

Introduction to Rhodium and Multiobjective Robust Decision Making

CEE 6200 Final Project

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Rhodium library in Python

Formulate
simulation-
optimization
problem

Generate
alternatives

Sample candidate
solutions over
possible states of
world

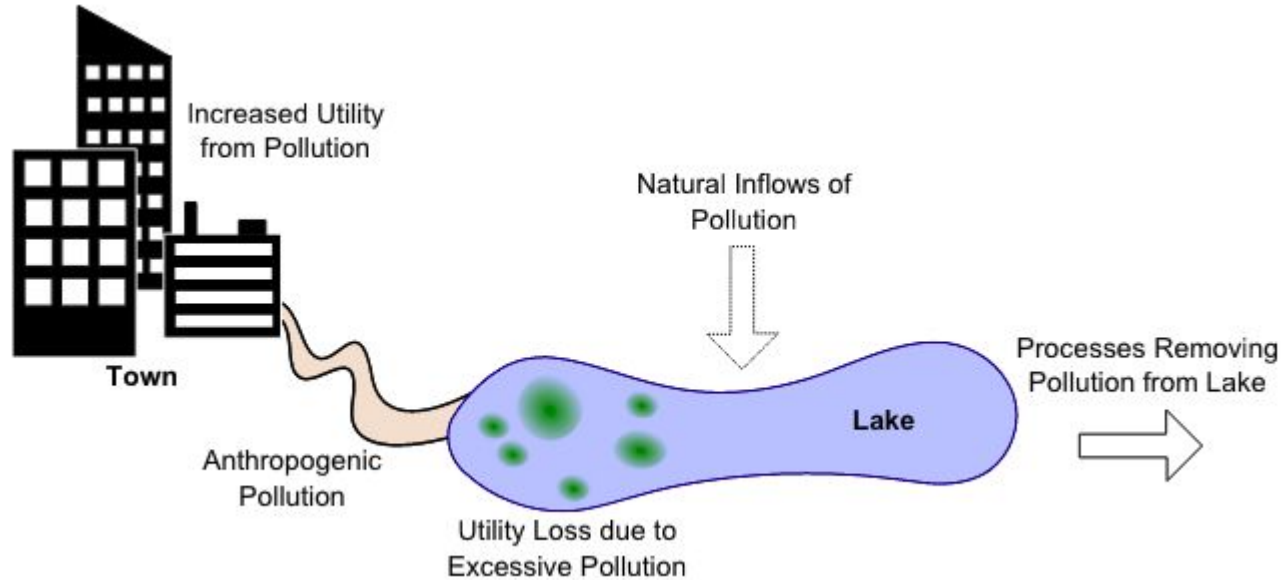
Perform sensitivity
analysis

MORDM framework has environmental relevance

- Ecosystem problems include watershed management
- Such problems often have multiple competing objectives
 - Maximum pollution limits
 - Economic utility
 - Managing wildlife



