

What is robustness?

Problem framing challenges for water systems planning under change

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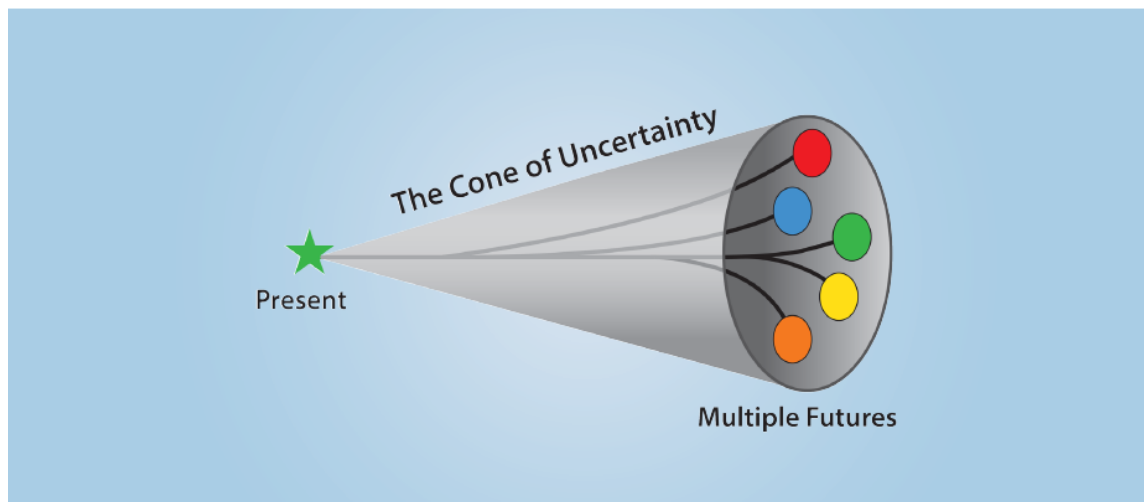
<http://reed.cee.cornell.edu>



Robustness: a climate risk sectoral example

EMBRACING UNCERTAINTY

A Case Study Examination of How Climate Change
is Shifting Water Utility Planning



Prepared for:

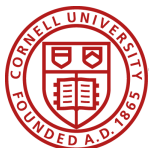
Water Utility Climate Alliance (WUCA)
American Water Works Association (AWWA)
Water Research Foundation (WRF)
Association of Metropolitan Water Agencies (AMWA)

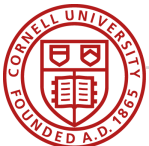
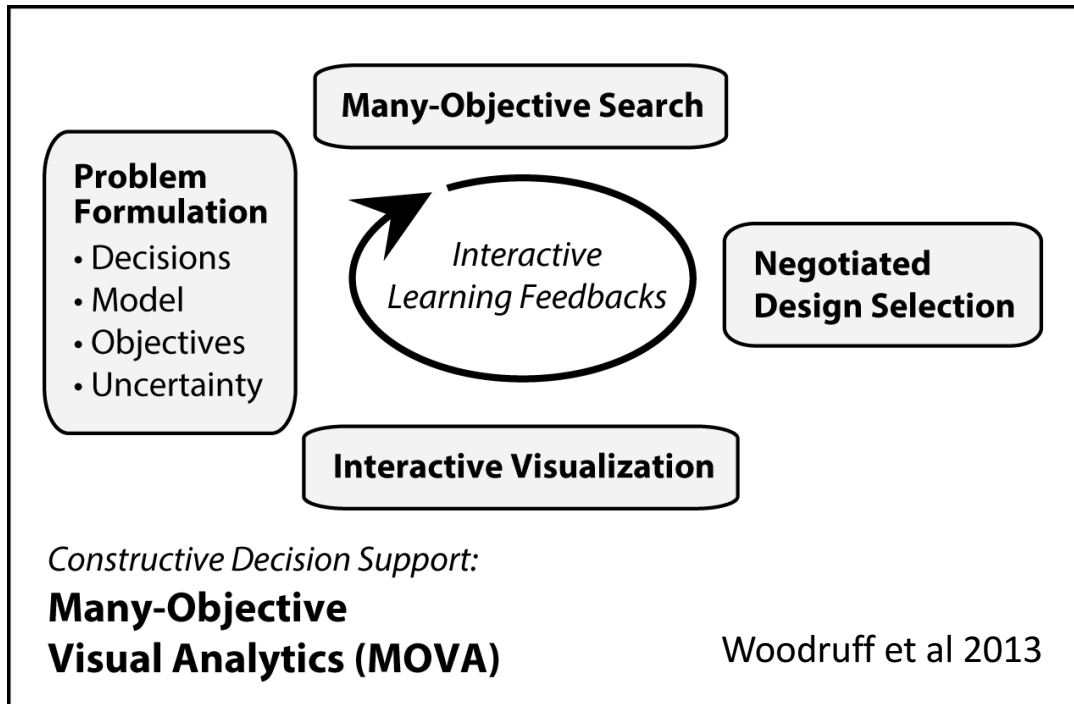
Project Manager: Laurna Kaatz, Denver Water

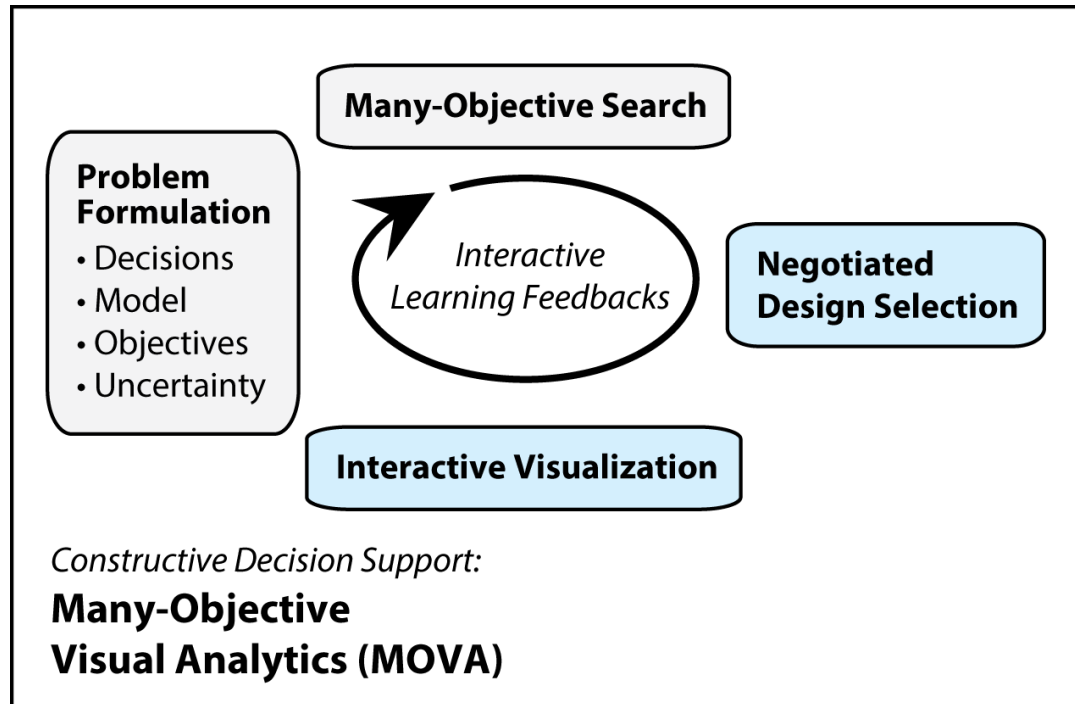
13 Case Studies

- UK & Australia
- Denver Water
- Bureau of Rec.
- CA DWR
- MWD
- many more

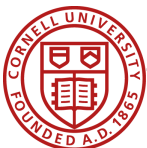
Seeking robustness
across possible futures







Many-Objective Robust Decision-Making for Multiple Stakeholders



Many-Objective Search

Problem
Formulation

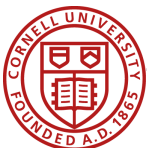
- Decisions
- Model
- Objectives
- Uncertainty

Constructive Decision
Many-Objective
Visual Analysis

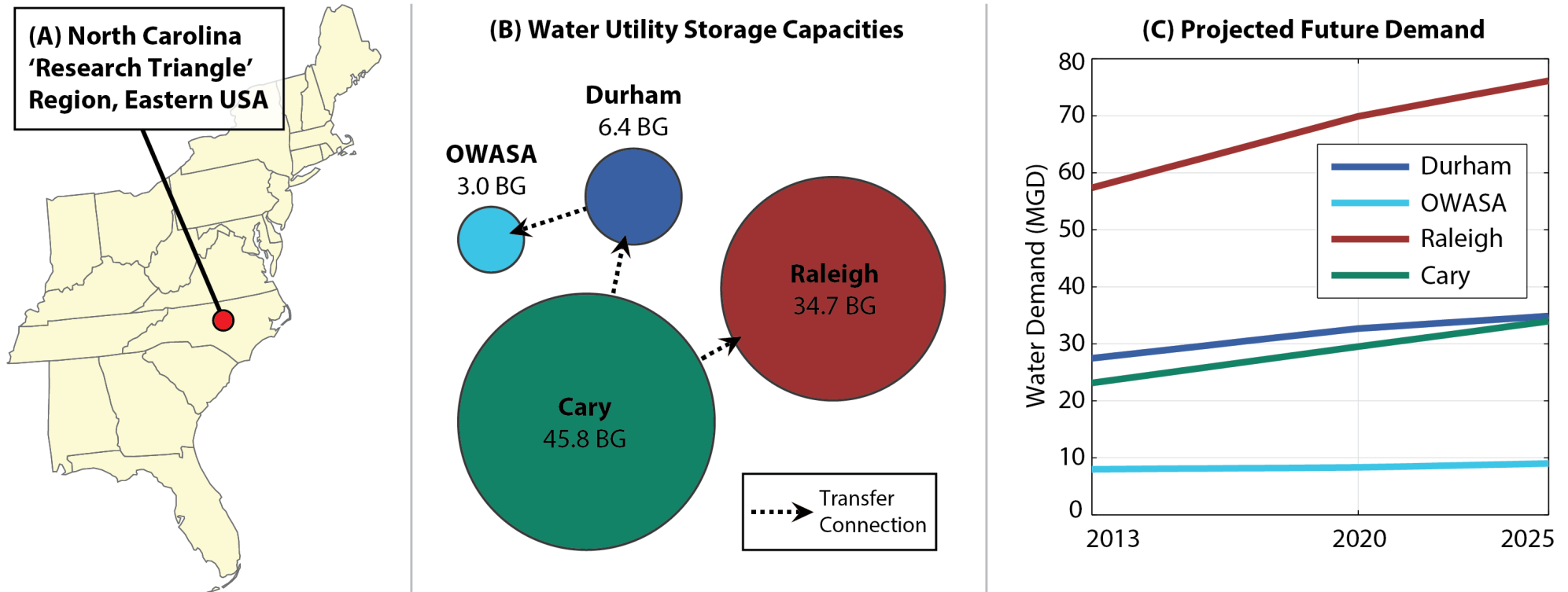
To summarize:

- (1) Rapidly explore multiple competing problem formulations (hypotheses)
- (2) Facilitate learning and visual tradeoff analysis
- (3) Ensure decisions and monitoring recommendations are robust to many futures

Evaluation of
Stakeholders'
Preferences,
Tradeoffs, and
Dependencies



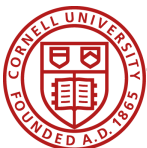
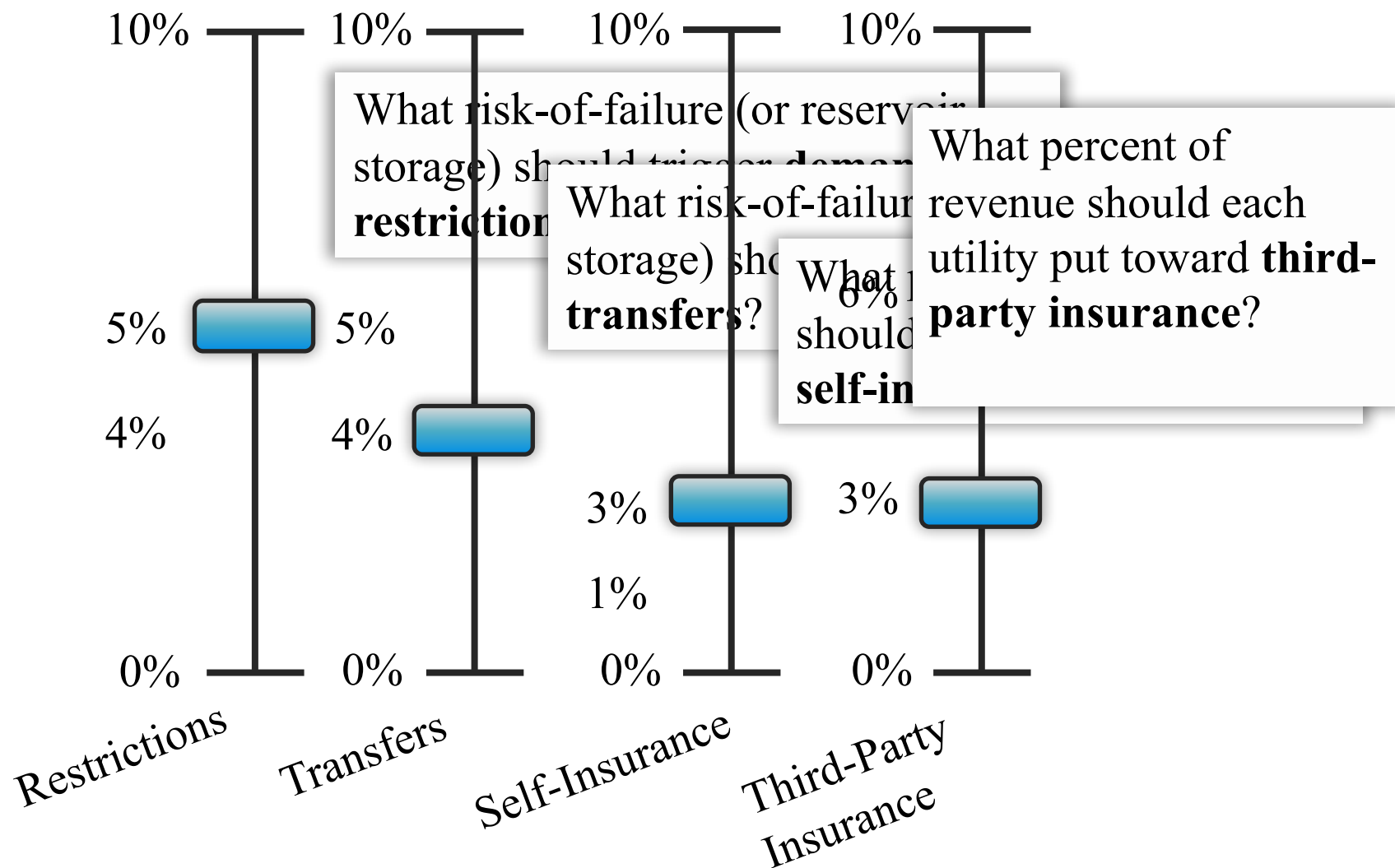
Overview of 'Research Triangle' Water Utilities: North Carolina, USA



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



Each utility has four **decision variables** to model drought management actions



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.



