

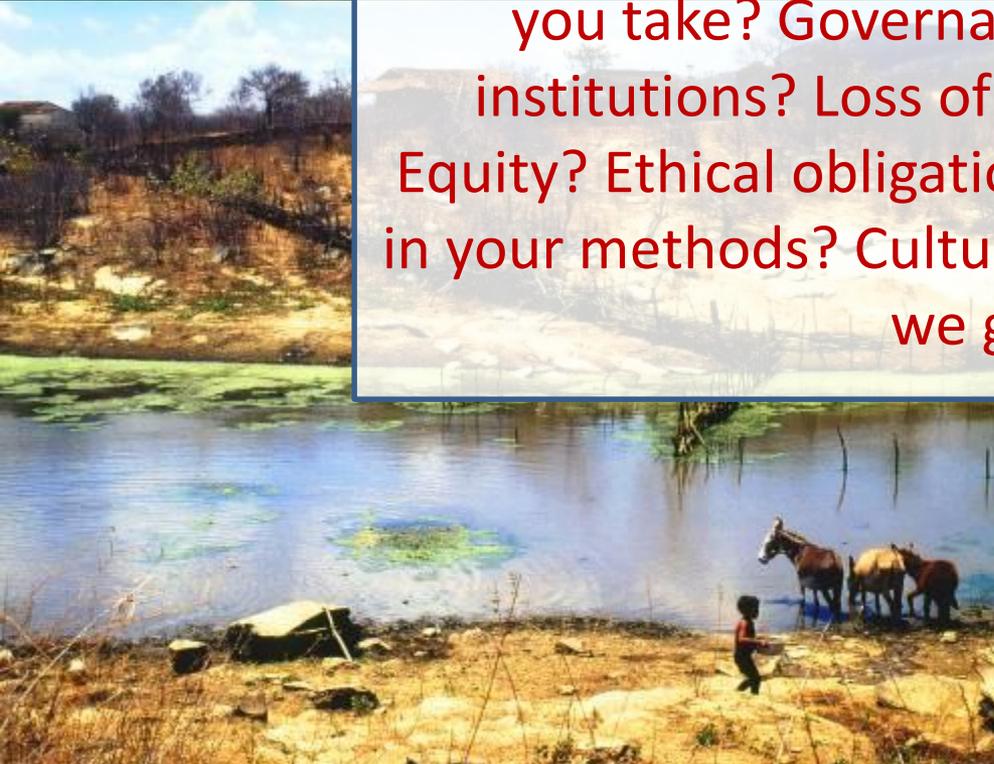
Challenges in Assessing , Predicting, and Managing Water Resources Systems

Patrick Reed
Cornell University
Civil & Environmental Engr.
Snowmass Impacts Session
July 2014

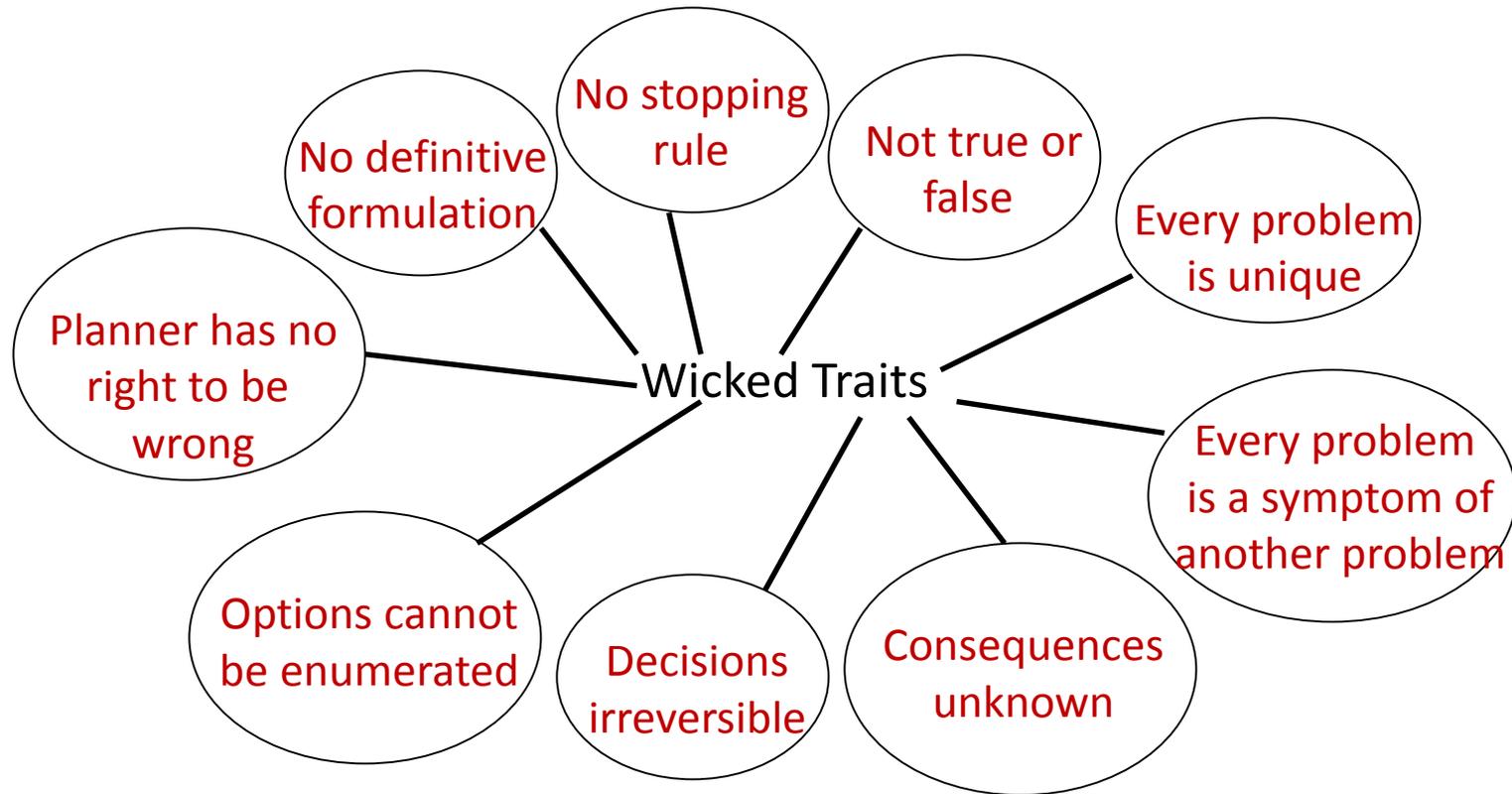




Ok, so what do you mean by impacts? Who are your stakeholders? Decision makers? Objectives? Constraints? Potential to do harm with modeling assumptions? What actions can you take? Governance issues? Stable institutions? Loss of human life matter? Equity? Ethical obligations? Tacit assumptions in your methods? Cultural bias? and on and on we go...



So let's start with a well regarded criticism model-based policy planning



Rittel & Webber, 1973 "Dilemmas in Planning", Policy Sciences



So defining “wicked water problems” is The Problem

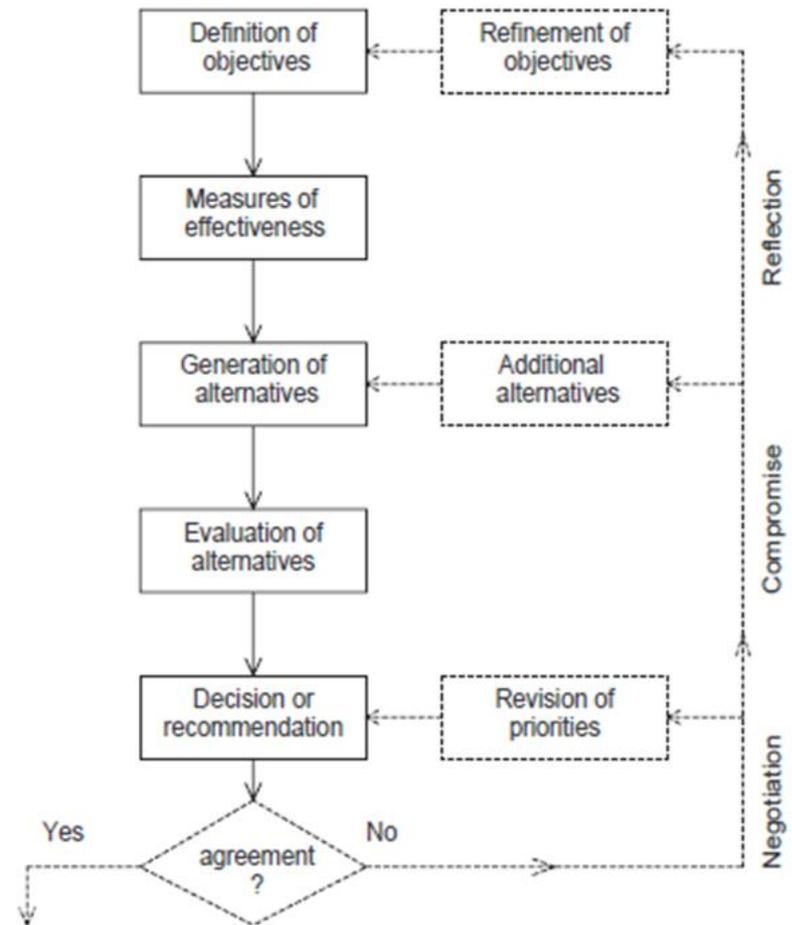
Maass et al 1962

- (1) Identify suite of objectives
- (2) Translate objectives into design criteria
- (3) Identify “optimal” designs
- (4) Evaluate the consequences of the designs

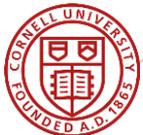
“For many water-resource activities in this country, for example, the design criterion (as set forth in the Flood Control Act of 1936) ...is so imprecise that it encourages engineers to design river systems for maximum physical outputs and confine economic considerations solely to cost minimization.”

Calling for consideration of both economic & engineered performance objectives. First detailed motivation for multiobjective design.