The Lake Problem



A town's inhabitants must decide how much phosphorus to release into a lake

Lake problem equations

- Rate of eutrophication modelled as

$$\frac{dP}{dt} = I - bP + \frac{qP^a}{K^a + P^a}$$

P = mass/concentration of P

q = rate of recycling of P

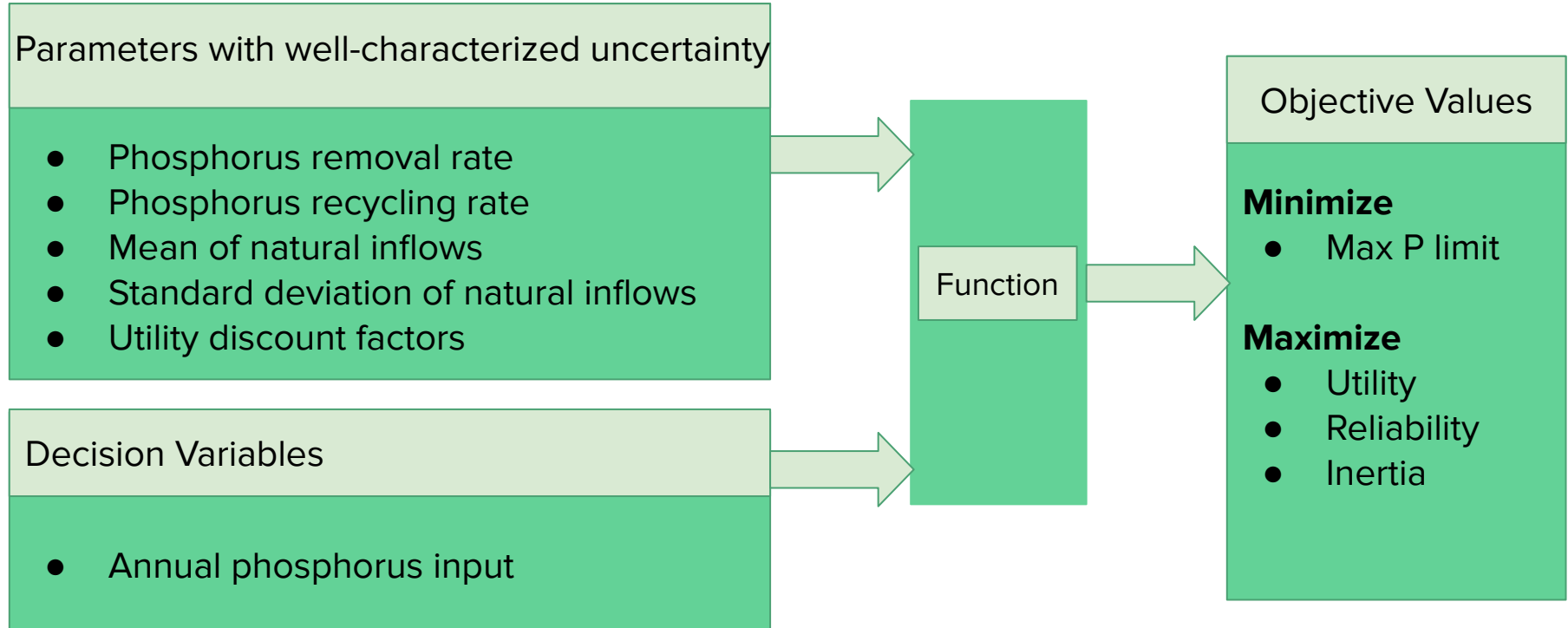
I = rate of P input from watershed

b = rate of P loss per unit time

a = dimensionless parameter describing the steepness of sigmoid curve

K = P value at which recycling reaches half the maximum rate

Lake problem formulation in Rhodium



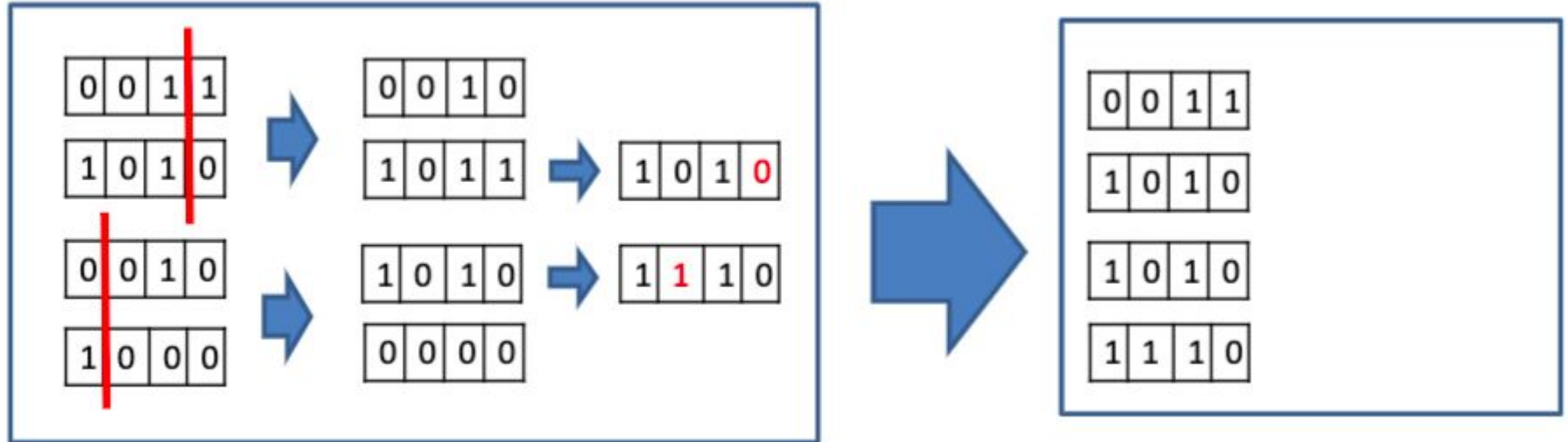
Lake problem formulation in Rhodium

- Model inputs

```
def lake_problem(pollution_limit,  
    b = 0.42,          # decay rate for P in lake (0.42 = irreversible)  
    q = 2.0,           # recycling exponent  
    mean = 0.02,       # mean of natural inflows  
    stdev = 0.001,     # standard deviation of natural inflows  
    alpha = 0.4,       # utility from pollution  
    delta = 0.98,      # future utility discount rate  
    nsamples = 100):  # monte carlo sampling of natural inflows)
```

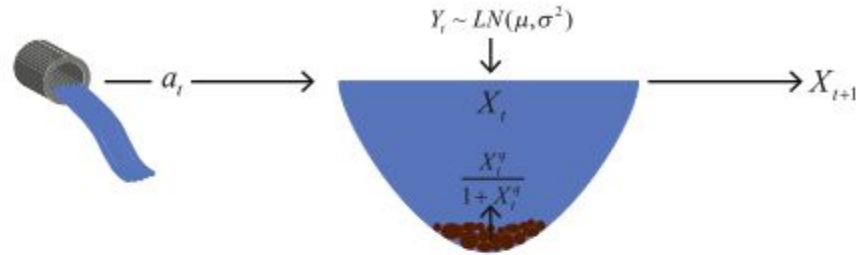
Optimize the Rhodium Model

- The Rhodium model uses the NSGA-II optimization algorithm
- “Levers” or parameters in the model are iteratively tweaked by the algorithm to find optimal policies

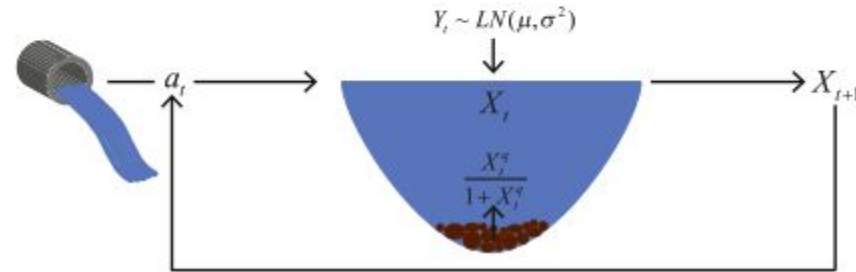


Closed Loop Direct Policy Search vs Open Loop Intertemporal Control Strategy

a) Open Loop Control



b) Closed Loop Control



Source: Quinn et al (2017)

Model Output

- Model output is a collection of optimal policies
- Each policy includes corresponding values for the 4 objective functions

		pollution_limit	max_P	utility	inertia	reliability
0	[0.02472156653486018, 0.0005092878026342444, 0...	0.230526	0.302243	0.838384	1.0000	
1	[0.025479198131745425, 0.08099822347194309, 0....	0.413127	0.447774	0.727273	0.9996	
2	[0.08281475714788841, 0.001261564498185647, 0....	0.217104	0.352261	0.808081	1.0000	
3	[0.08111582993527067, 0.001261564498185647, 0....	0.911485	0.431152	0.757576	0.9526	
4	[0.08111582993527067, 0.001261564498185647, 0....	0.329243	0.382202	0.797980	1.0000	
5	[0.08270864663396345, 0.00022583728504802186, ...	0.704168	0.414815	0.808081	0.9743	
6	[0.04011701826649017, 0.05257271958858992, 0.0...	0.527062	0.444348	0.747475	0.9937	
7	[0.07755534965011637, 0.001261564498185647, 0....	0.826528	0.426855	0.767677	0.9622	
8	[0.025479198131745425, 0.06834231918517299, 0....	0.272270	0.340484	0.818182	1.0000	
9	[0.025479198131745425, 0.06834231918517299, 0....	0.311525	0.416324	0.767677	0.9994	

