Modeling Workgroup Meeting Quarterly Review

Optimization update

Kalyanmoy Deb, Pouyan Nejadhashemi, Gregorio Toscano, and Hoda Razavi.



Overview

- Objective 2: Development of Efficient Multi-objective Optimization Procedures
 - Oct 1, 2021 to September 30, 2023 (24 months)
- Up-to-date status of the project: Optimization approaches
- Innovization Study (extension)
- Submitted papers in this quarter.
- Conclusions and future work

Current status of the project

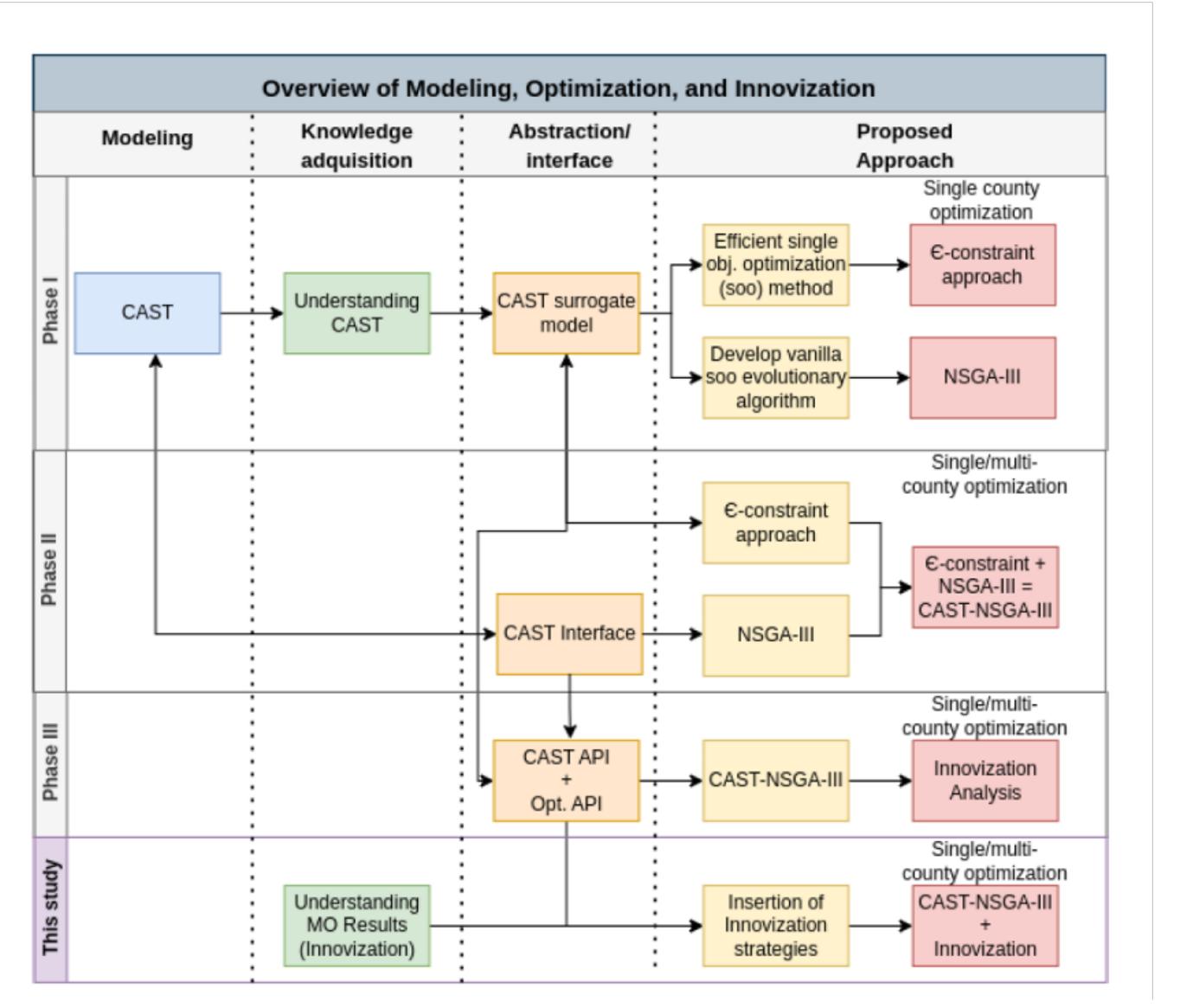
Calendar Year	2020)		2021				2022				2023				2024				2025				202
Calendar Quarter	Q2	Q3	Q4	Q1	Q2	Q3	Q4	Q1	Q2	Q3	Q4	Q1	Q2	Q3	Q4	Q1	Q2	Q3	Q4	Q1	Q2	Q3	Q4	Q1
Project Year		Ye	ar 1	Year 2			ear 2	Year 3					Ye	ar 4			Ye	ar 5		Year 6		ar 6		
Task 1: Development of an efficient single-objective hybrid optimization procedure																								
1.1: Understanding CAST modules and effect of BMPs on objectives and constraints																								
1.2: Development of a simplified point-based structured single- objective optimization procedure																								
1.3: Development of a hybrid customized single-objective optimization procedure																								
1.4: Verification and validation with CBP users and decision-makers and update of optimization procedure																								_
Task 2: Development of efficient multi-objective (MO) optimization procedures															·									
2.1: Develop generative MO optimization using hybrid optimization procedure developed at Task 1																								
2.2: Develop simultaneous MO customized optimization using population-based evolutionary algorithms																								
2.3: Comparison of generative & simultaneous procedures and validation with CBP users & decision-makers															_									
2.4: Develop an interactive multi-criterion decision-making aid for choosing a single preferred solution			Ļ						ii ii			1												
Task 3: Scalability Studies and Improvements using Learning Engine and Parallel Computing																								
3.1: Comparative study to choose a few best performing methods																								
3.2: Scalability to State and Watershed level Scenarios																								
3.3: "Innovization" approach for improving scalability																								
4.4: Distributed computing approach for improving scalability																								\vdash
Task 4: User-friendly and routine applications with enhanced optimization procedures																								
4.1: User-friendly optimization through a dashboard																								
4.2: Surrogate-assisted optimization procedures													1											\Box
4.3: Robust optimization method for handling uncertainties in variables and parameters																								
4.4: Sustainable watershed management practices		1			╢	t	+	1	1	+	T							\vdash	+-	 	\parallel			

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Project Year		Υe	ar 1			Υe	ar 2			Ye	ar 3			Ye	ar 4			Ye	ar 5			Ye	ar 6	
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Algorithm development

