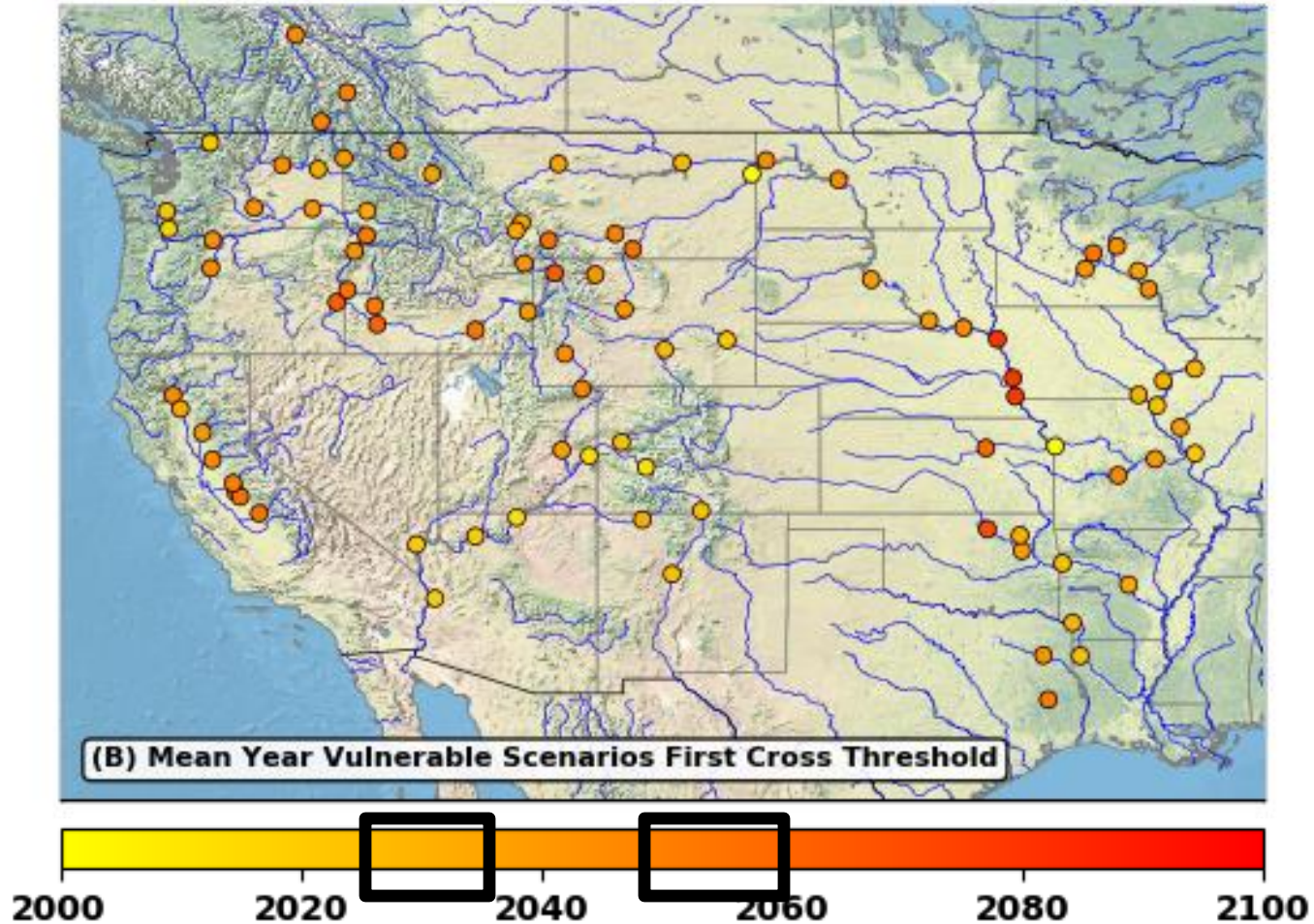


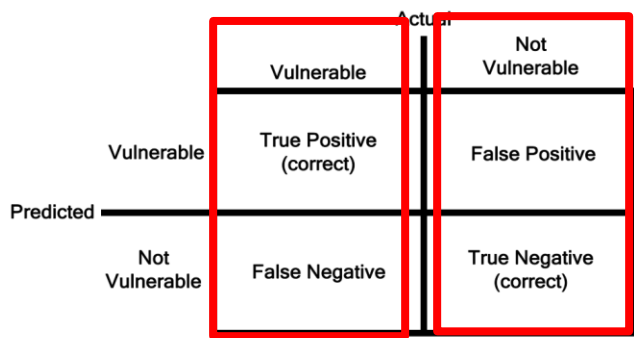
Thresholds represent if-then rules (observation → action)