Lake problem optimization data summarized in a data frame

Using Rhodium to Explore Optimization Data

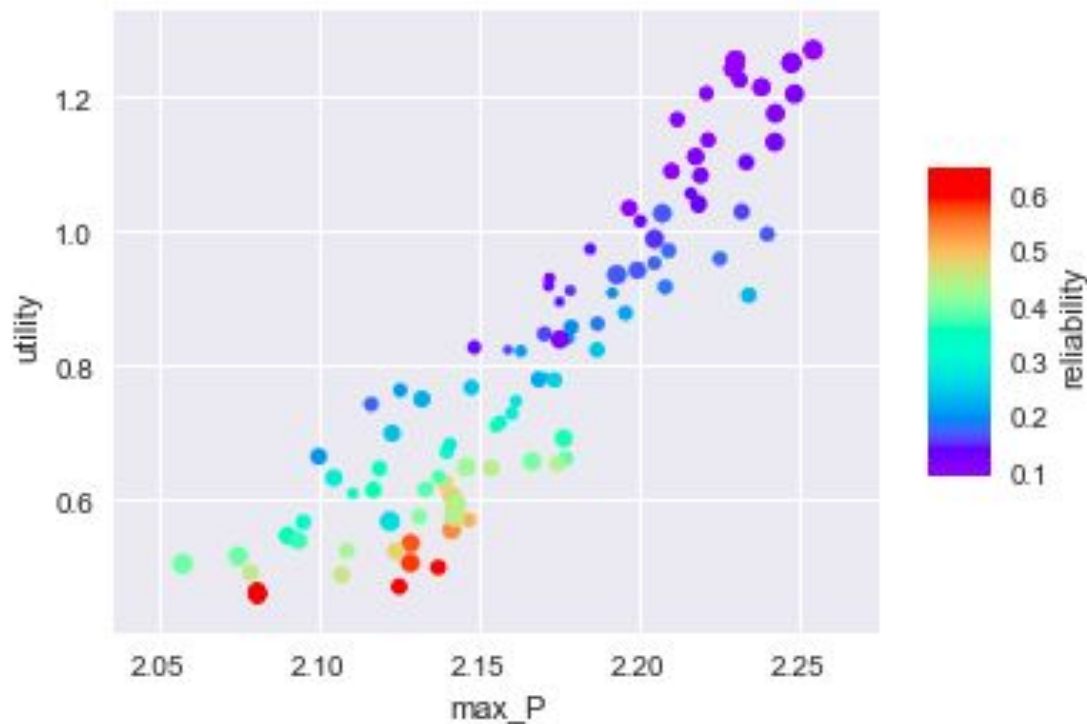
- **2D scatter plot of the Pareto front**