Overview



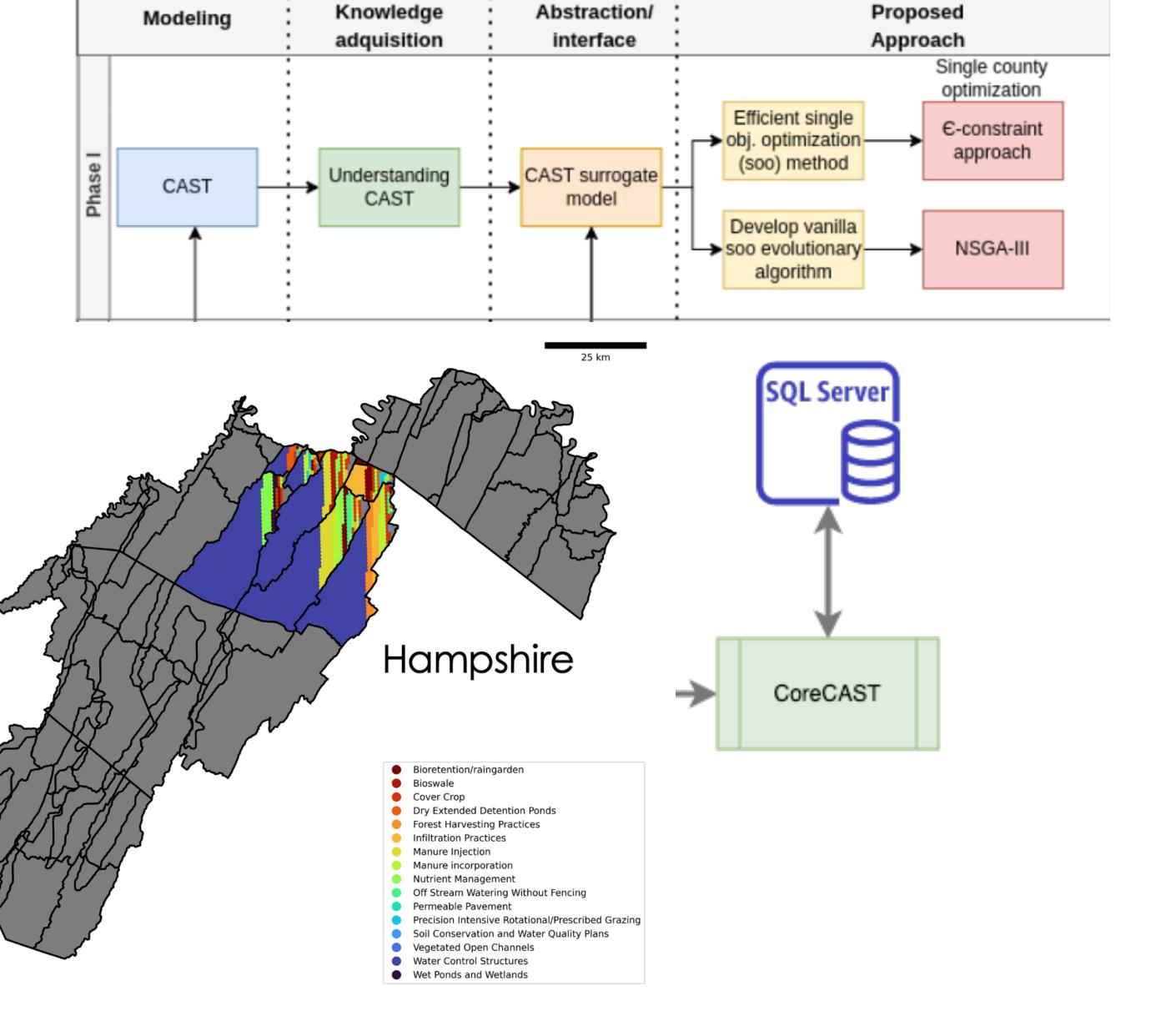
Phase I

SQL Server

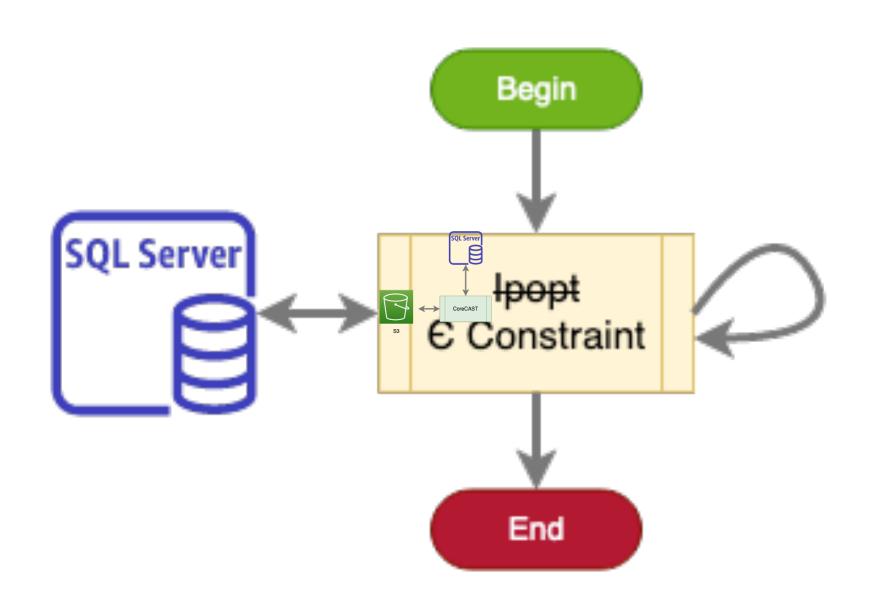
Begin

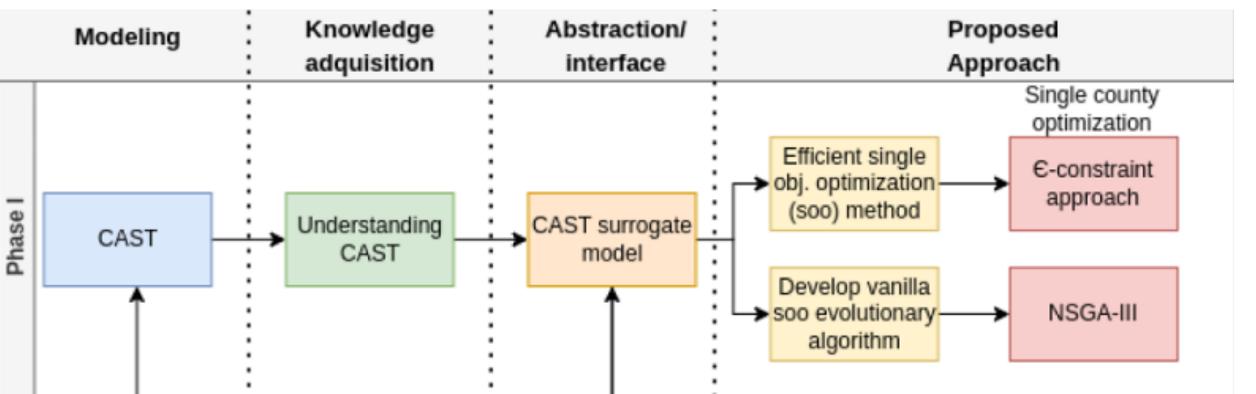
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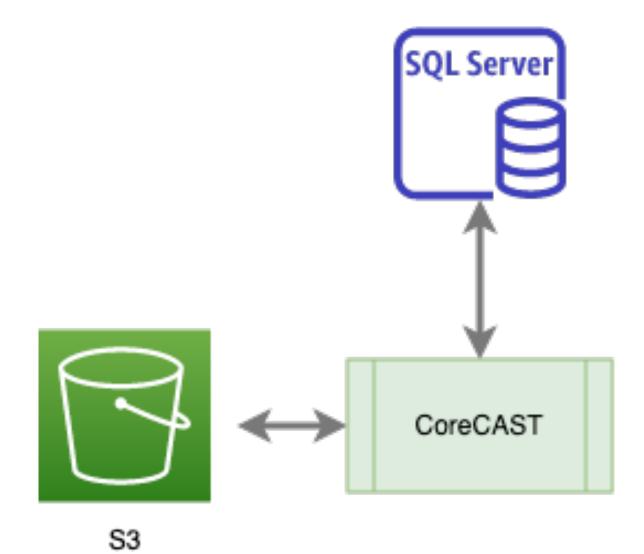
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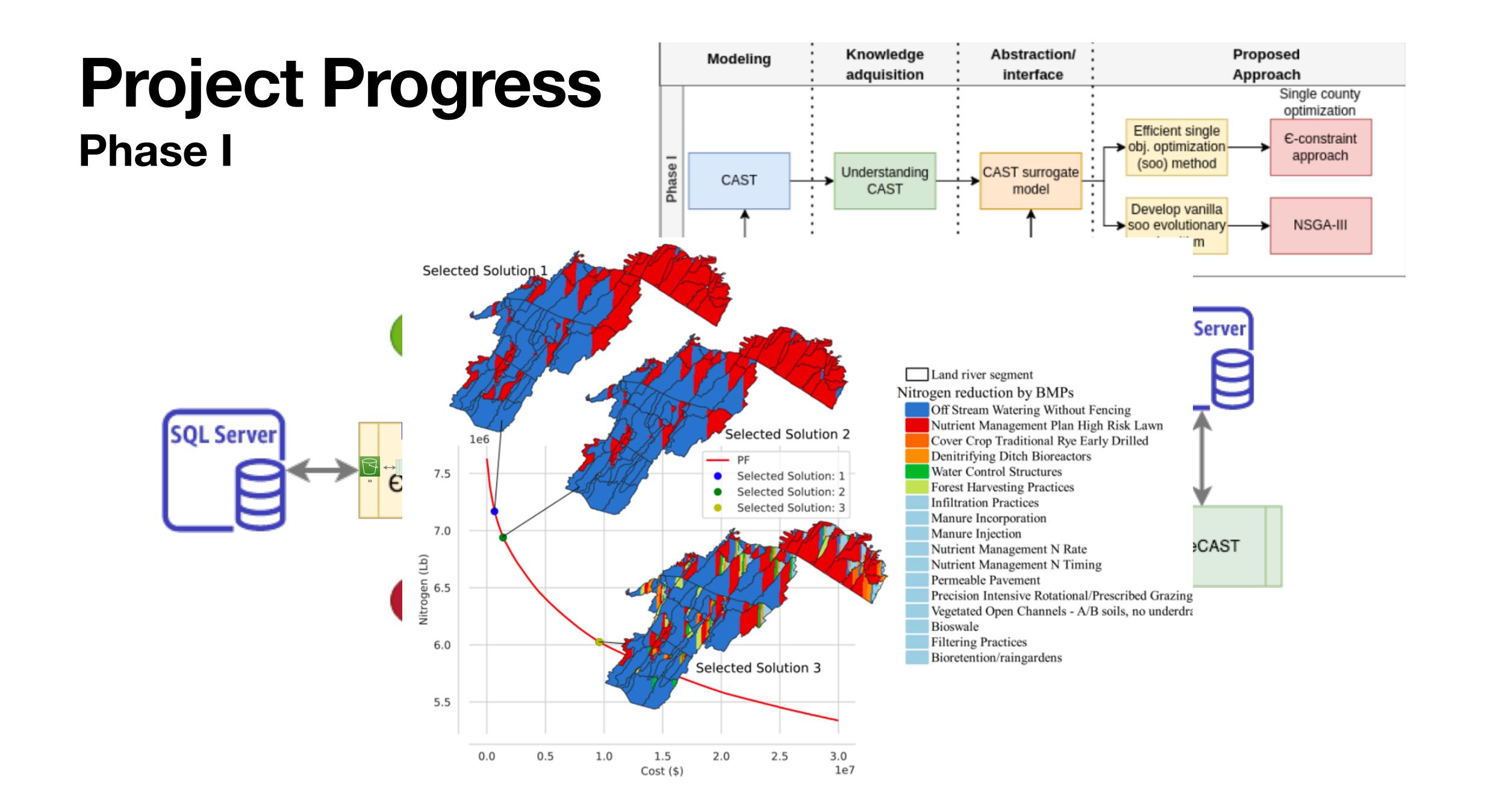


