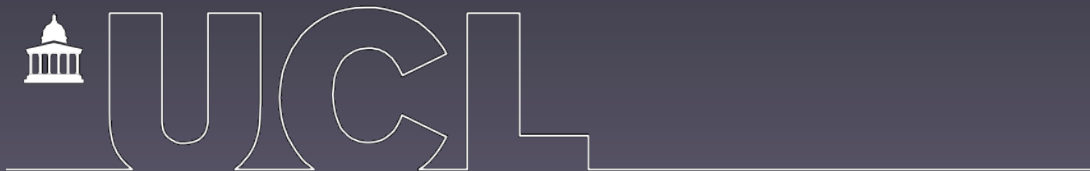


Search Based Requirements Selection and Optimisation

Yuanyuan Zhang

CREST Centre

University College London



Agenda

Background

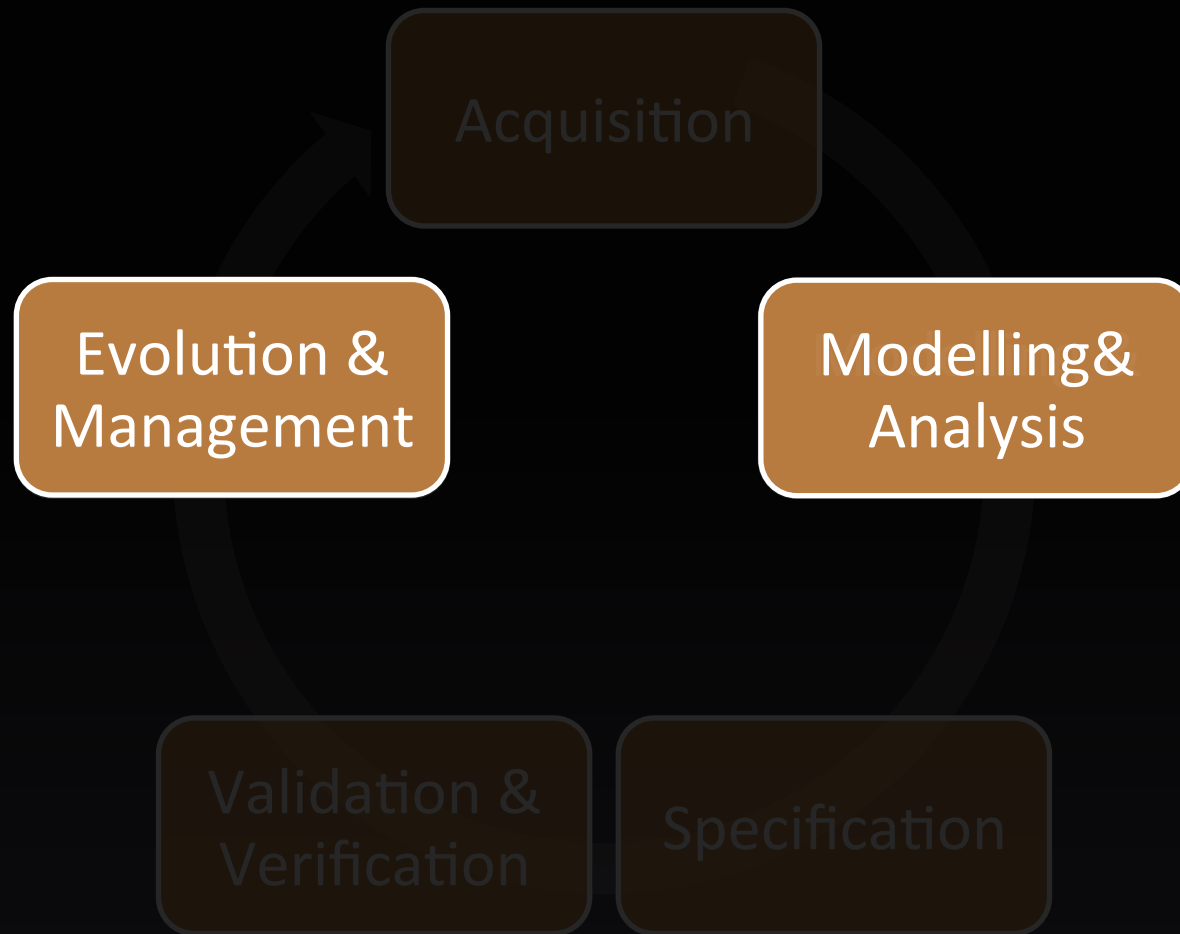
Problem

Solution

Empirical Study

Conclusion

Requirements Engineering Process



Requirements Selection & Optimisation

Task

Using prioritisation, visualisation, and optimisation techniques helps decision maker to select the optimal or near optimal subset from all possible requirements to be implemented.