X-axis: Max P limit

Y-axis: Utility

Color: Reliability

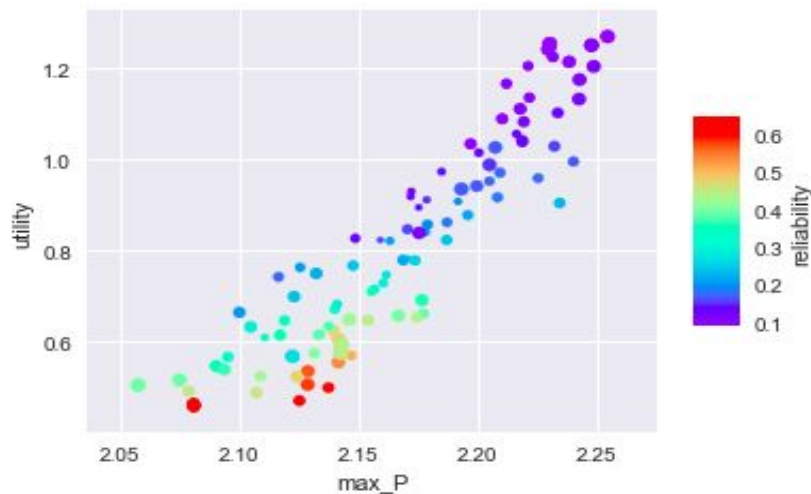
Size: Inertia



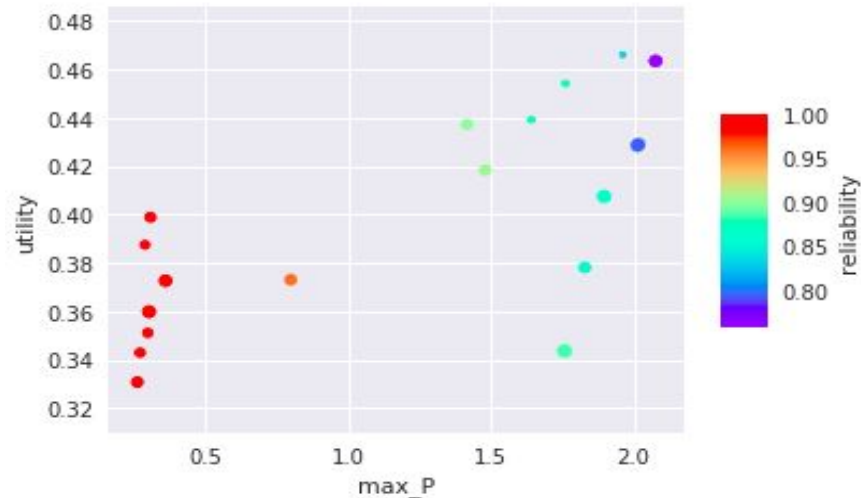
Pareto Optimal Policies

Added Optimization Constraints to Narrow Results

- Added a reliability ≥ 0.75 constraint

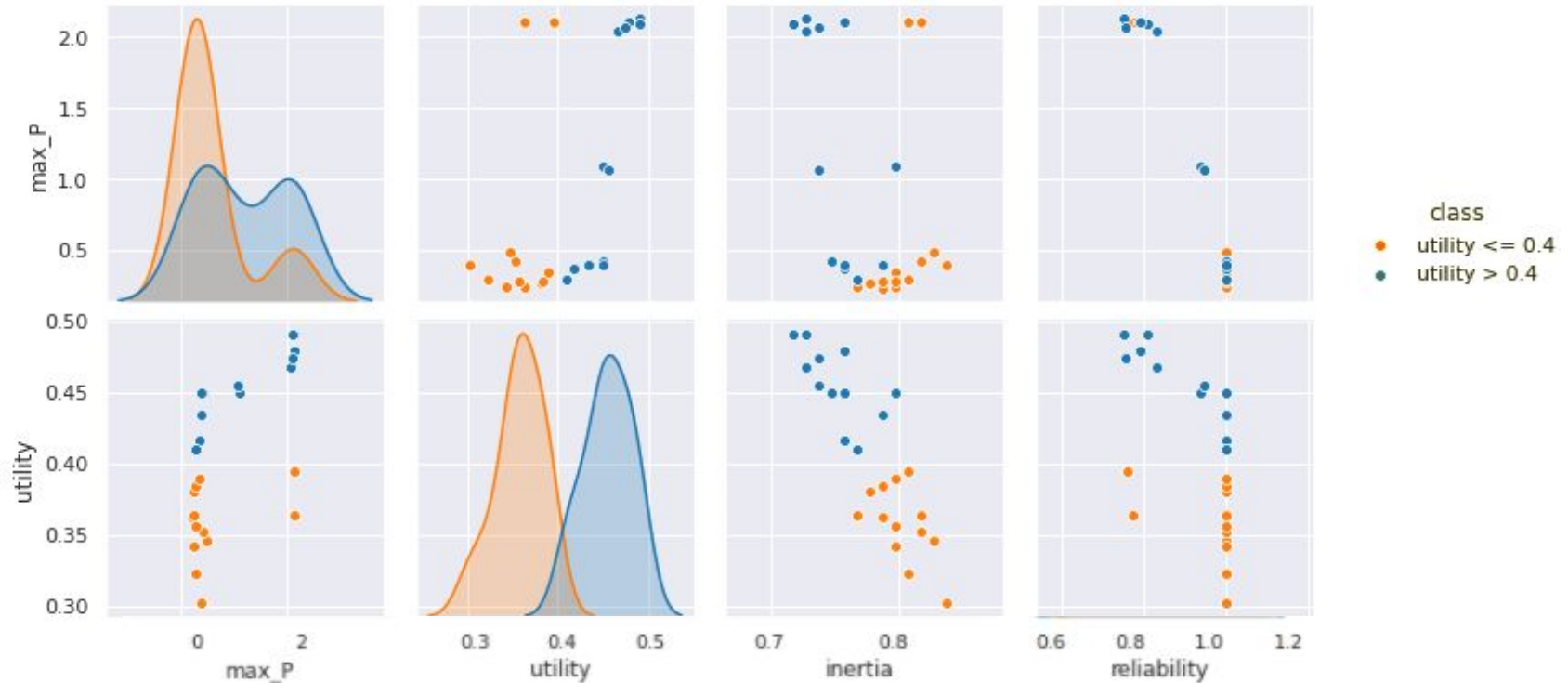


Scatter plot of pareto optimal policies



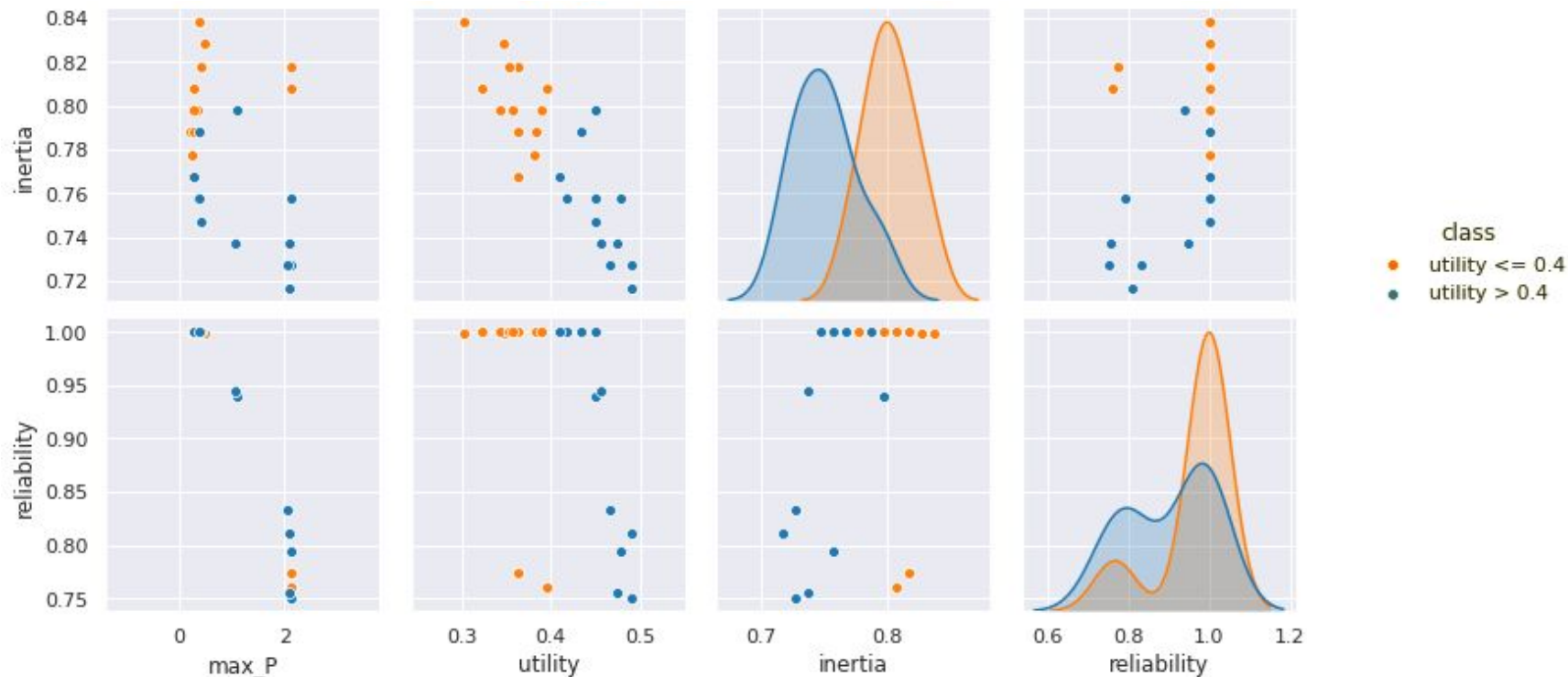
Scatter plot of pareto optimal policies from constrained problem