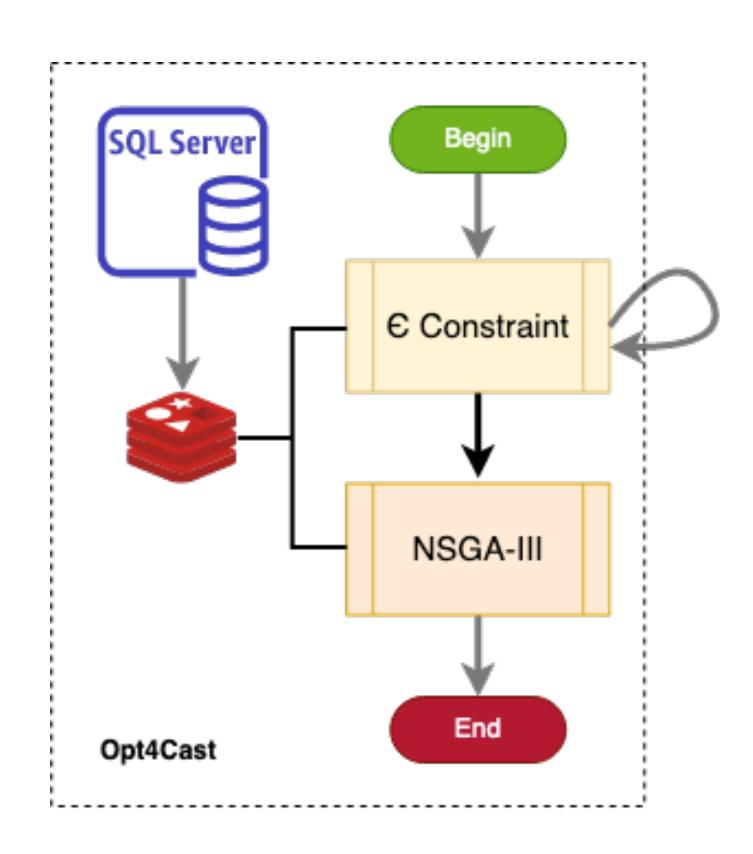
Project Progress Phase I

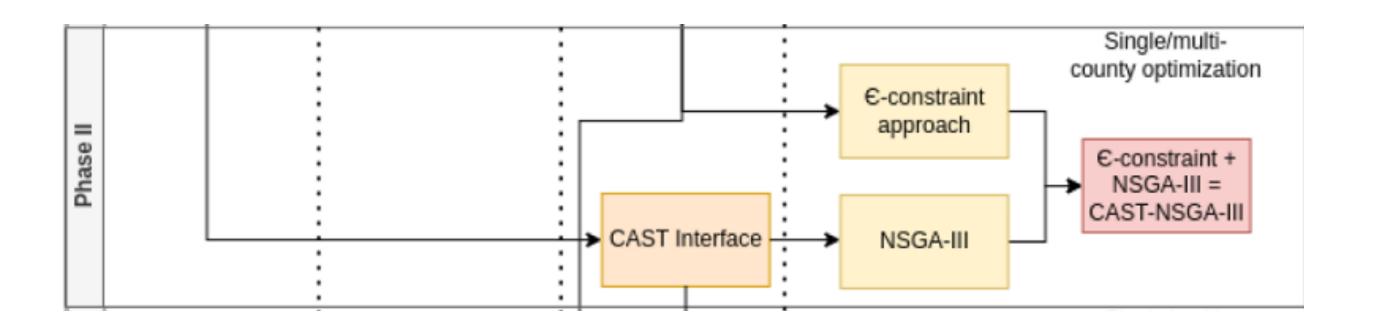


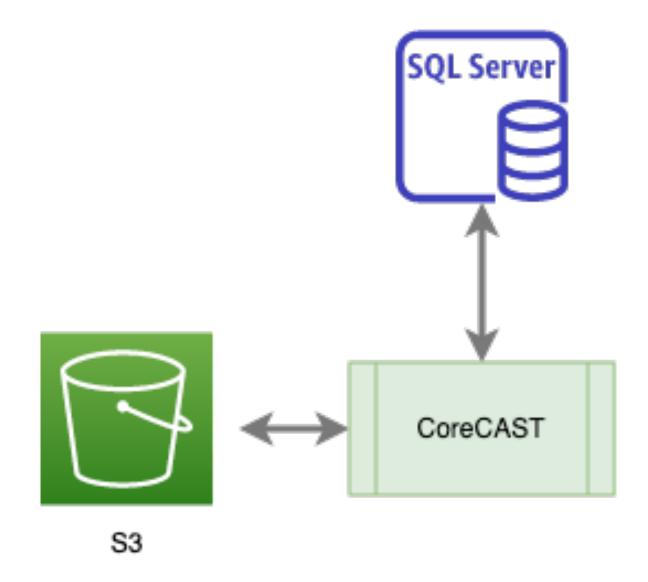


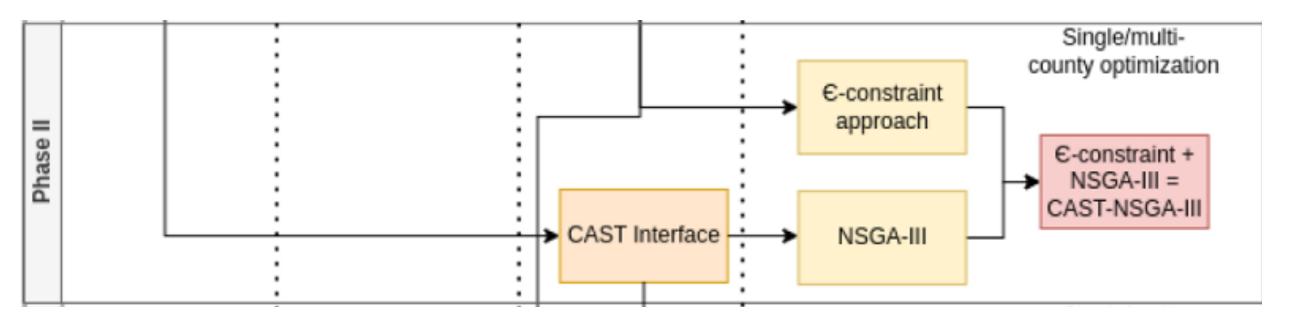


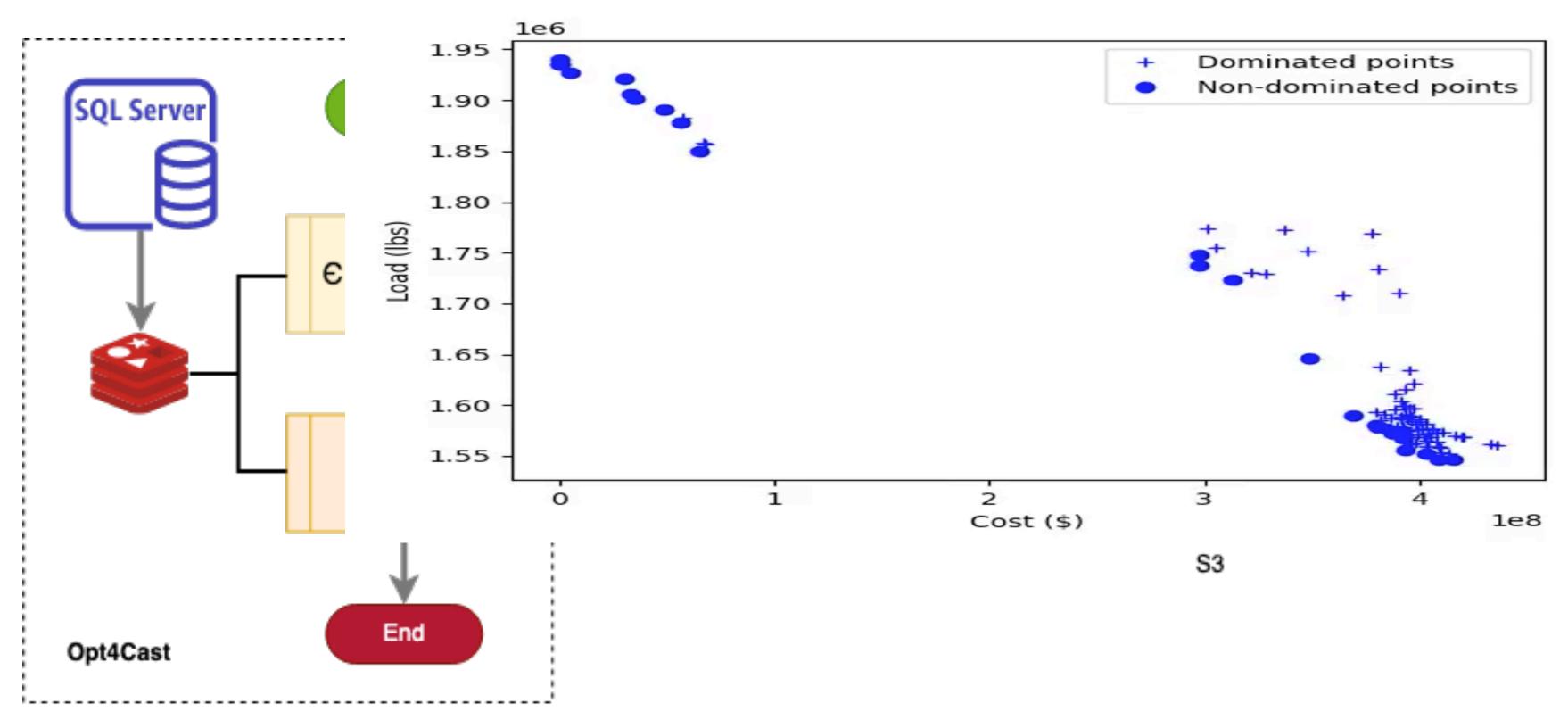


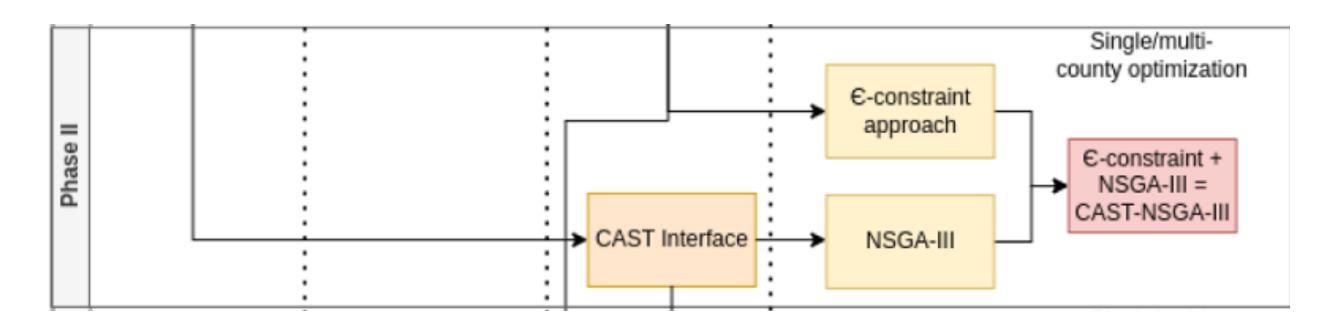


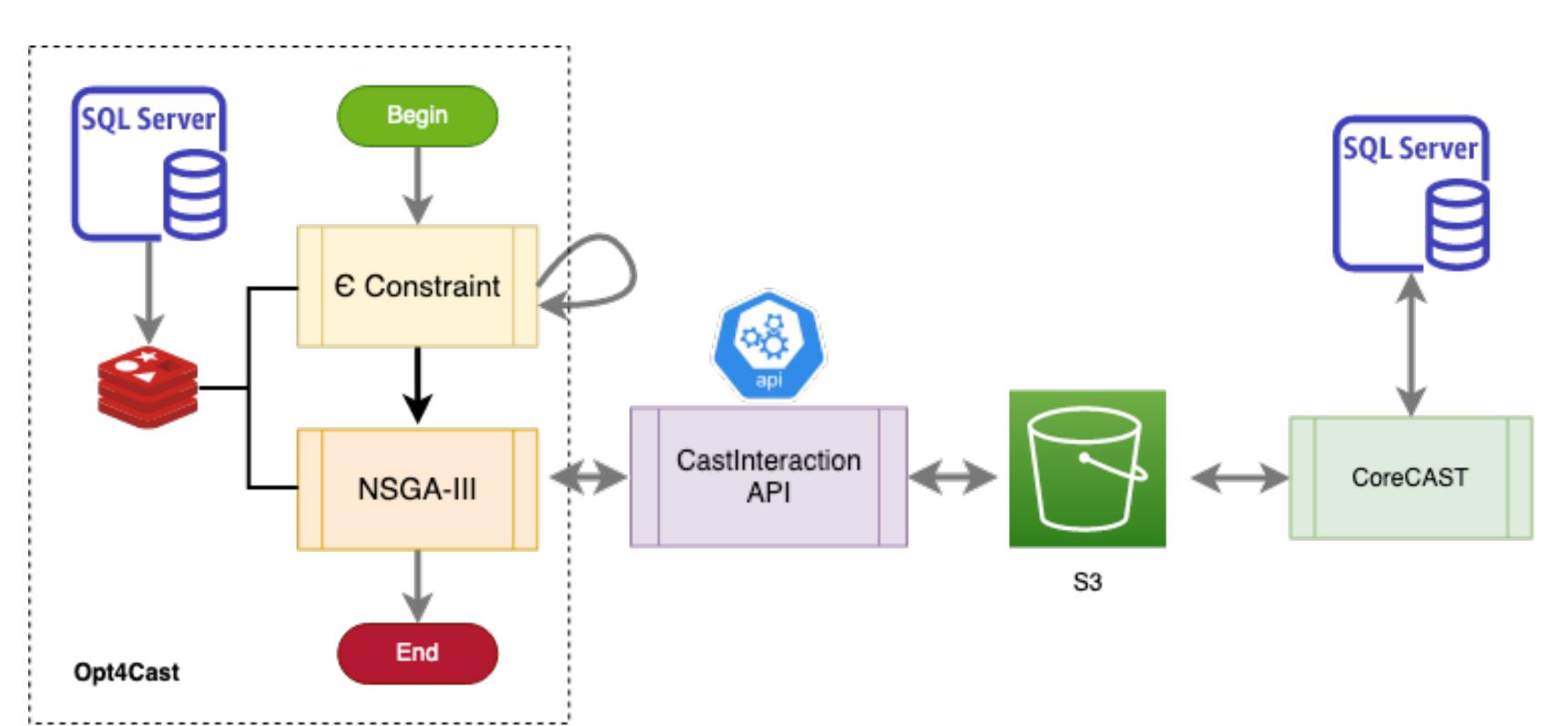


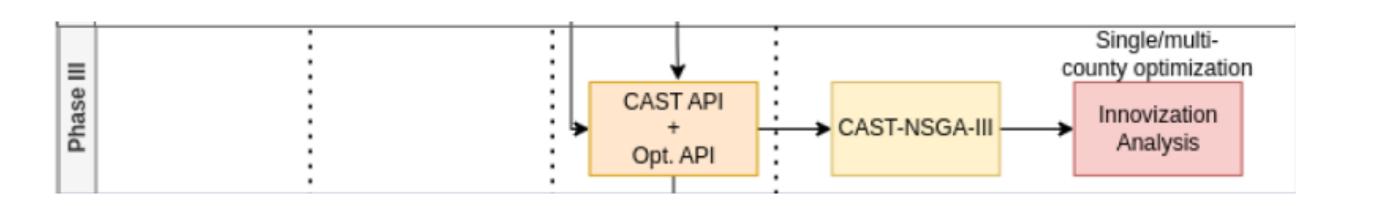