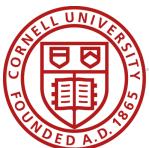
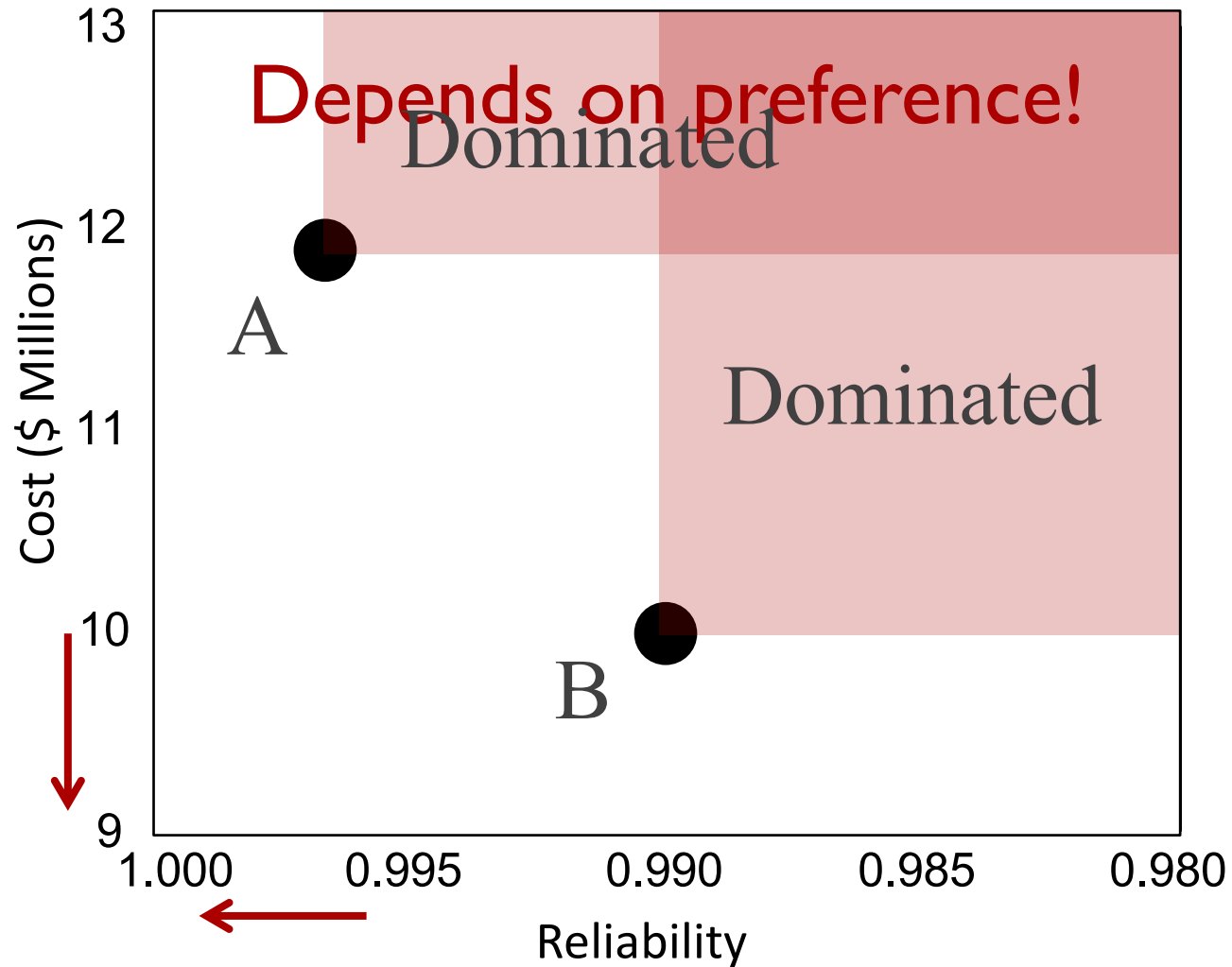
What portfolio complexity is needed?

Multiple formulations tested – a
“constructive” approach (Tsoukias 2008)

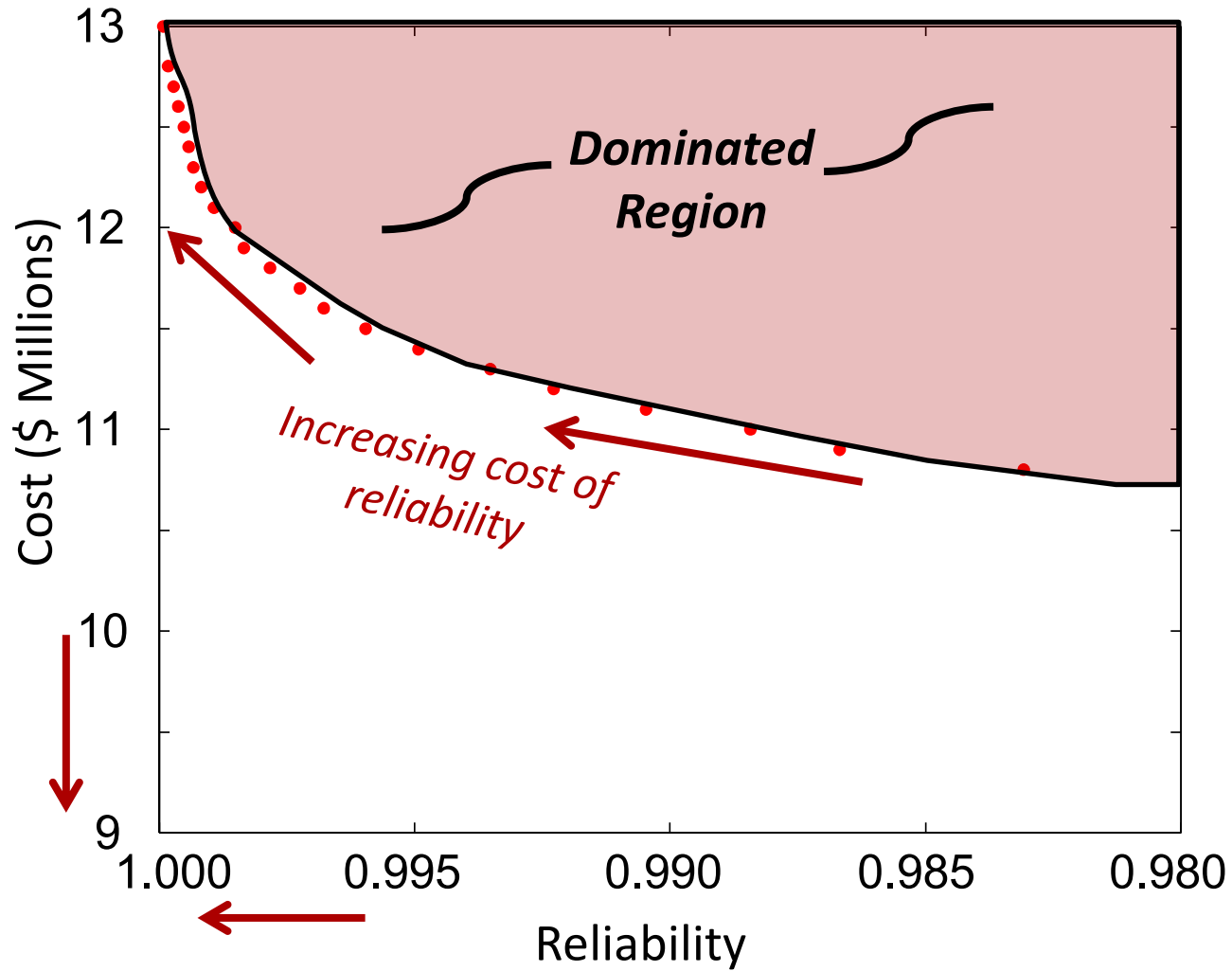
1. Restrictions only (status quo)
2. Restrictions + Transfers
3. Restrictions + Transfers + Self-insurance
4. Restrictions + Transfers + Self-insurance + Third-party Insurance



Multi-objective: which solution is better?



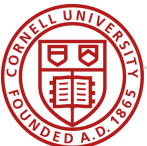
Looking for non-dominated solutions (tradeoff)



Multi-Objective Evolutionary Optimization

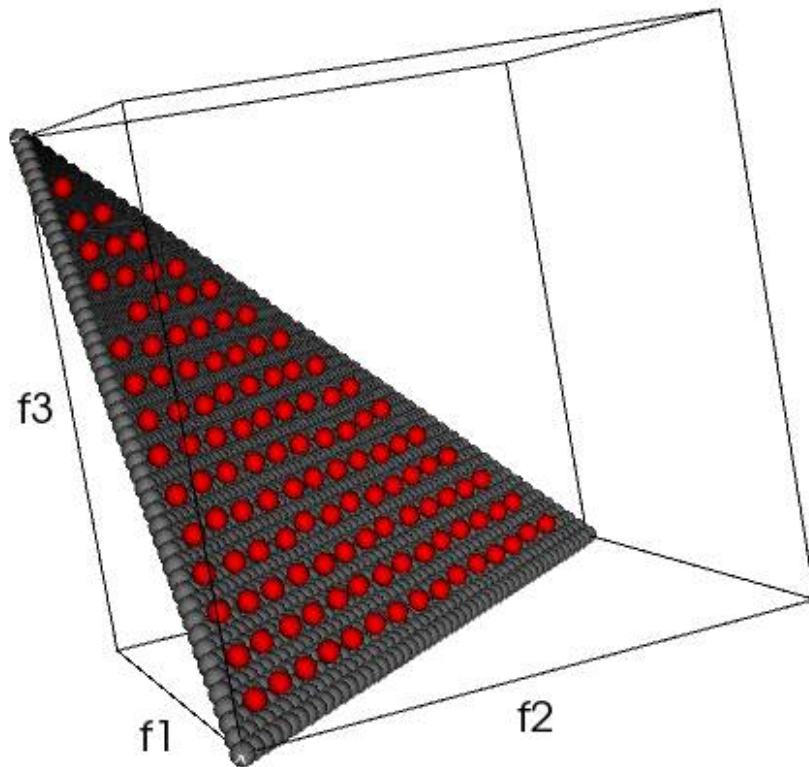
Heuristic method: flexibility
for stochastic problems with
unknown gradients

Search balances convergence
and diversity



Multi-Objective Evolutionary Optimization

Three-objective Test Problem

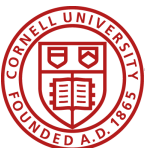


Heuristic method: flexibility for stochastic problems with unknown gradients

Search balances convergence and diversity

Borg MOEA: efficient, reliable broad range of applications

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].



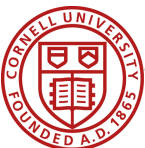
High-Performance Computing (HPC) lets us answer questions in minutes instead of days



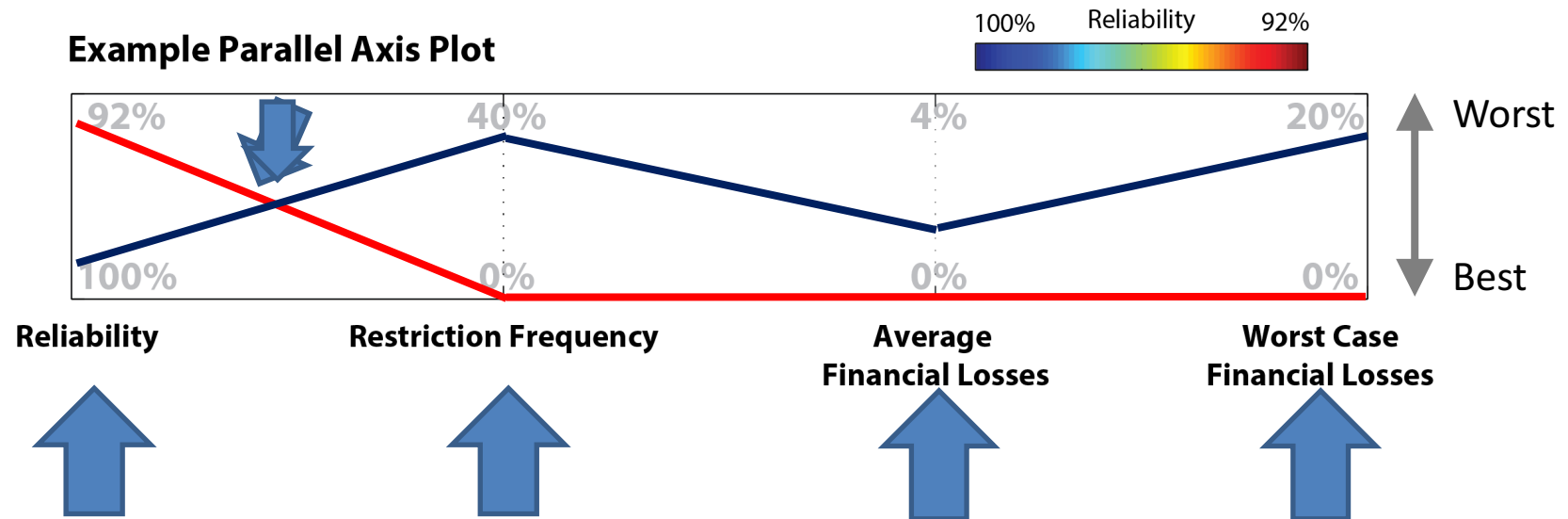
Reed, P. and Hadka, D., "Evolving Many-Objective Water Management to Exploit Exascale Computing", Water Resources Research, In-Press.

XSEDE

Extreme Science and Engineering
Discovery Environment



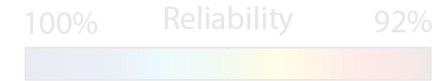
Parallel axis plots help stakeholders visualize tradeoffs between conflicting objectives



- 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



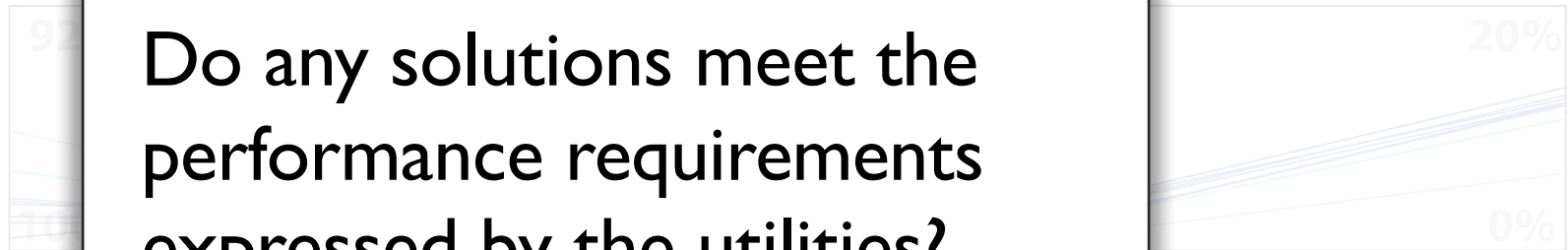
Regional Portfolio: Pareto-Optimal Solutions



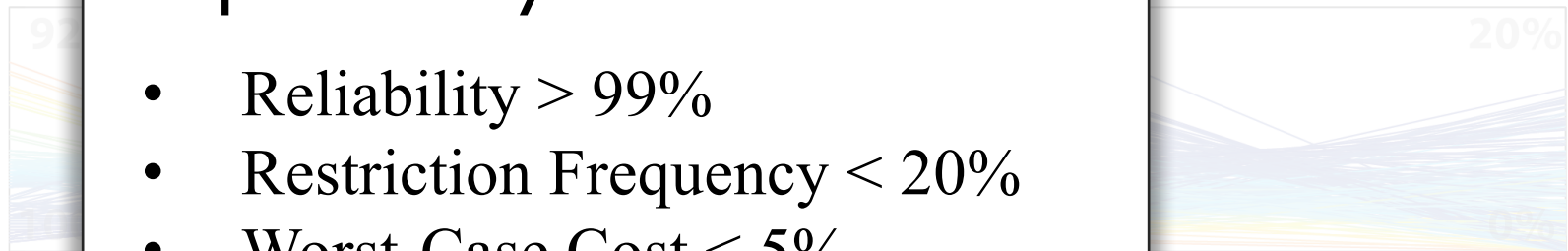
Formulation 1
No Transfers or Mitigation
(8 Solutions)



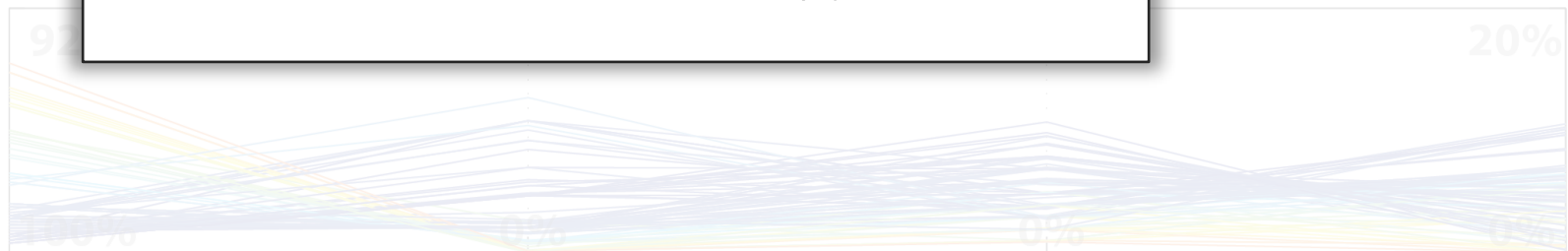
Formulation 2
Add Transfers
(9 Solutions)



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?