# Year when false negative rate falls below 50% (i.e., when a negative classification becomes better than random)

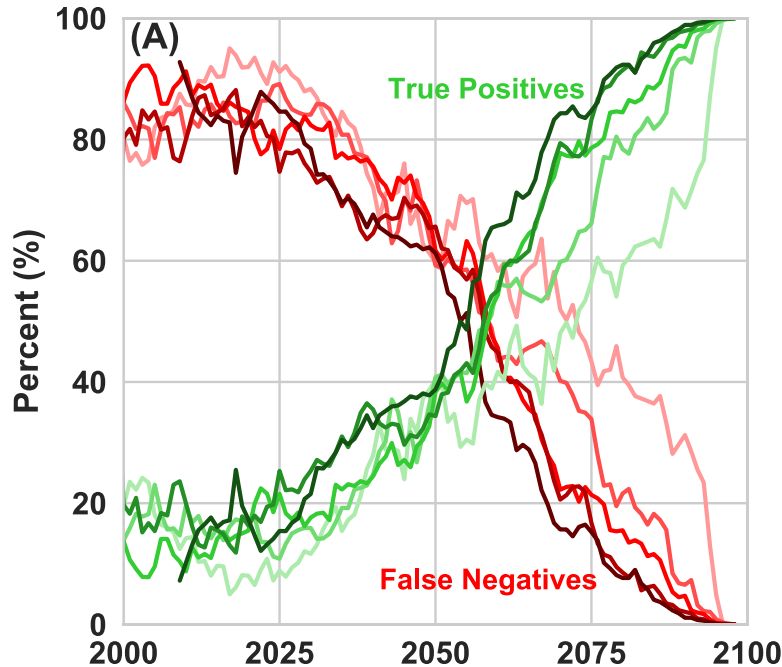


# Average year when “vulnerable” scenarios are first identified (basin-specific patterns)

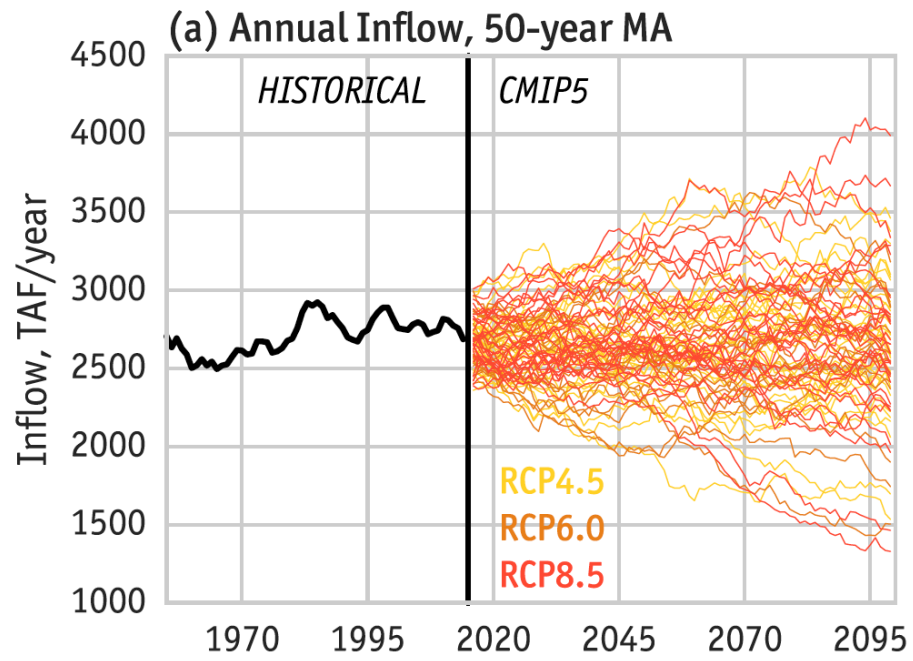




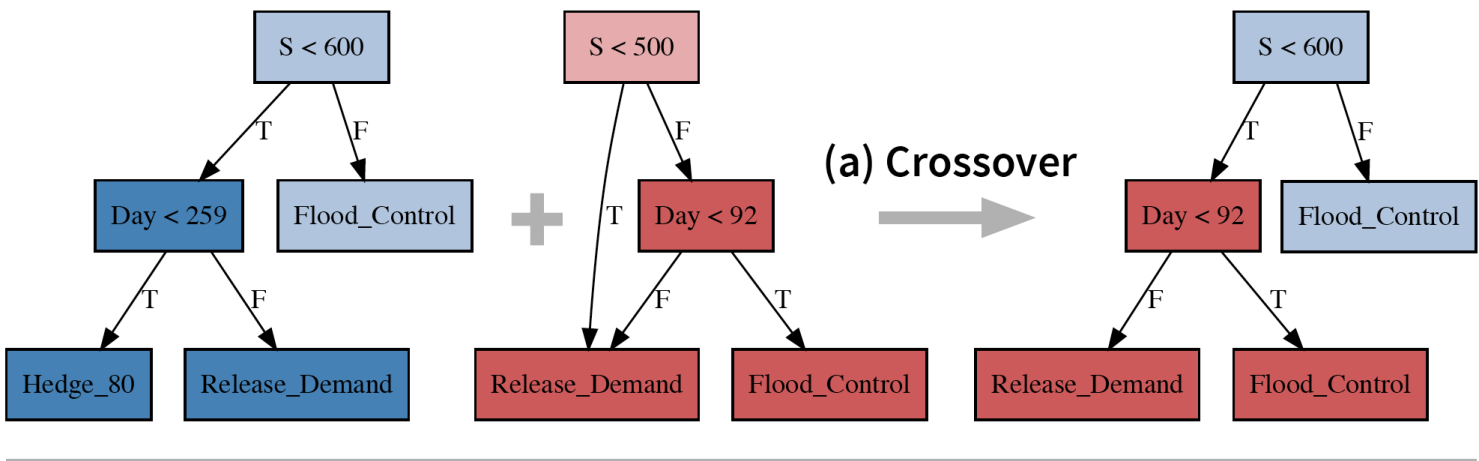
# Sensitivity of results to moving window length



- Short moving window: respond quickly to changes
- Long moving window: wait-and-see, lower error rates



