

# Decision Support for Wicked Problems

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# How to Use Quantitative Information to Make Good Policy Choices?

In many policy areas, quantitative information is often indispensable to making good policy choices





**Climate change** 



Health





Defense

Economic policy

It may seem obvious that quantitative analysis can best inform policy by making predictions of the future

But predictions -- a core principle of the scientific method -- often complicate the use of quantitative information when:

- Uncertainties are deep
- People disagree

Fortunately, there is another way,

which creates new and interesting analytic challenges

### **Traditional Policy Analysis Begins with a Consensus Understanding of the Future**



Predi



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## Many Policy Challenges Usefully Understood as Complex and "Wicked" Problems

Characteristics of "wicked" problems include:

- Not well bounded,
- Framed differently by various groups and individuals,
- · Large to existential scientific uncertainties,
- · Non-linear dynamics, and
- · Not well understood until after the formulation of a solution



Complex and complicated systems have different implications for understanding and acting:

- Behavior of complex systems can be understood, but difficult to predict
- Complex systems often best managed by probe and respond, rather than analyze and act

Cynefin framework

# "Predict then Act" Can Break Down When Uncertainties are Deep



Under conditions of deep uncertainty:

- Uncertainties are often underestimated
- Competing analyses can contribute to gridlock
- Misplaced concreteness can blind decision makers to surprise

#### <u>Deep uncertainty</u> occurs

when the parties to a decision do not know or do not agree on the likelihood of alternative futures or how actions are related to consequences

### To Inform Decisions Under Deep Uncertainty, Invert the Order of Traditional Analysis



Kalra, N., S. Hallegatte, R. Lempert, C. Brown, A. Fozzard, S. Gill and A. Shah (2014). <u>Agreeing on</u> <u>Robust Decisions: A New Process for Decision Making Under Deep Uncertainty</u>. WPS-6906, World

# **Robust Decision Making (RDM) Provides Such an "Agree on Decision" Approach**

- Robust decision-making (RDM) is an iterative, multi-scenario, multi-objective decision analytic framework that aims to:
  - Help identify potential robust strategies,
  - Characterize the vulnerabilities of such strategies, and
  - Display the tradeoffs among them.
- RDM rests on a simple concept
  - Rather than use computer models and data as predictive tools,
  - RDM runs model thousands to millions of times to:
    - Stress test proposed policies against a wide range of futures, and
    - Uses the results to help decision identify policy-relevant scenarios and strategies robust across those scenarios.

Robust Decision Making (RDM) is one common approach for decision making under deep uncertainty (DMDU)

Vincent A. W. J. Marchau Warren E. Walker Pieter J. T. M. Bloemen Steven W. Popper *Editors* 

#### Decision Making under Deep Uncertainty From Theory to Practice

DMDU CCE OPEN Dringer

# **Outline**

- Robust Decision Making (RDM)
- Stress Tests
- Supporting Deliberation
- Observations

# "Scenario Discovery" Algorithms Illuminate Vulnerabilities of Proposed Plans

Sacramento Council of Governments (SACOG) aims to improve mobility and equity while meeting stringent greenhouse reduction requirements

SACOG's 2016 Regional Transportation Plan Sustainable Community Strategy



- SACOG's 2016 plan aims to meet these goals by 2035 with:
  - Targeted transportation investments and
  - More compact growth
  - Used RDM to stress test this plan over many futures
    - Used simple model fit to agencies travel demand model
    - Results helped to identify key vulnerabilities and potential responses

Lempert, R., J. Syme, G. Mazur, D. Knopman, G. Ballard-Rosa, K. Lizon and I. Edochie (2020) Meeting Climate, Mobility, and Equity Goals in Transportation Planning Under Wide-Ranging Scenarios: A Demonstration of Robust Decision Making. Journal of the American Planning Association

# Under What Conditions Would SACOG Meet Its Mobility, Equity, and Climate Goals for 2035?

Run model over many plausible futures



SACOG meets goals when mobility is high and emissions are low

What combination of factors best distinguish the green and grey cases?