“Prescriptive Multi-objective Decision Support”

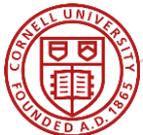


Bruen (2008) “Systems Analysis – a new paradigm and decision support tools for the water framework directive.” HESS vol. 12

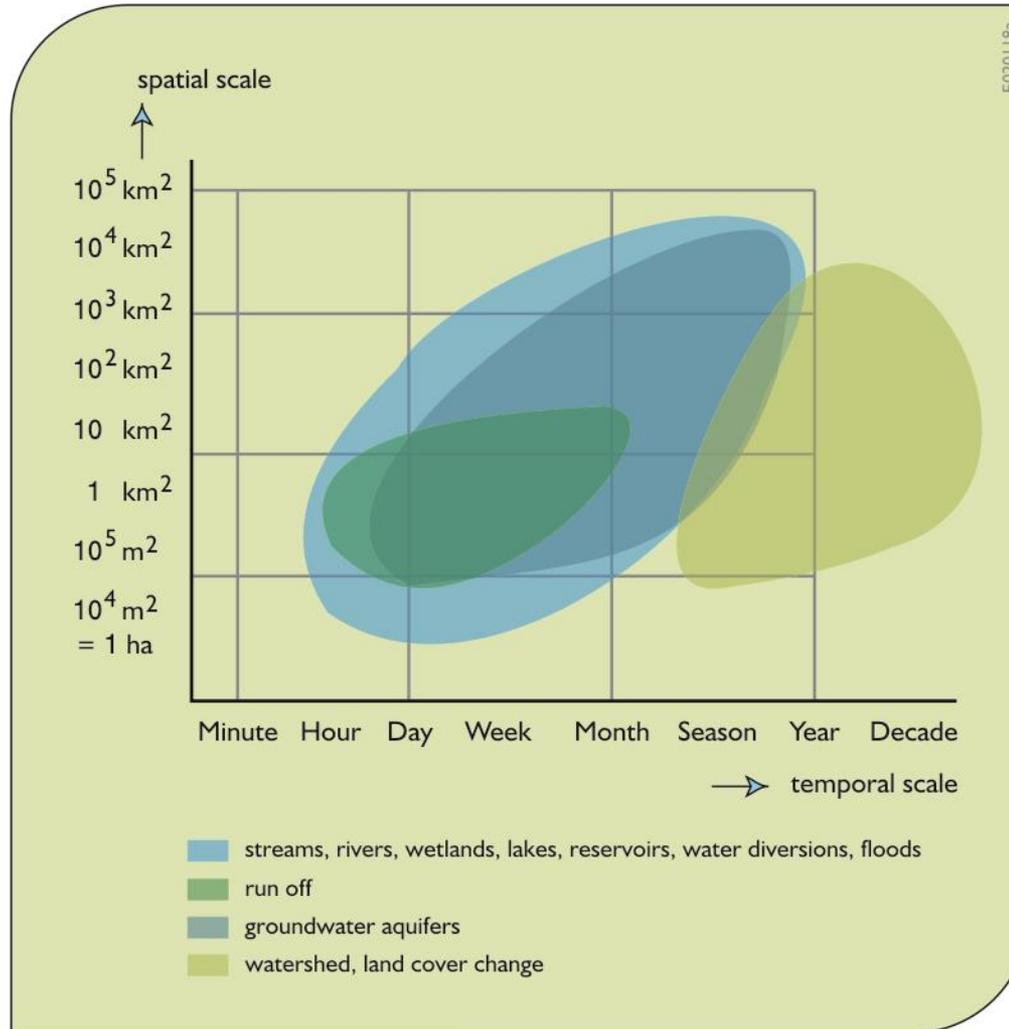


“ The ultimate point of disagreement ... was the role of policy analysts: to simply *make the relevant tradeoffs known* to policymakers ... or, to *make normative recommendations...* ”

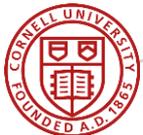
— Banzhaf (2009), on the 1960's debate between the Harvard Water Program and Resources for the Future (RFF)



Space & time Scales are critical to properly posing problems

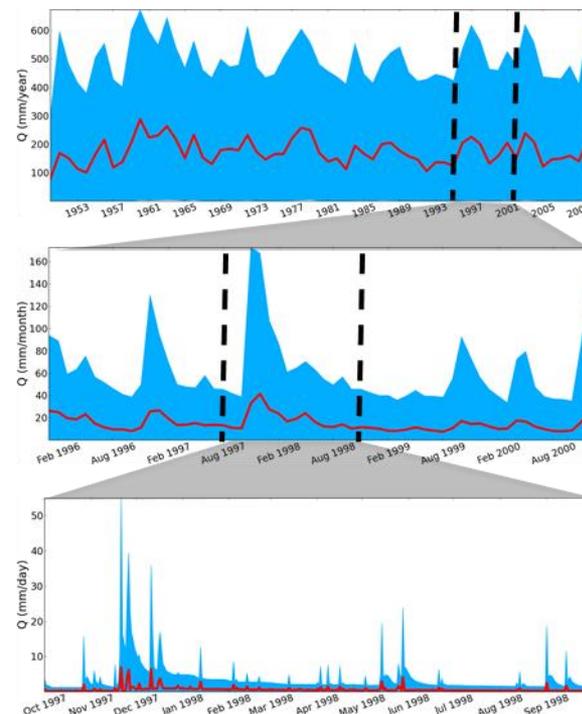
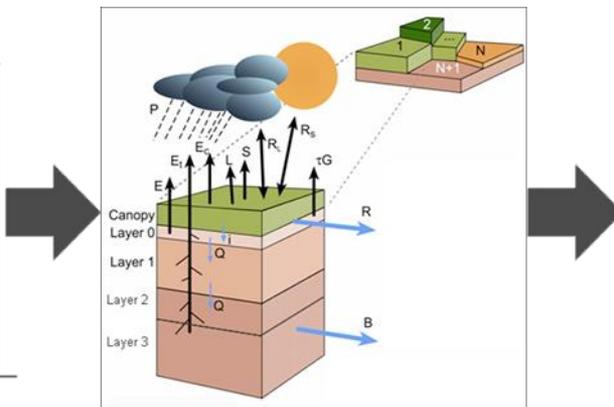
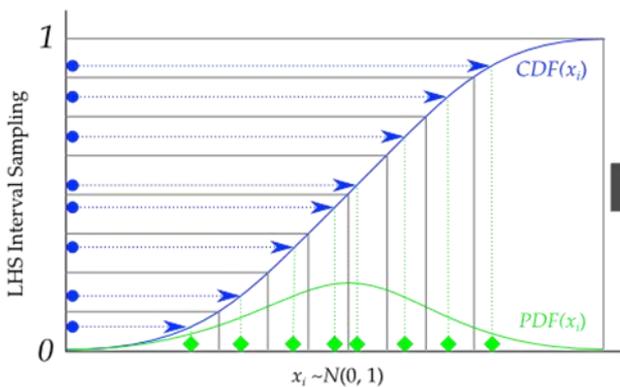


Loucks, D. P. and E. van Beek (2005). Water Resources Systems Planning and Management United Nations Educational, Scientific and Cultural Organization.

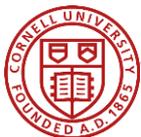


Uncertainty & Observation Limits Change Over Space and Time Scales!!

Let's play with a 10,000 LHS 1-degree VIC Monte Carlo for the globe (~300 years of computation & 1.5 petabytes of output)



Chaney, Herman, Reed, and Wood, (In-Prep)

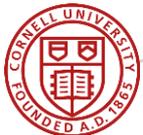
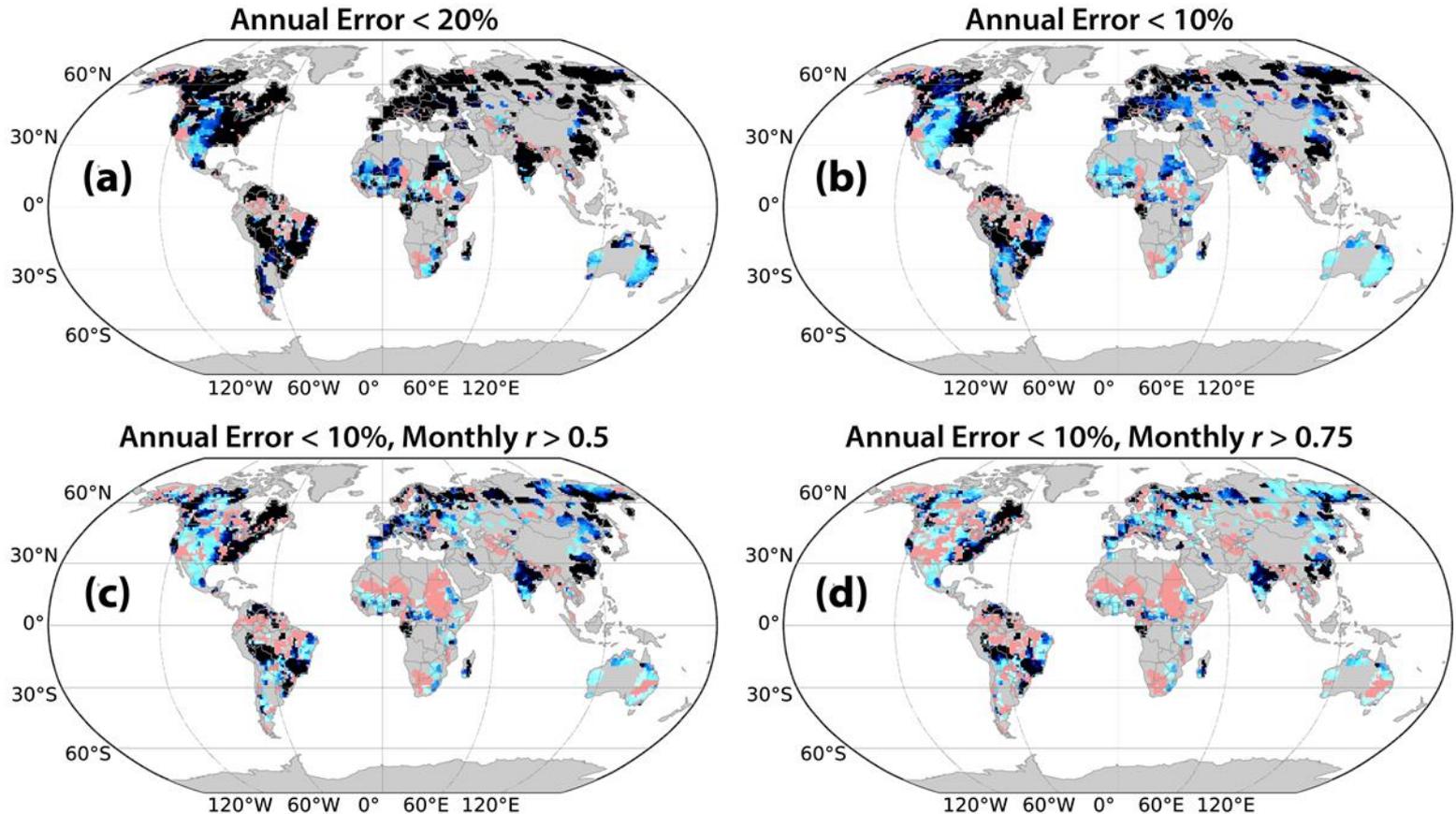
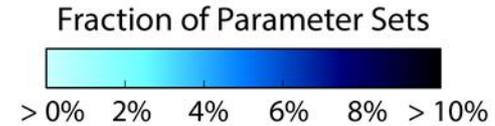


GRDC Observations Limited in Value & Coverage When Assessing Uncertain Hydrologic Extremes