Pairwise Comparison of Objective Values for Pareto Optimal Policies



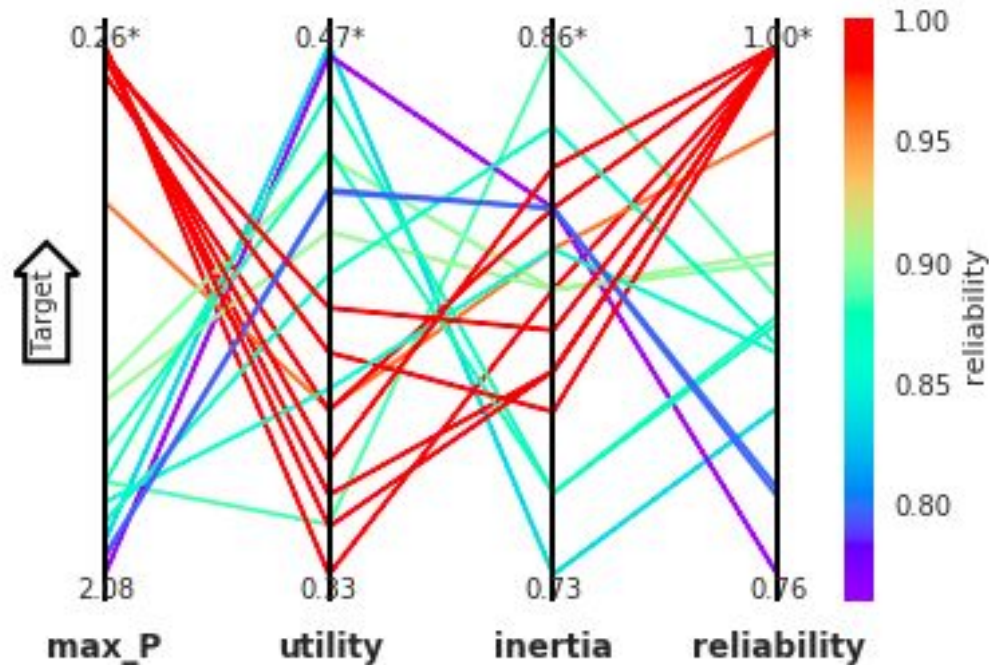
Pairwise comparisons of objectives

Pairwise Comparison of Objective Values for Pareto Optimal Policies



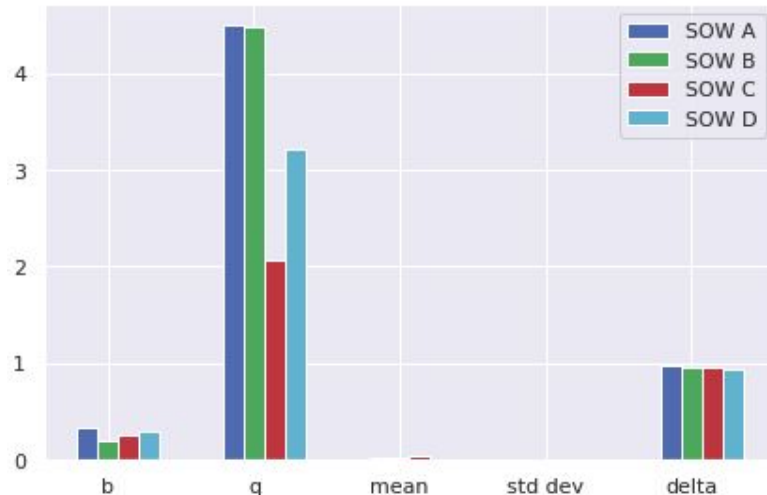
Pairwise comparisons of objectives

Parallel Axis Plot of Objective Values for Pareto Optimal Policies



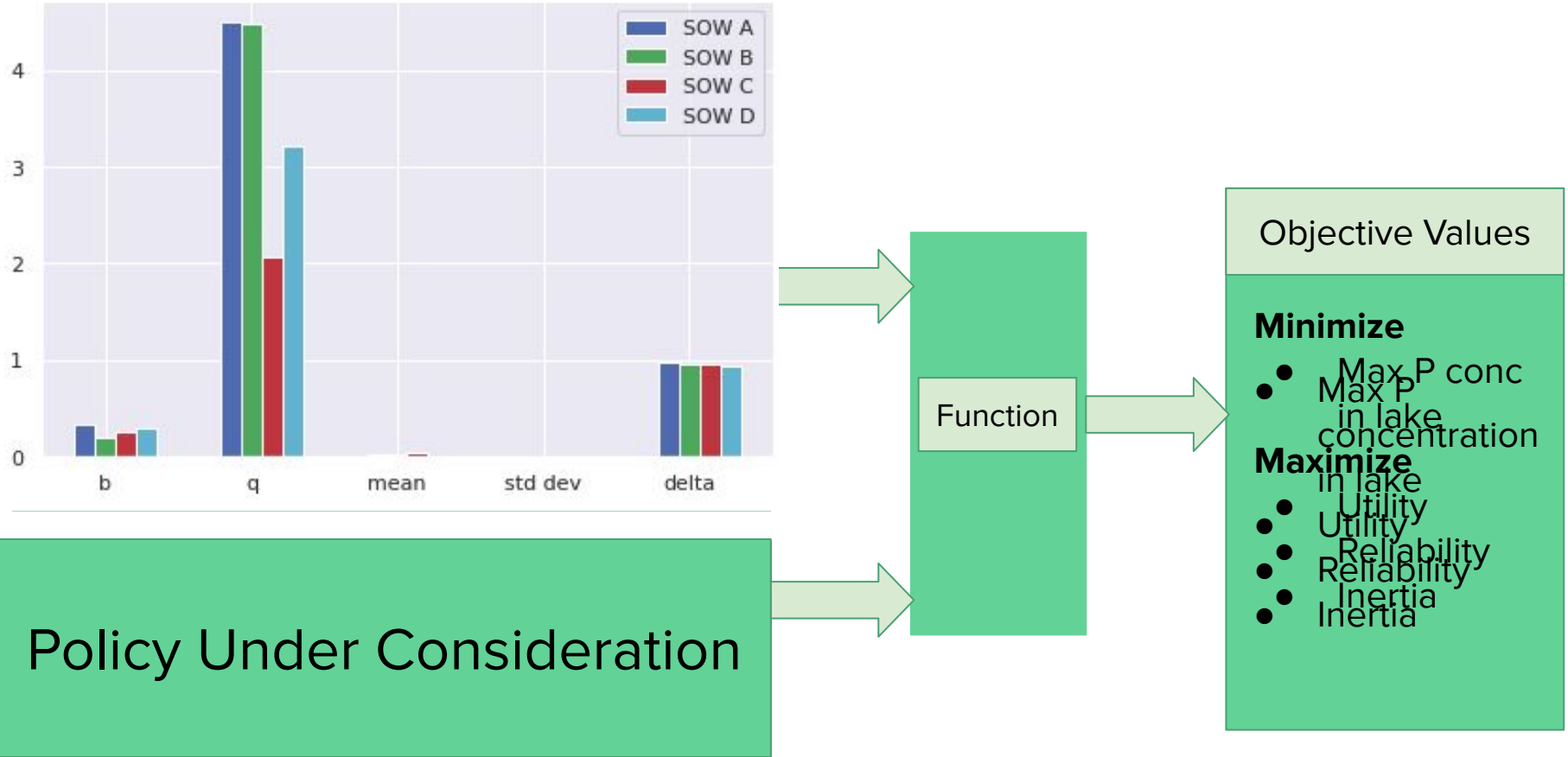
Putting the RDM in MORDM

- Generate multiple states of the world by defining the uncertain parameters in the model
- We generated 1000 states of the world by defining uniform distributions for recycling exponent, decay rate, mean of natural P inflows, standard deviation of P inflows, and future utility discount rate