1. Indicate policy-relevant cases in database of simulation results



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- 2. Statistical analysis finds low-dimensional clusters with high density of these cases



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driving forces of interest to decision makers

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3. Clusters represent scenarios and driving forces of interest to decision makers Approach provides measures of merit for scenario quality

Density:

 How many cases inside the scenario are *policy-relevant*? (e.g. 75%)

Coverage:

 How many of all the *policy-relevant* cases do the scenarios include? (e.g. 82%)

Interpretability:

 Is the number of scenarios and driving forces sufficiently small to understand? (e.g. 1 scenario with two driving forces)

On scenario discovery: Bryant & Lempert (2010). Thinking inside the box: a participatory, computer-assisted approach to scenario discovery <u>Technological Forecasting and Social Change</u>

Kwakkel (2017) "The Exploratory Modeling Workbench: An open-source toolkit for exploratory modeling, scenario distrovery, and (multi-objective) robust decision making." <u>Environmental Modelling & Software</u>

# Under What Conditions Would SACOG Meet Its Mobility, Equity, and Climate Goals for 2035?

#### Run model over many plausible futures



Green = case meets all SACOG goals Grey = case misses some SACOG goals

#### "Scenario discovery" classification algorithms show key drivers of success



- Green bars show parameter variation ranges that best differentiate futures that meet and miss goals
- Variables without green bars are not a key driver/differentiator for meeting or missing goals

Scenario discovery generates scenarios that can be used in stakeholder engagements

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# Participatory Processes Can Prove Important to Effective Policy Action

Participatory process:

- Can enhance legitimacy and inclusion
- Combine scientific and local knowledge
- Connect estimates of consequences with participants' values



#### Deliberation with analysis is:

- A participatory learning process in which stakeholders deliberate over problem framings supported by analytic products responsive to those framings
- Useful when new problem framings emerge through stakeholder interactions with analytics and each other
- Useful for the "frame reflection' often needed to address wicked problems

# **Disagreements Over Values Often Become Disagreements Over Facts**

### Public understanding of science as function of "scientific intelligence"

There is "solid evidence" of recent global warming due "mostly" to human activity such as burning fossil fuels





Kahan, D.M., *Climate-Science Communcation and the Measurement Problem.* Advances in <u>Political Psychology</u>, 2015. 36

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Climate scientists believe that if the North Pole icecap melted as a result of human-caused global warming, global sea levels would rise



Kahan, D.M., *Climate-Science Communication and the Measurement Problem.* Advances in <u>Political Psycholog</u>y, 2015. 36

### How Can Analytics Engage Multiple Worldviews?

- Wicked problems engage people with different worldviews
  - A world view is a comprehensive conception of the world, comprising a correlated set of values, beliefs, and policy preferences that shapes how one understands, judges, and acts in the world
- Approaches for addressing wicked problems mostly envision:
  - Iterative learning processes
  - Frame reflection, in which participants are encouraged to see problem through each others' eyes
  - "Clumbsy" solutions, which contain elements of different worldviews
- Utopia-dystopia matrices help operationalize these ideas



# Simple Example Introduces Analytic Tools

#### Imagine a Hypothetical Town on Shore of a Pristine Lake

- Proposed economic development could:
  - Increase pollution into the lake
  - Enhance welfare for some current and new residents
  - Damage welfare for other current residents

- Deep uncertainties include:
  - Pollution tipping point for lake
  - Pollution intensity of development
  - Effectiveness of alternative policies



#### What strategy should the town pursue?

### RDM Suggests Compromise Strategies Across Worldviews

#### Identify three worldviews from the cultural theory of risk

|   | Hierarchists                                | Egalitarians                                       | Individualists                                     |  |  |
|---|---|--|--|--|--|
| Expectations<br>about lake<br>and economy | • Lake and economy are benign within bounds | • Lake is fragile, but<br>the economy is<br>benign | • Like is benign, but<br>the economy if<br>fragile |  |  |
|   | Lake and Economy                            | Lake & Traditional<br>Sector                       | Lake Economy                                       |  |  |
| Appropriate                               | • Favor regulation and                      | Distrusts training                                 | <ul> <li>Distrust regulation</li> </ul>            |  |  |
| actions town                              | training                                    | • Favors regulation                                | <ul> <li>Favor training</li> </ul>                 |  |  |
| might take                                | <ul> <li>Favor adaptive</li> </ul>          | • Distrust adaptive                                | <ul> <li>Favor adaptive</li> </ul>                 |  |  |
|   | management                                  | management   | management   |  |  |
| Goals                                     | Achieve appropriate                         | • Traditional sector is                            | <ul> <li>Maximize autonomy</li> </ul>              |  |  |
|   | balance among                               | sacred, protect at all                             | and opportunity                                    |  |  |
|   | economic and                                | costs  | • Favor economy over                               |  |  |
|   | environmental                               | <ul> <li>Minimize pollution</li> </ul>             | environment  |  |  |
|   | objectives                                  | Distrust tradcoffs                                 |  |  |  |