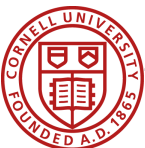
- Reliability $> 99\%$
- Restriction Frequency $< 20\%$
- Worst-Case Cost $< 5\%$

Reliability

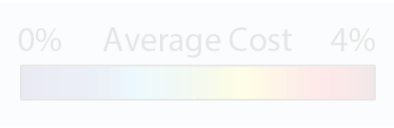
Restriction Frequency

Average Cost

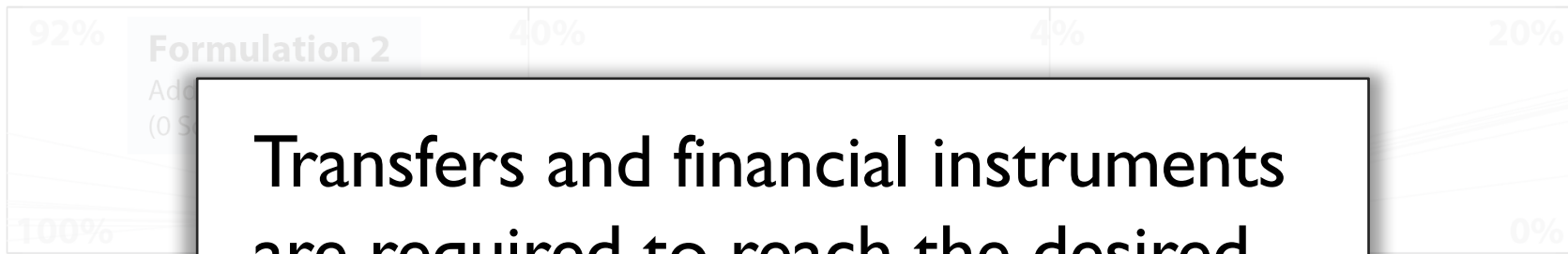
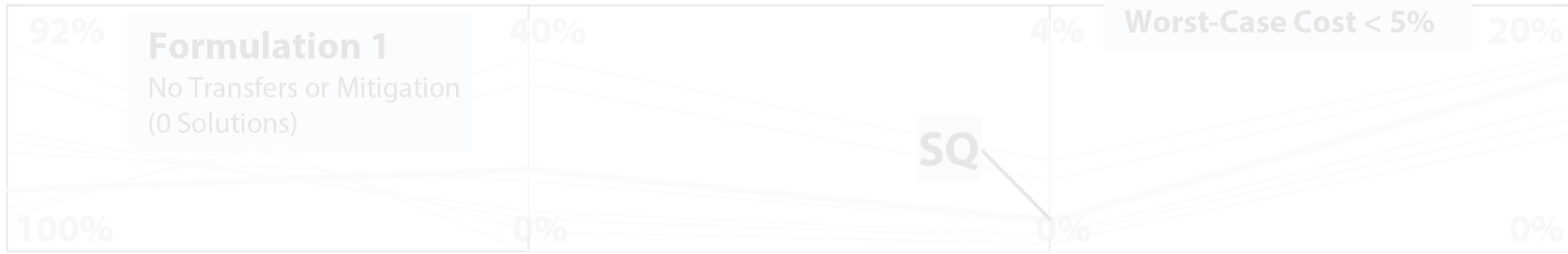
Worst Case Cost



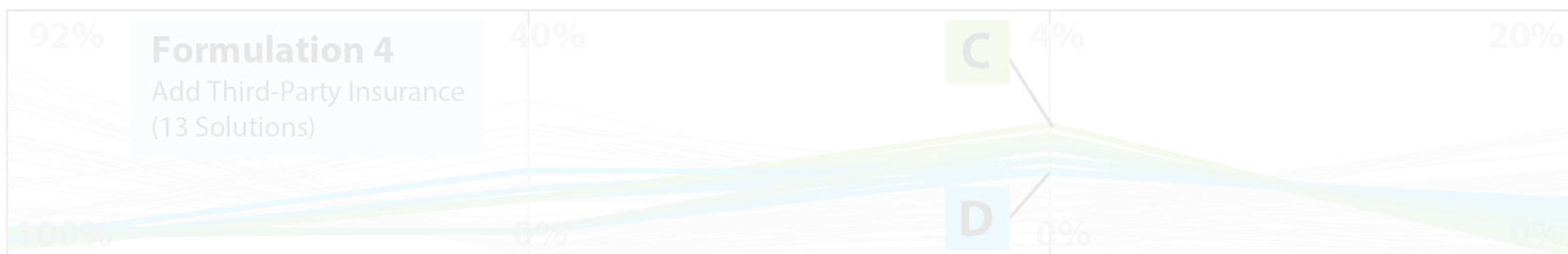
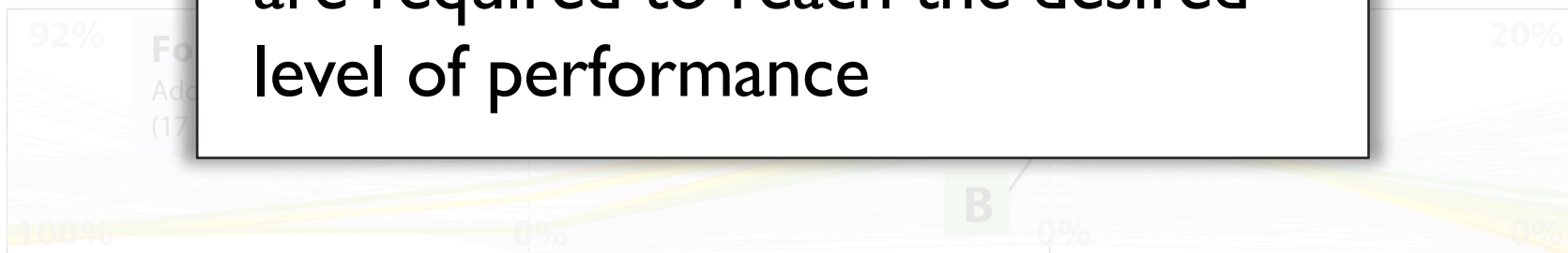
Multi-Objective Optimization: High Reliability Solutions



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



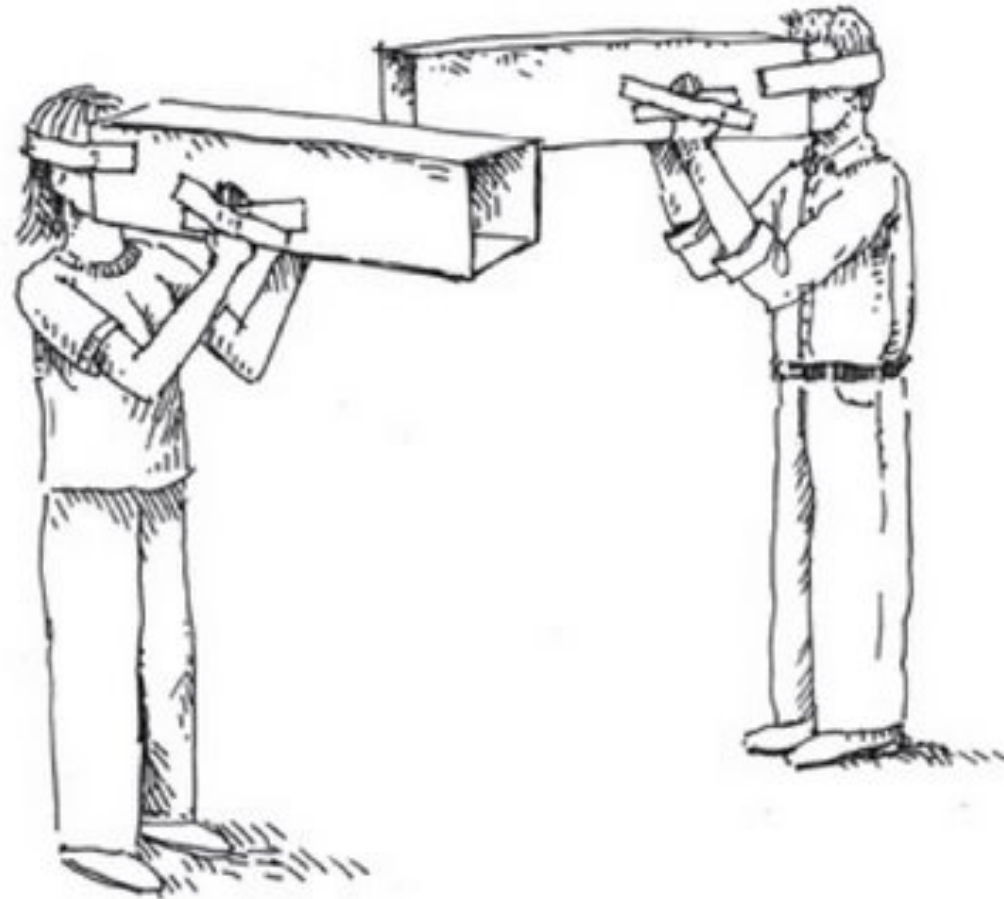
**Transfers and financial instruments
are required to reach the desired
level of performance**



Reliability Restriction Frequency Average Cost Worst Case Cost



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



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



Many-Objective Robust Decision-Making for Multiple Stakeholders

What are the decision-relevant consequences of the choices we make when analyzing robustness?

Problem Formulation

- Decisions
- Model
- Objectives
- Uncertainty

Many-Objective Search

Interactive Learning Feedbacks

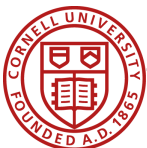
Negotiated Design Selection

Constructive Decision Support
Many-Objective Visual Analytics (MVA)

Evaluation of stakeholders' vulnerabilities, tradeoffs, and dependencies

robustness

Selection of robust solutions for individual and collective stakeholders



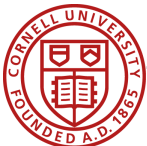
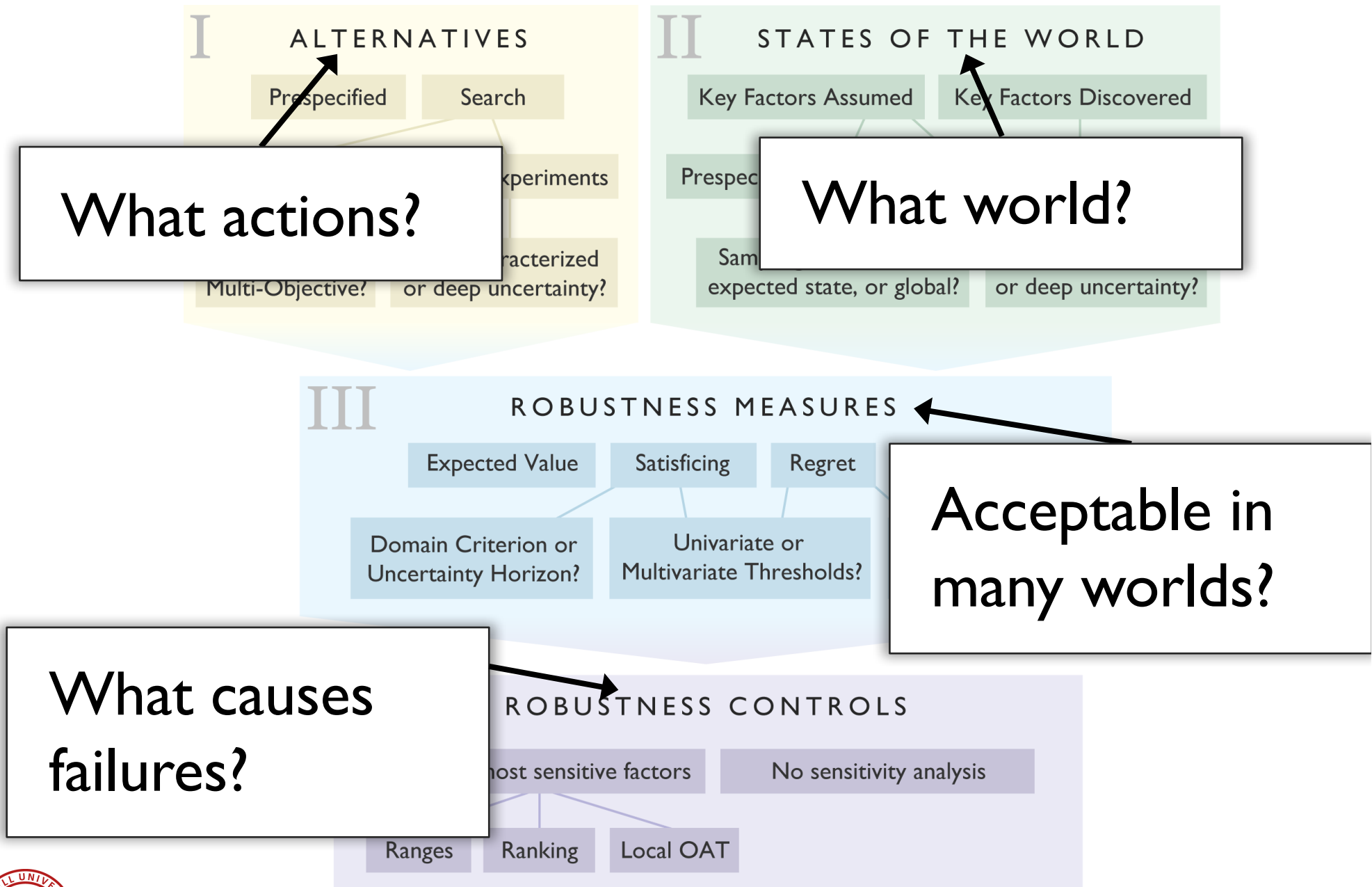
What do robustness analyses have in common?

Evaluate **alternatives** in multiple
states of the world...

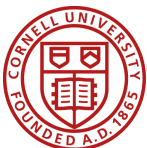
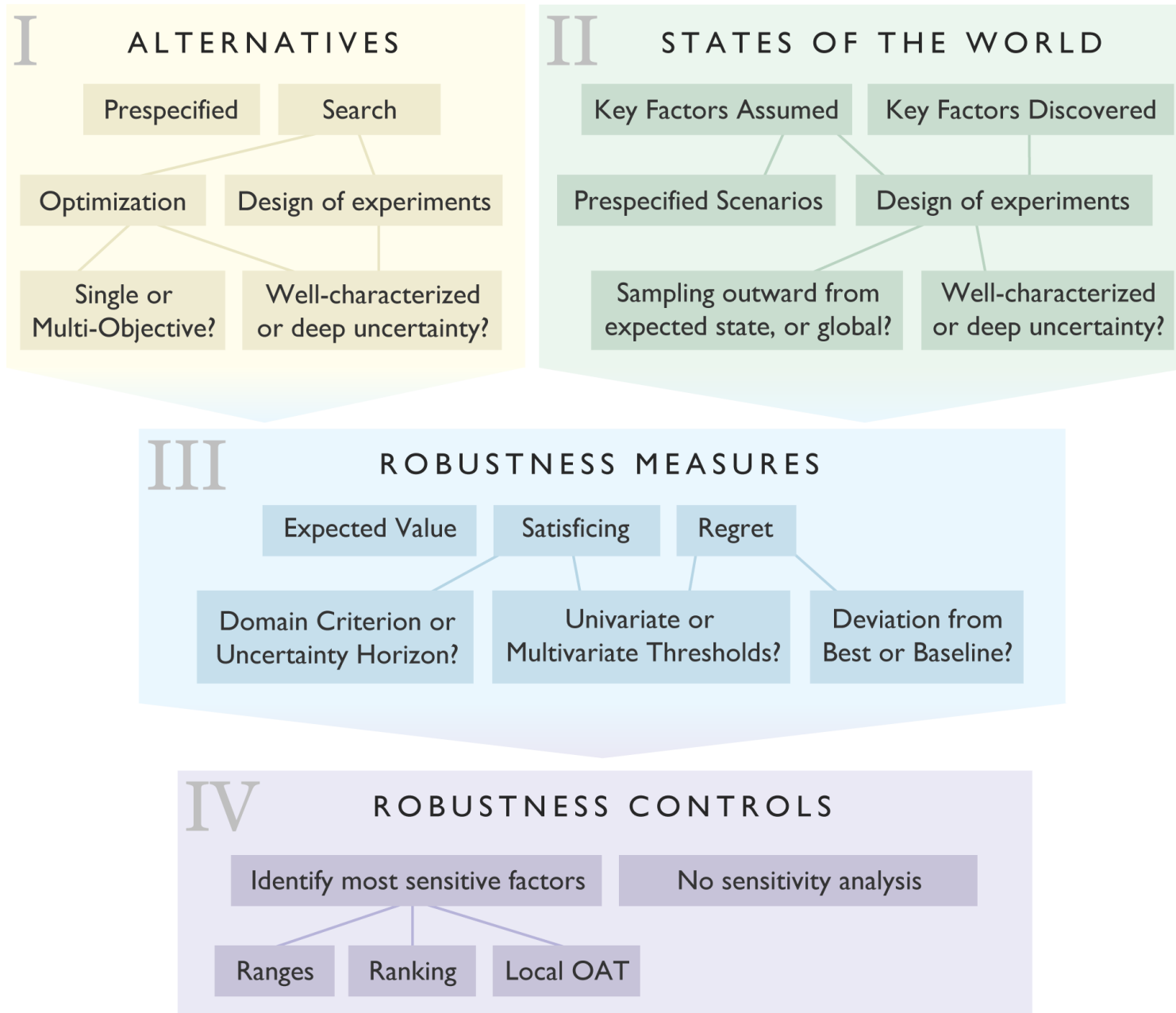
Quantify **robustness measures** and
determine **sensitive uncertainties**