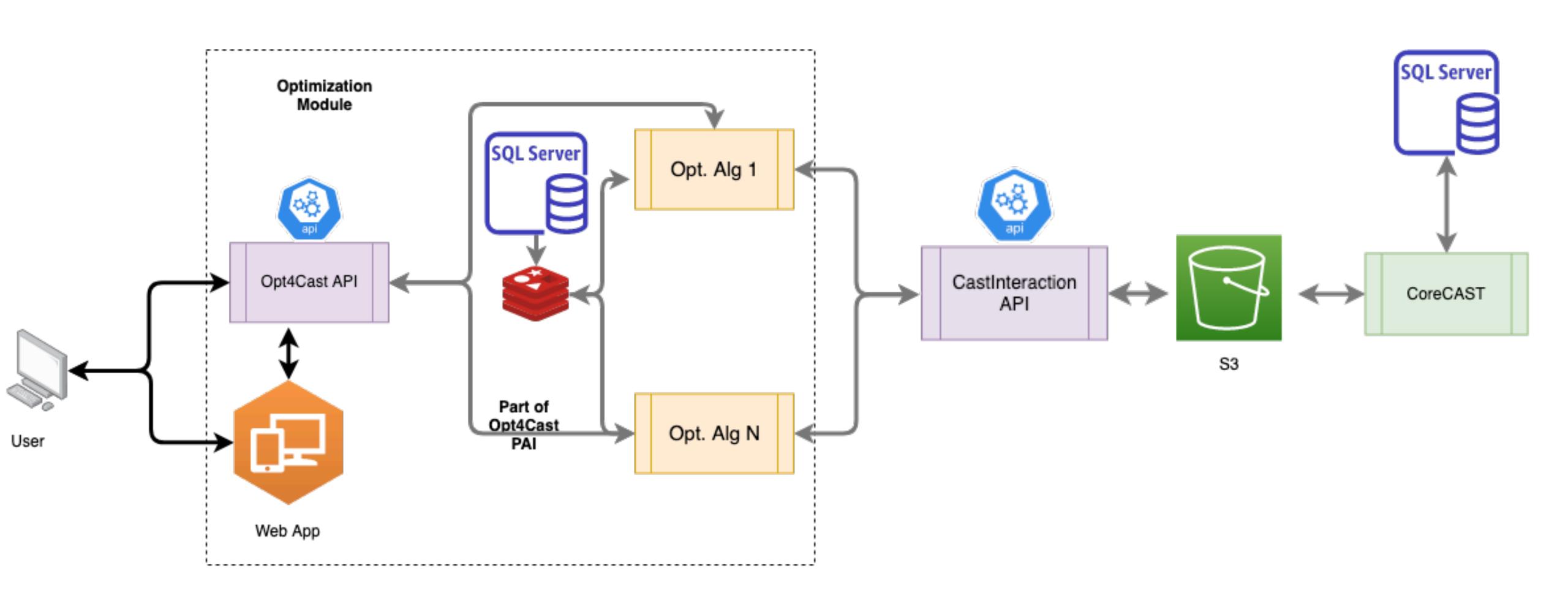




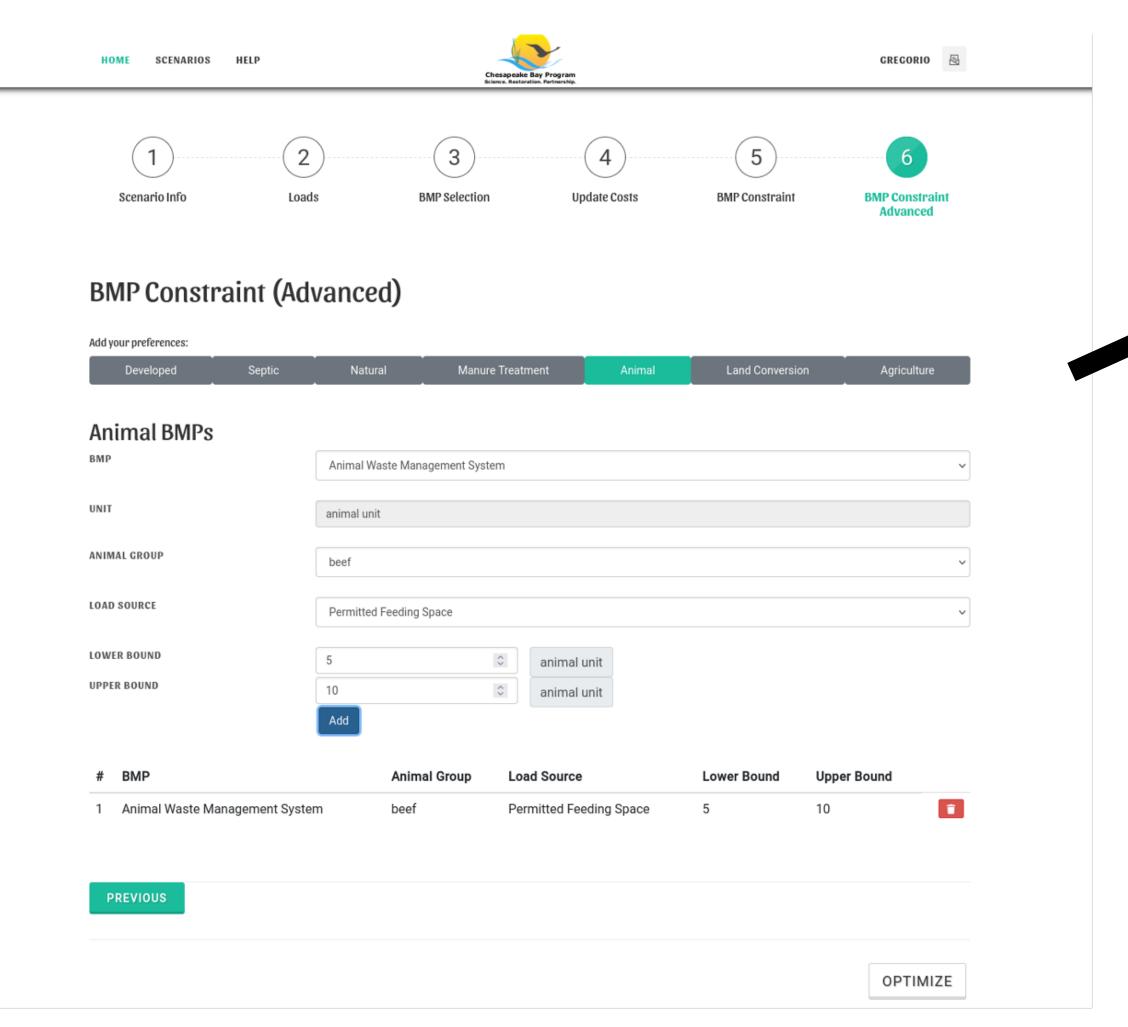


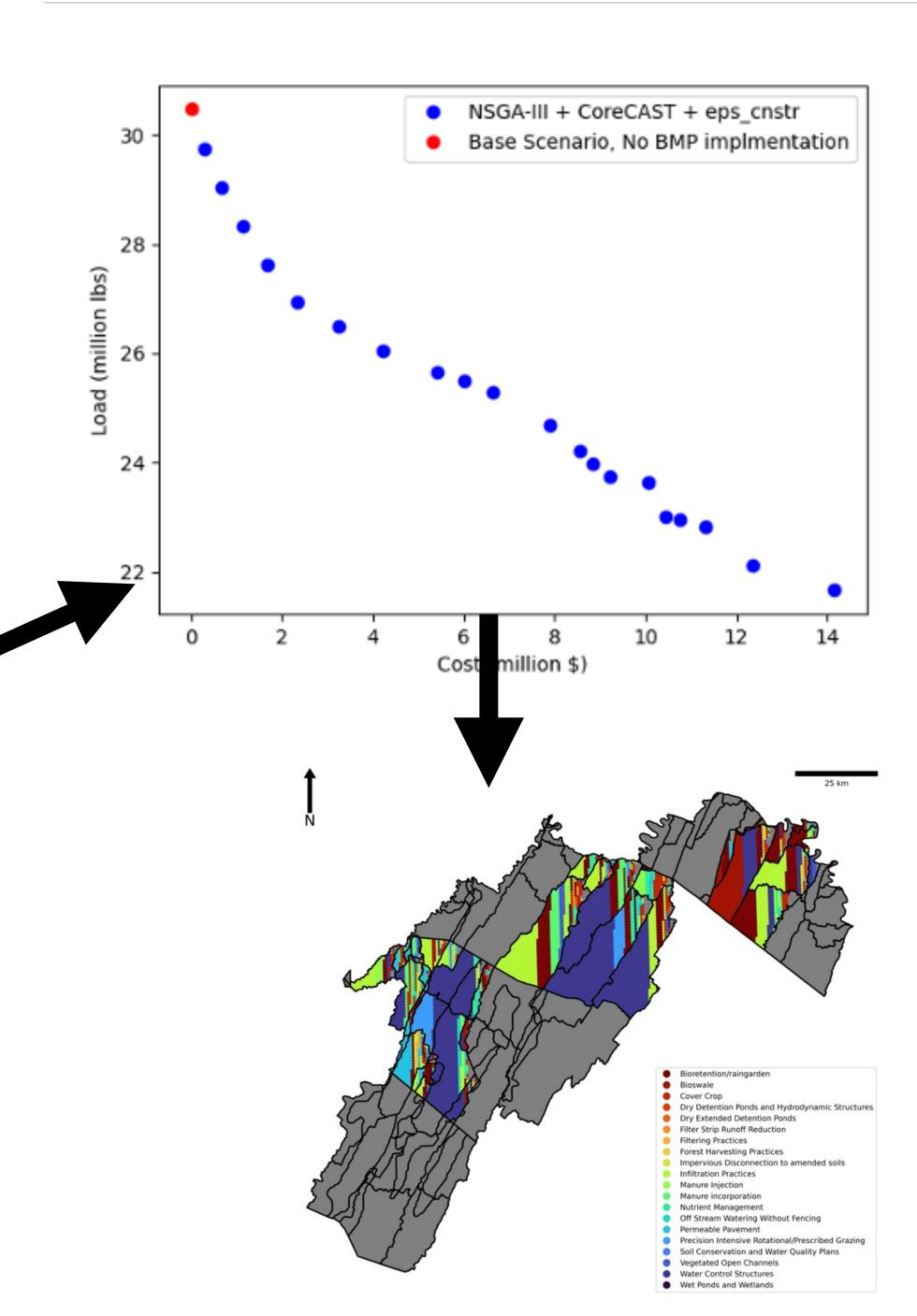




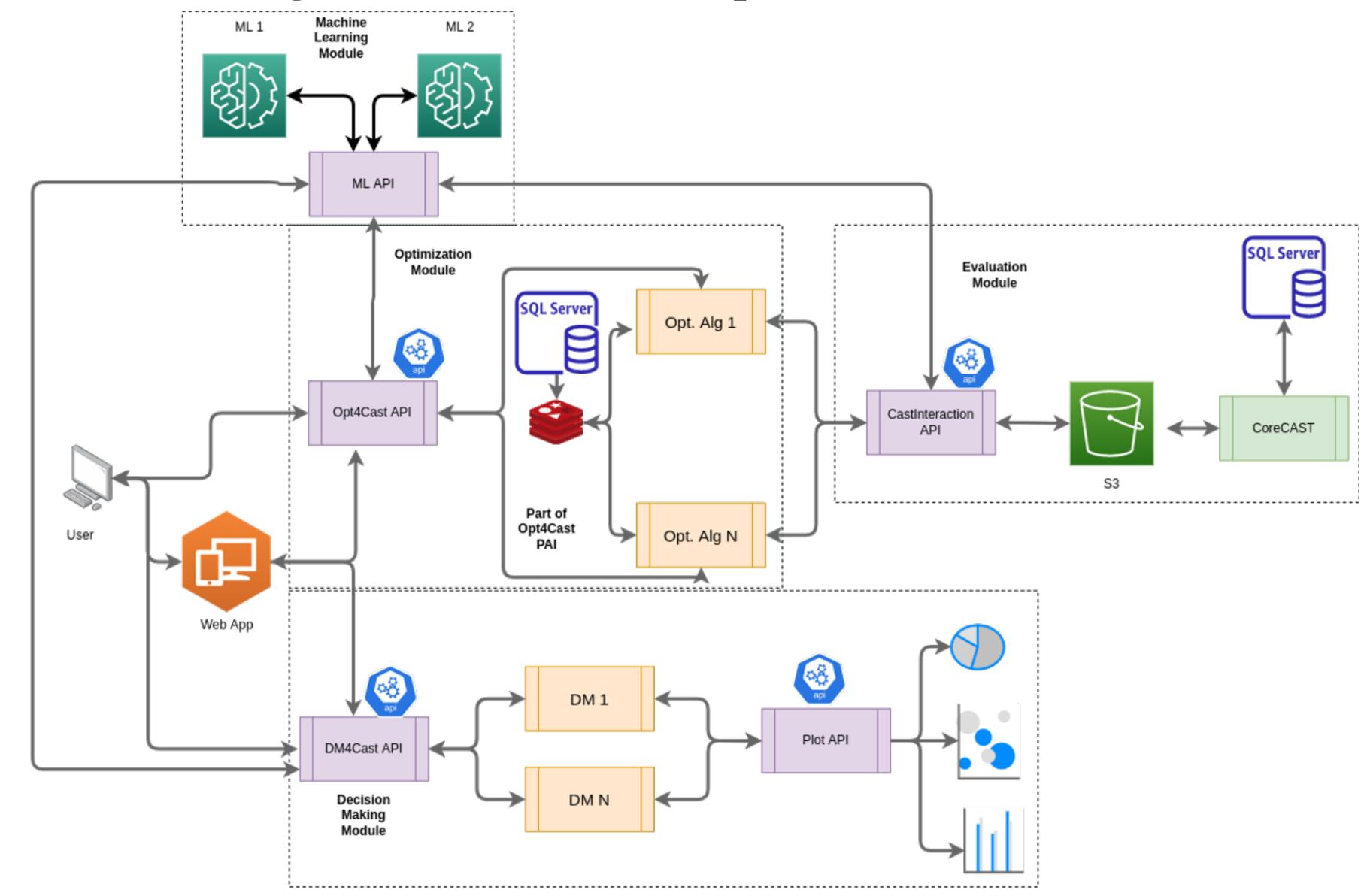


Project Progress Phase II

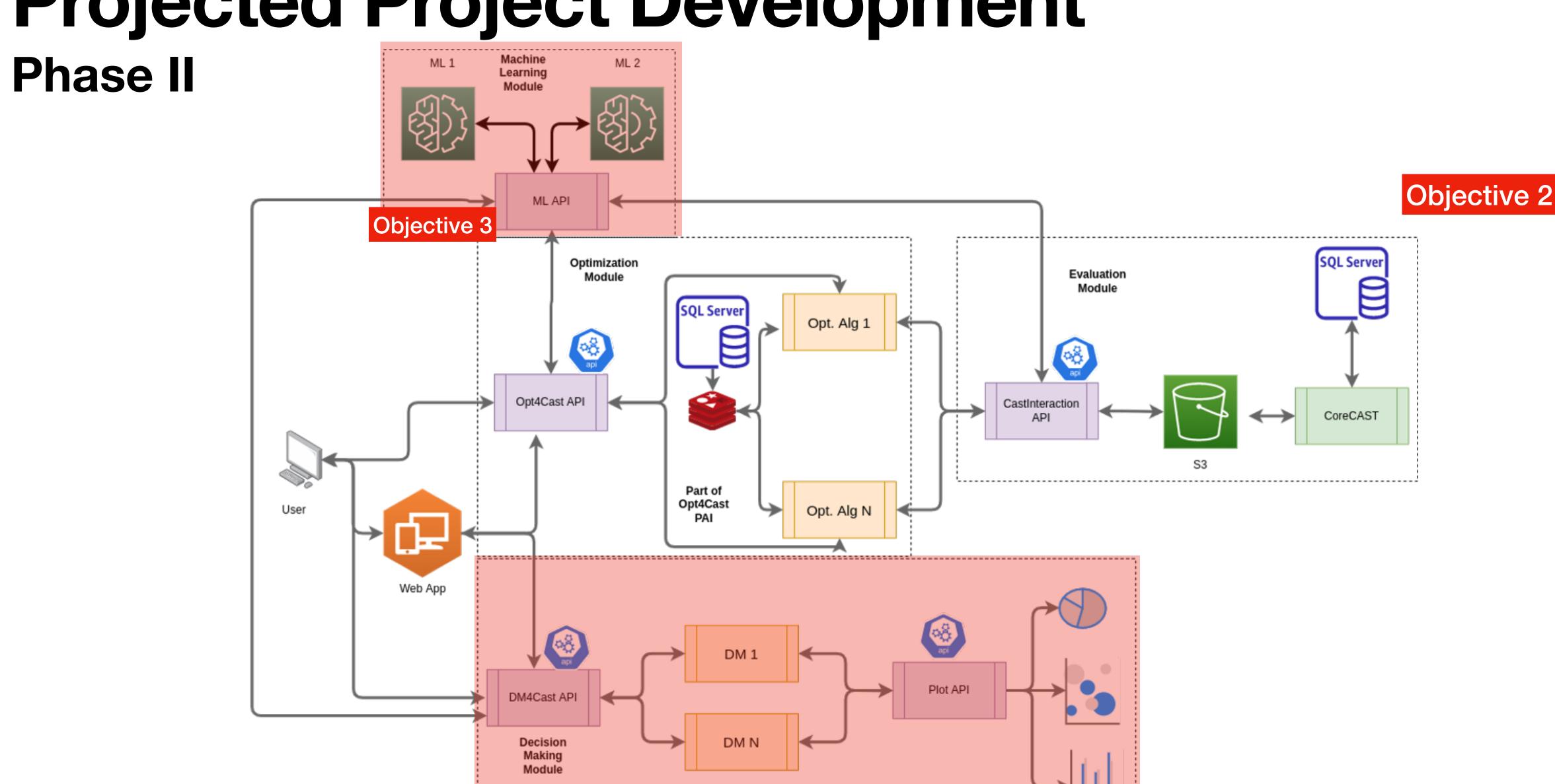




Projected Project Development

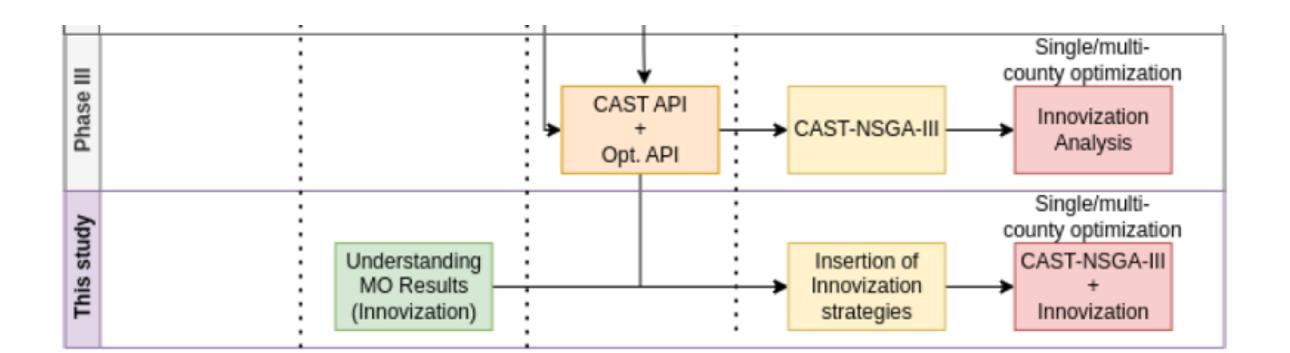


Projected Project Development



Phase III

What are the benefits of optimization?