VIC Parameter Ensemble

Fraction of parameter sets meeting error criteria

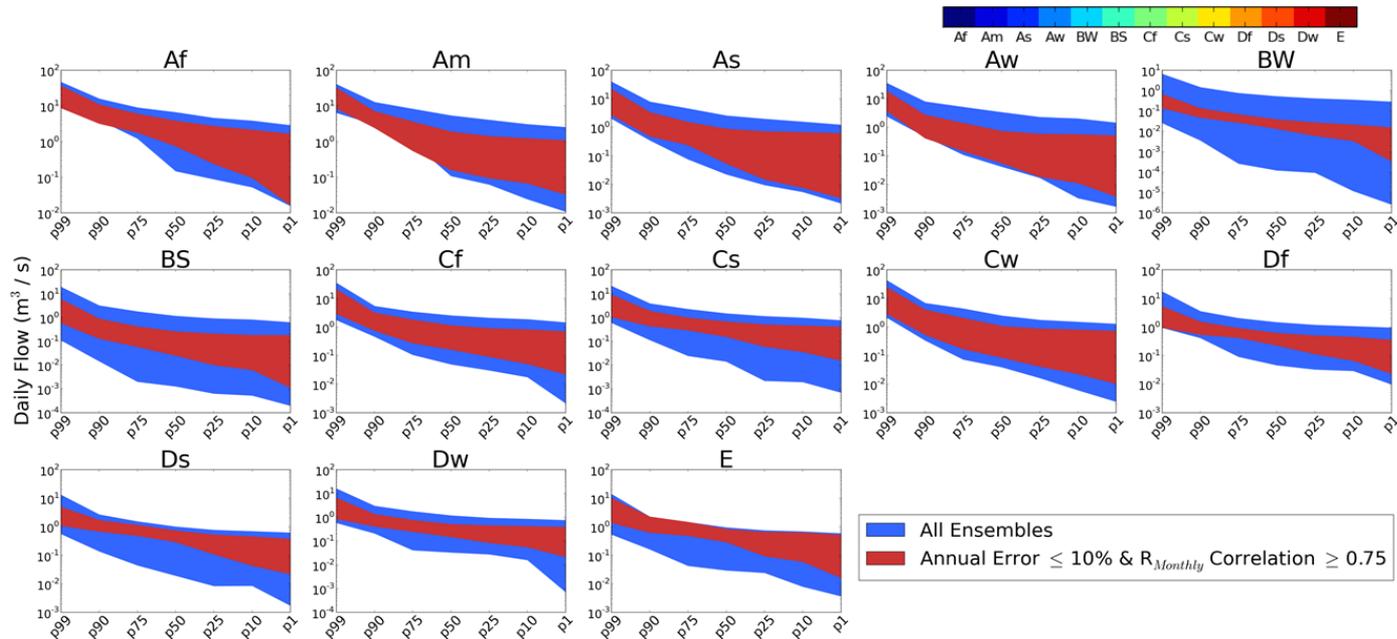
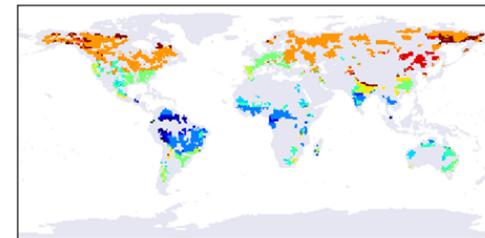
■ No parameter sets meet criteria



GRDC observations may not help reduce uncertainty in floods & droughts in a lot of places

Flow Duration Curve: Climate

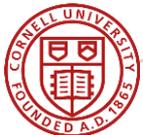
- Spread in the flow duration curve (Daily)
- Large differences between climates



By the way, the observations themselves are bit uncertain when they exist...



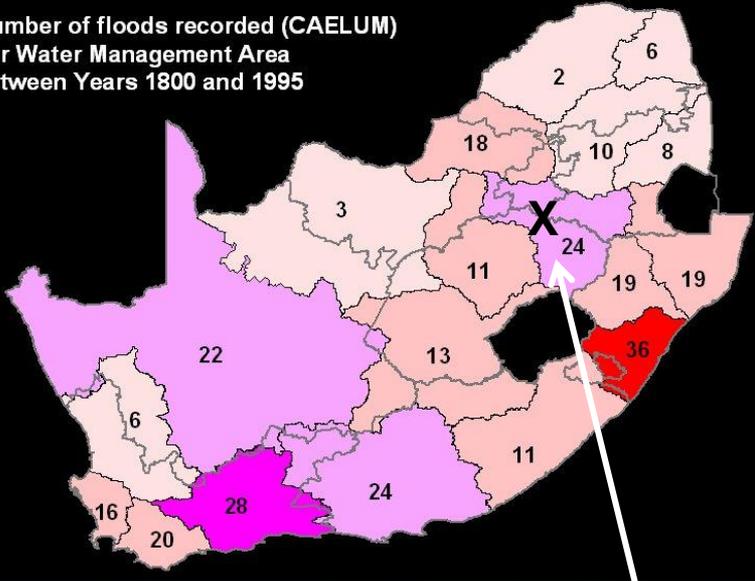
Courtesy of K. van Werkhoven



Some fun with the trivial notion of mass balance

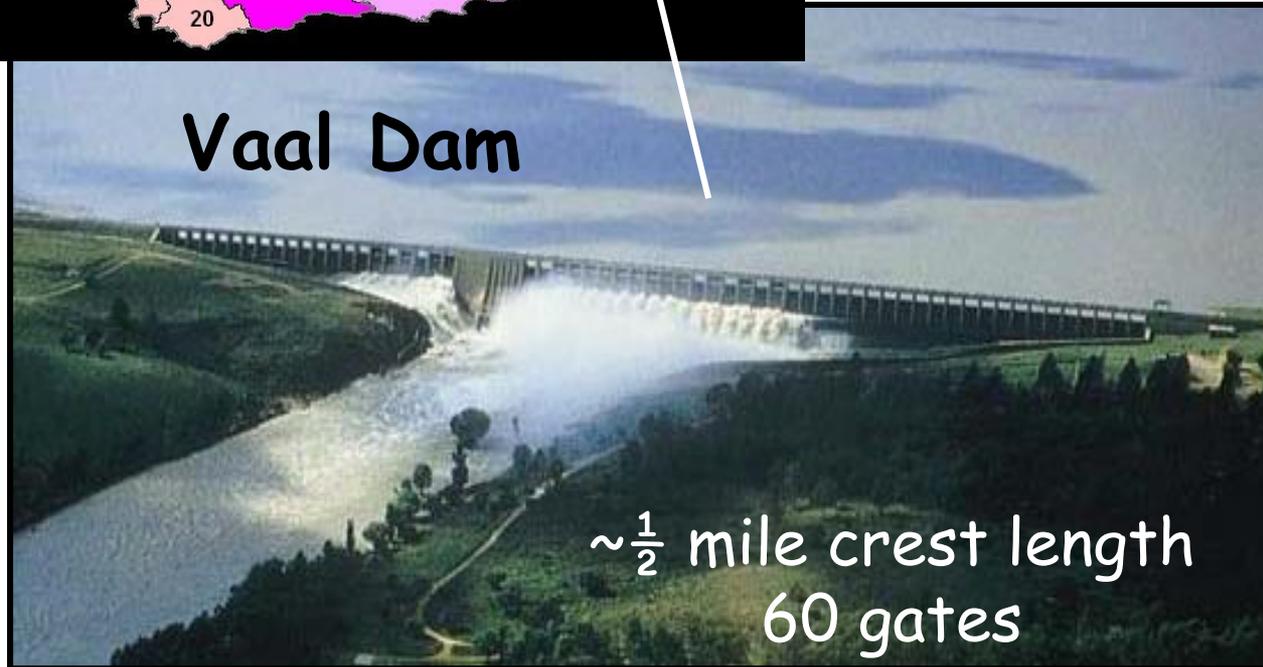
Flood Hazard Areas

Number of floods recorded (CAELUM)
per Water Management Area
between Years 1800 and 1995



- ✓ 56 km from Johannesburg
- ✓ Highly developed downstream
- ✓ Ensure safety of the structures
- ✓ Minimization of damages downstream
- ✓ Ensuring the dams are 100% full after event

Vaal Dam



Capacity

100%

"Full" Storage = 2603 mil m³

+26%

Additional Flood Absorption
Capacity = 677 mil m³

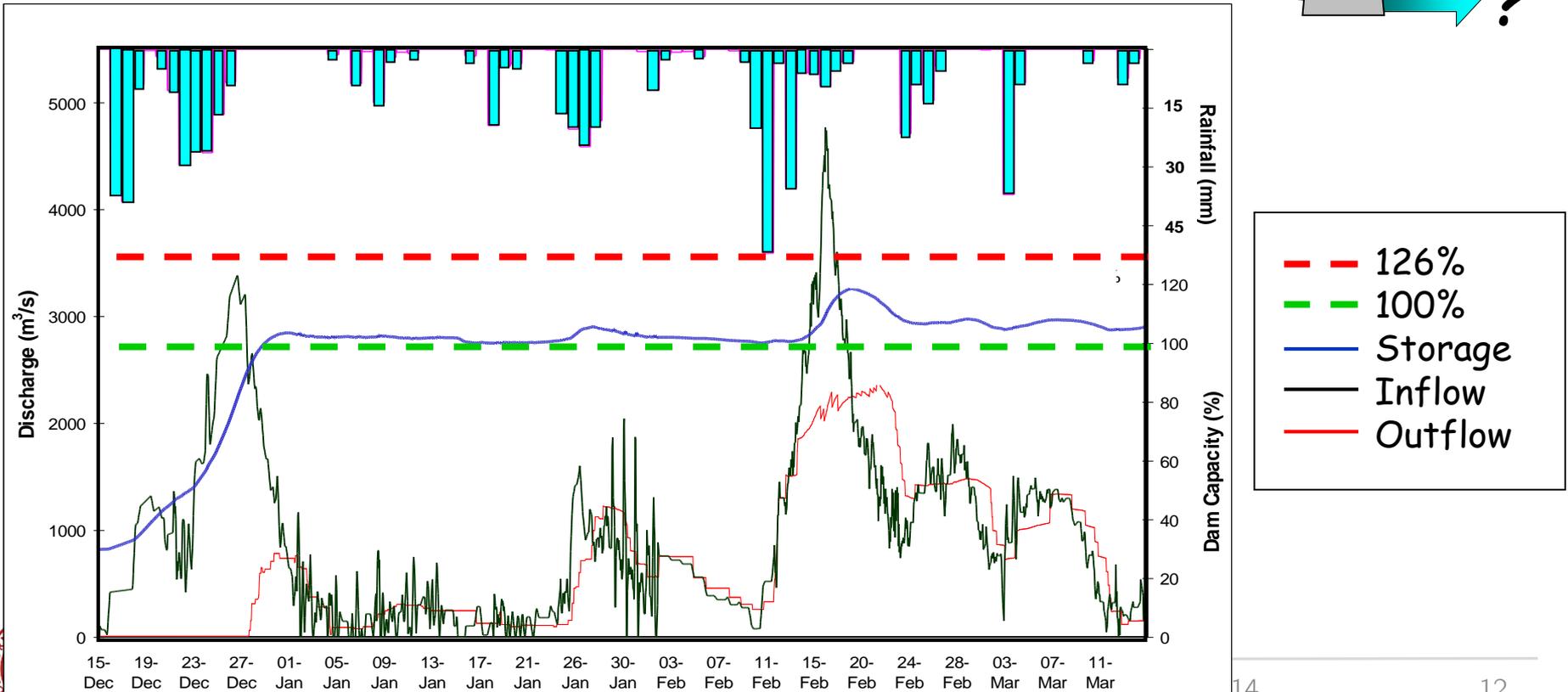
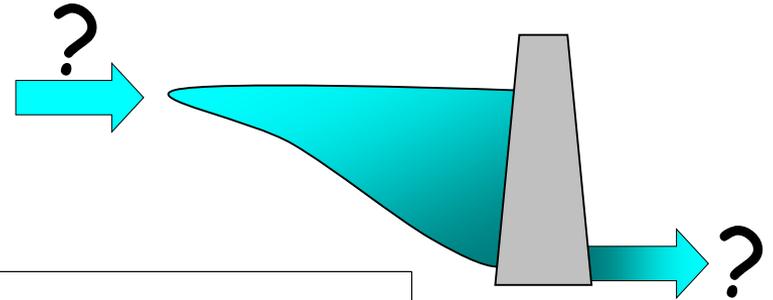
Courtesy of K. van Werkhoven

It's awesome to be lucky in the absence of river forecasting institutions?

forecasting institutions?