Taxonomy of Robustness Frameworks



Taxonomy of Robustness Frameworks



Taxonomy of Robustness Frameworks

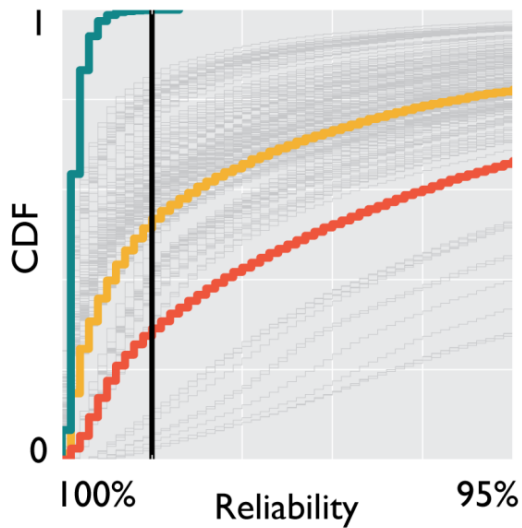


Discovering solutions through search **improves robustness** relative to prespecified alternatives

I Performance CDFs over Uncertain States of the World

Multi-objective performance for Durham

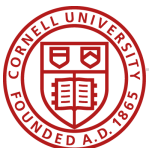
- Prespecified Solution
- Robust Solution from Search
- Robust Solution with Reduced Demand Growth
- Pareto-approximate set (Search)
- Stakeholder Requirement



Degrading Performance →

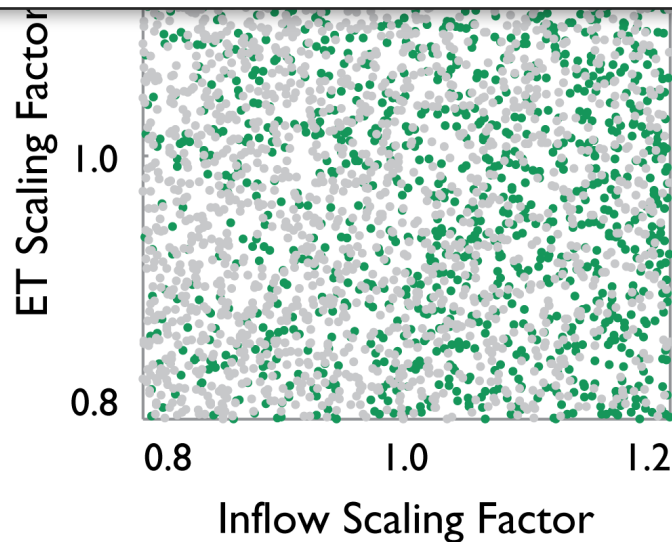


Taxonomy of Robustness Frameworks



An *a priori* focus on climate/hydrologic factors may fail to capture system vulnerabilities

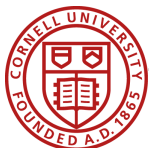
Rate of demand growth controls success more so than hydrology



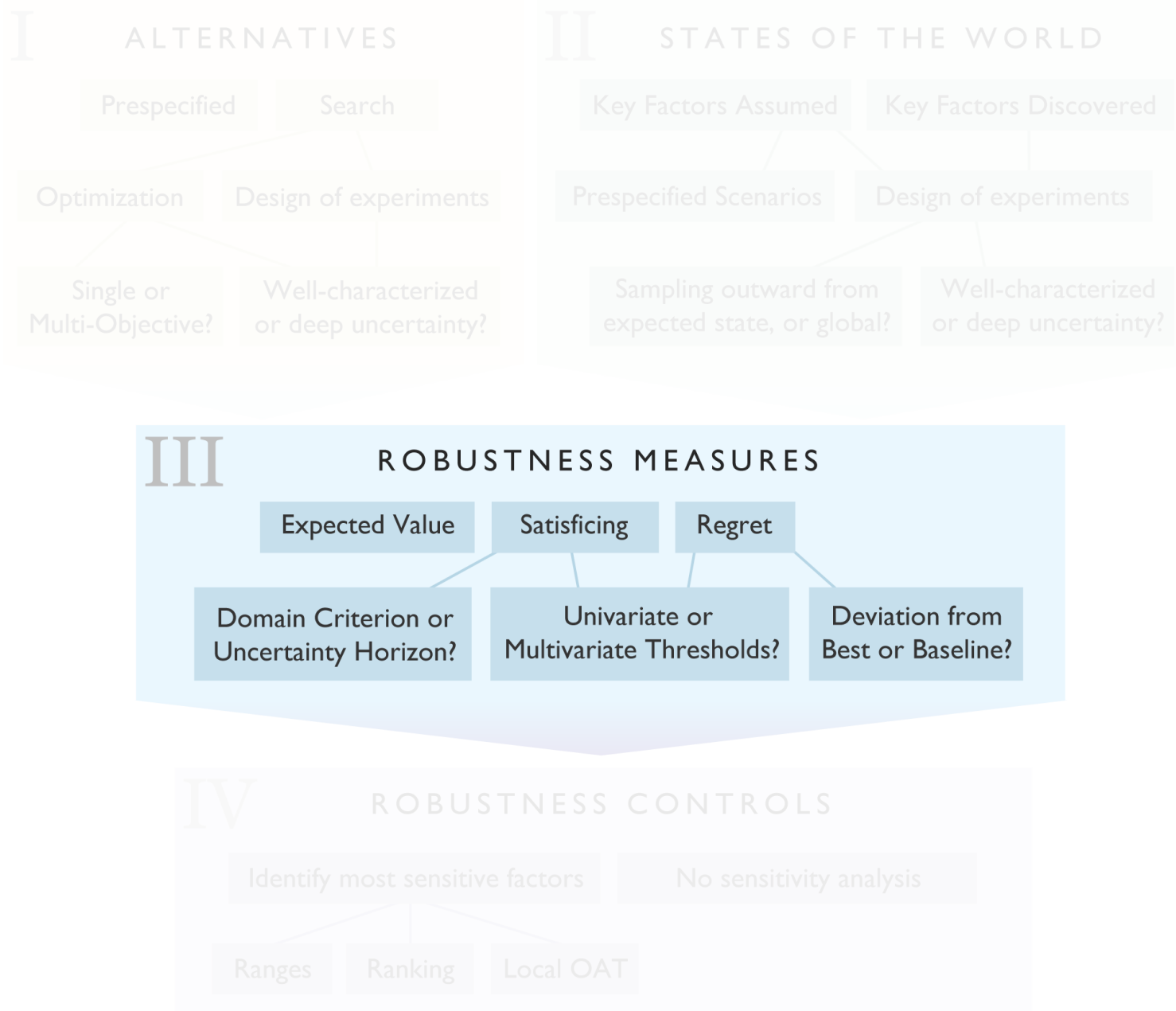
ents?

Cost < 5%

- Succeeds
- Fails



Taxonomy of Robustness Frameworks



Domain Criterion

$$= \frac{n_1}{n_1 + n_2}$$

Demand	⊗	⊗	⊙	⊙
	⊗	⊗	⊙	⊙
	⊗	⊙	⊙	⊙
	⊙	⊙	⊙	⊙
	Rainfall			

Number of failures

Distance

References: Lempert and Collins (2007), Schneller and Sphicas (1983), Hipel and Ben-Haim (1999)



How to measure robustness?

Which solutions would each measure choose from our Pareto front?

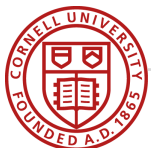
Definitions adapted from Lempert and Collins (2007)



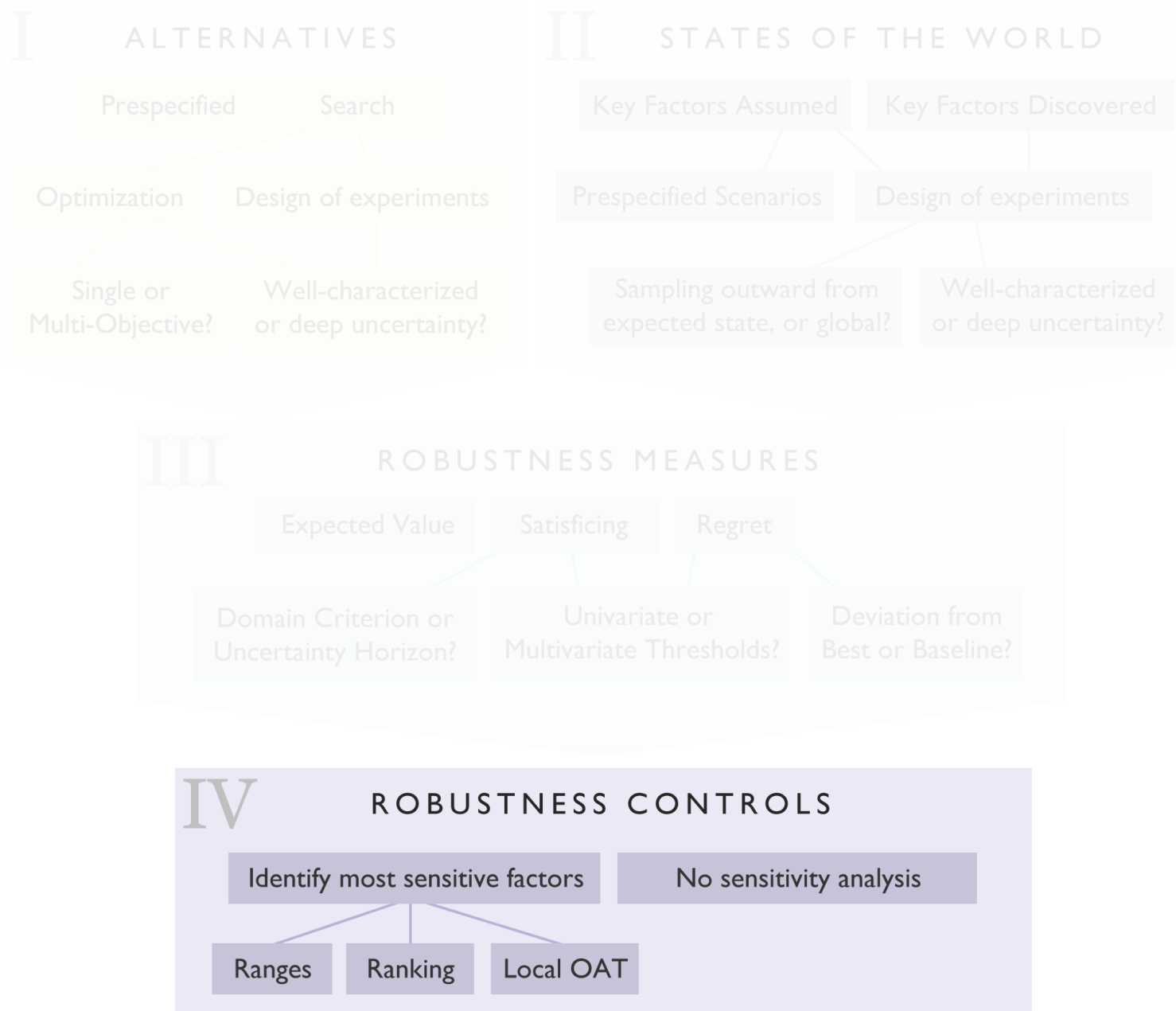
Recall requirements:

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

Only the multivariate satisficing measure (**SI**) meets these



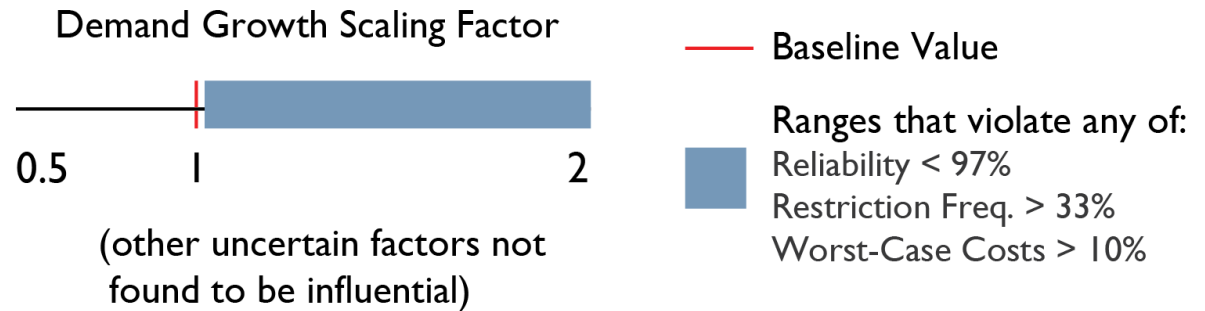
Taxonomy of Robustness Frameworks



Ranges (PRIM) or ranking (Sobol)?

The methods
complement
each other, not
exclusive

Factor Mapping (Patient Rule Induction Method)



Key Points

- (1) Pre-specified decision alternatives can suffer from a status quo bias, ignore full tradeoff context, and may fail to meet performance requirements (e.g., high reliability)
- (2) Robustness-based decision frameworks can be classified according to several interchangeable ideas
- (3) We need to better understand how methodological choices impact the selection of a “robust” solution, including the quantification of robustness and sensitivity analysis approaches



Questions on Section I?

- (1) Pre-specified decision alternatives can suffer from a status quo bias, ignore full tradeoff context, and may fail to meet performance requirements (e.g., high reliability)
- (2) Robustness-based decision frameworks can be classified according to several interchangeable ideas
- (3) We need to better understand how methodological choices impact the selection of a “robust” solution, including the quantification of robustness and sensitivity analysis approaches