- Identify the best solutions for the problem at hand.
- Generate knowledge to solve future problems.

Innovization Analysis

•What is innovization?

Learning from optimization results and introducing new ideas, products, and services different from the existing ones.

What Innovization can do to CBPO?

- Provide information for better decision-making for BMP selections (farmers)
- Identify the high priority areas for BMP implementation (regulators)
- Help with resources allocation (policymakers)

Innovization Experiment

BMP Selection ranking methodology based on Land use

•Overall goal: learn from optimization results to:

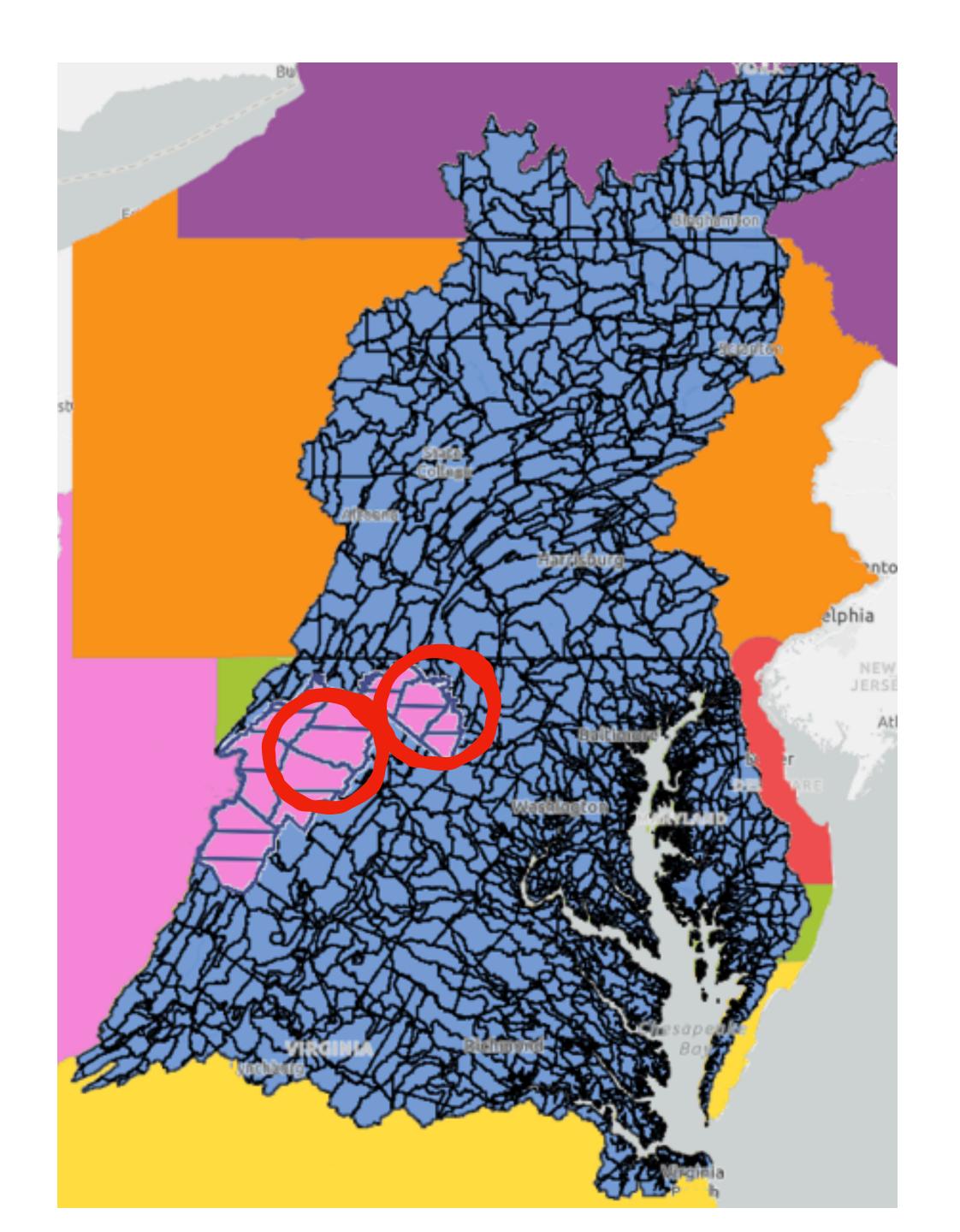
A.Examine different ranking methodologies to **identify the top BMPs**, B.Identify the **similarities and differences** between top-ranked BMPs, C.Provide recommendations to **improve the optimization process**.

Innovization experiment

BMP Selection ranking methodology based on Land use

In West Virginia, we identified the top two counties with the highest areas of urban and agricultural land uses.

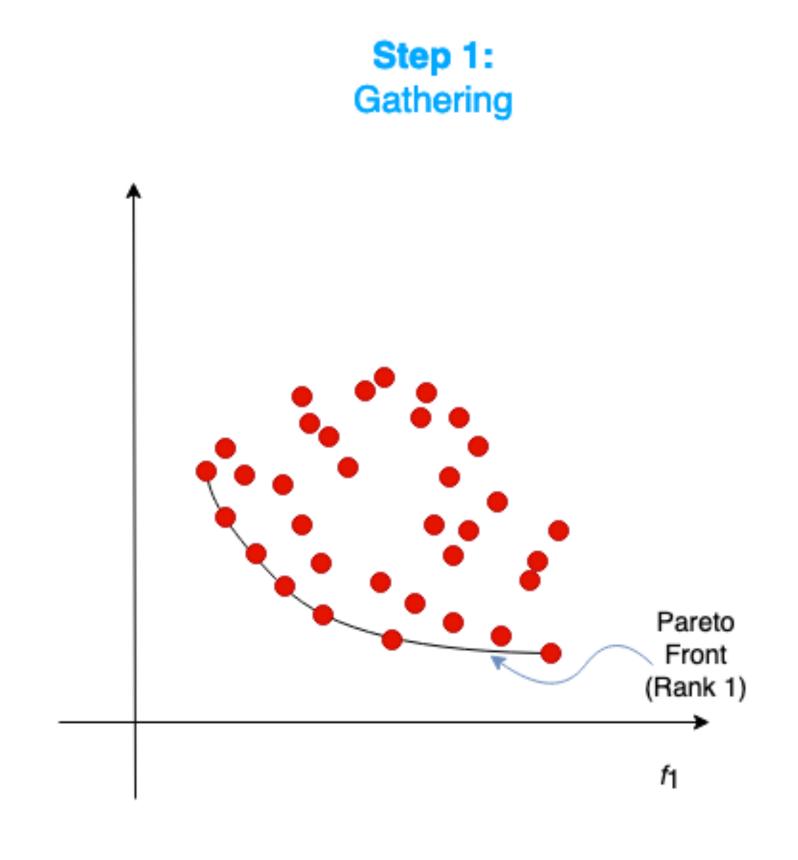
- (Berkeley and Mineral): Urban dominated
- •(Jefferson and Hardy): Agricultural dominated



Innovization Experiment

BMP Selection ranking methodology based on Land use

- We performed 11 runs of our CoreCAST-optimization algorithm.
- Each run evaluated 3,000 scenarios (1,000 scenarios epsilon constraint +. 2000 scenarios NSGA-III
- The output solution of each execution consists of 20 solutions.
- We gathered all 220 solutions (from the 11 runs)

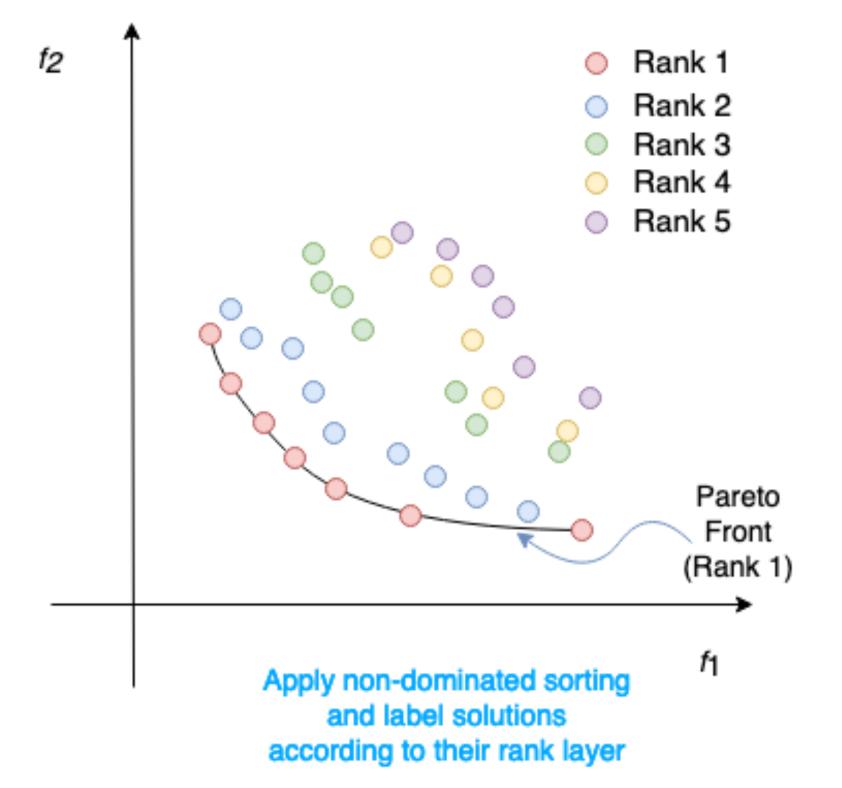


Innovization Experiment

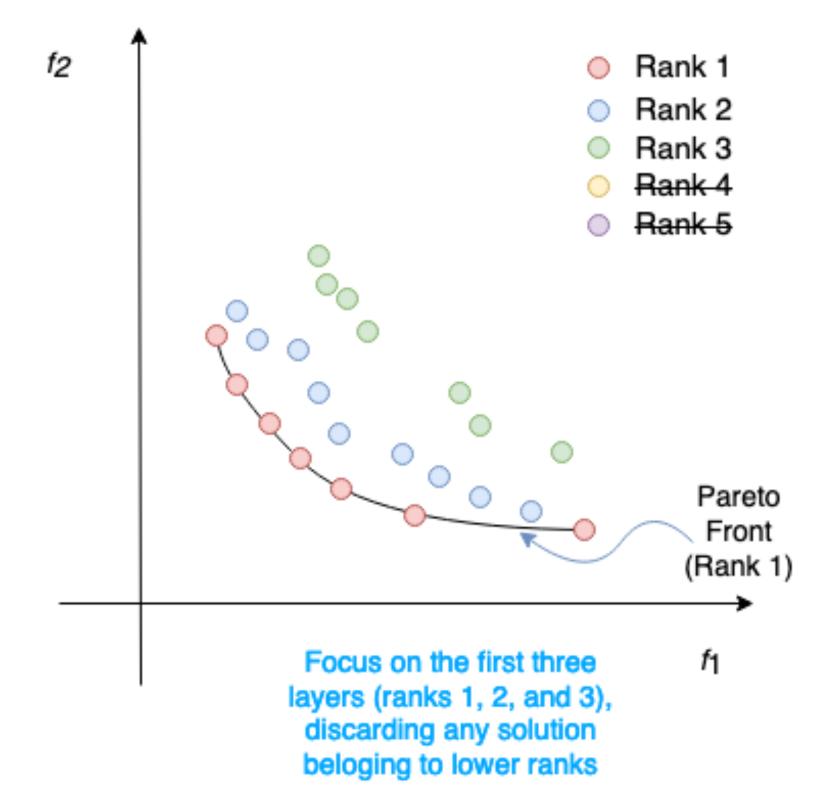
BMP Selection ranking methodology based on Land use