February 1996

- ✓ Highest inflow ever recorded at 4800 cms
- ✓ **Already at full capacity**
- ✓ Maximum release 2300 cms
- ✓ Storage peaked at 118.5%



Ok, let's ask a few IAM relevant questions after our mass balance toy problem.

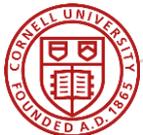
Can IAMs capture major extremes and potential major infrastructure failures that are likely to become more common and dominate impacts?

Is it appropriate to evaluate actual climate change impacts including loss of life using deterministic scenario driven frameworks?

Do IAMs need to deal with the fact that the structure of our “water balance” is changing with an endogenous dependence on human actions?

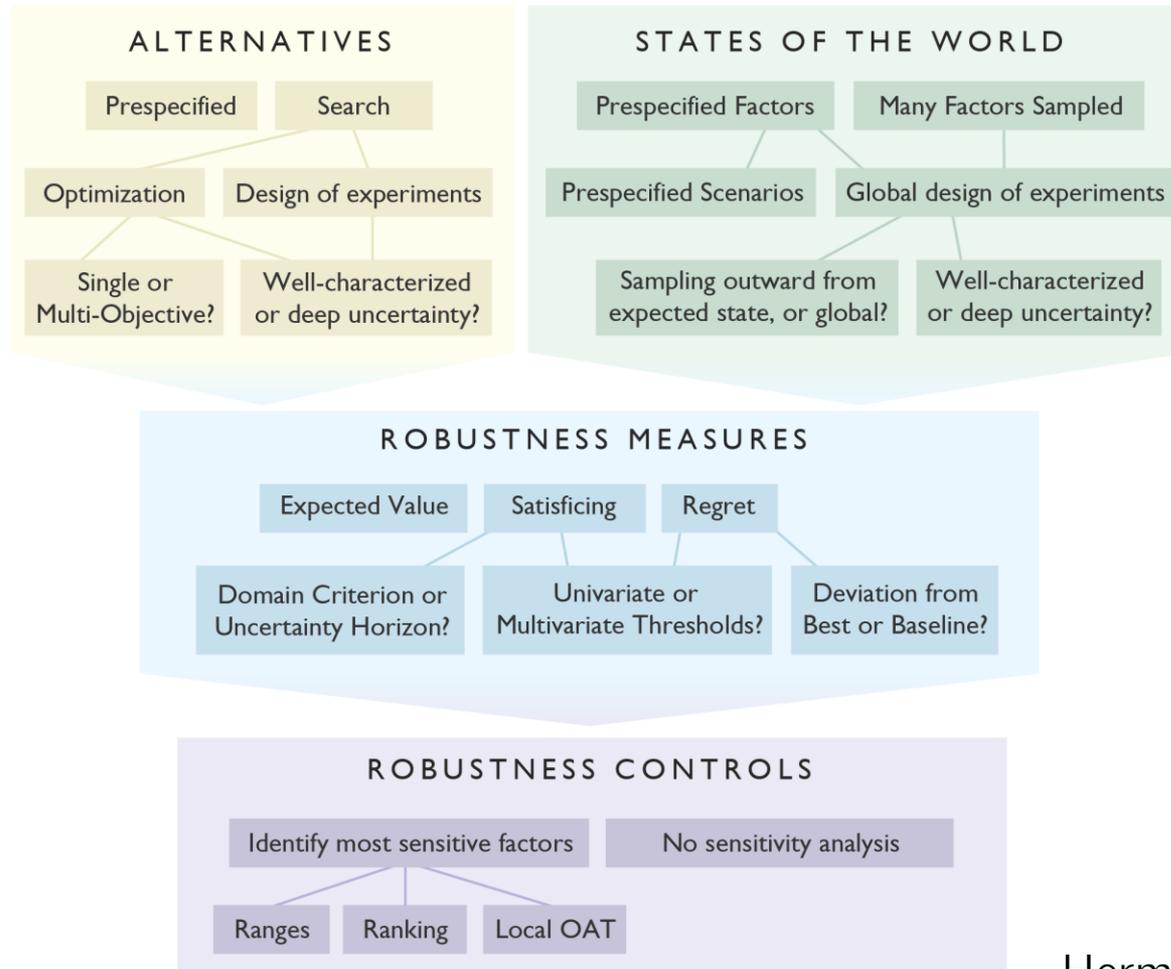
Do IAMs have/need observations for evolving global depletion rates of aquifers, water quality concerns, failures in allocative institutions, locally evolving & competing multi-sector demands, changing infrastructures, financial risks, etc.?

What are the appropriate roles of normative economics? Tacit assumptions? Who is represented in measures of efficiency & equity? Whose impacted negatively by those measures? Willingness to support ex post monitoring/analysis? Culpable for unintended consequences for poor predictions for real systems?

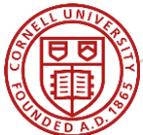


These types of questions are motivating a shift in water resources planning towards “robustness”

Classification of Robustness Frameworks

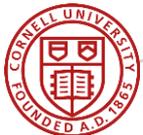


Herman et al. (In-Prep)



Multi-objective Decision Support under Deep Uncertainty

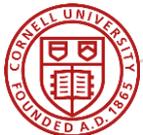
1. Defining stakeholders' decisions and objectives
2. Optimization and preference elicitation
3. Are these solutions robust to uncertainty?
4. Which uncertainties cause critical failures?
5. Balancing robustness across stakeholders



Overview of 'Research Triangle' Water Utilities: North Carolina, USA



- Transition from water abundance to scarcity
- Storage/demand ratios allow intra-regional transfers

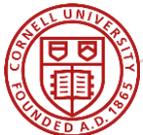


Planning through 2025—new reservoir capacity is expensive and raises environmental concerns.

The utilities seek cooperative agreements to use existing infrastructure more efficiently.



UNC
GILLINGS SCHOOL OF
GLOBAL PUBLIC HEALTH



Utilities' options for cooperative agreements

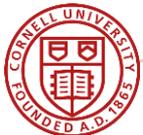
Usage Restrictions: Conserve water, lose revenue

Water Transfers: Maintain reliability without restrictions

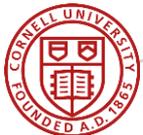
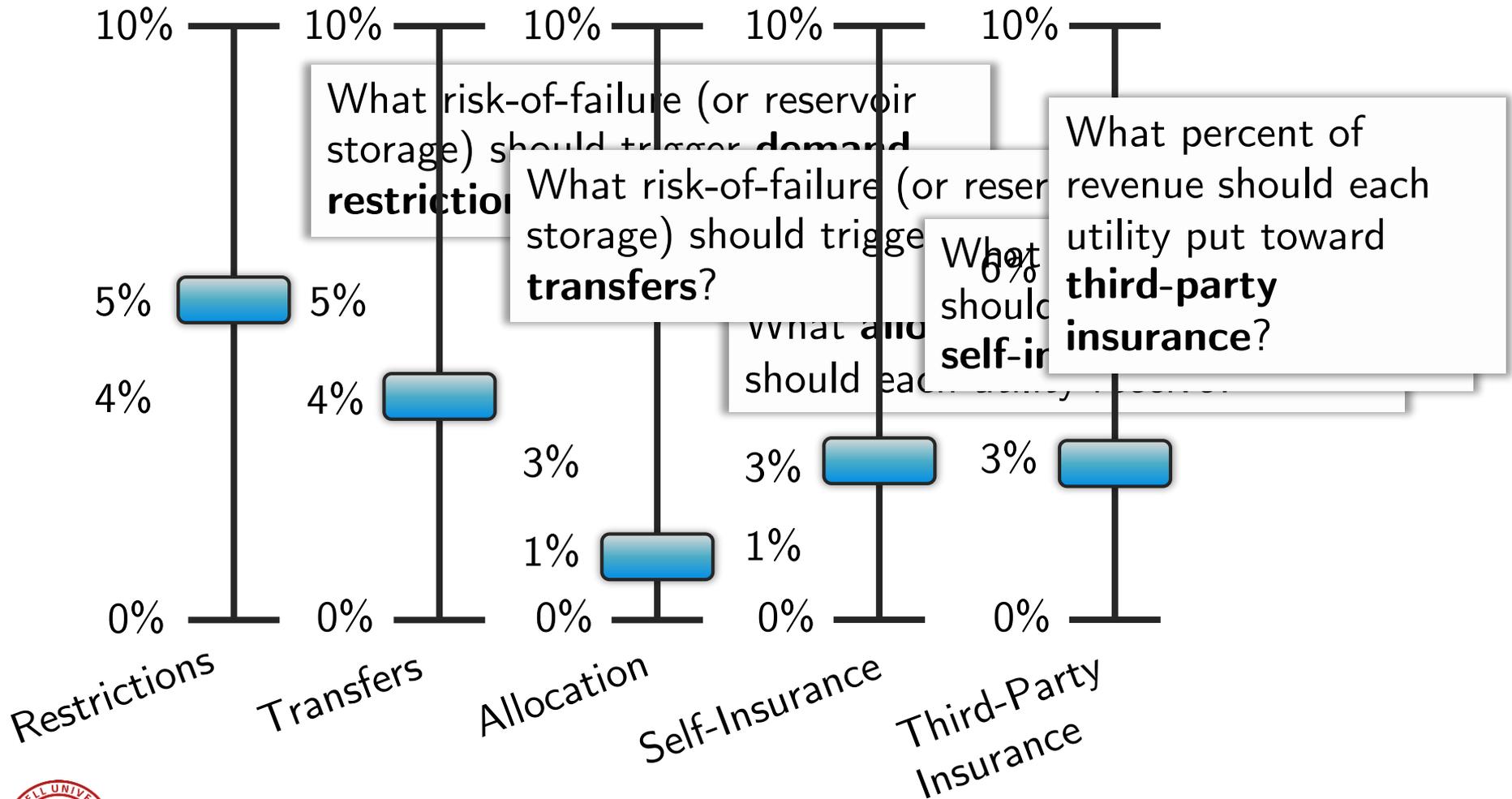
Financial instruments: Insure against lost revenue due to restrictions and transfers

→ Self-Insurance and Third-Party Insurance

(Zeff and Characklis, WRR 2013)