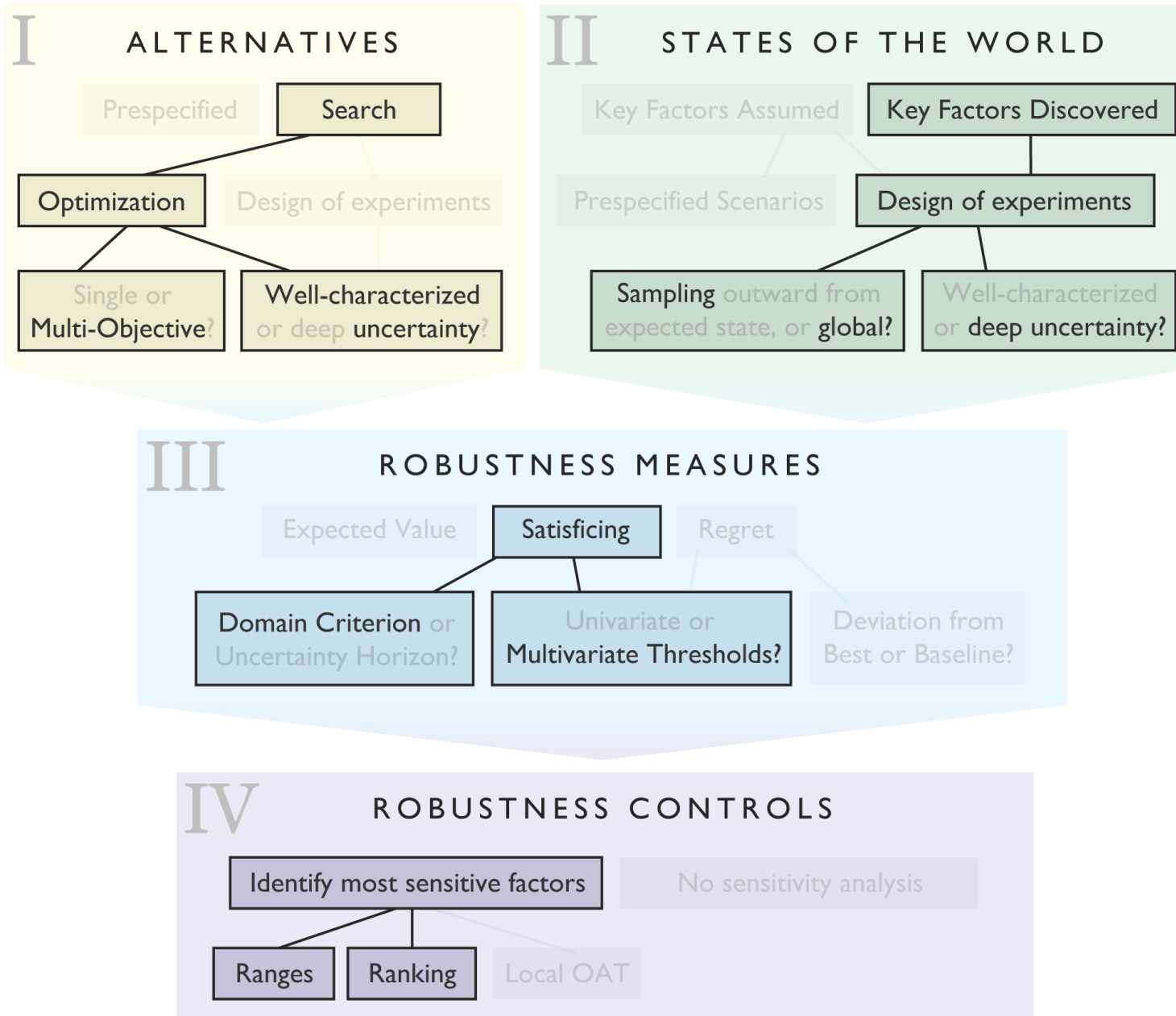


Section II: Language, Other Applications, & Tools

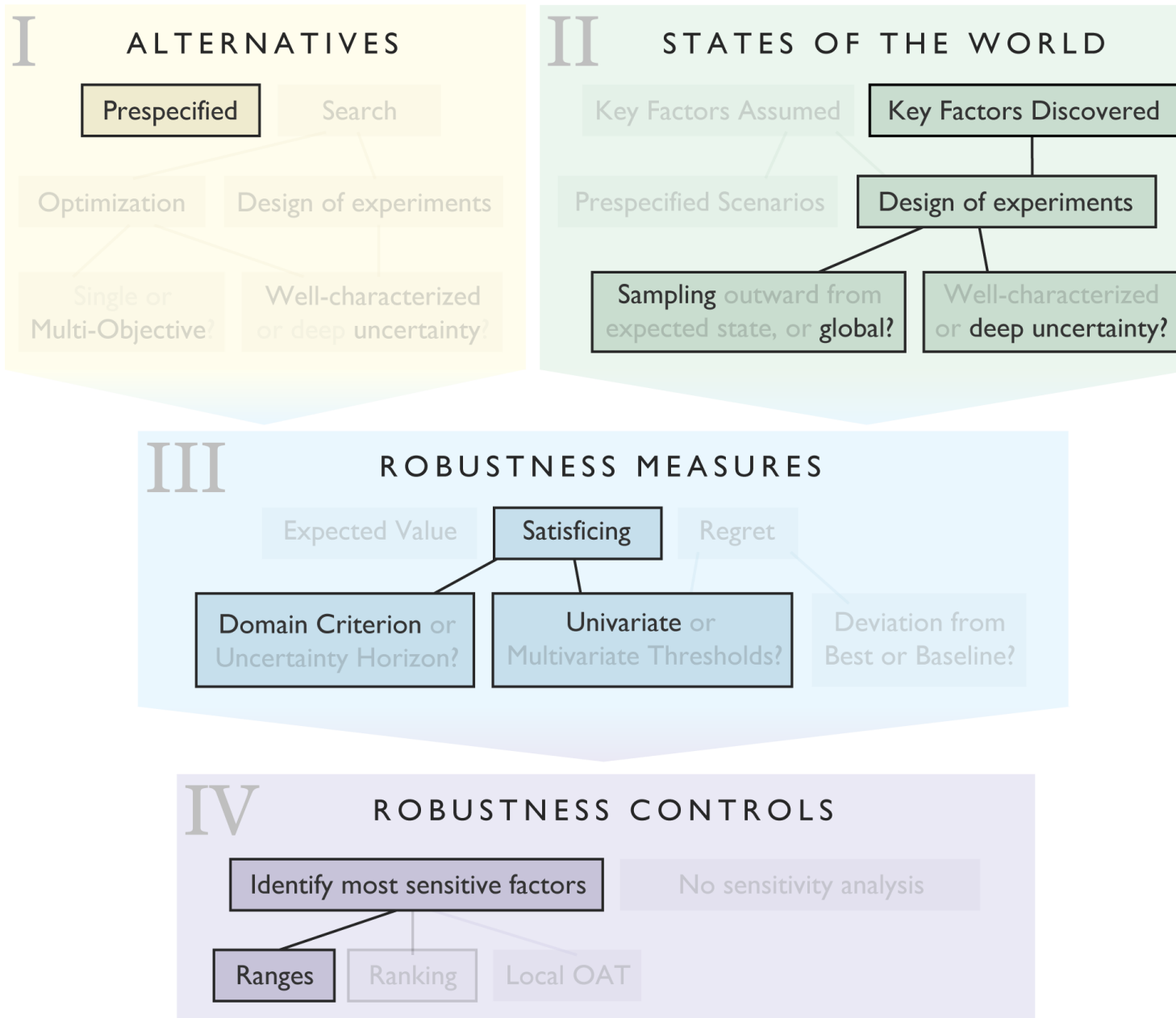
- ① What do robustness-based decision frameworks have in common?
- ② How do methodological choices impact decision recommendations?
- ③ How can we expand the value & impacts of our advances?



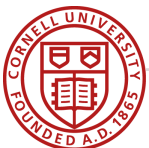
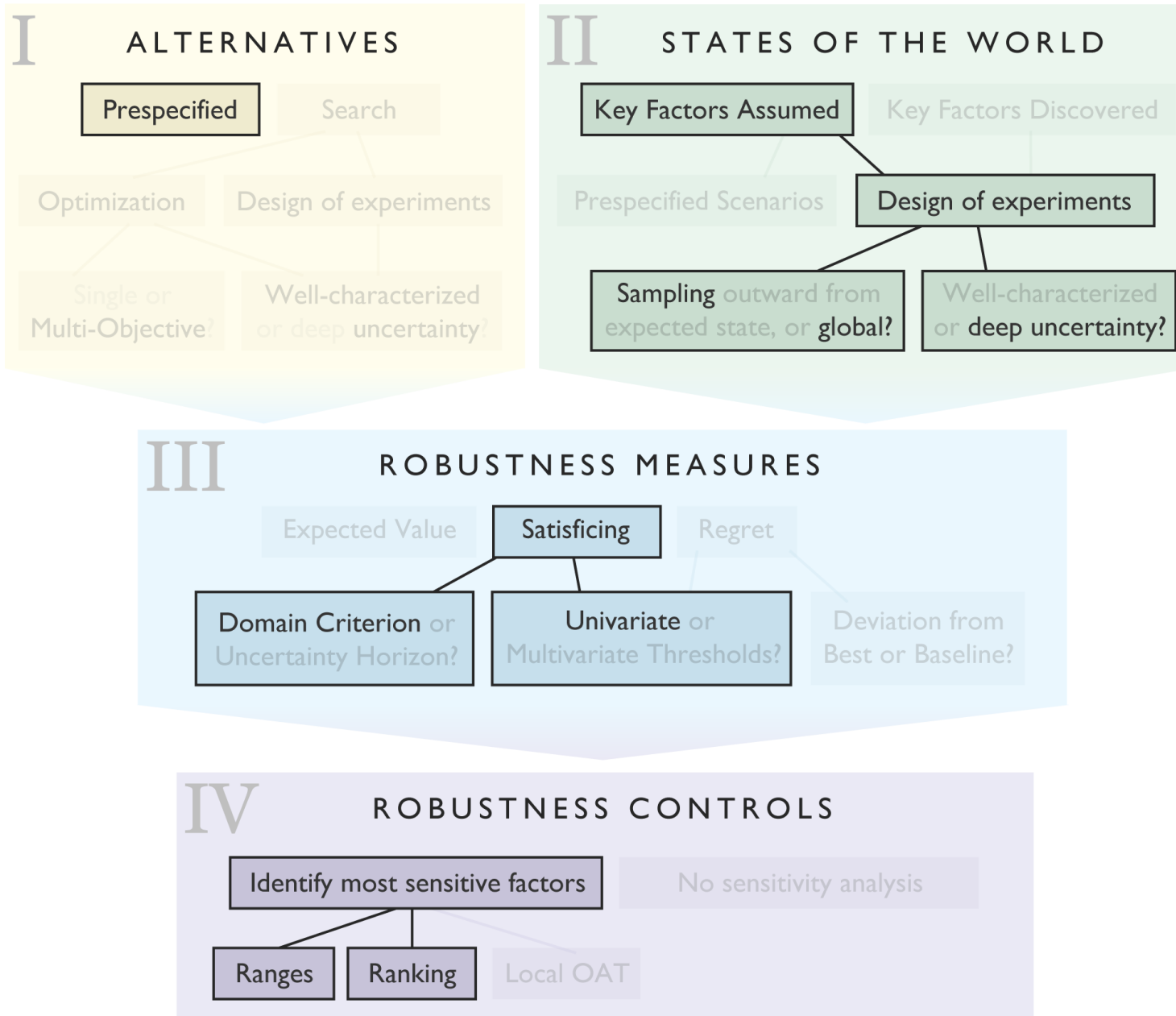
Many-Objective Robust Decision Making



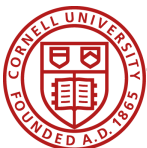
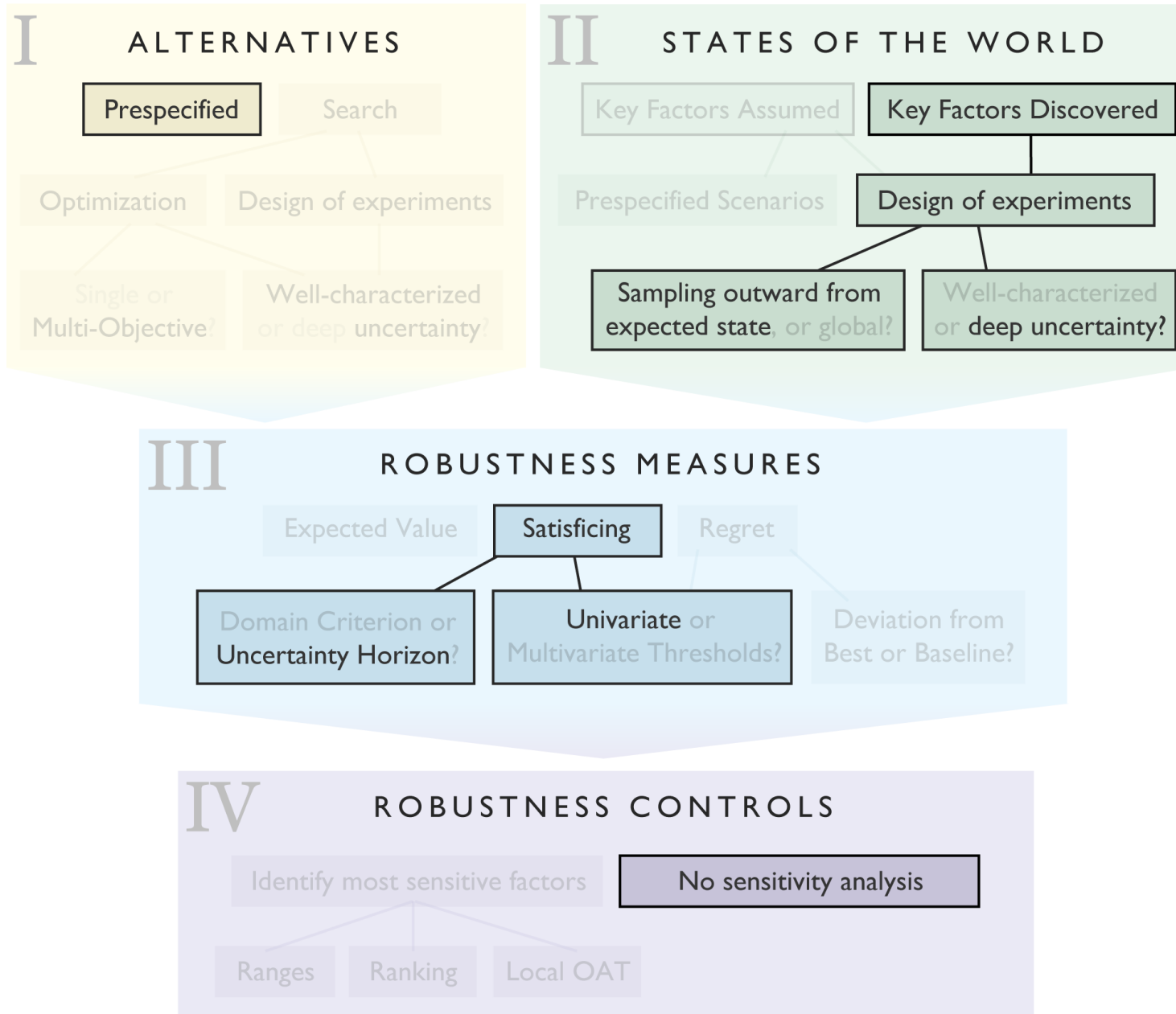
Robust Decision Making



Decision Scaling



Information-Gap

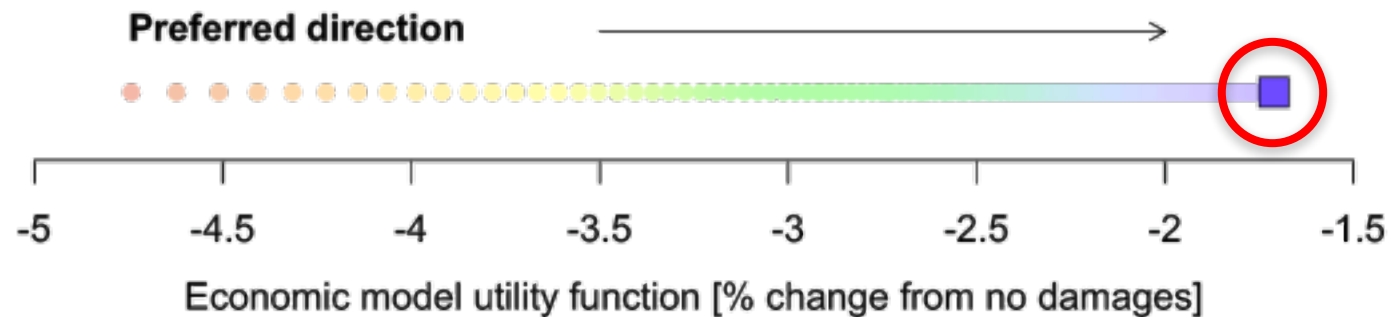


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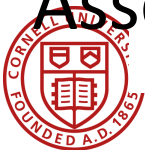
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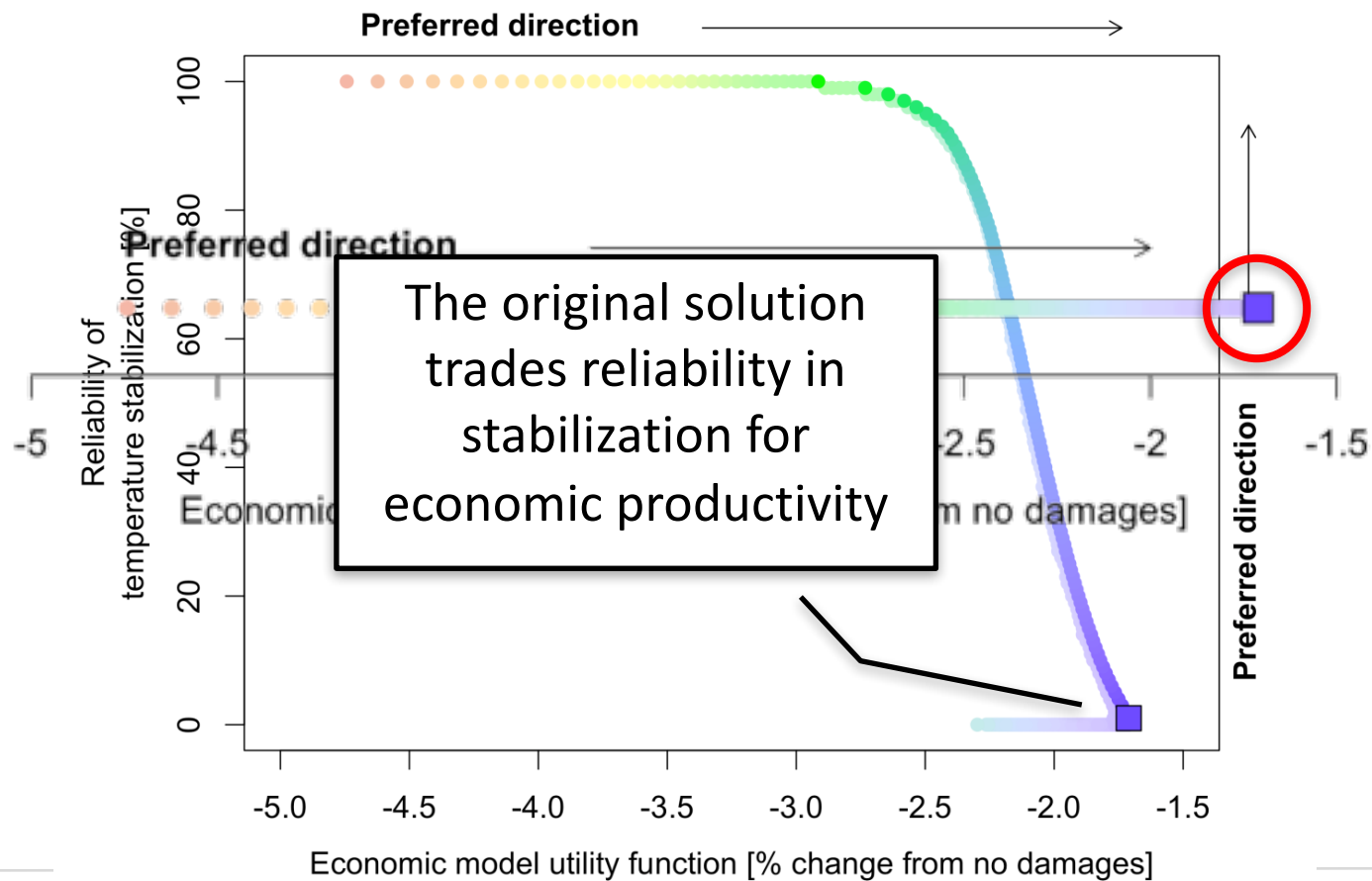
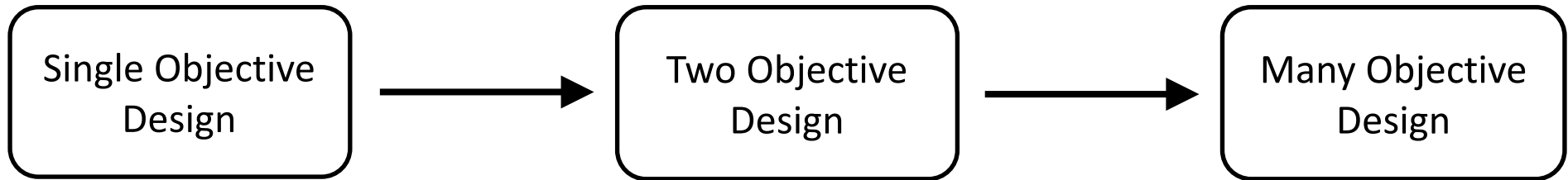
② How do methodological choices impact decision recommendations?