Requirements Interaction Management

“the set of activities directed towards the discovery, management, and disposition of critical relationships among sets of requirements.”

Requirements Change



Unpredictable change



Predictable change

Problem

Requirements Selection



Background

Problem

Solution

Empirical Study

Conclusion

Problem

Requirements Selection



Background

Problem

Solution

Empirical Study

Conclusion

Problem

Requirements Selection



Background

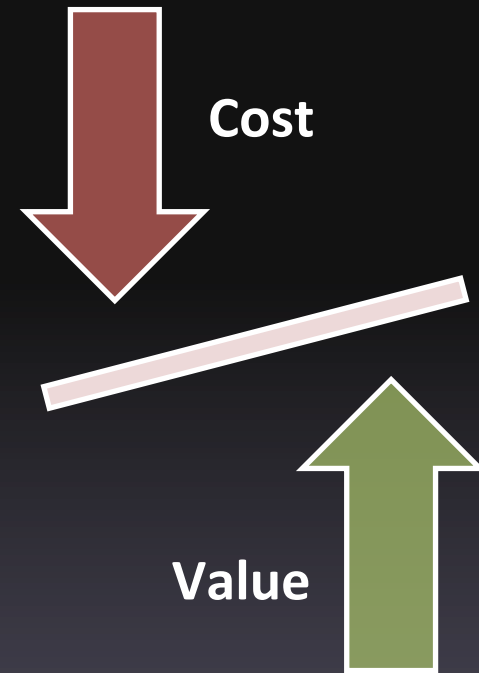
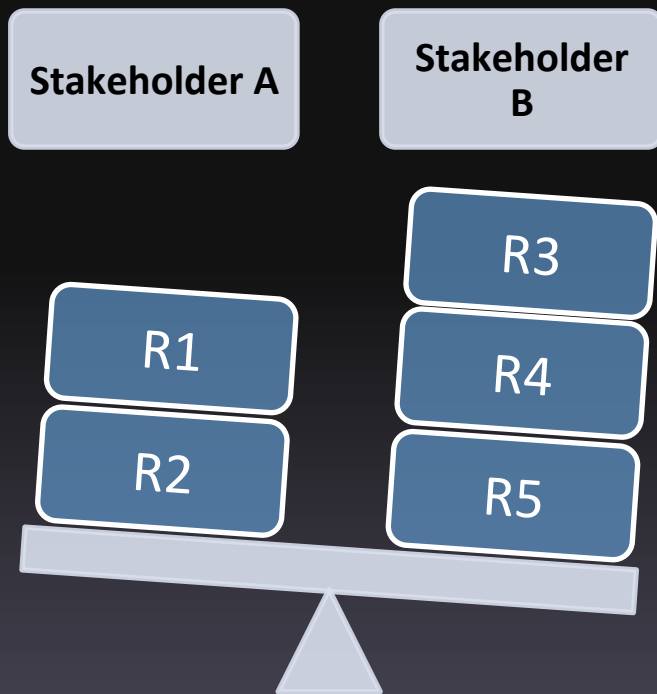
Problem