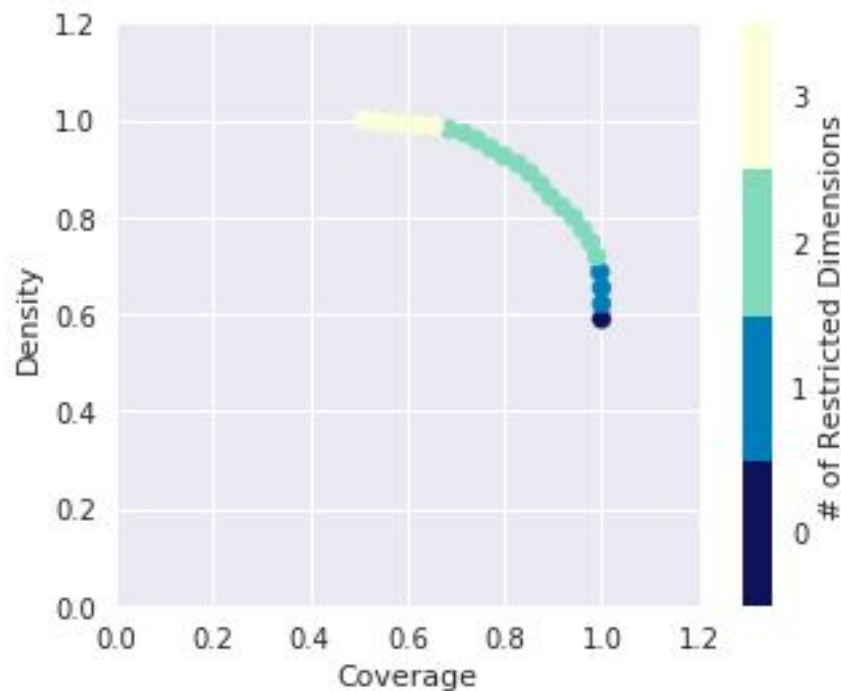
Visual representation of 4 states of world that were generated using Rhodium framework