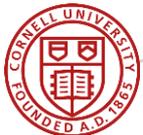
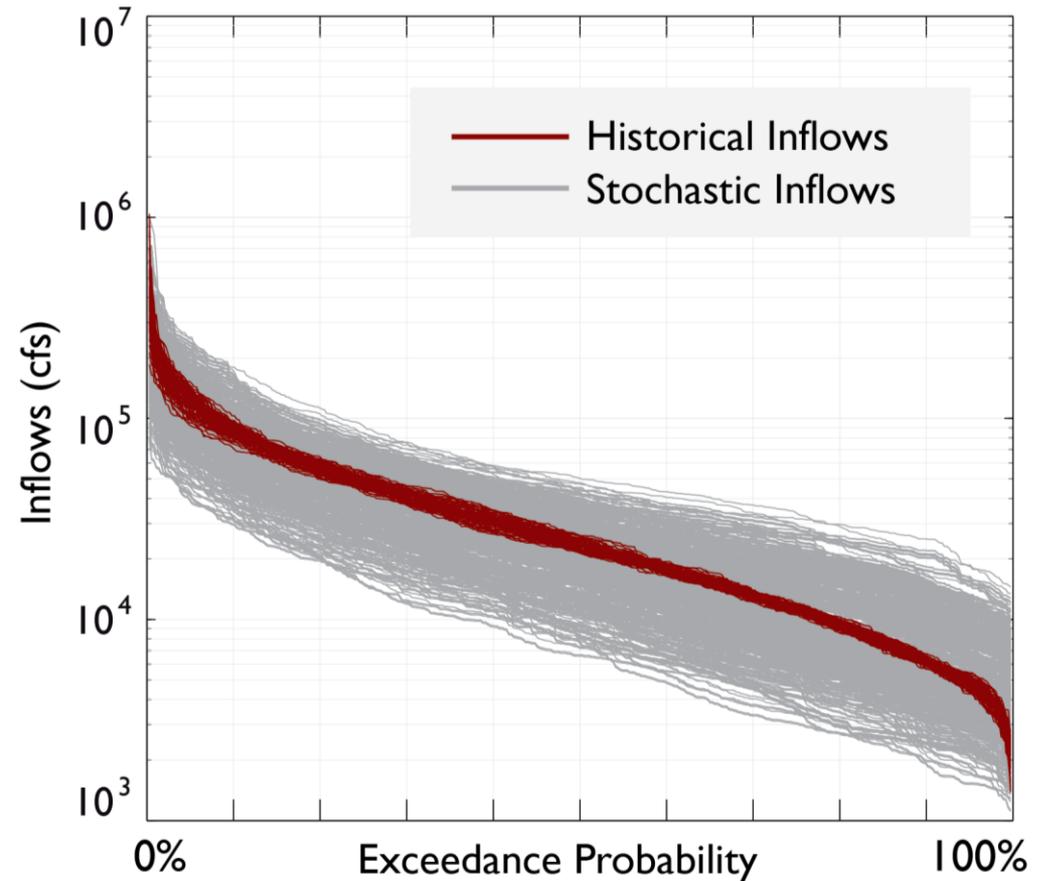
The utilities have defined five decision variables



Future reservoir inflows are uncertain!

Model includes an ensemble of future inflows generated from historical data

Force solutions to perform well across a broader range of hydrologic conditions



Four objectives defined by the utilities

Reliability (Max): # years where reservoir storage $> 20\%$

Restriction Frequency (Min):

years with drought conservation measures enacted

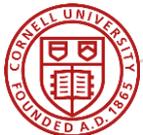
Average Financial Losses (Min):

Revenue reductions + costs due to drought management

Worst-Case Financial Losses (Min):

Financial losses in the 1% worst scenario

The worst-performing utility is optimized such that others will perform as well or better.



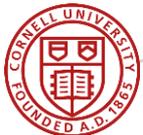
What **formulation** meets stakeholder requirements? A constructive approach

Constructive *decision aiding*:

- Iterative learning process
- Stakeholder feedback
- Multiple formulations posed

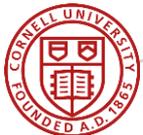
— Tsoukias (2008), From decision theory to decision aiding methodology

1. Restrictions only (status quo)
2. Restrictions + Transfers
3. Restrictions + Transfers + Self-insurance
4. Restrictions + Transfers + Self-insurance + Third-party Insurance

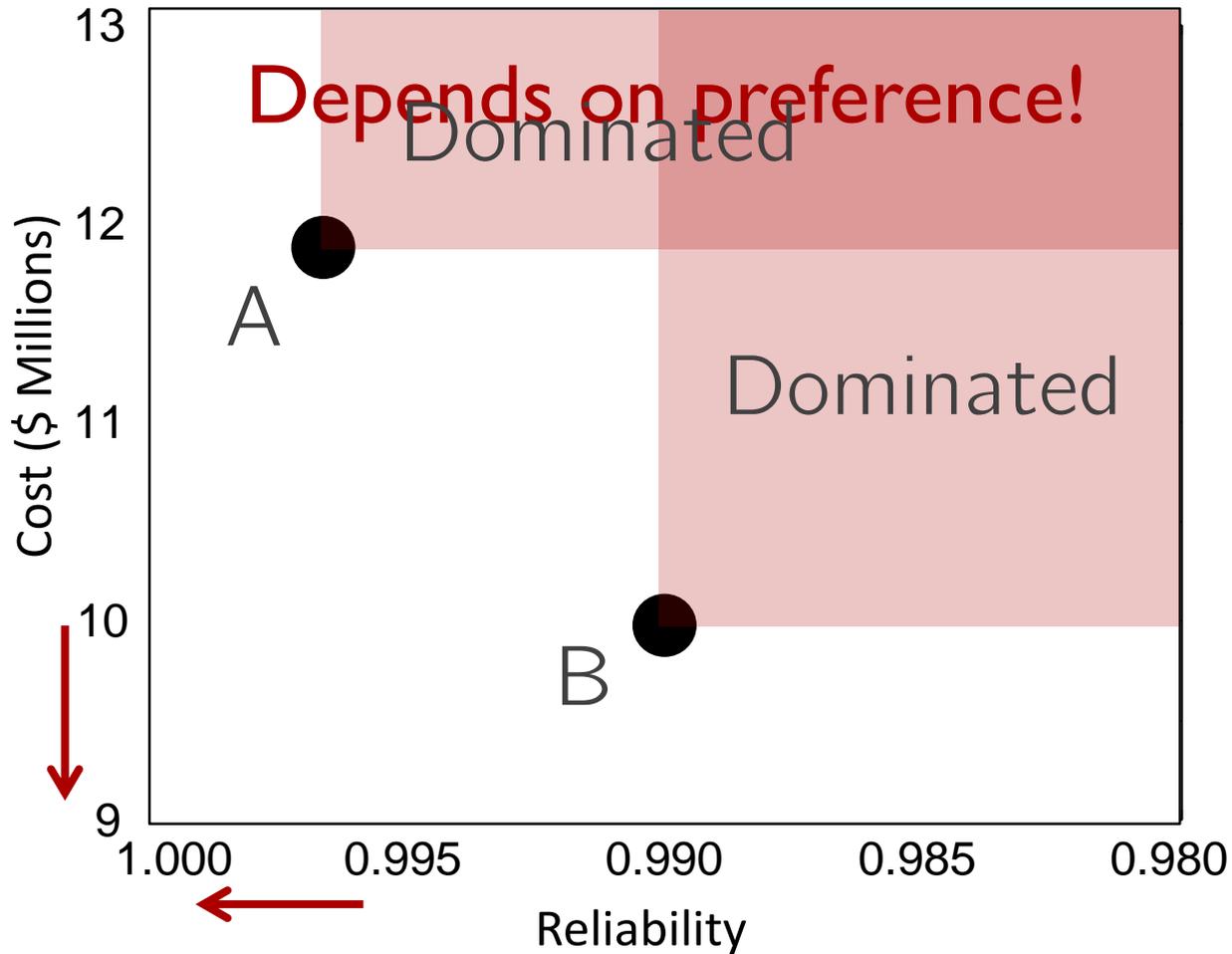


Multi-objective Decision Support under Uncertainty

1. Defining stakeholders' decisions and objectives
2. Optimization and preference elicitation
3. Are these solutions robust to uncertainty?
4. Which uncertainties cause critical failures?
5. Balancing robustness across stakeholders



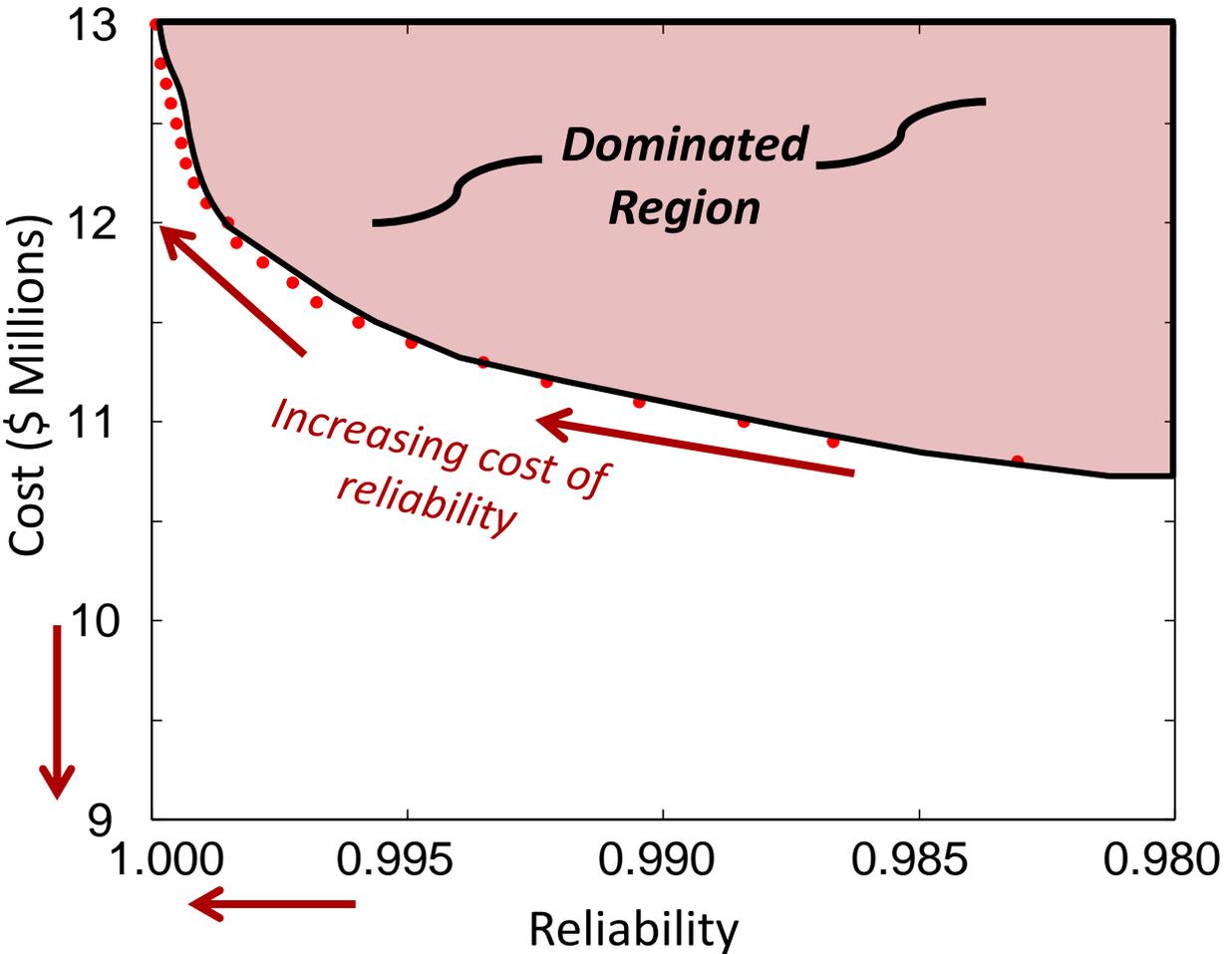
Multi-objective: which solution is better?