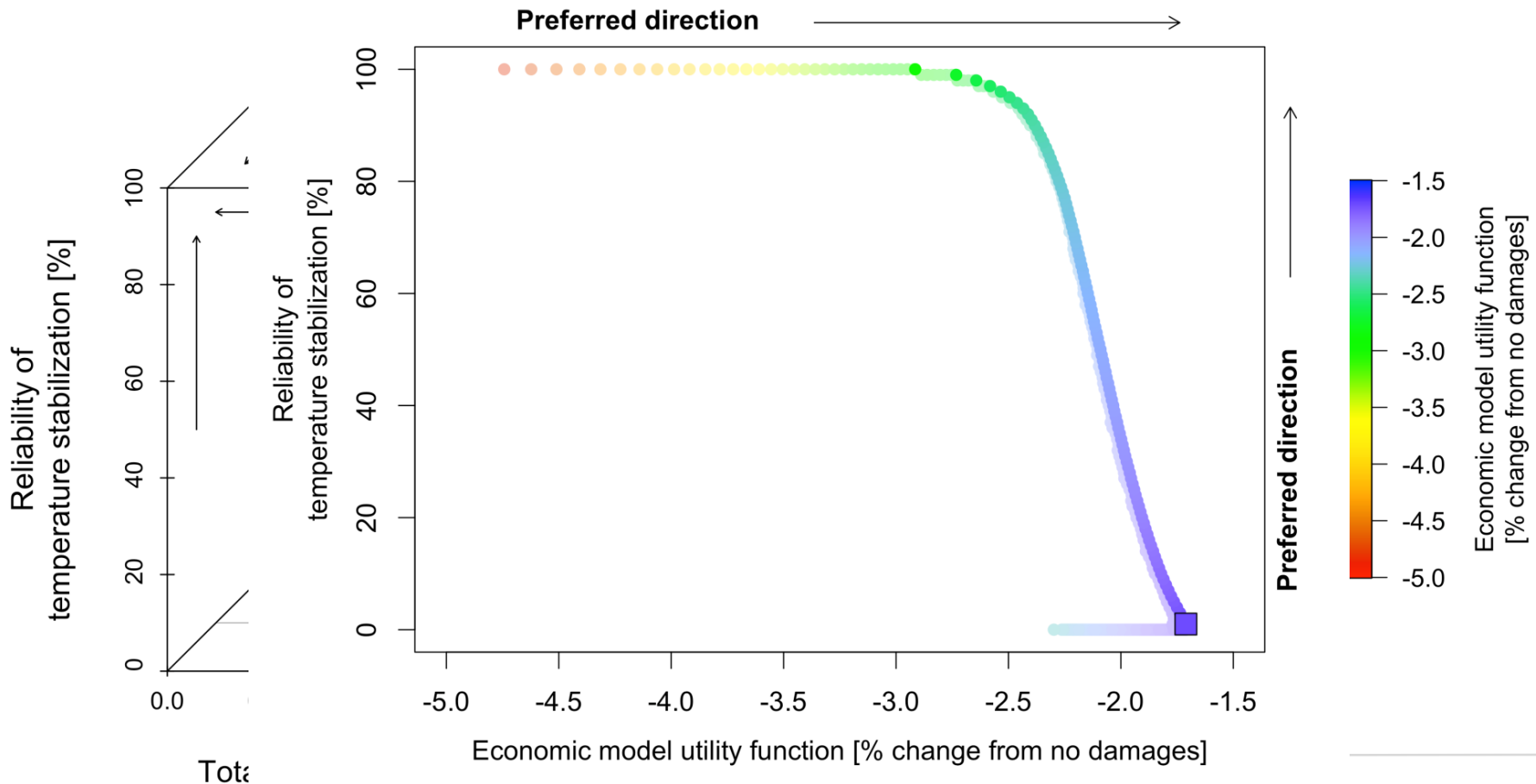
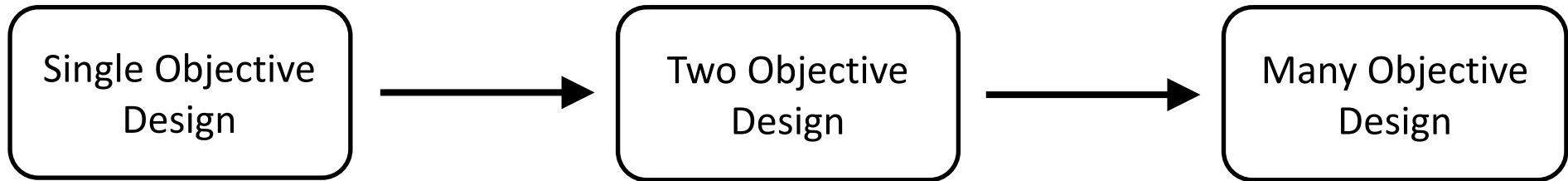
Mitigation vs Adaption Tradeoffs: DICE Integrated Assessment Model



② How do methodological choices impact decision recommendations?



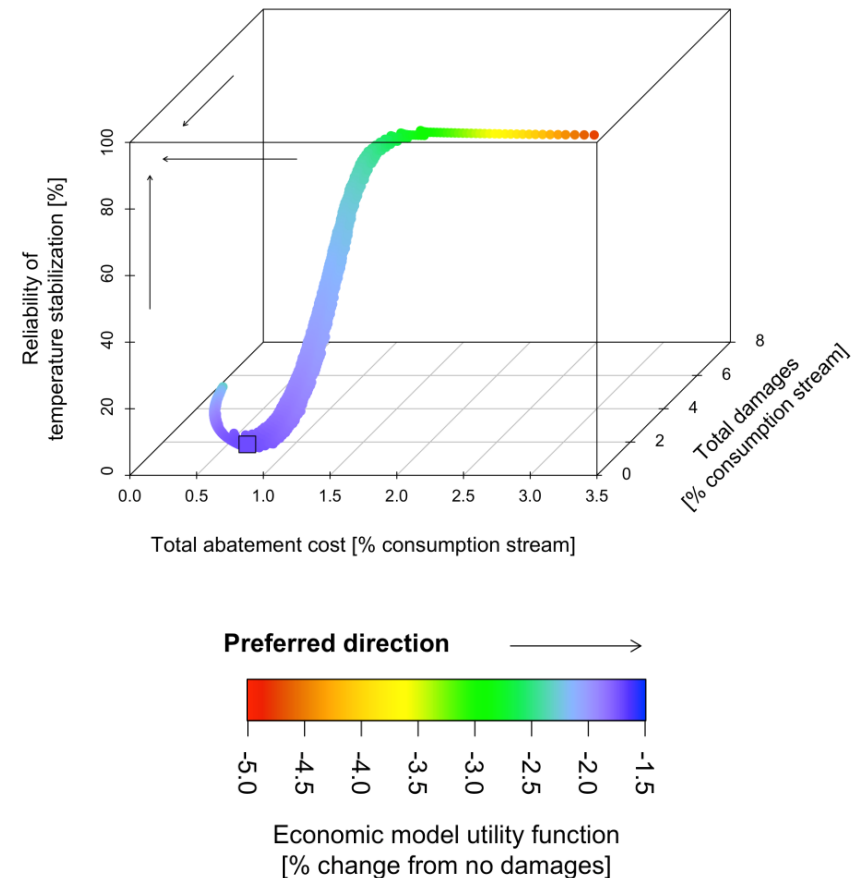
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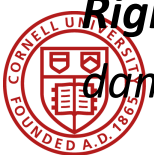
② How do methodological choices impact decision recommendations?

Climate risk management benefits from explicit representation of societal trade-offs

- Methods:** Sample uncertain distribution of climate sensitivity and quantify key trade-offs.
- Findings:** Reliability of geophysical stabilization not well accounted for with traditional utility representation of preference.
- Relevance:** **Decision analysts can use the trade-offs to better inform the negotiated policies and their consequences.**
- Linkages:** Insights feed into (multi-objective) robust decision making framework.

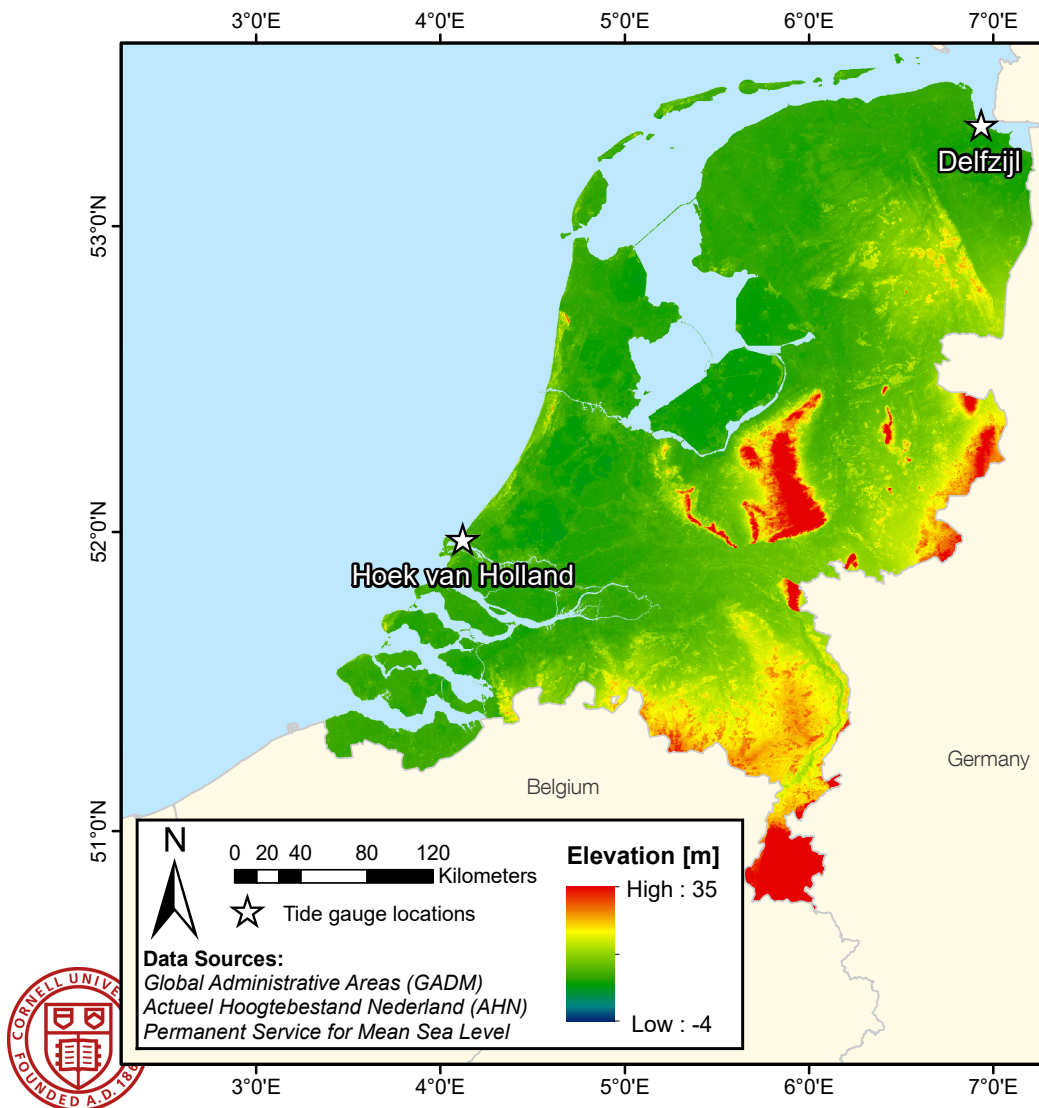


Right: Explicit trade-offs between reliable stabilization, abatement costs, and damages given uncertainty in climate sensitivity.



② How do methodological choices impact decision recommendations?

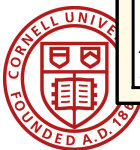
Classic Climate Risk Adaptation Example (van Dantzig, 1956)



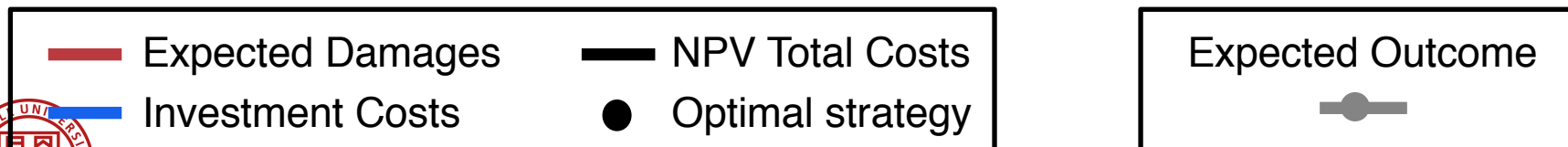
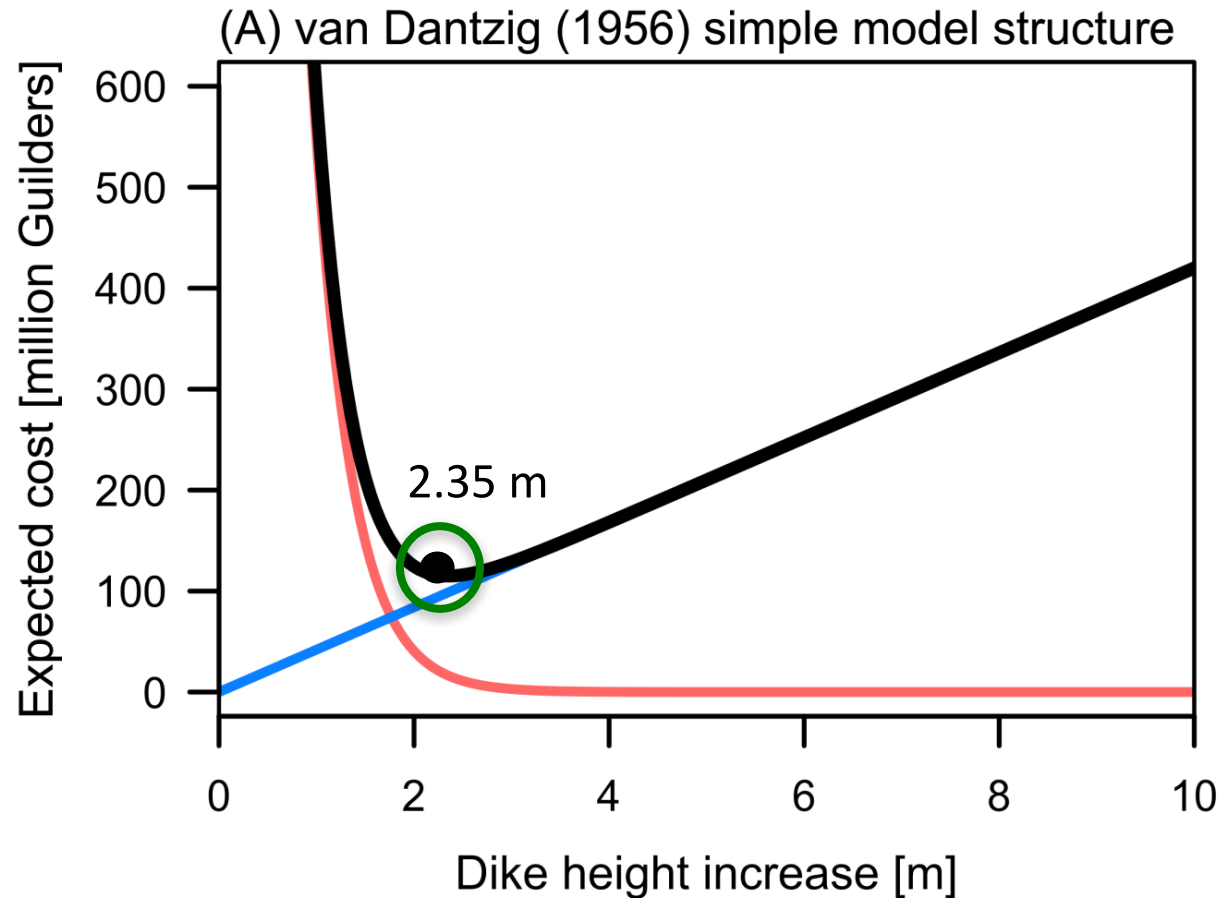
What dike height?
Global sea level rise?
Storm surge?