 We filtered the solutions regarding non-dominance layers

Step 2:
Apply non-dominated sorting



Step 3: Discard lower ranks



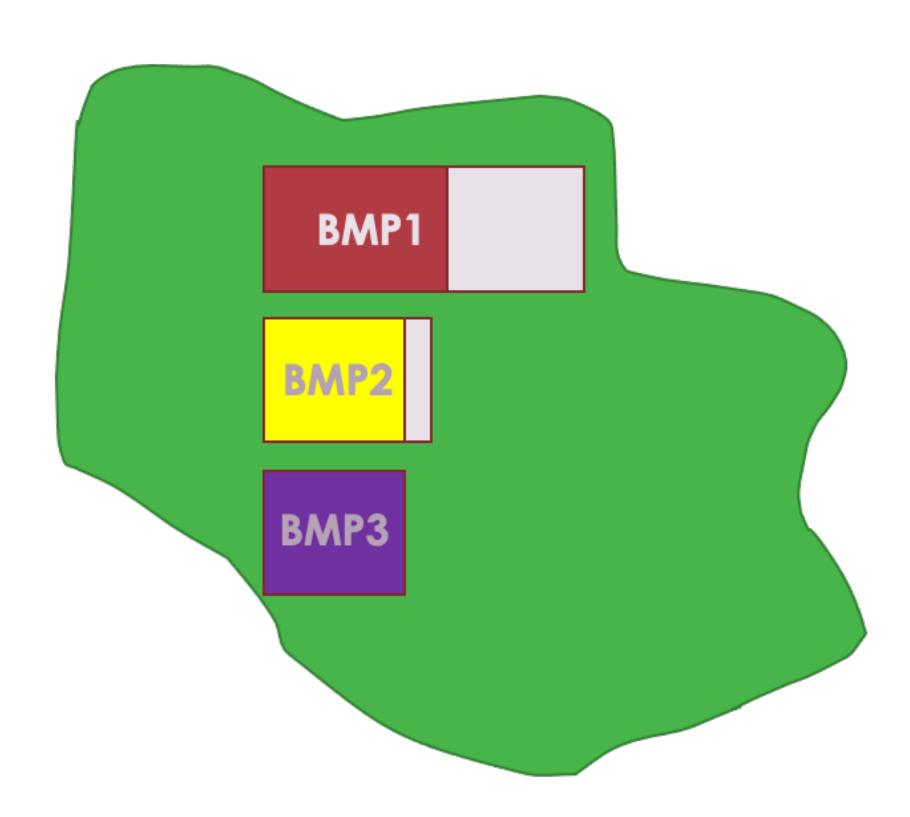
BMP Selection ranking methodology based on Land use

Three ranking strategies from the combined Pareto front:

- Strategy 1: Rank the top BMPs based on the implementation acreages;
- Strategy 2:
- Strategy 3:

Ranking methodology 1:





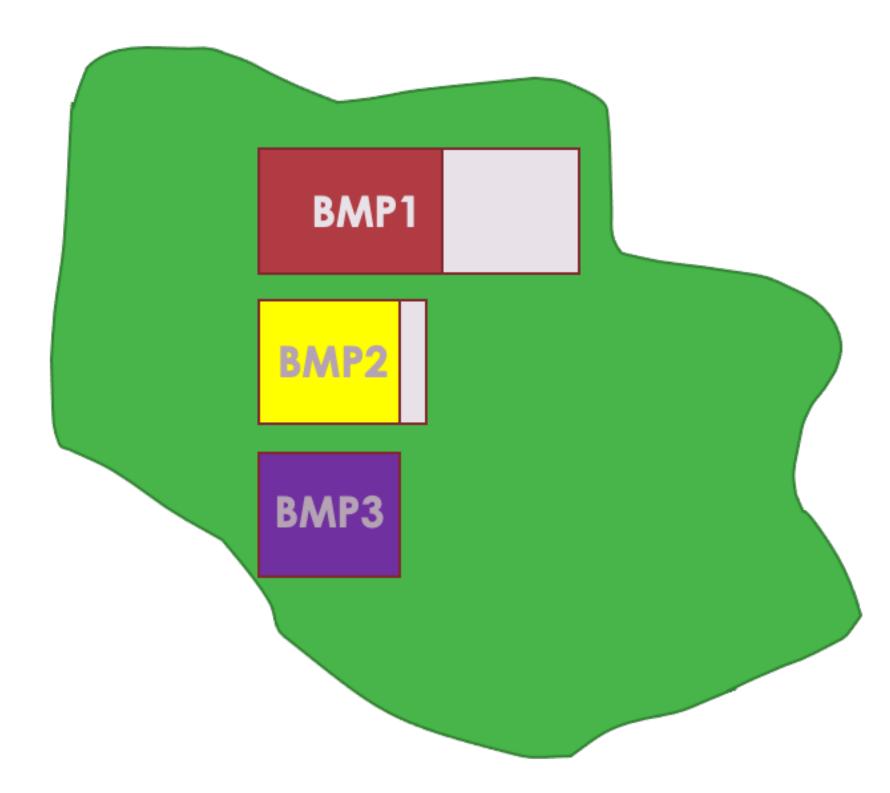
BMP Selection ranking methodology based on Land use

Three ranking strategies from the combined Pareto front:

- Strategy 1:
- **Strategy 2:** Rank the top BMPs based on the percentage of **maximum allowable acreages**;
- Strategy 3:

Ranking methodology 2:

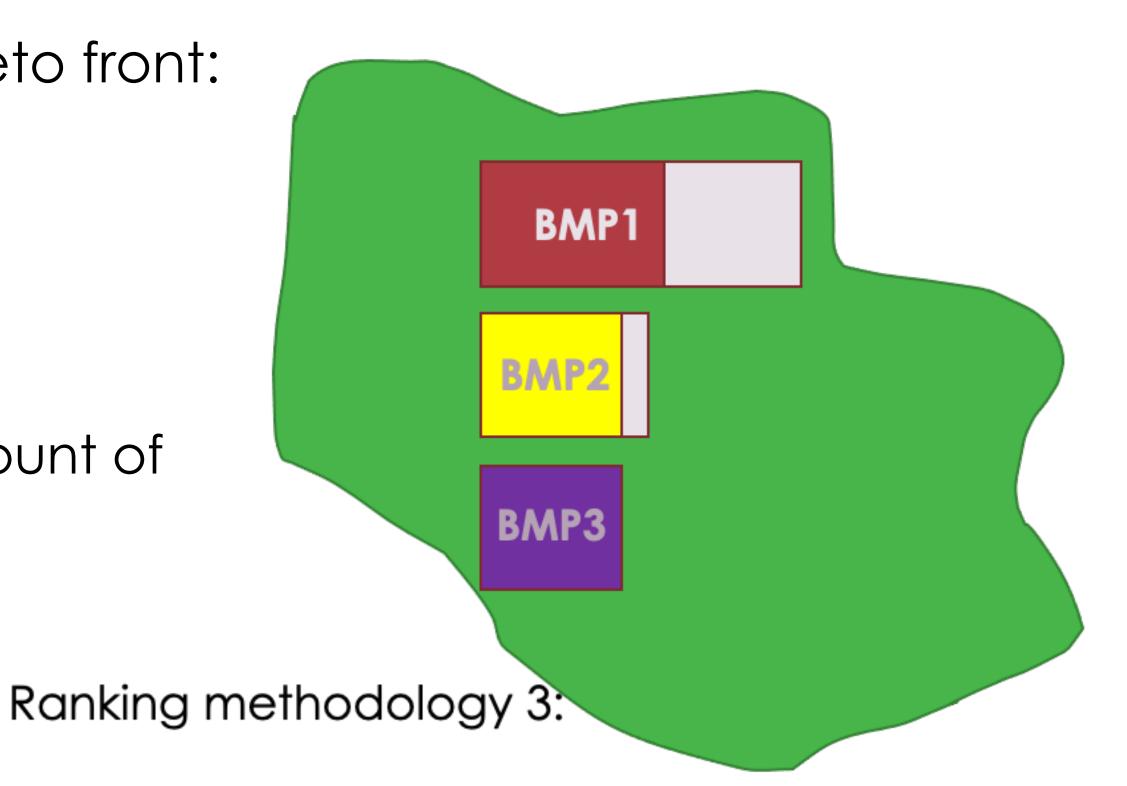
BMP3 BMP2 BMP1



BMP Selection ranking methodology based on Land use

Three ranking strategies from the combined Pareto front:

- Strategy 1:
- Strategy 2:
- Strategy 3: Rank the top BMPs based on the amount of nitrogen reduction per dollar spent.



BMP2 (\$12/lb N) BMP3 (\$15/lb N) BMP1(\$24/lb N)

BMP Selection ranking methodology based on Land use

Three ranking strategies from the combined Pareto front:

- Strategy 1: Rank the top BMPs based on the implementation acreages;
- **Strategy 2:** Rank the top BMPs based on the percentage of **maximum allowable acreages**;
- Strategy 3: Rank the top BMPs based on the amount of nitrogen reduction per dollar spent.

Ranking methodology 1: Ranking methodology 2: Ranking methodology 3:

BMP1

BMP2

BMP3

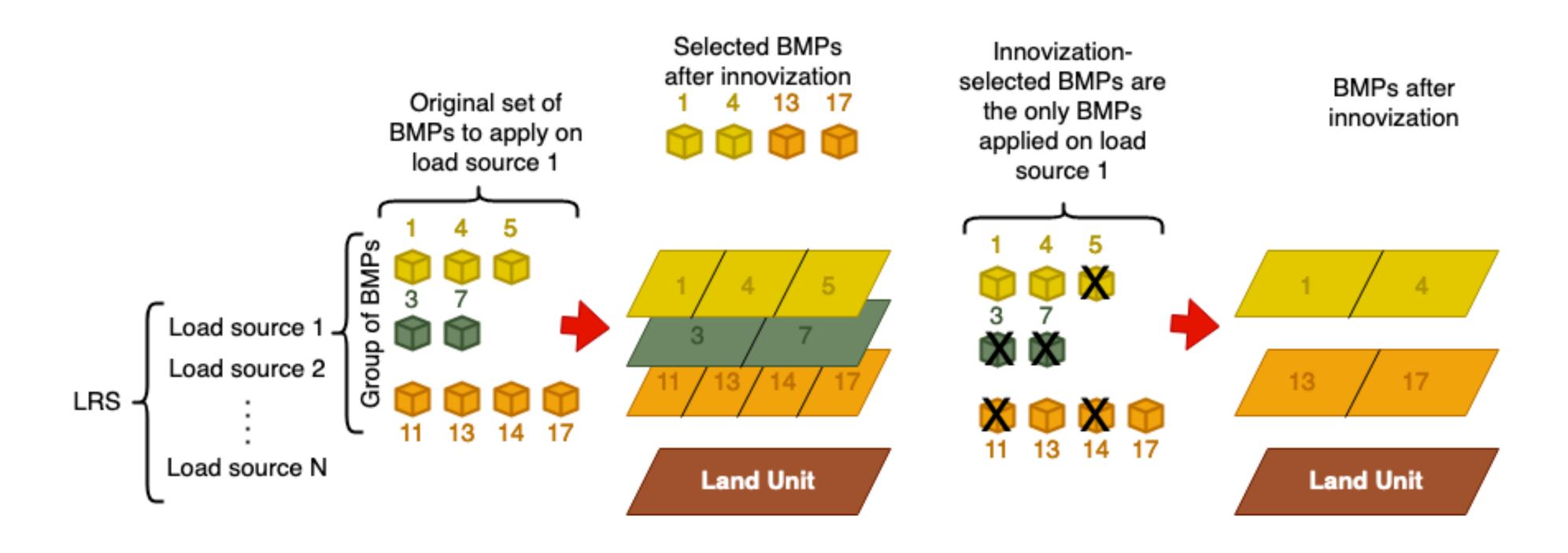
 BMP1
 BMP3
 BMP2 (\$12/lb N)

 BMP2
 BMP3 (\$15/lb N)

 BMP3
 BMP1 (\$24/lb N)