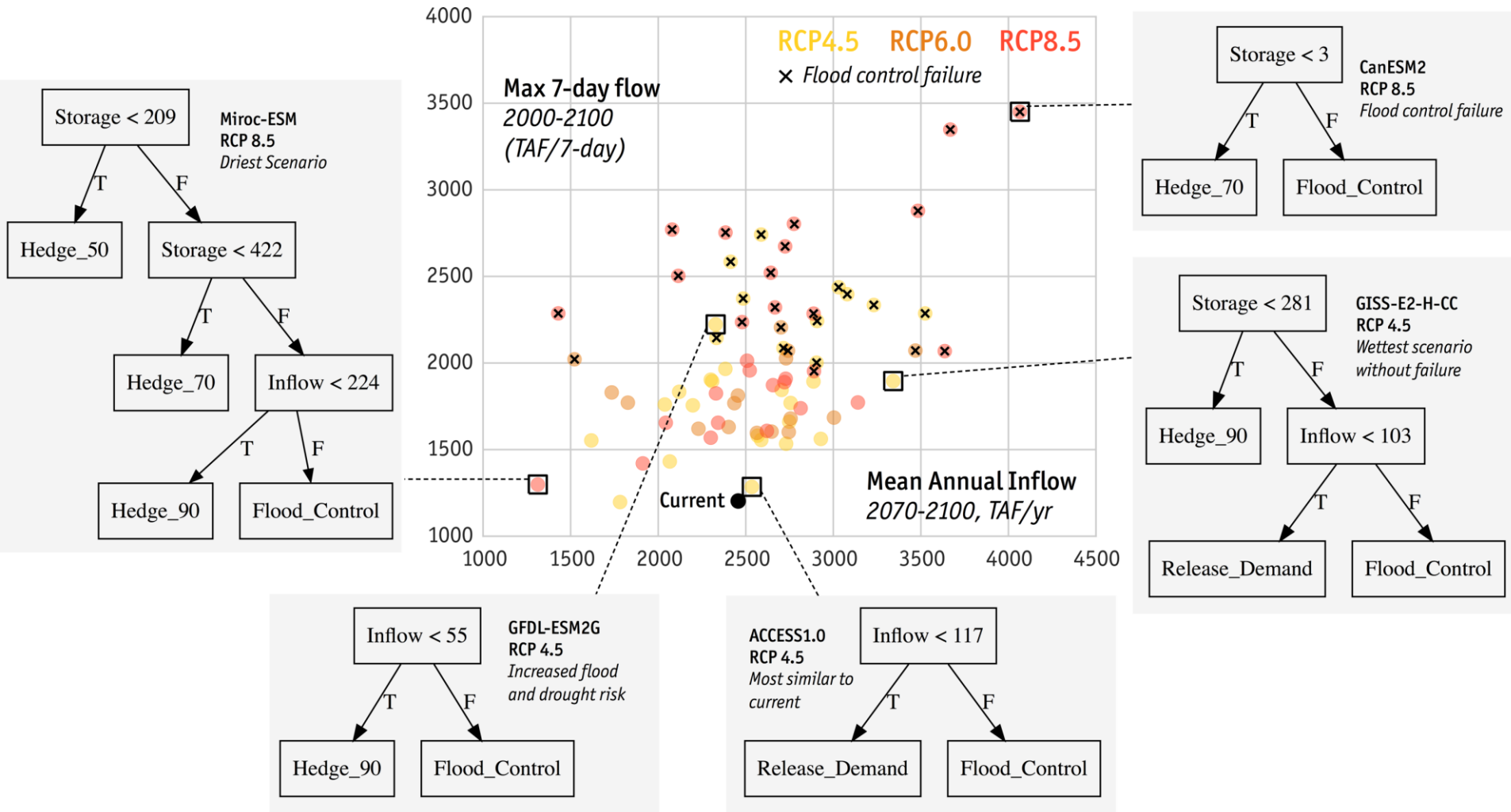




Genetic programming customized for binary trees

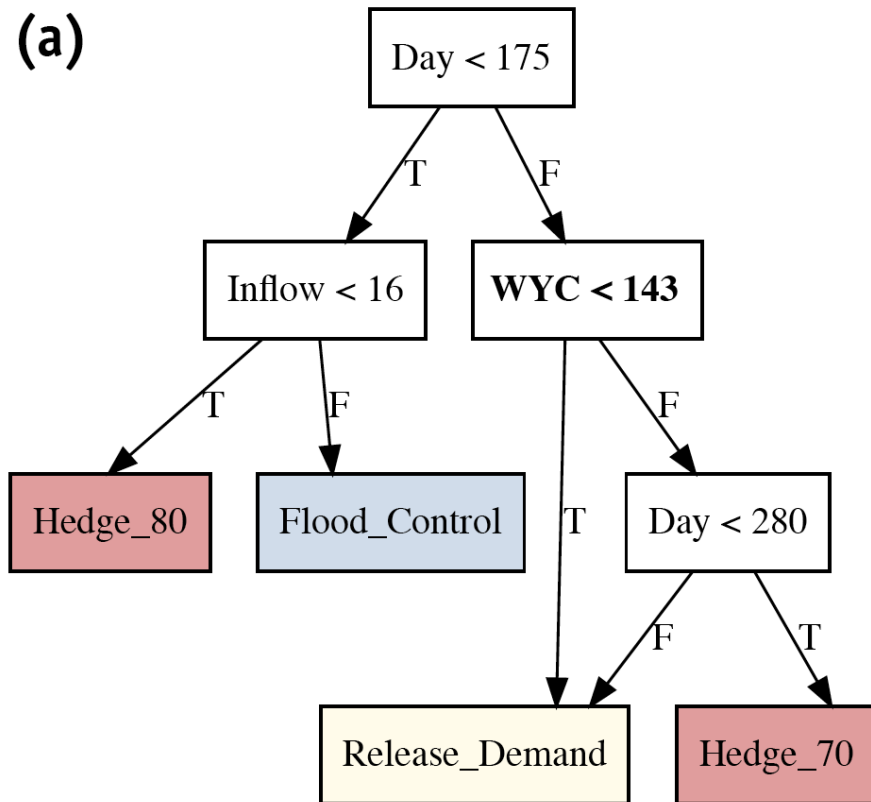
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# (2) Policies trained on CMIP5 scenarios



### (3) Policy trained on all CMIP5 scenarios

(a)



Long-term indicator variables: MA  
Inflow, flood risk, WY centroid

# Problem: Validation