Economics-Engineering-
Earth Science: A
Synthesis Problem

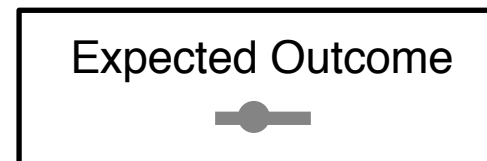
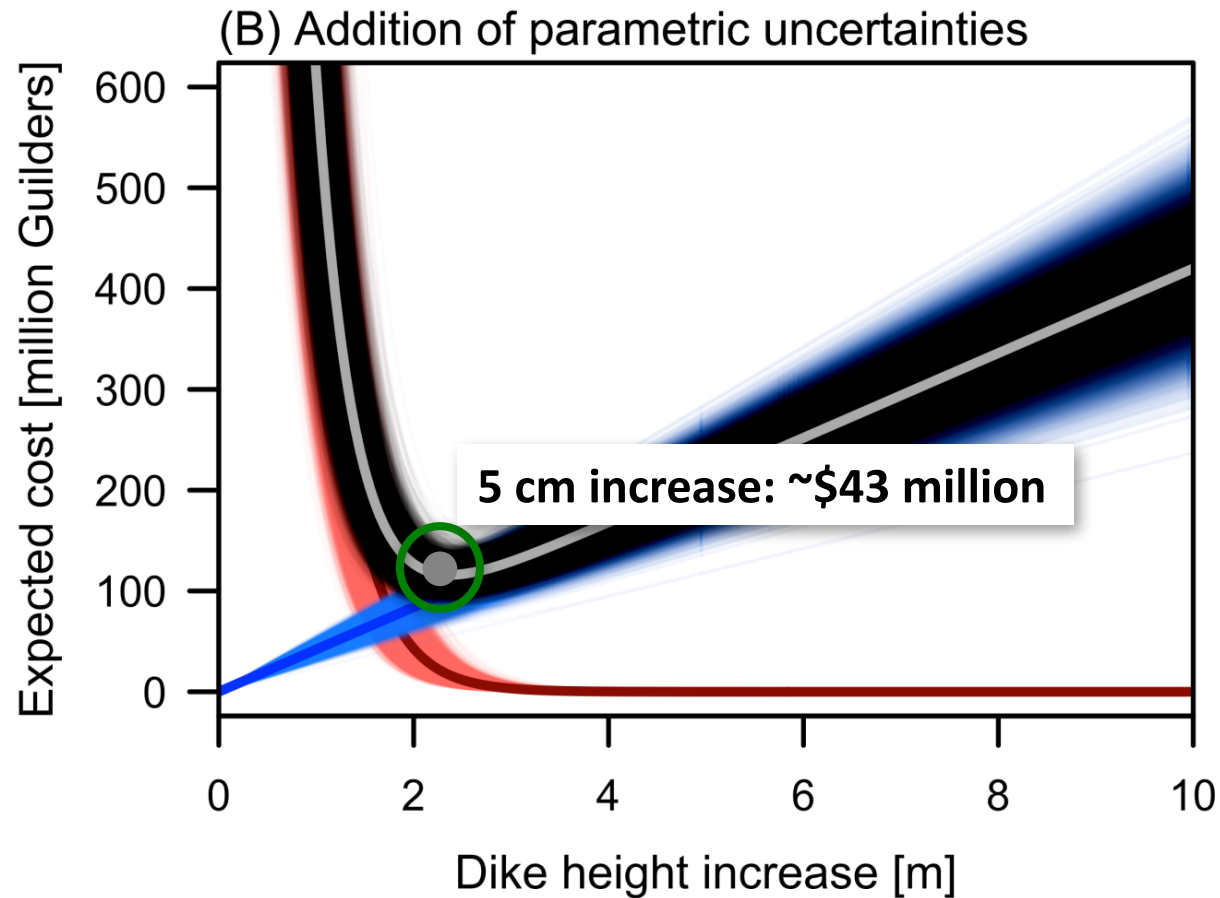
Left: Map showing land surface elevation above mean sea level, the Netherlands.



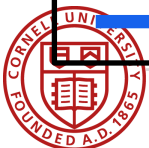
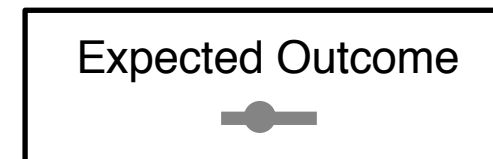
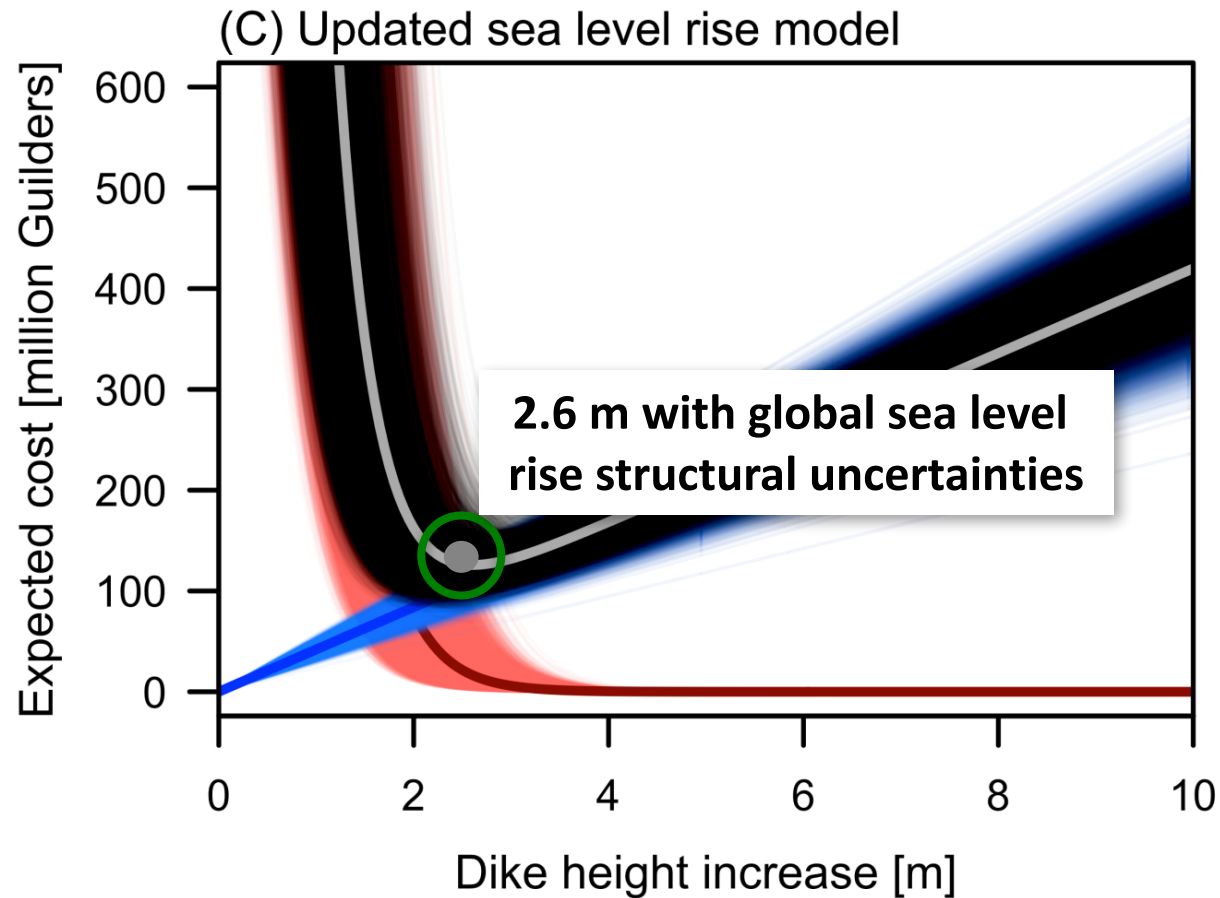
② How do methodological choices impact decision recommendations?



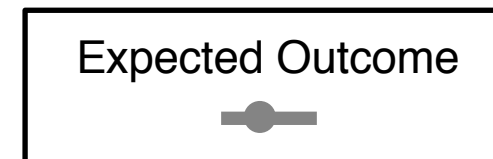
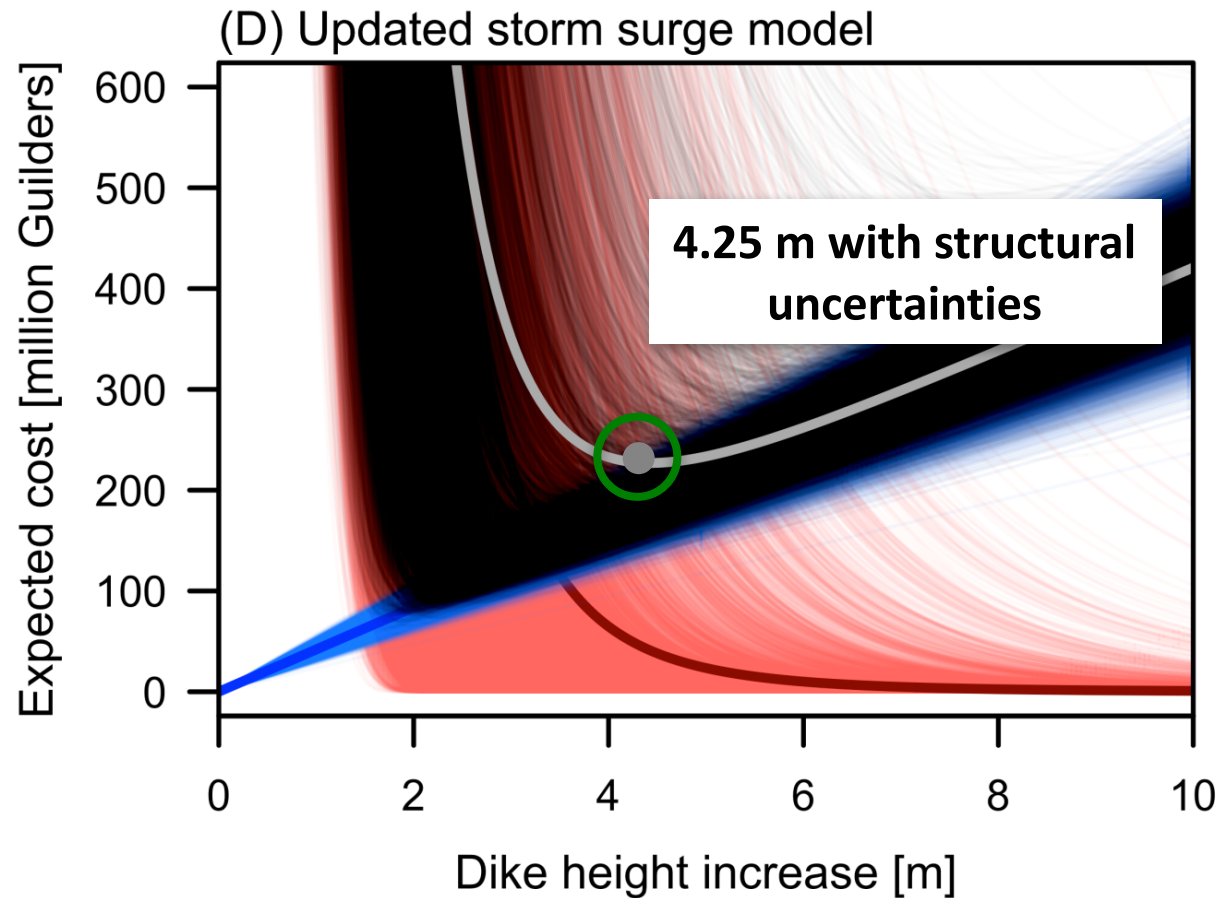
② How do methodological choices impact decision recommendations?



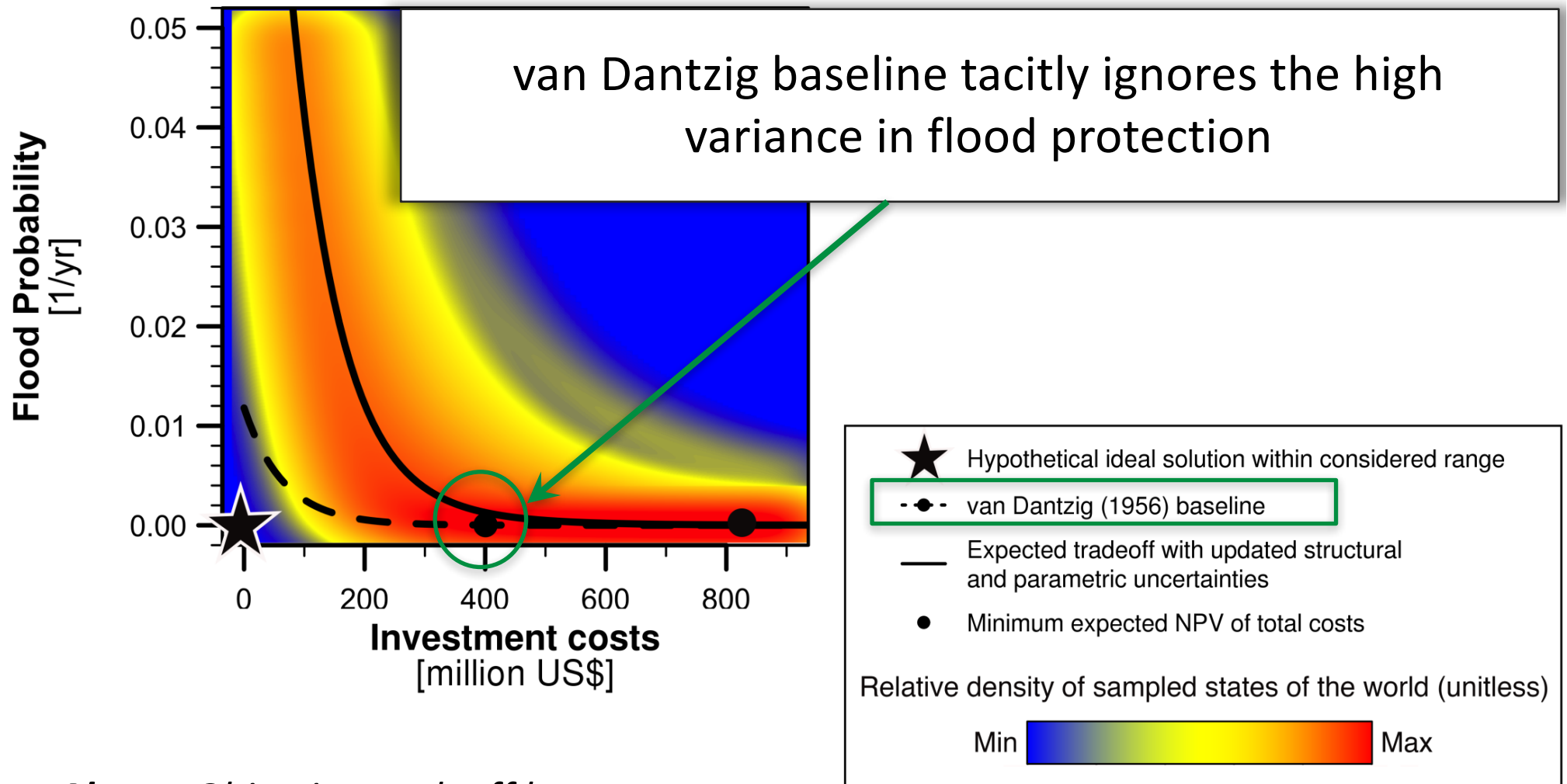
② How do methodological choices impact decision recommendations?