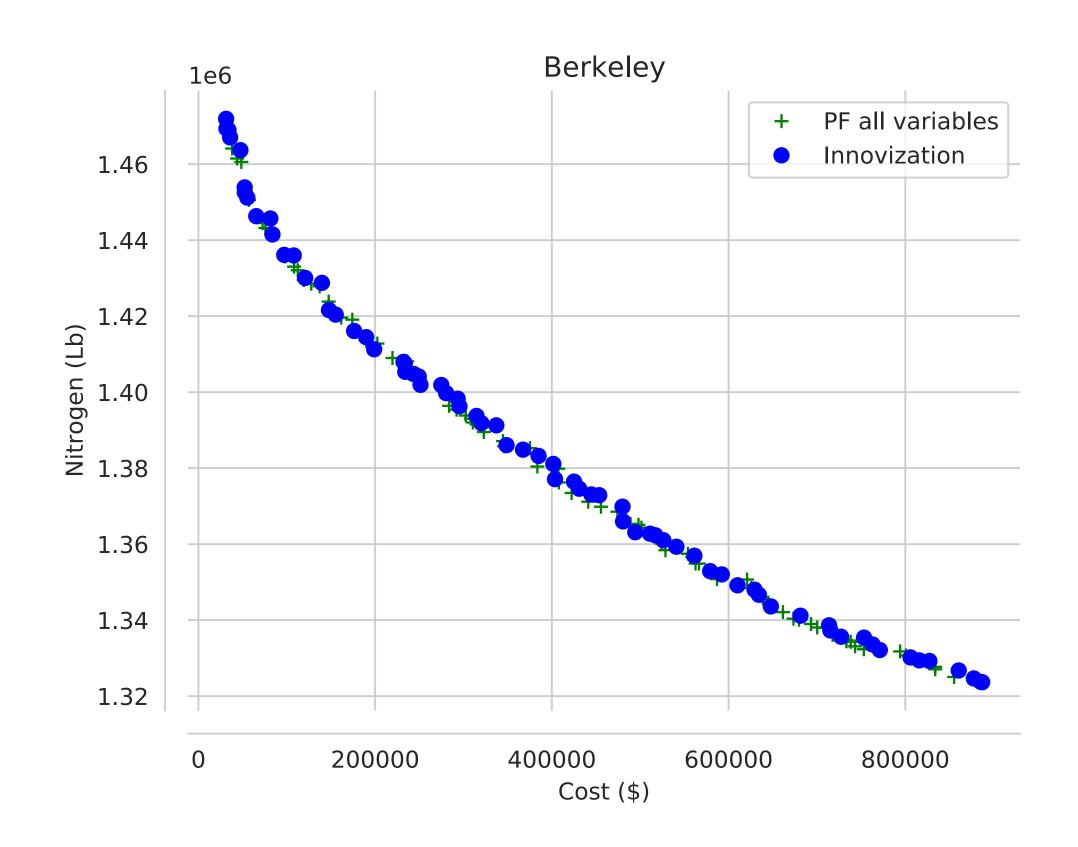
Re-optimization

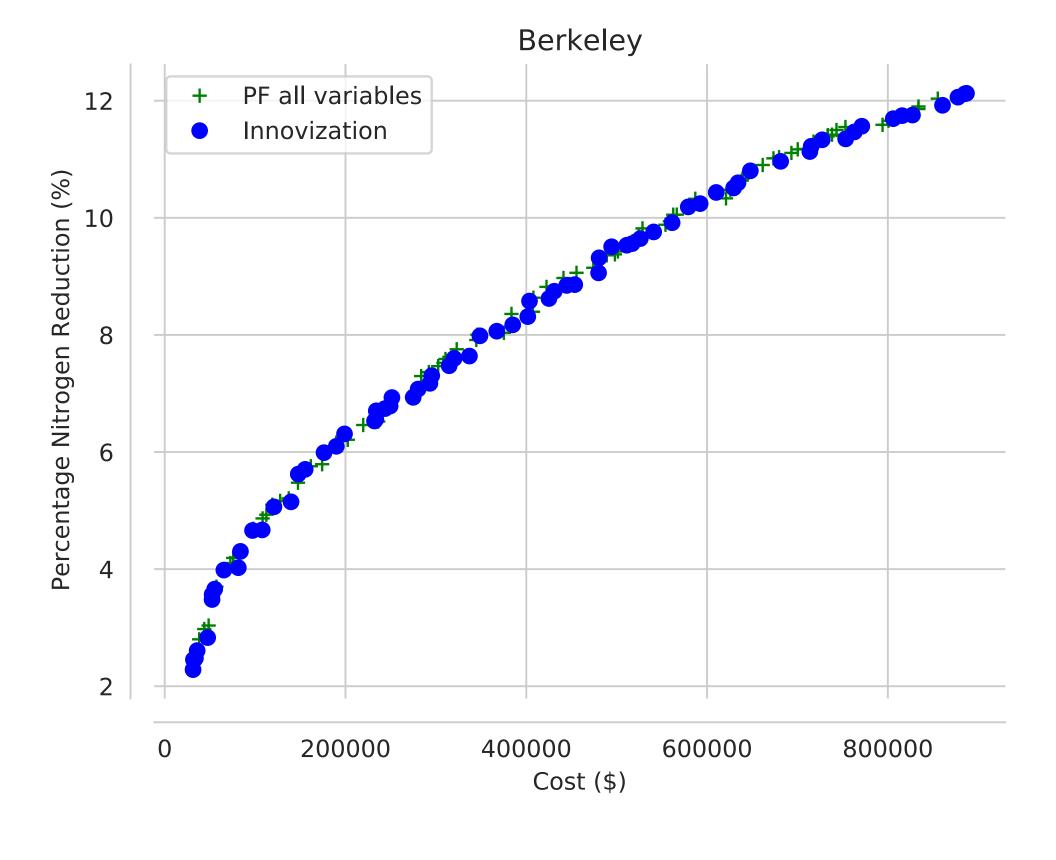
Example of method to use a re-optimization-based innovization



Results for Berkeley County

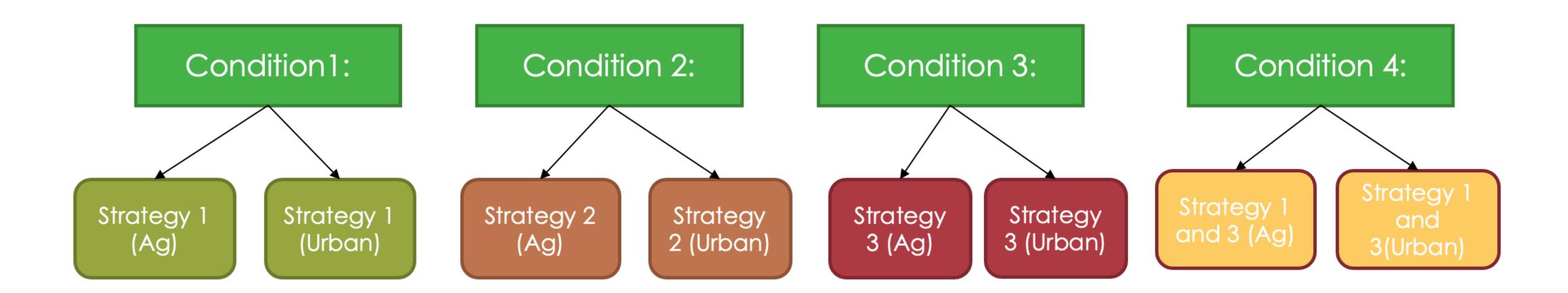
Variable reduction of 30%





Four Conditions for Re-Optimization

- Strategy 1: Implementation area
- Strategy 2: Max implementation area (available area)
- Strategy 3: Dollar / Nitrogen

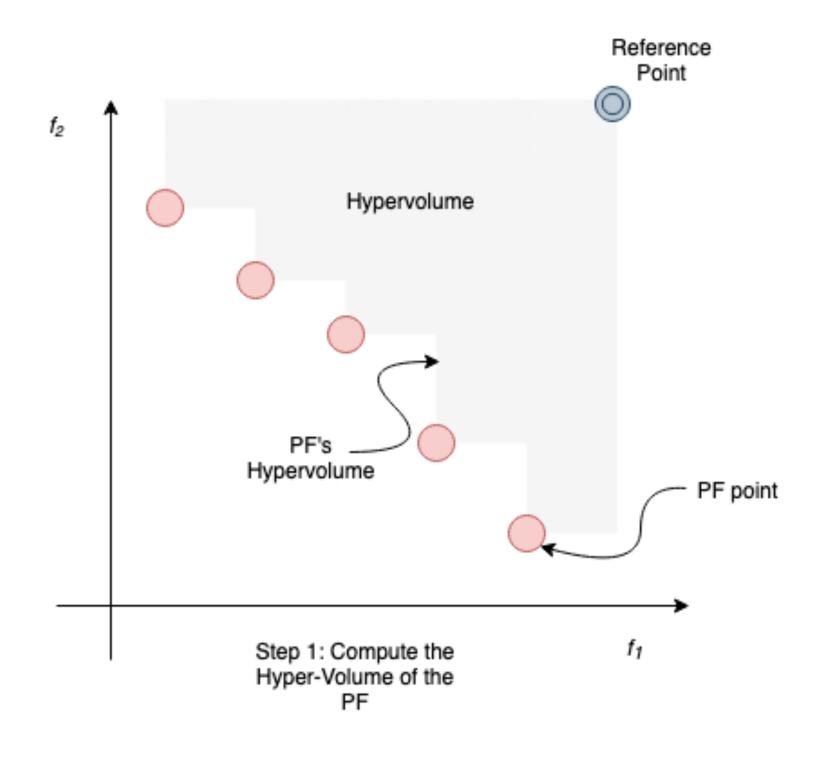


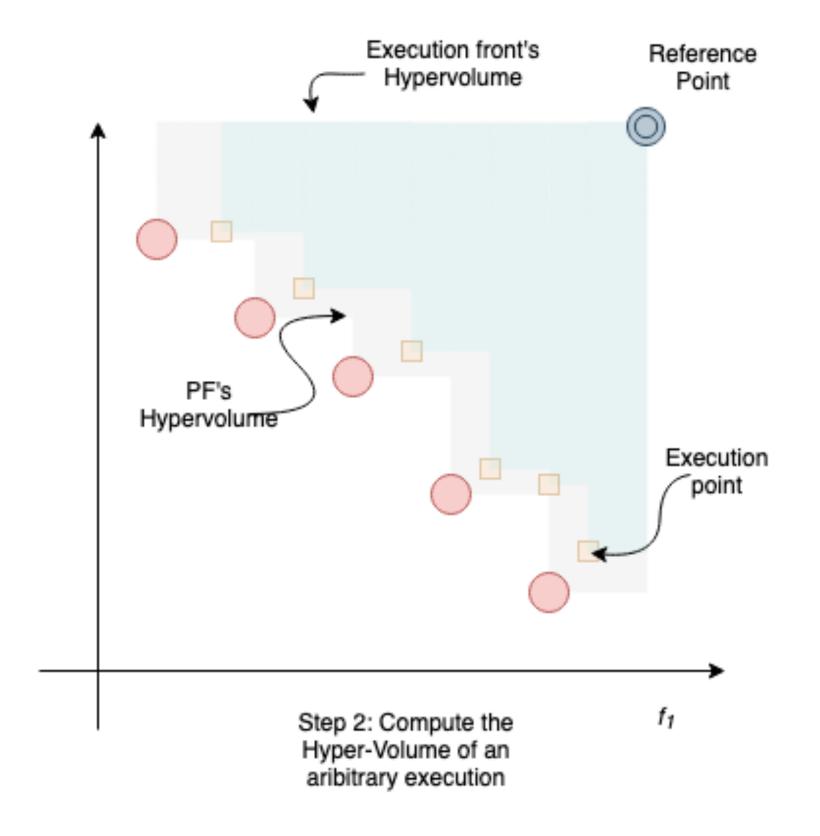
Variable reduction.

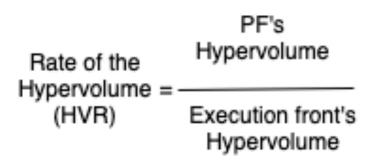
		% of all		% of all		% of all		% of all
	Berkeley	vars	Mineral	vars	Hardy	vars	Jefferson	vars
All								
Variables	14090	100	20260	100	18607	100	12303	100
Condition 1	537	3.81	810	4.00	777	4.18	488	3.97
Condition 2	654	4.64	990	4.89	946	5.08	592	4.81
Condition 3	681	4.83	1020	5.03	959	5.15	600	4.88
Condition 4	717	5.09	1080	5.33	1011	5.43	632	5.14