Kasprzyk et al. 2009



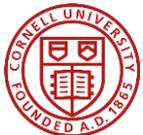
Looking for non-dominated solutions (tradeoff)



Multi-Objective Evolutionary Optimization

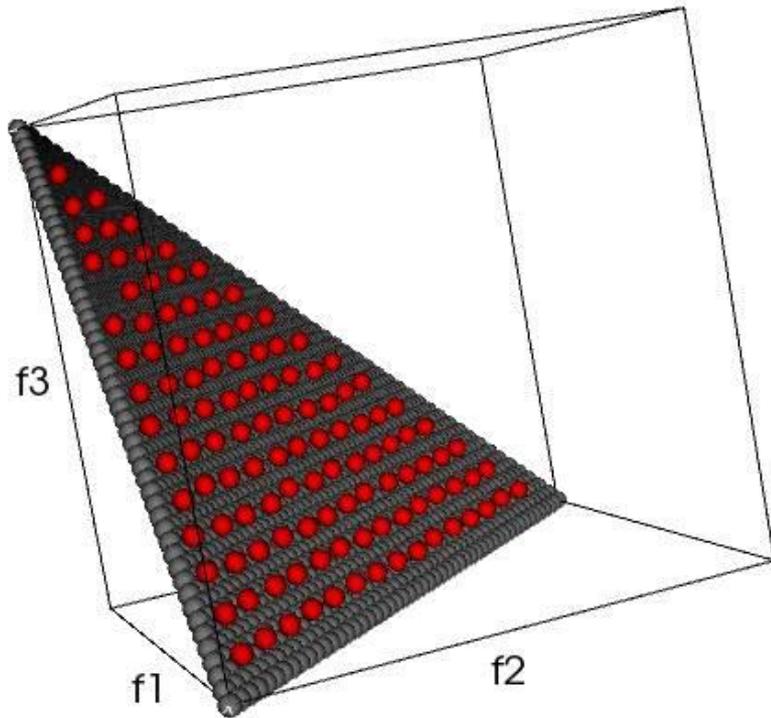
Heuristic method: flexibility
for stochastic problems with
unknown gradients

Search balances
convergence and diversity



Multi-Objective Evolutionary Optimization

Three-objective Test Problem

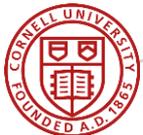


Heuristic method: flexibility for stochastic problems with unknown gradients

Search balances convergence and diversity

Borg MOEA: efficient, reliable performance for water portfolio planning

Reed, P.M., D. Hadka, J.D. Herman, J.R. Kasprzyk, and J.B. Kollat. 2013. Evolutionary Multiobjective Optimization in Water Resources: The Past, Present, and Future. *Advances in Water Resources*, 51, 438–456. [Invited Submission for 35th Anniversary Issue].



High-Performance Computing (HPC) lets us answer questions in minutes instead of days

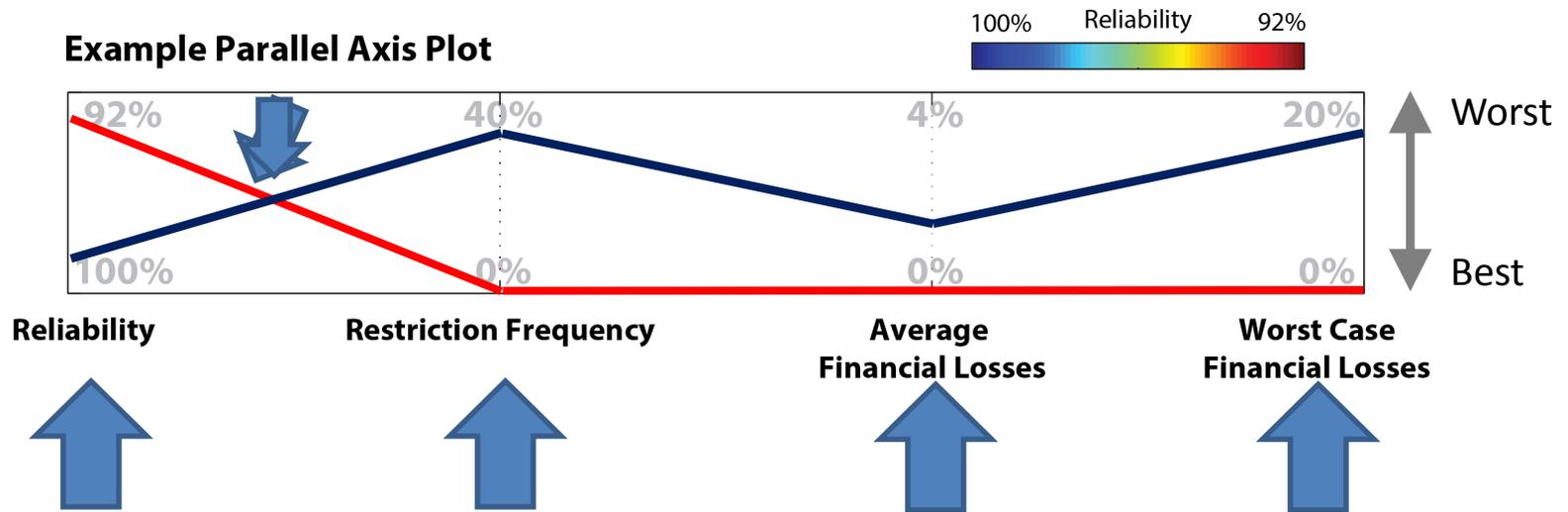


TACC Stampede Cluster

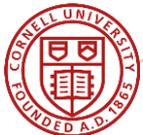
XSEDE
Extreme Science and Engineering
Discovery Environment



Parallel axis plots help stakeholders visualize tradeoffs between conflicting objectives



- Each line represents one solution
- X-Axis shows the four objectives to be optimized
- Y-Axis shows the objective value (performance)
- Crossing lines indicate tradeoffs



Regional Portfolio: Pareto-Optimal Solutions



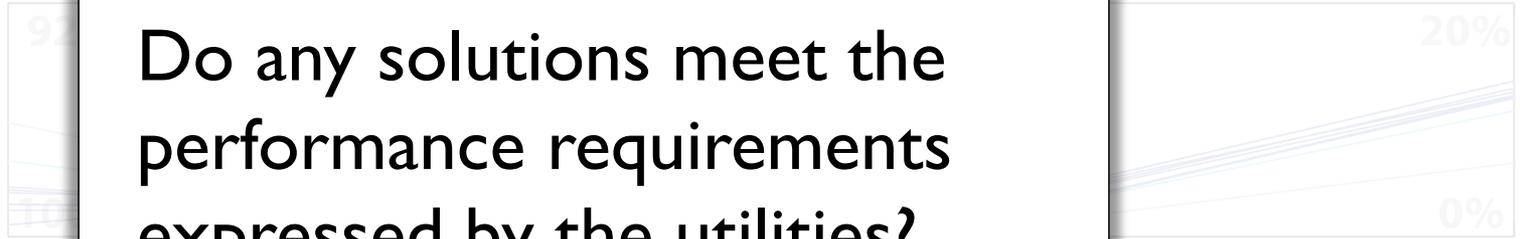
Formulation 1

No Transfers or Mitigation
(8 Solutions)



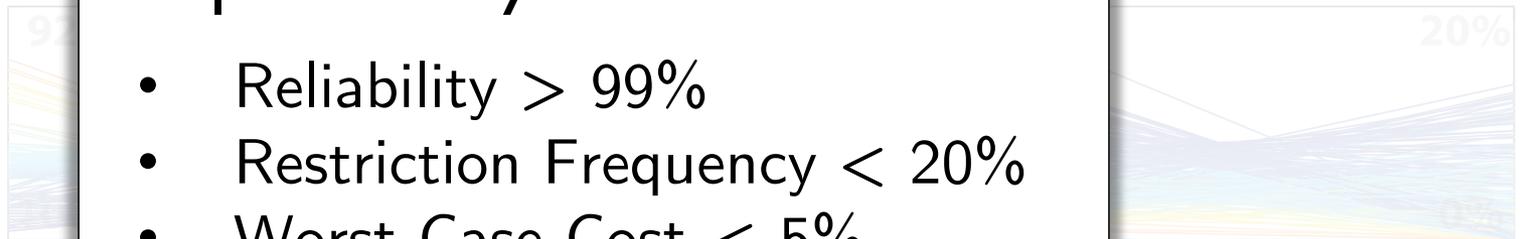
Formulation 2

Add Transfers
(9 Solutions)



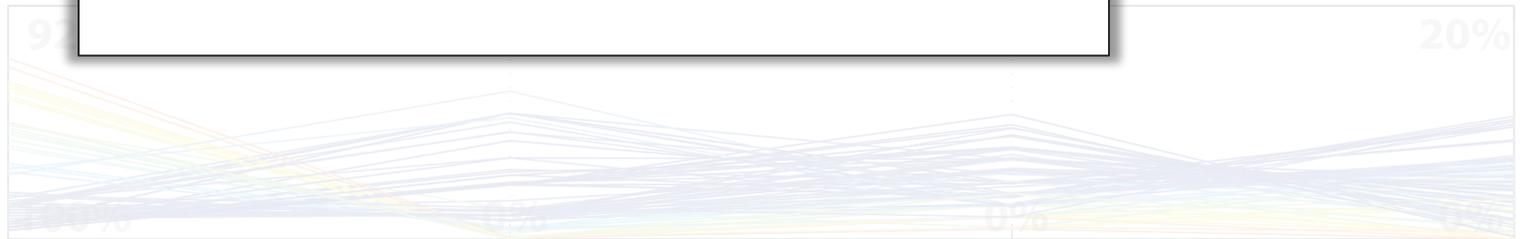
Formulation 3

Add Self-Insurance
(215 Solutions)



Formulation 4

Add Third-Party Insurance
(84 Solutions)



Do any solutions meet the performance requirements expressed by the utilities?