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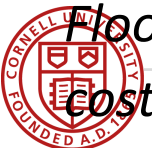
② How do methodological choices impact decision recommendations?



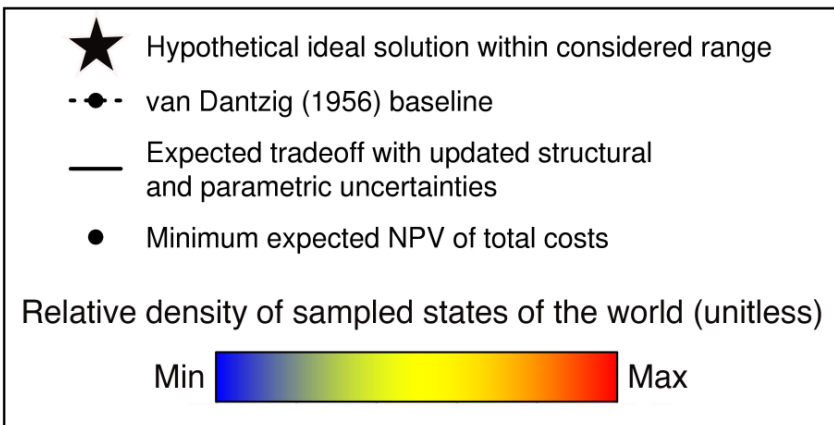
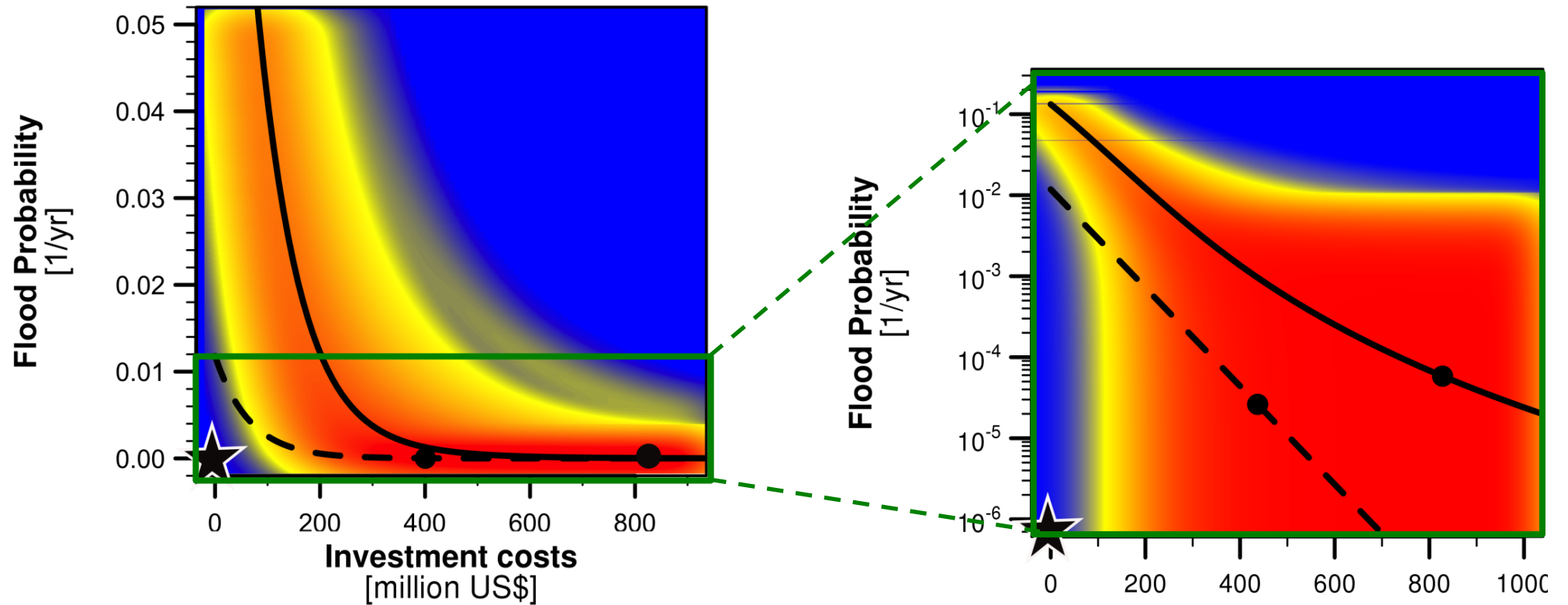
Above: Objective tradeoff between

Flood Probability and Investment

costs.



② How do methodological choices impact decision recommendations?



Tail-area behavior yields a severe variance in the reliability of a given investment

Section II: Language, Other Applications, & Tools

- ① What do robustness-based decision frameworks have in common?
- ② How do methodological choices impact decision recommendations?
- ③ How can we expand the value & impacts of our advances?



③ How can we exploit the SCRiM SRN to expand the value & impacts of our advances?



OpenMORDM

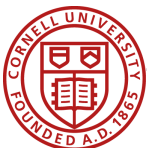
Multiobjective Robust Decision Making in R

Developed by Penn State (David Hadka, Klaus Keller) and Cornell (Jon Herman, Patrick Reed)

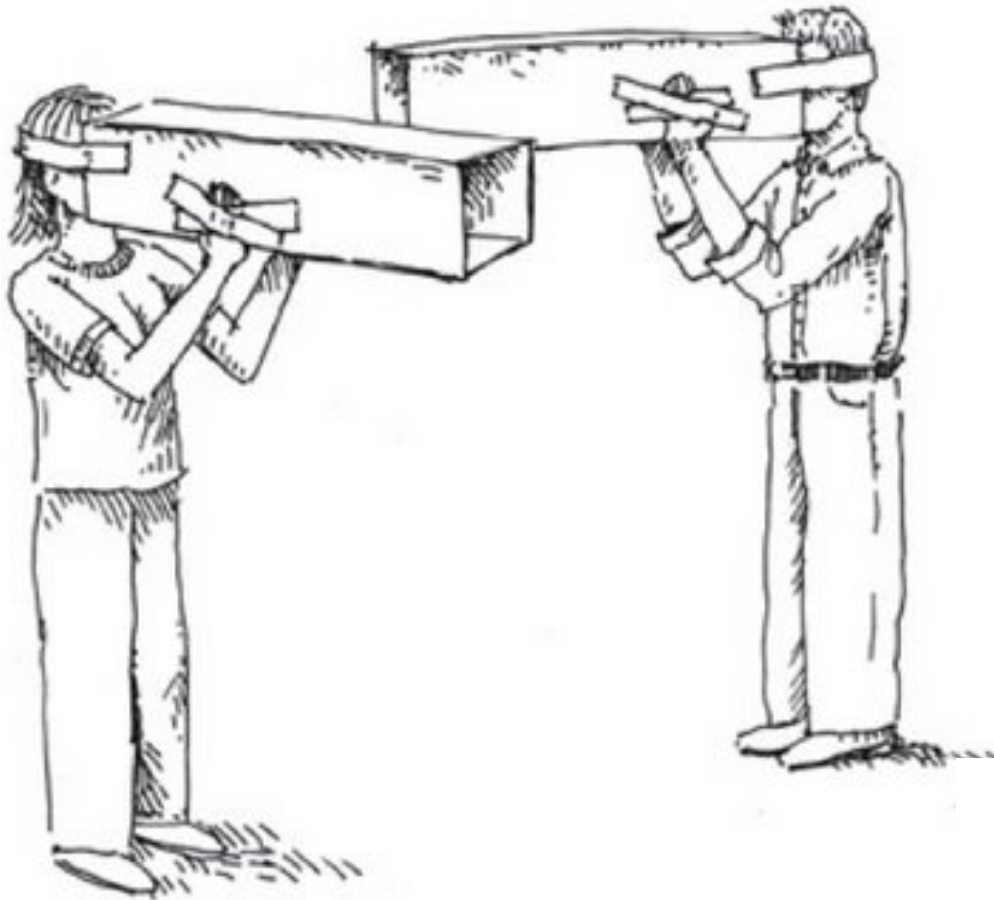
What is OpenMORDM?

R library for Multiobjective Robust Decision Making (MORDM)
Exploring systems with deep uncertainties, identify vulnerabilities,
understand tradeoffs between competing goals

Free and Open Source - <http://github.com/OpenMORDM>



3 How can we exploit the SCRiM SRN to expand the value & impacts of our advances?



Which actions are robust? How to decide?

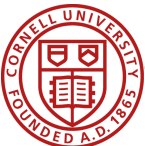
Sample deeply uncertain states of the world (*climate sensitivity, sea-level rise, etc.*)



OpenMORDM

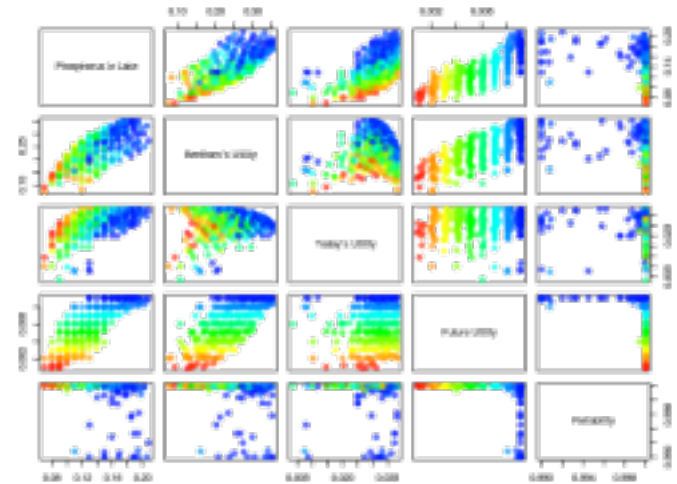
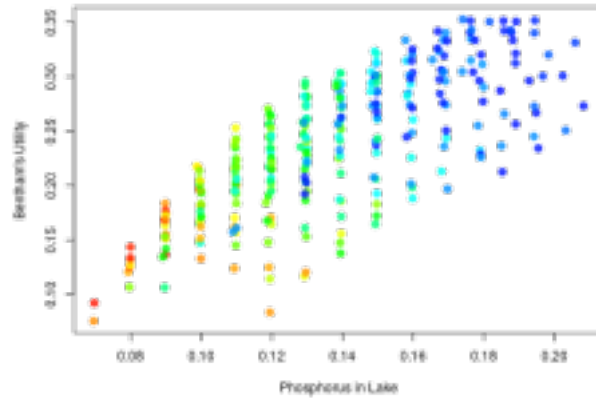
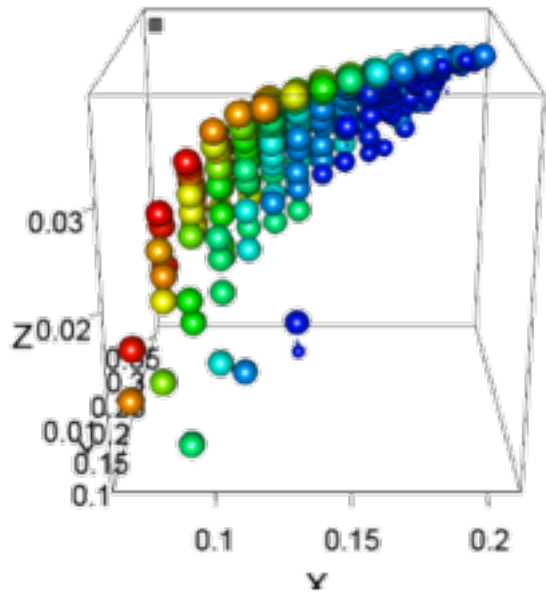
Multiobjective Robust Decision Making in R

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

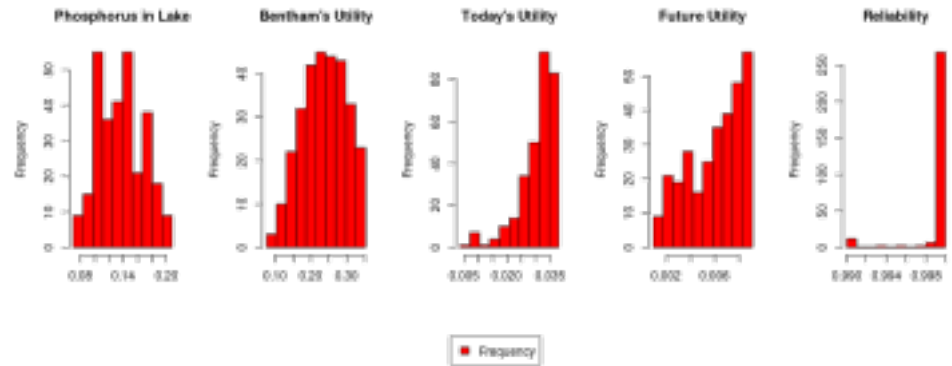
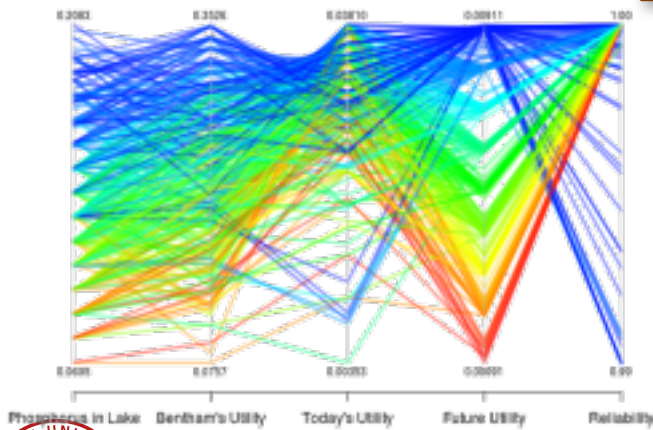


3

How can we exploit the SCRiM SRN to expand the value & impacts of our advances?



Interactive, high-dimensional visualization of datasets



3

How can we exploit the SCRiM SRN to expand the value & impacts of our advances?

The screenshot shows the homepage of the MOEA Framework. At the top is a navigation bar with links for Home, Features, Examples, Downloads, Documentation, and Support. Below this is the MOEA Framework logo, which consists of three black circles connected by lines, followed by the text "MOEA Framework" and "A Free and Open Source Java Framework for Multiobjective Optimization". The main content area is divided into two columns. The left column contains a section titled "A Framework for Innovation" with a paragraph of text describing the framework's capabilities and a list of "Key Features" including fast implementations, extensibility, parallelization, and professional support. Below this is a "Gallery" section with several small thumbnail images. The right column contains a "Quick Links" section with links for Demo Application, Compiled Binaries, Source Code, and Documentation. Below these links are social media links for GitHub, showing 27 forks and 46 stars, and a "License" section indicating it is licensed under the GNU Lesser General Public License. At the bottom of the page, there is a search bar with the text "Find in page" and options for "Highlight All" and "Match Case".



<http://moeaframework.org/>

3

How can we exploit the SCRiM SRN to expand the value & impacts of our advances?

BorgMOEA

Home Get It! Publications

Welcome.

The Borg Multiobjective Evolutionary Algorithm (MOEA) is a state-of-the-art optimization algorithm developed by David Hadka and Patrick Reed at the Pennsylvania State University. Borg is freely available for academic and non-commercial use. Use this site to learn more about the Borg MOEA and request access to its source code.

Many-Objective


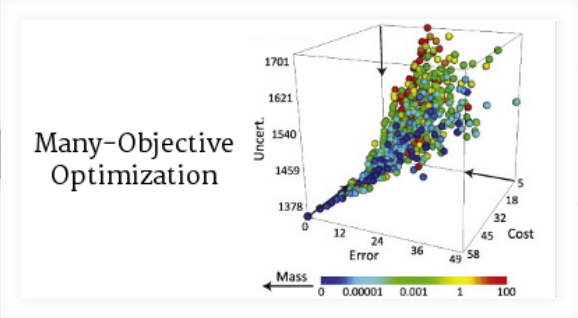
Borg efficiently captures the tradeoffs between many conflicting performance objectives, providing decision makers with detailed insight into their problem characteristics.

Adaptive Search

Borg uses an ensemble of search operators, auto-adapting their use at runtime to tailor itself to your optimization problem.

High-Performance

Written in efficient, high-performance ANSI C, the Borg MOEA wastes little time when solving your problem. Runs on Unix, Linux, Windows, and Mac.





<http://BorgMOEA.org/>

Thank you & any questions?

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H.B. Zeff, PhD candidate, University of North Carolina Chapel Hill

Water, Sustainability & Climate Grant: 2014-67003-22076



United States Department of Agriculture
National Institute of Food and Agriculture