Performance Measure

Ratio of the Hypervolume



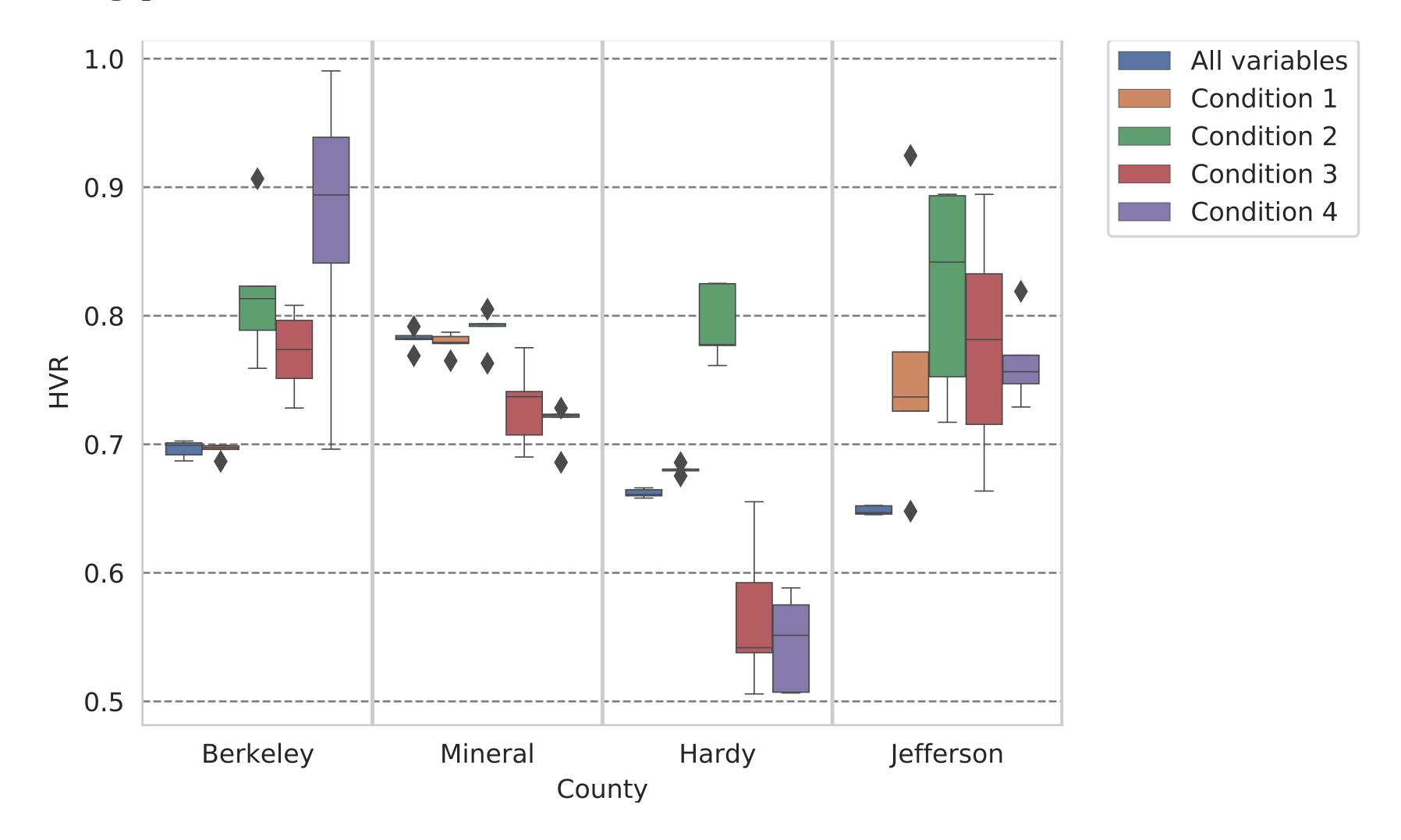




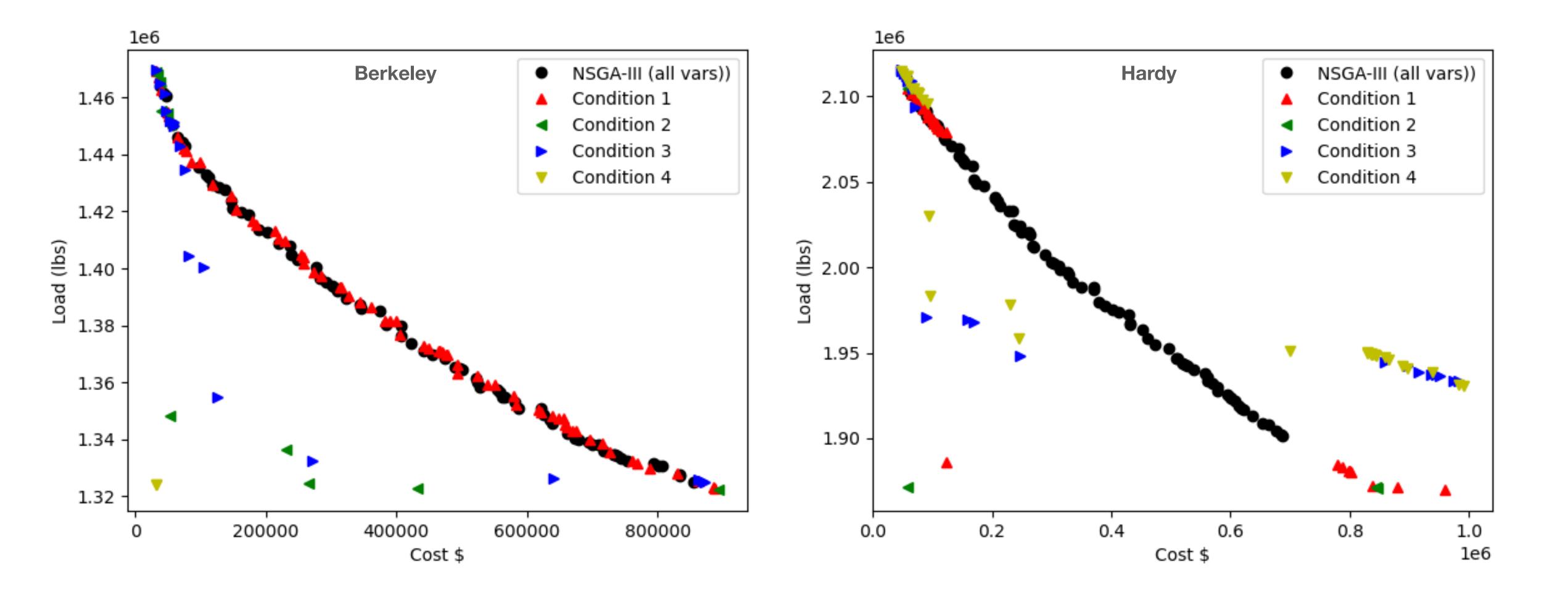
Step 3: Compute the HVR

Performance Measure

Ratio of the Hypervolume

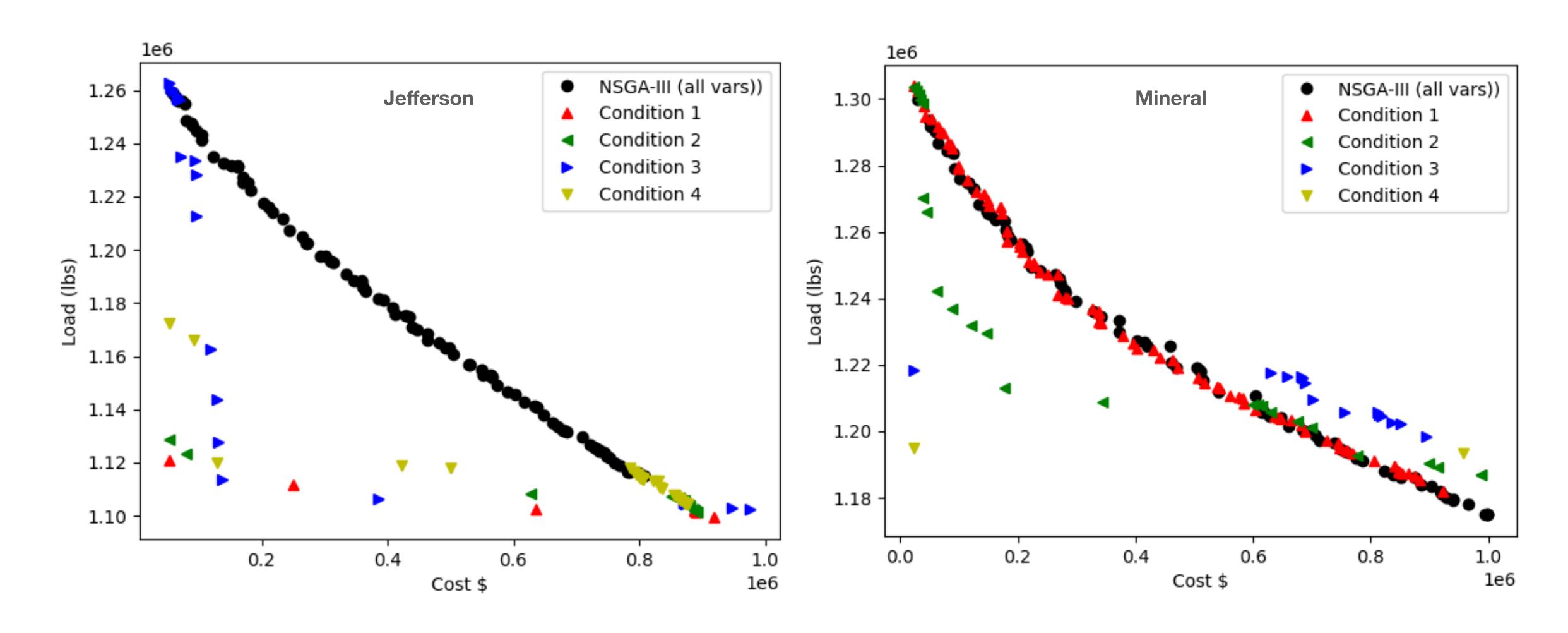


Re-optimization-based Innovization Berkeley and Hardy

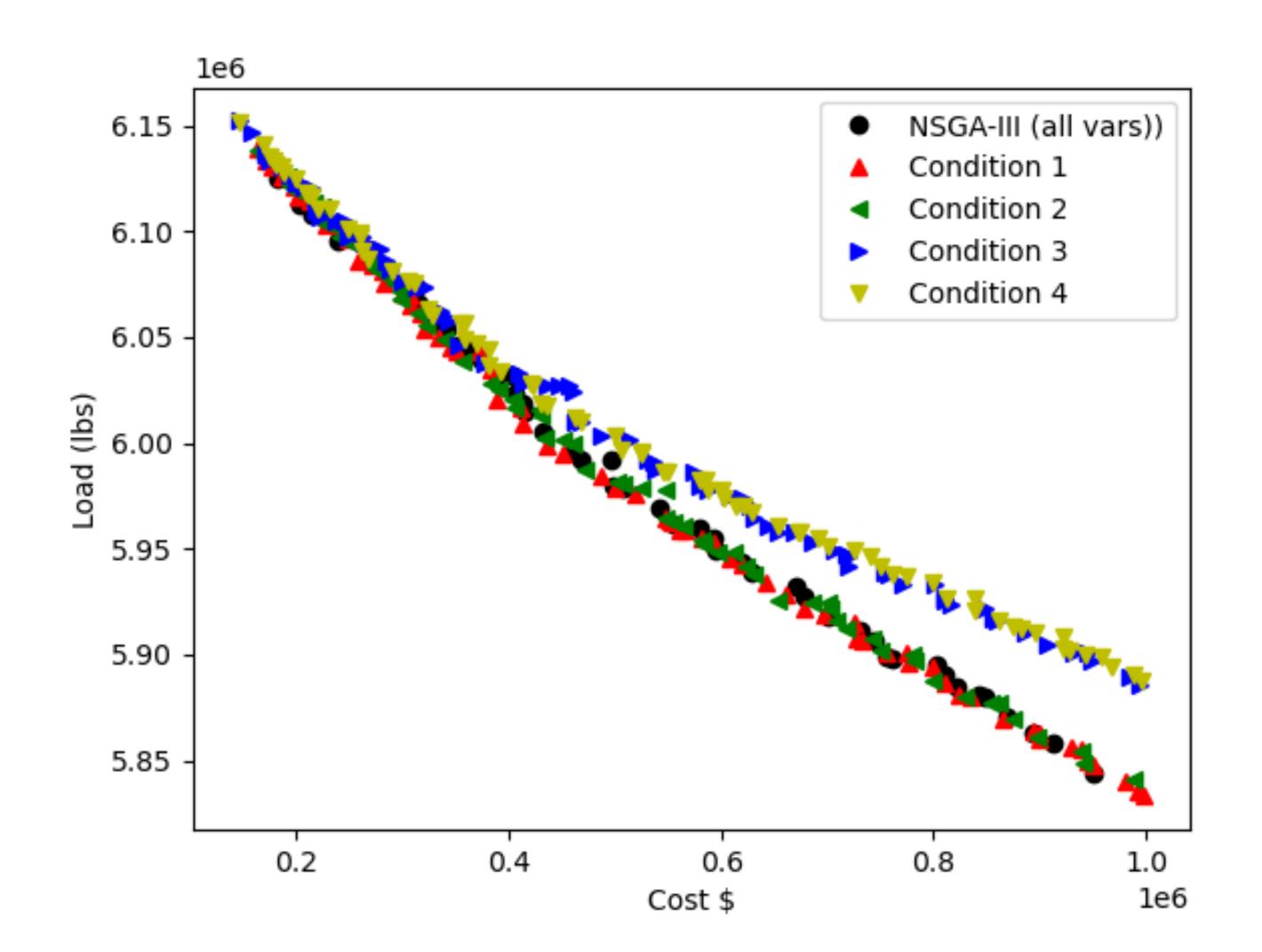


Re-optimization-based Innovization

Jefferson and Mineral



Multi-County Results



Submission of Papers (this quarter)

- Gregorio Toscano, Juan Hernández, Julian Blank, Pouyan Nejadhashemi, Kalyanmoy Deb, and Lewis Linker. Utilizing Innovization to Solve Large-scale Multi-objective Chesapeake Bay Watershed Problem Efficiently. Submitted to the 2023 Congress on Evolutionary Computation (submitted in January 2023)
- Gregorio Toscano, Hoda Razavi, Pouyan Nejadhashemi, Kalyanmoy Deb, and Lewis Linker, 2023. Large-scale Multi-objective Optimization for Watershed Planning and Assessment. IEEE Transactions on Systems, Man, and Cybernetics: Systems (submitted in March 2023).

Conclusions and Future Work

- We have developed multi-objective methods that accept users' preferences and find several solutions in a single run.
- Such a tool will help us reduce the time to evaluate and analyze our optimization algorithms.
- Innovization can help us to perform more efficient search.
- The results are promising, and we are planning to incorporate these results in the design of our future approaches.
- Replicate study with rest of BMPs.

Thank you!