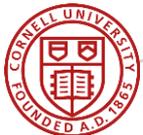
- Reliability $> 99\%$
- Restriction Frequency $< 20\%$
- Worst-Case Cost $< 5\%$

Reliability

Restriction Frequency

Average Cost

Worst Case Cost



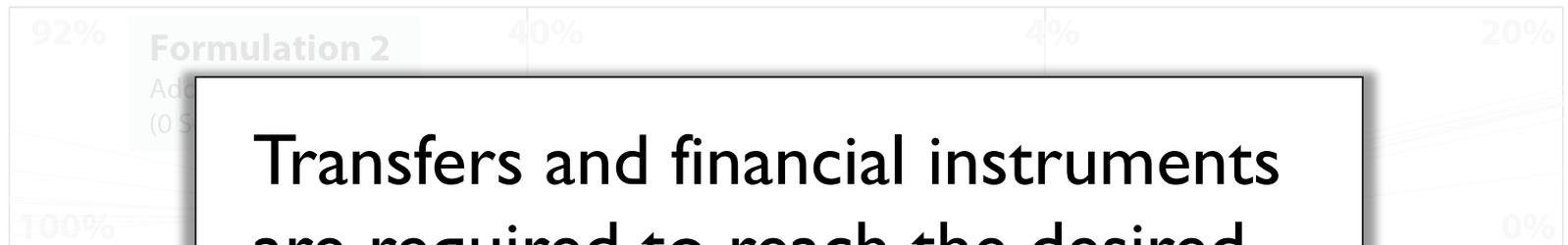
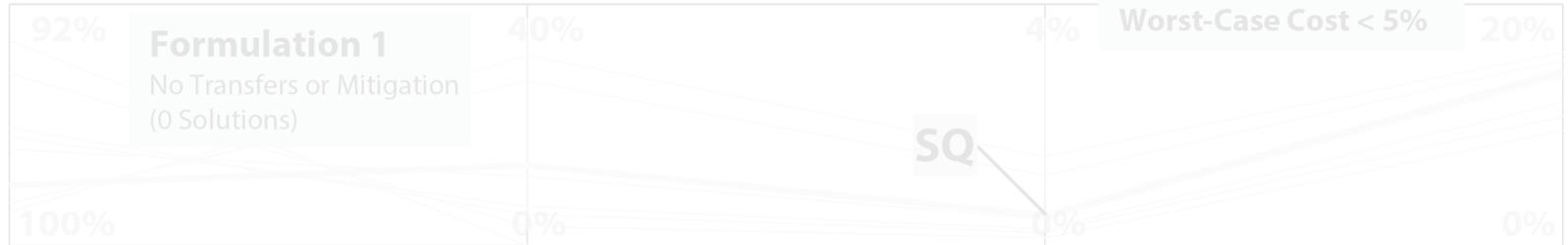
Multi-Objective Optimization: High Reliability Solutions

0% Average Cost 4%



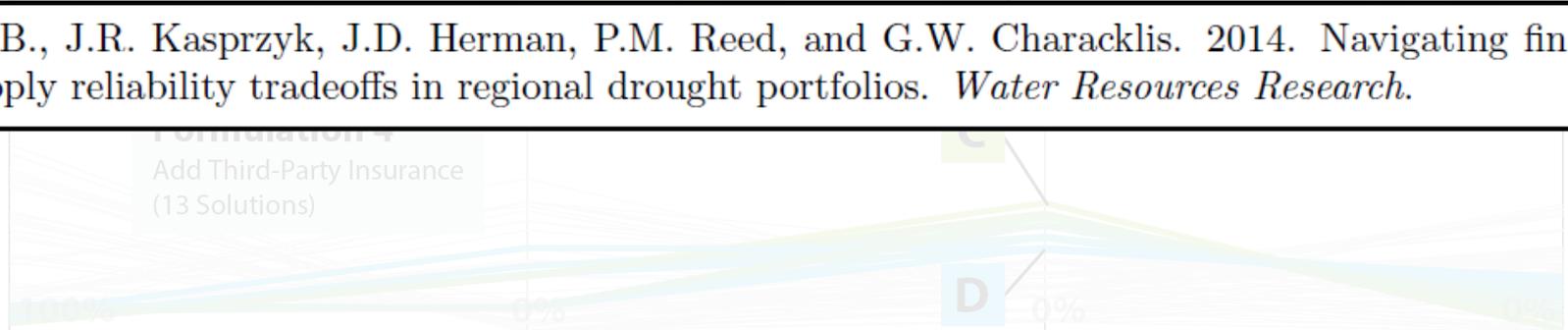
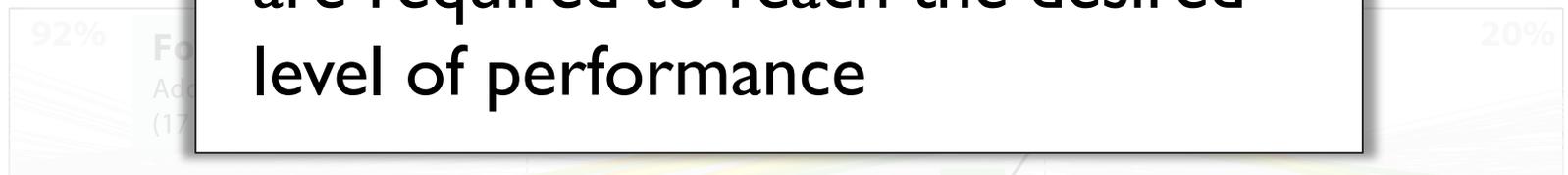
Highlighted Solutions:

- Reliability > 99%
- Restriction Freq. < 20%
- Worst-Case Cost < 5%



Transfers and financial instruments are required to reach the desired level of performance

Zeff, H.B., J.R. Kasprzyk, J.D. Herman, P.M. Reed, and G.W. Characklis. 2014. Navigating financial and supply reliability tradeoffs in regional drought portfolios. *Water Resources Research*.

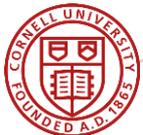


Reliability Restriction Frequency Average Cost Worst Case Cost

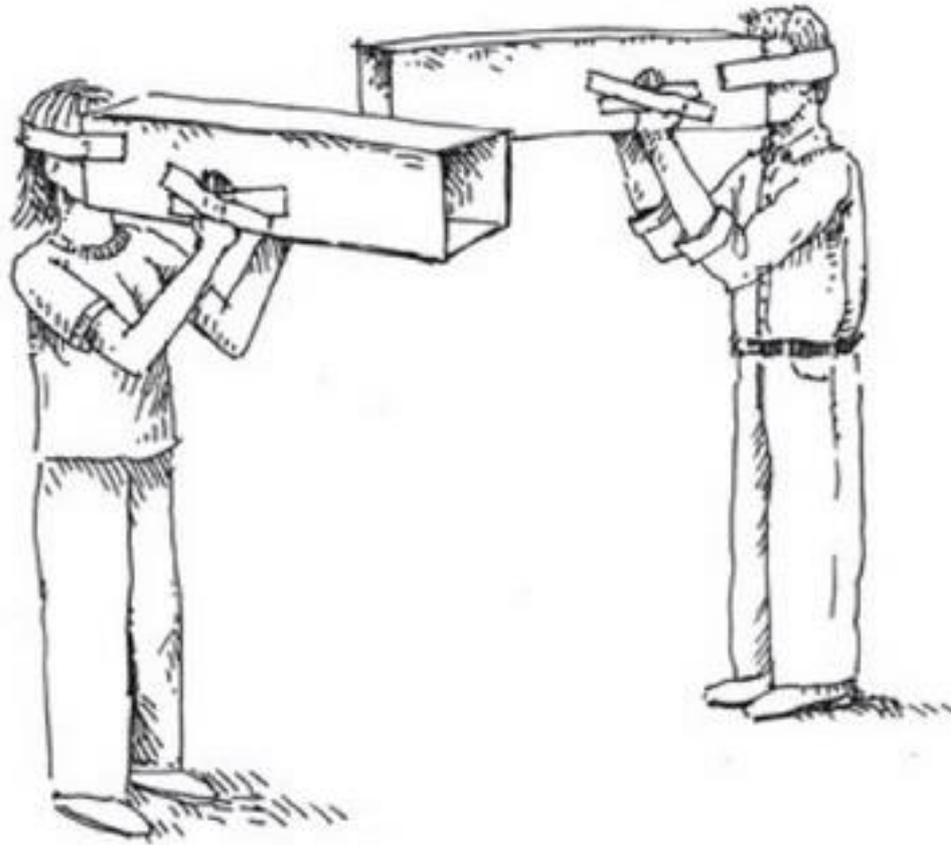


Multi-objective Decision Support under Uncertainty

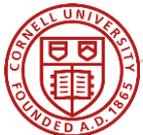
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Optimizing to a single Monte Carlo future: what if we're wrong?



<http://www.hockscqc.com/articles/tunnelvision/tunnel-vision.jpg>



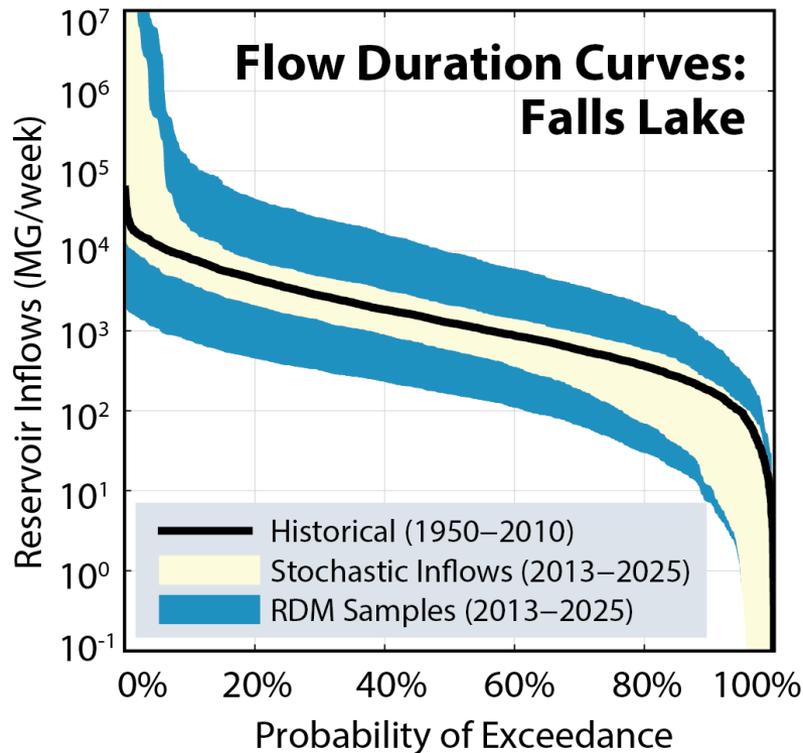
“

Shift from asking the *prediction question* –
‘how likely is this scenario?’
to asking the *decision impact question* –
‘how likely would this scenario need to
be to affect one’s choice of strategy?’ ”

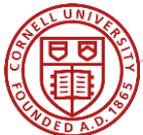
— Bryant and Lempert (2010)



Construct alternative **states of the world** from a set of uncertainties, including inflow and demand multipliers



11 other uncertainties including demand seasonality, reservoir evaporation, transfer prices, etc.

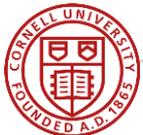


Comparing alternatives under deep uncertainty

Robustness:

The number of sampled states of the world in which a solution satisfies the elicited performance requirements

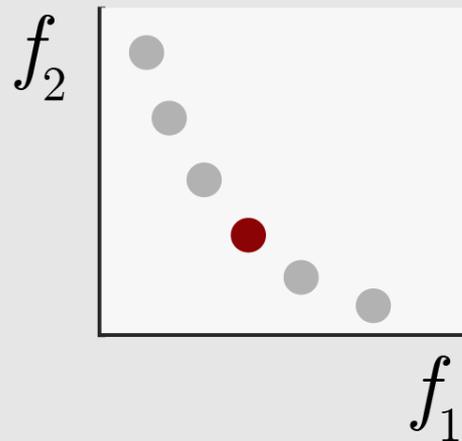
- Seeking “satisficing rather than optimality” (Lempert 2002 PNAS)
- Satisficing concept dates to Simon (1959 Am. Econ. Rev.)