Process

- 1) Conduct RDM analysis for each worldview
- Examine how solution for each world view performs from perspective of other worldviews
- 3) Search for solutions that perform well across worldviews



## **Find Policy Solutions for Hierarchist Worldview**



- 1. Express policy solutions as a controller with four parameters:  $x_{set}$ : Target level for pollution
  - b: Amount of regulation for pollution past target level
  - c: Amount of regulation for level of warning
  - q: Amount of training for a given unemployment rate
  - Use multi-objective robust decision making (MORDM) algorithms to find non-dominated solutions over many objectives



### **Find Solutions for All Worldviews**



## How do we compare solutions across worldviews?



- Measure comparative performance using regret
- Calculate regret as the Euclidean distance between:
  - Solution from one world view and
  - closest point on pareto surface for another world view

Distance<sub>s</sub> (f) = 
$$\sqrt{\sum_{j}^{J} [O_{s,j}(f) - O_{r^{*},j}(f)]^{2}}.$$



### **Display Comparisons in Utopia-Dystopia Matrices**

|          |               | Regret using Strategy's World View |      |         |        |              |     |            |      |
|----------|---------------|------------------------------------|------|---------|--------|--------------|-----|------------|------|
| 1        |               | Environment                        |      | Economy |        | Unemployment |     | Inequality |      |
| Strategy | World<br>View | Rel                                | Poll | Econ    | Employ | Trad         | New | Unemp      | Inc  |
| Н        | Е             | 82%                                | 41%  | 5%      |        | 88%          | 17% | 74%        |      |
| Н        | Ι             | 0%                                 | -1%  | 14%     |        | 0%           | 2%  | -1%        |      |
| Е        | Н             | 0%                                 | -1%  | 0%      | 0%     |              |     | 2%         | 0%   |
| Е        | Ι             | 0%                                 | -6%  | 33%     | -7%    |              |     | 14%        | -42% |
| Ι        | Η             | 21%                                |      | 3%      |        |              |     |            |      |
| Ι        | Е             | 83%                                |      | 26%     |        |              |     |            |      |





### **Seeking Robust Compromise Strategies**

- 1. Identify new controller, more acceptable to Egalitarians:
  - Regulate pollution at constant level until pollution intensity drops below a threshold level
- 2. Find strategies that simultaneously perform well in multiple world views

|          |               | Relative Regret using Realized World View's objectives |      |         |        |              |      |            |     |
|----------|---------------|--|------|---------|--------|--------------|------|------------|-----|
|          |               | Environment  |      | Economy |        | Unemployment |      | Inequality |     |
| Strategy | World<br>View | Rel  | Poll | Econ    | Employ | Trad         | New  | Unemp      | Inc |
| H/E      | Е             | 0%   | -2%  | 3%      | 0%     |              |      | 11%        | 0%  |
|          | Н             | 0%   | -23% | 18%     |        | -98%         | 499% | 424%       |     |
| H/I      | Ι             | 0%   |      | 2%      |        |              |      |            |     |
|          | Η             | 0%   | 21%  | -2%     |        | 45%          | -24% | 64%        |     |



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# RDM Uses Models as Exploratory, Not Predictive Tools

Consider two fundamentally different ways to use computer simulation models

#### Consolidative models:

- Gather all relevant knowledge into a single package which, once validated, can be used as a surrogate for the real world
- Used for prediction

#### Exploratory models:

- Map assumptions onto consequences, without privileging any one set of assumptions
- Supports iterative problem-solving

Exploratory modeling can:

- Help generate hypotheses
- Identify special cases
- Assess common properties across many cases -in particular identify robust strategies

# Emerging Computational Tools Facilitate RDM and DMDU

- Classification algorithms help extract decision-relevant information from large ensembles of model runs
  - Opportunities for machine learning and other algorithm development?
- Robust optimization algorithms help identify strategies robust across scenarios
- Visualization, often interactive, helps users navigate among multiple futures
- High performance computing makes each of these steps possible

Potential for an interesting approach to human-machine collaboration



Lempert, Popper, Bankes (2003)

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# **RDM Helps People Make Better Decisions, Not Better Predictions**

#### **Basic principles**

- 1. Consider multiple futures, not one single future, in your planning. Choose these futures to stress test your organization's plans
- 2. Seek robust plans that perform well over many futures, not optimal plans designed for a single, best-estimate future
- 3. Make your plans flexible and adaptive, which often makes them more robust



### Plan over multiple



Some DMDU aphorisms:

- Premature aggregation is the root of all evil in decision support (J. Kwakkel)
- Seek certainty in your decisions, not in your predictions

# **QUESTIONS?**

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