Overview of Transition from Optimization to Scenario Discovery

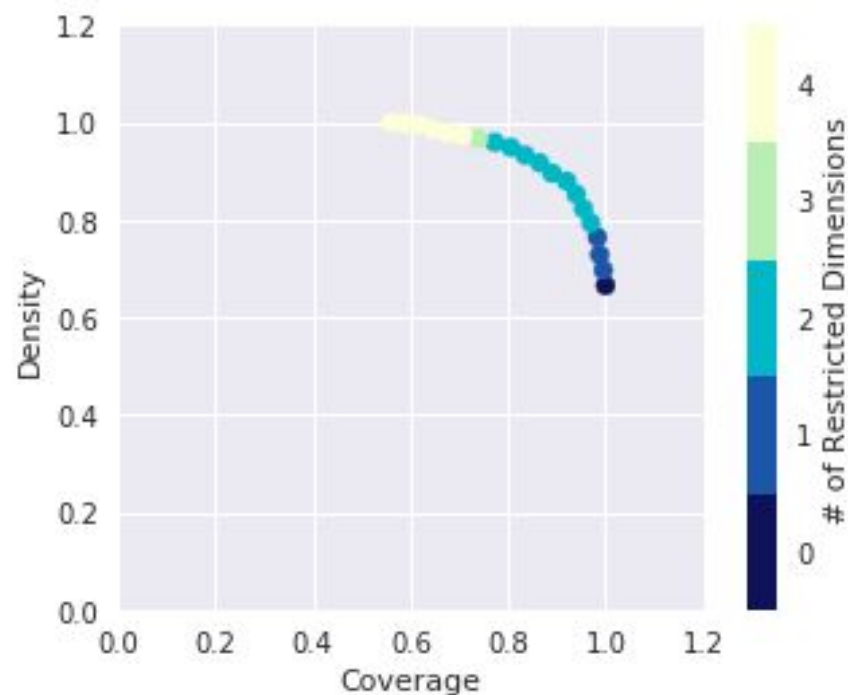


Sensitivity Analysis

Patient Rule Induction Method to Identify Key Uncertainties



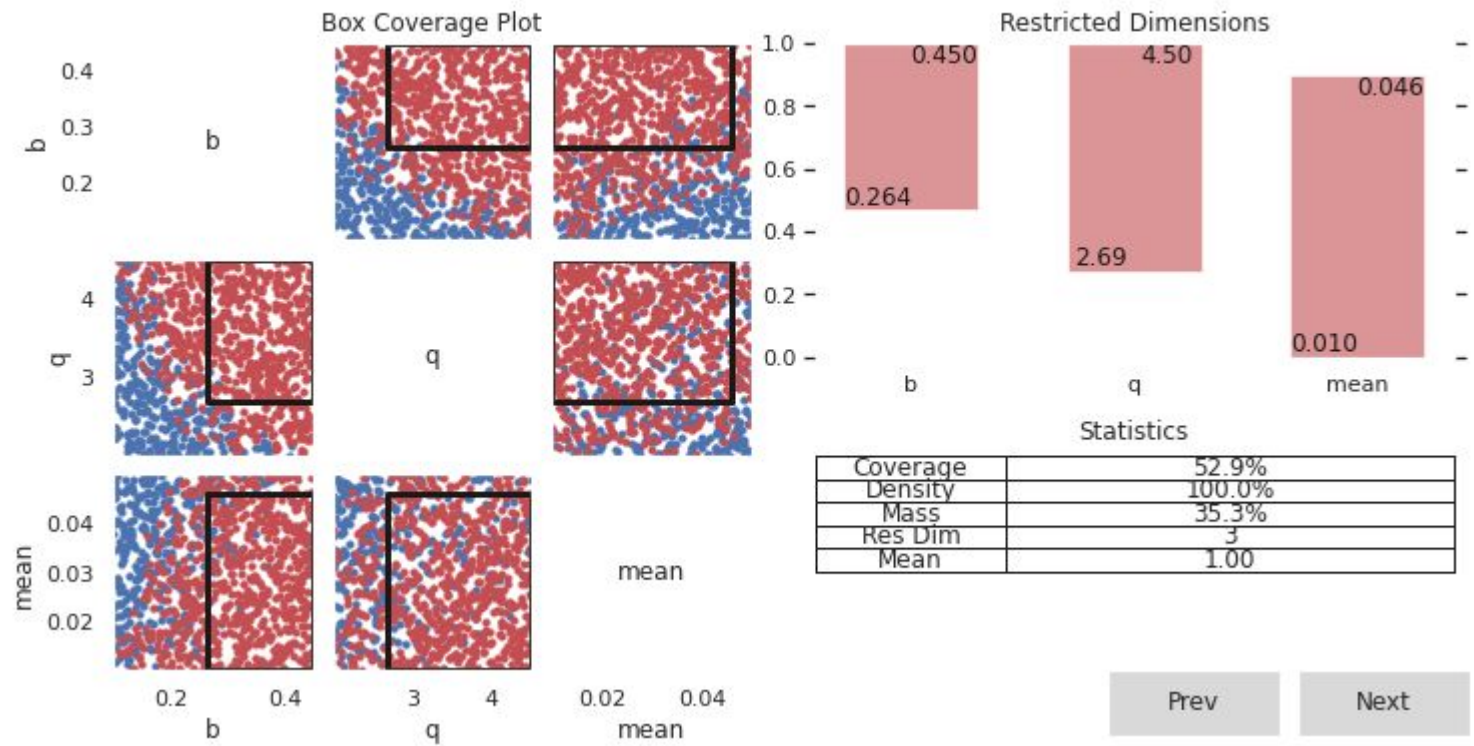
PRIM Results with Policy of Annual
pollution limit of .02



PRIM Results with Policy of Annual
pollution limit of .01

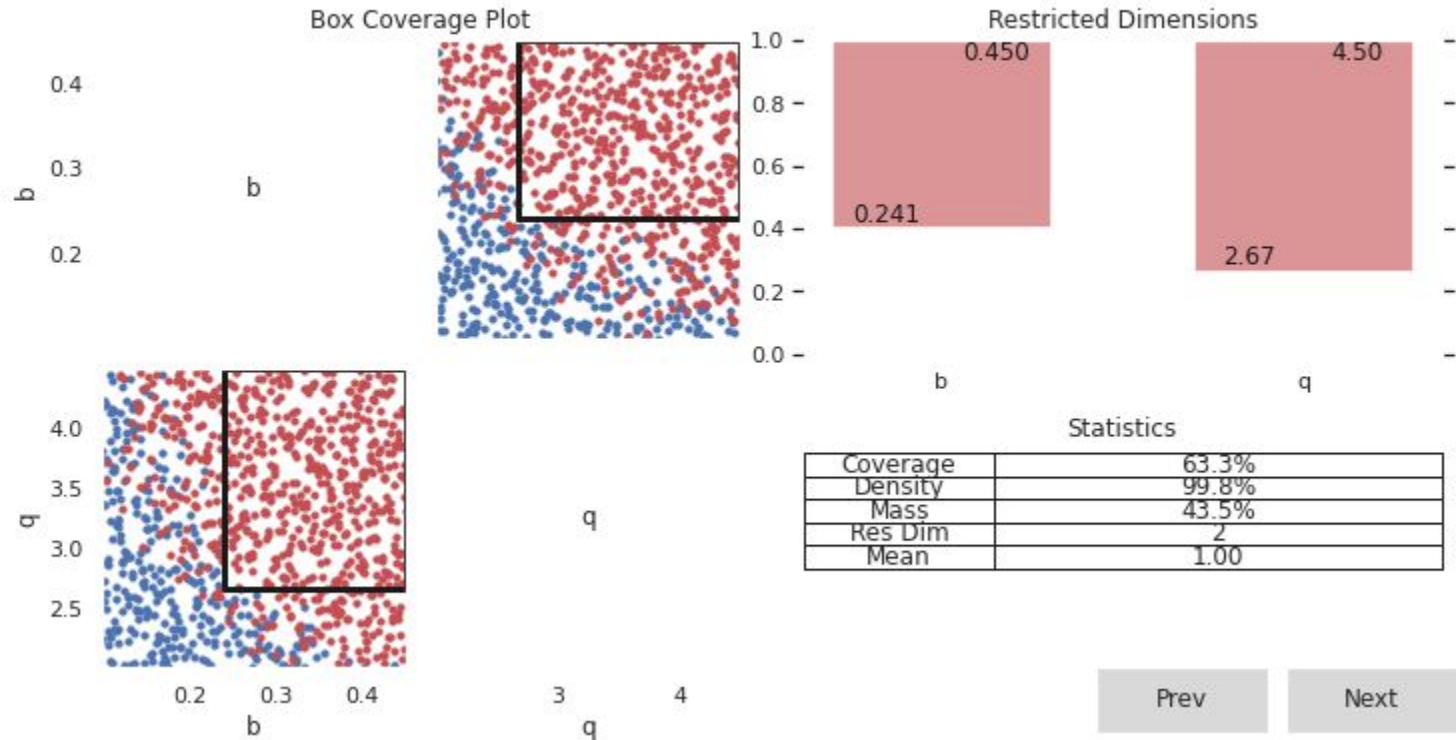
Patient Rule Induction Method to Identify Key Uncertainties