Re-evaluate projected “optimal” solutions in uncertain states of the world

Many-Objective Robust Decision Making (MORDM)

Near-optimal solutions
in expected future



Individual Solution Robustness across Sampled States of the World

Fraction of states of the world in which solutions satisfy the conditions:
Reliability > 99%
Restriction Frequency < 20%
Worst-Case Costs < 5%

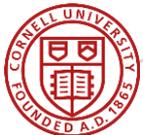


Our projected “optimal” solutions are **not robust** and may cause **tension** between stakeholders

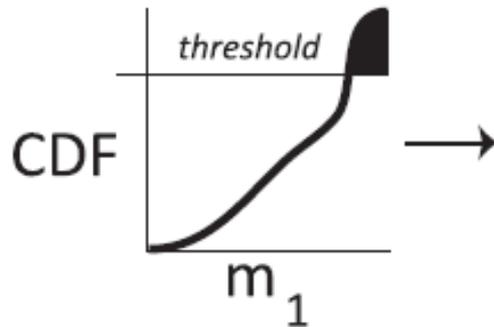


Multi-objective Decision Support under Uncertainty

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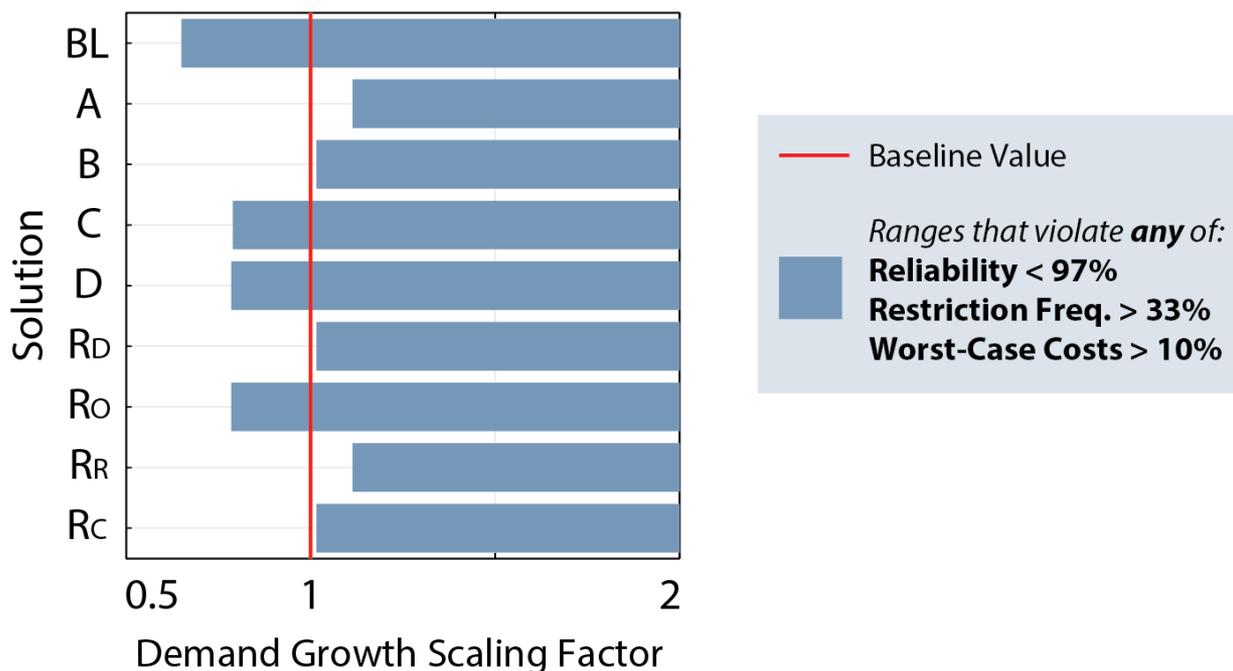
Global sensitivity analysis: Patient Rule Induction Method (PRIM)



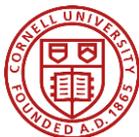
Step 1: Set percentile performance threshold using one or more measures, (m)

The **rate of demand growth** is the factor most likely to violate the utilities' failure thresholds

PRIM: Factors Causing Critical Failures

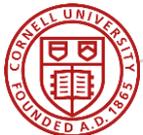


Other factors such as climate, seasonal demand patterns, etc. not found to be influential



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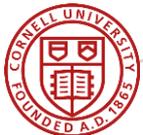
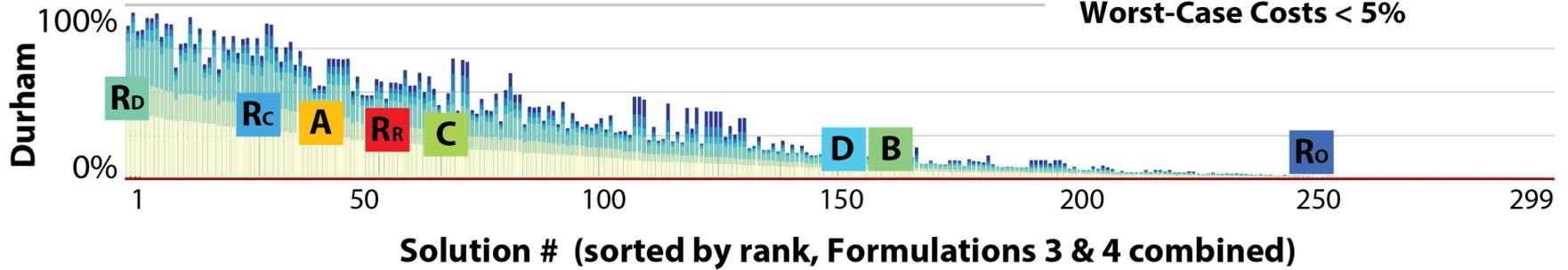
Robustness after Demand Growth Mitigation

Restriction on Demand Growth Scaling Factor:

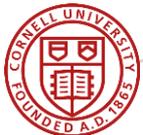
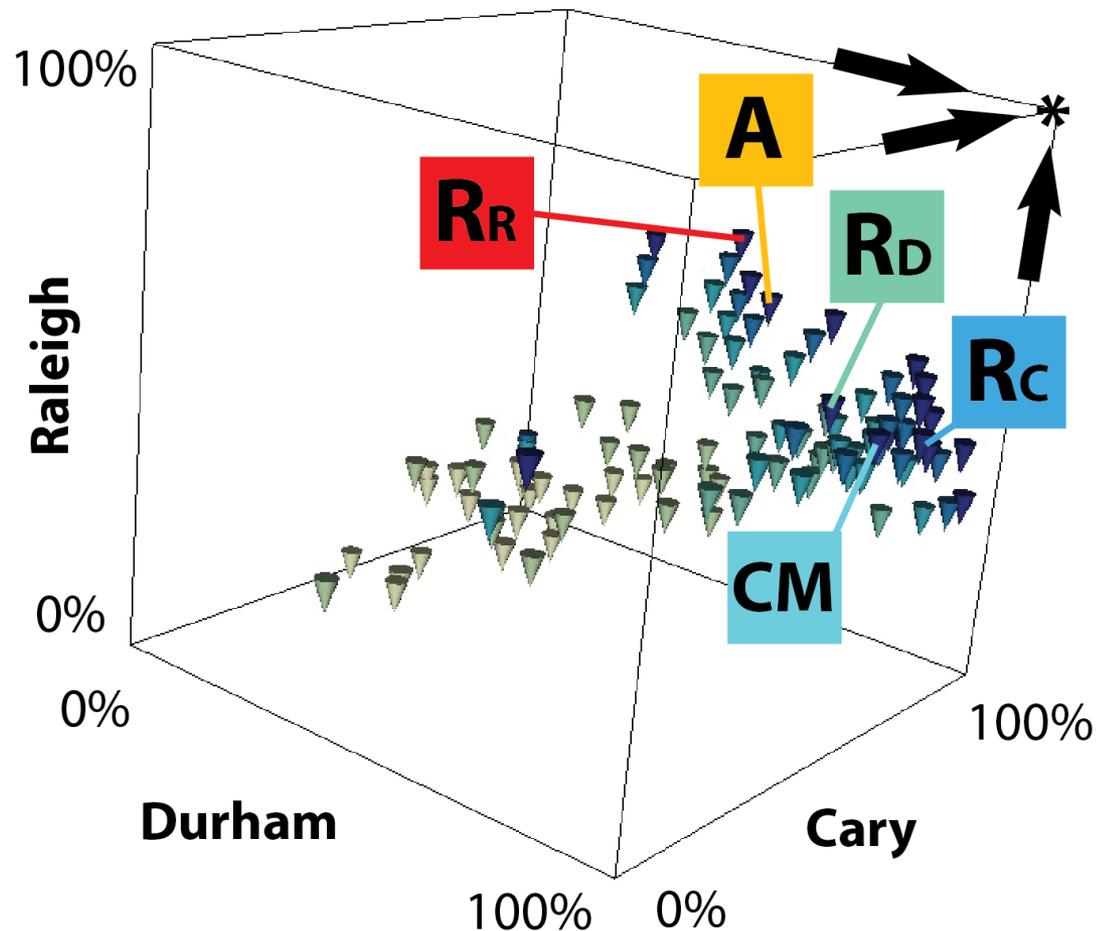
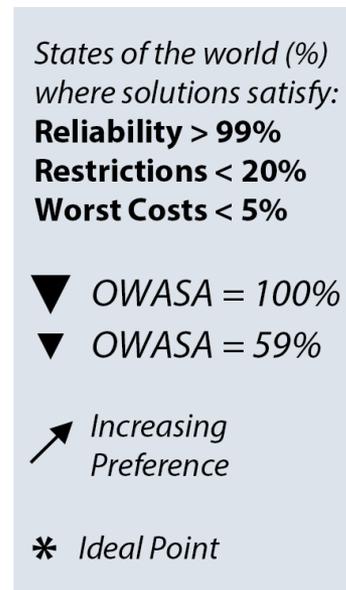
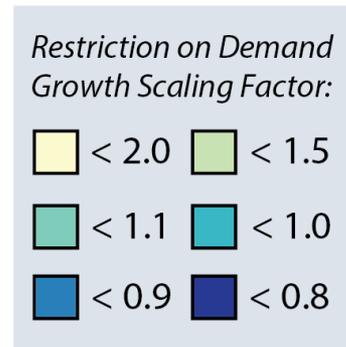


Fraction of states of the world in which solutions satisfy the conditions:

- Reliability > 99%**
- Restriction Frequency < 20%**
- Worst-Case Costs < 5%**



Robustness tradeoffs between stakeholders suggest a compromise solution (CM)



Key Points Highlighted for IAM Discussion

Water is a “wicked problem” with a long history of impacts assessments that should not be ignored.

“Change” is the operative word and climate may not be the headline issue in every future decade.

Predictions under change pose some serious questions about our current knowledge & modeling of these systems.

Uncertainty/risk are critical drivers. Deep uncertainty is a current major focus.

Observations & information systems matter!! Including them in policy analysis makes the “math” pretty difficult though.

Governance, institutions, incentives, evolving demands...not a bad idea to carefully understand the validity of your modeling assumptions.