Peeling/Pasting Trajectory 20

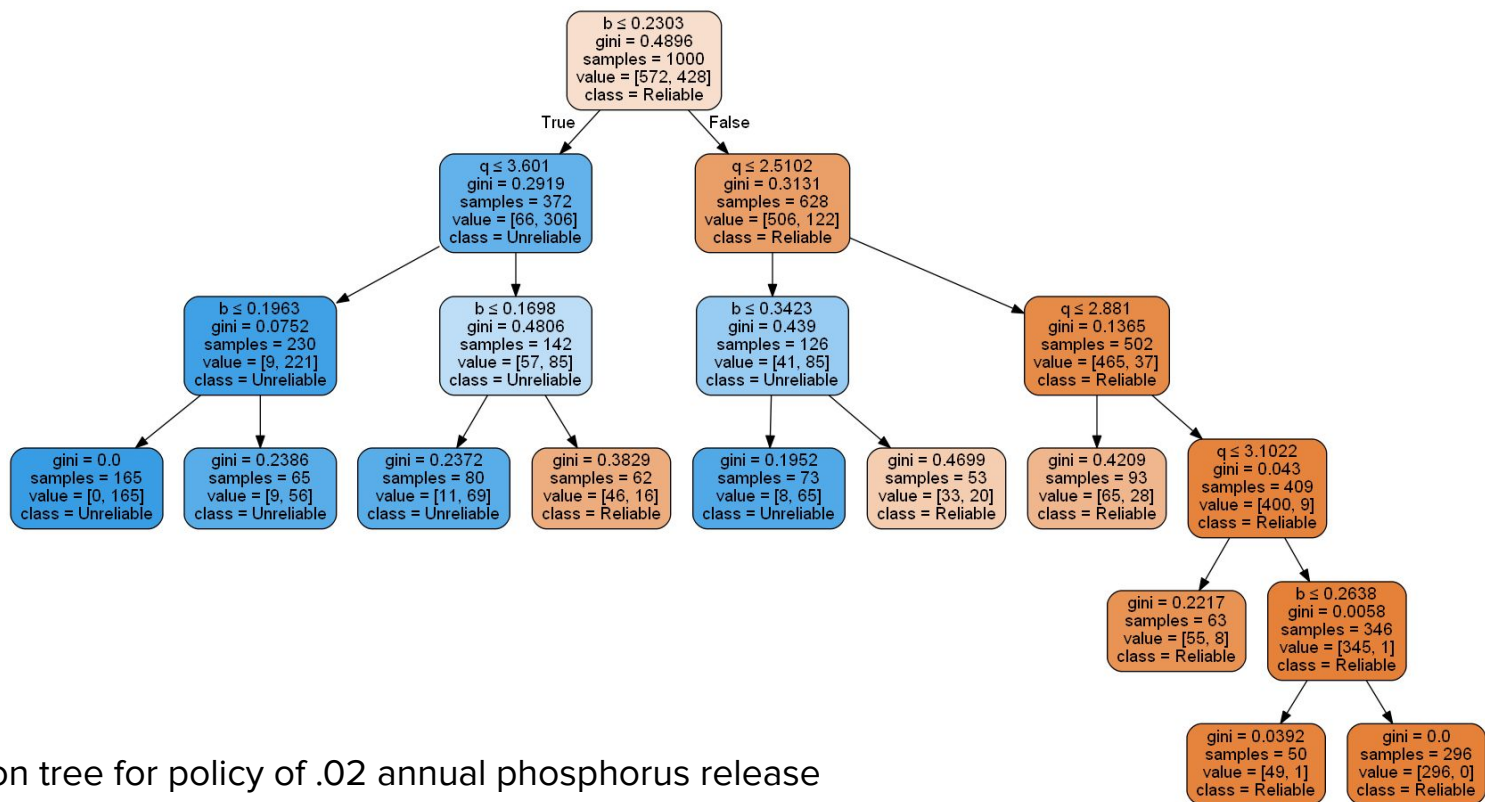


Patient Rule Induction Method to Identify Key Uncertainties

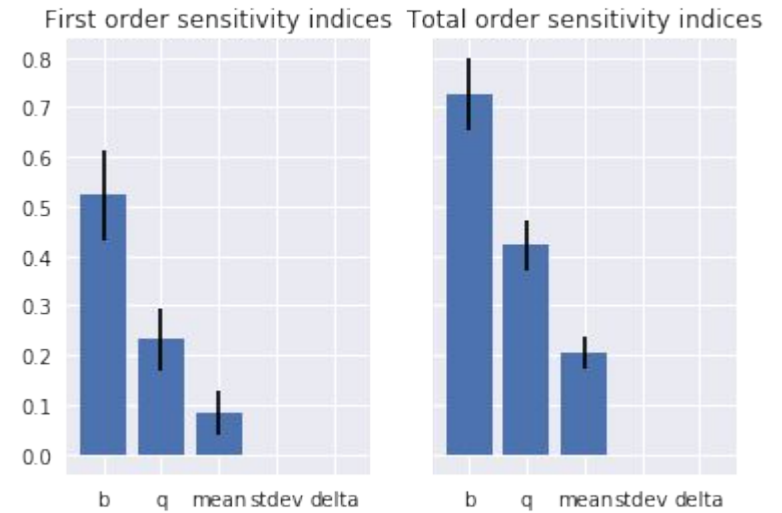
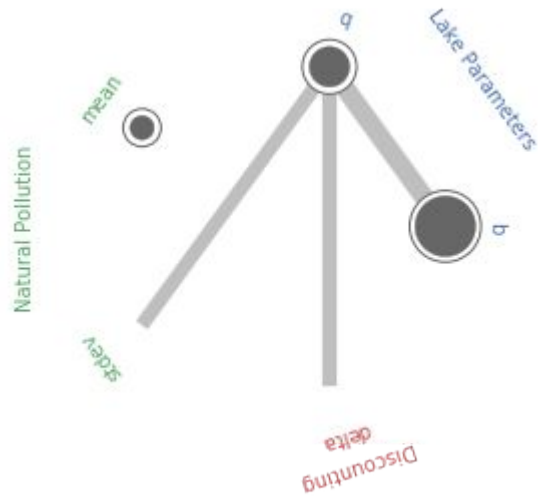
Peeling/Pasting Trajectory 16



Classification and Regression Trees



Sobol Sensitivity Analysis



Conclusion

- The decay rate (b) and recycling rate (q) seem to have the most influence
- Rhodium is very useful in multiobjective modelling and data visualization for environmental problems
- This model did not take into account the cost to implement any policy
- Jupyter notebooks aren't the best place to run Rhodium
- It might be worthwhile to use other algorithms besides NSGA-II
- States of the world were generated by assigning uniform distributions to uncertain parameters

Thank you

Sources

Direct policy search for robust multi-objective management of deeply uncertain socio-ecological tipping points (Quinn et al 2017)

Rhodium: Python Library for Many-Objective Robust Decision Making and Exploratory Modeling

Many Objective Robust Decision Making for Complex Environmental Systems Undergoing Change (Kasprzyk et al 2012)

Many-objective robust decision making for managing an ecosystems with deeply uncertain threshold response (Singh et al 2015)

Management of Eutrophication for Lakes Subject to Change (Carpenter et al 1999)