Solution

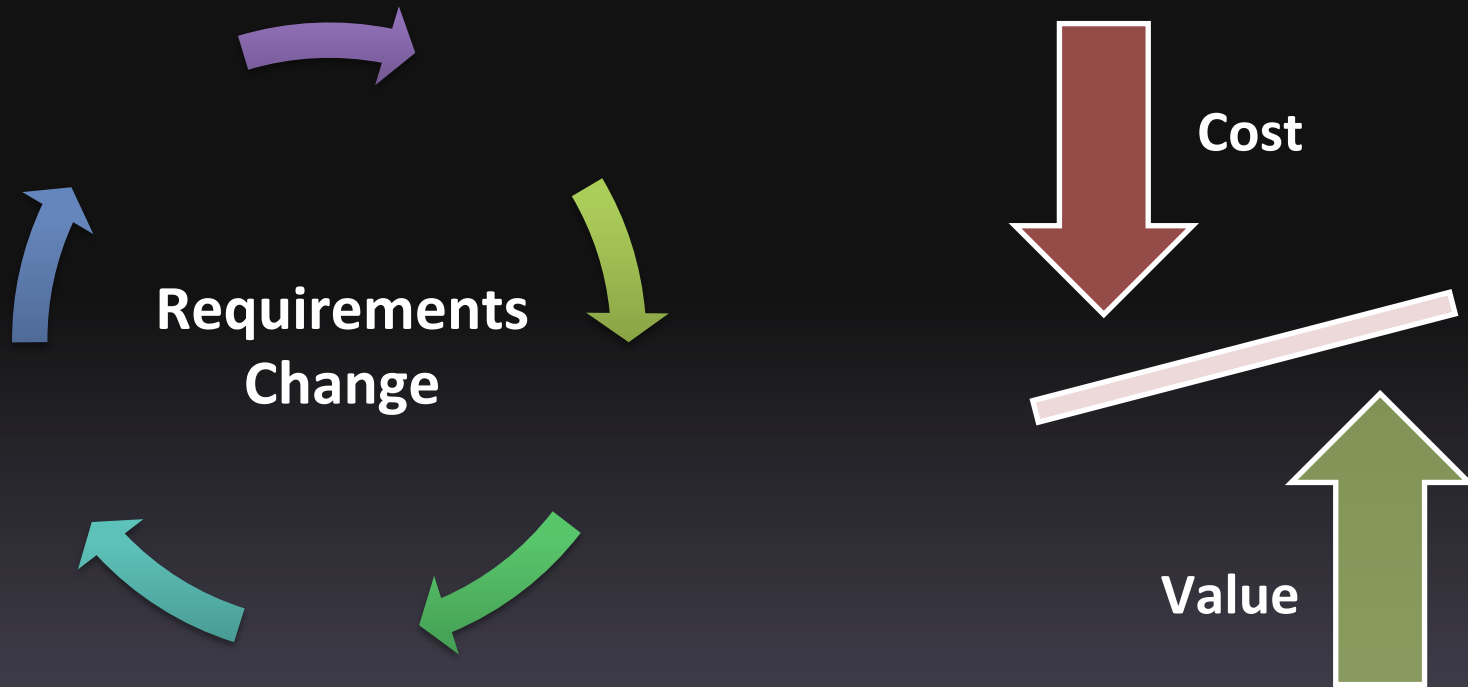
Empirical Study

Conclusion

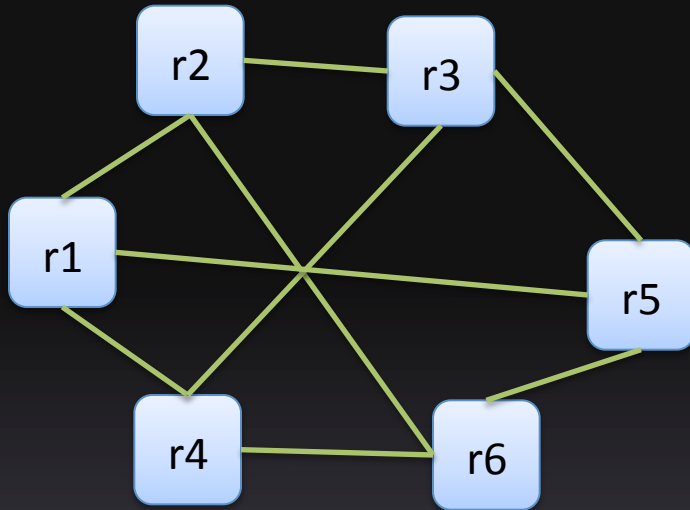
Goals



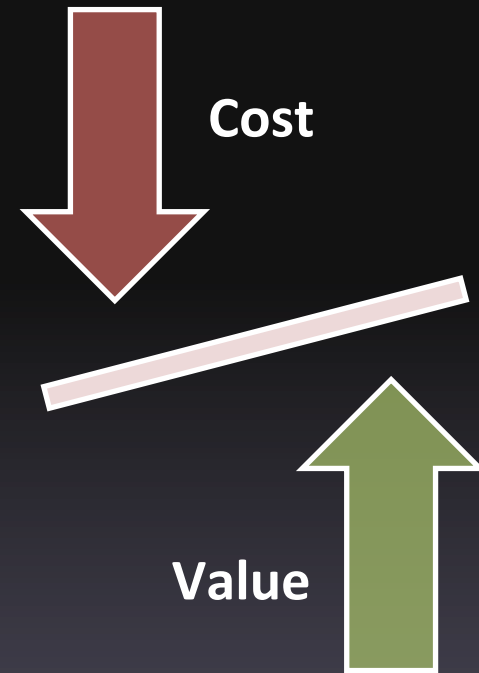
Goals



Goals



**Requirements
Dependency**



Search based Requirements Optimisation

Use of meta-heuristic algorithms to automate and optimise requirements selection process

- Choose appropriate representation of problem
- Define problem specific fitness function (to evaluate potential solutions)
- Use search based techniques to lead the search towards optimal points in the solution space

Why Search Based Approach?

Robustness

Scalability

Sensitivity analysis

Insight

Feedback & Explanation of results

...

Model

Stakeholder:

$$C = \{c_1, \dots, c_j, \dots, c_m\}$$



Weight:

$$Weight = \{w_1, \dots, w_j, \dots, w_m\}$$



Requirements:

$$R = \{r_1, \dots, r_i, \dots, r_n\}$$



Cost:

$$Cost = \{cost_1, \dots, cost_n\}$$



Model

- Each stakeholder c_j assigns a *value* to requirements r_i :

$$value(r_i, c_j)$$

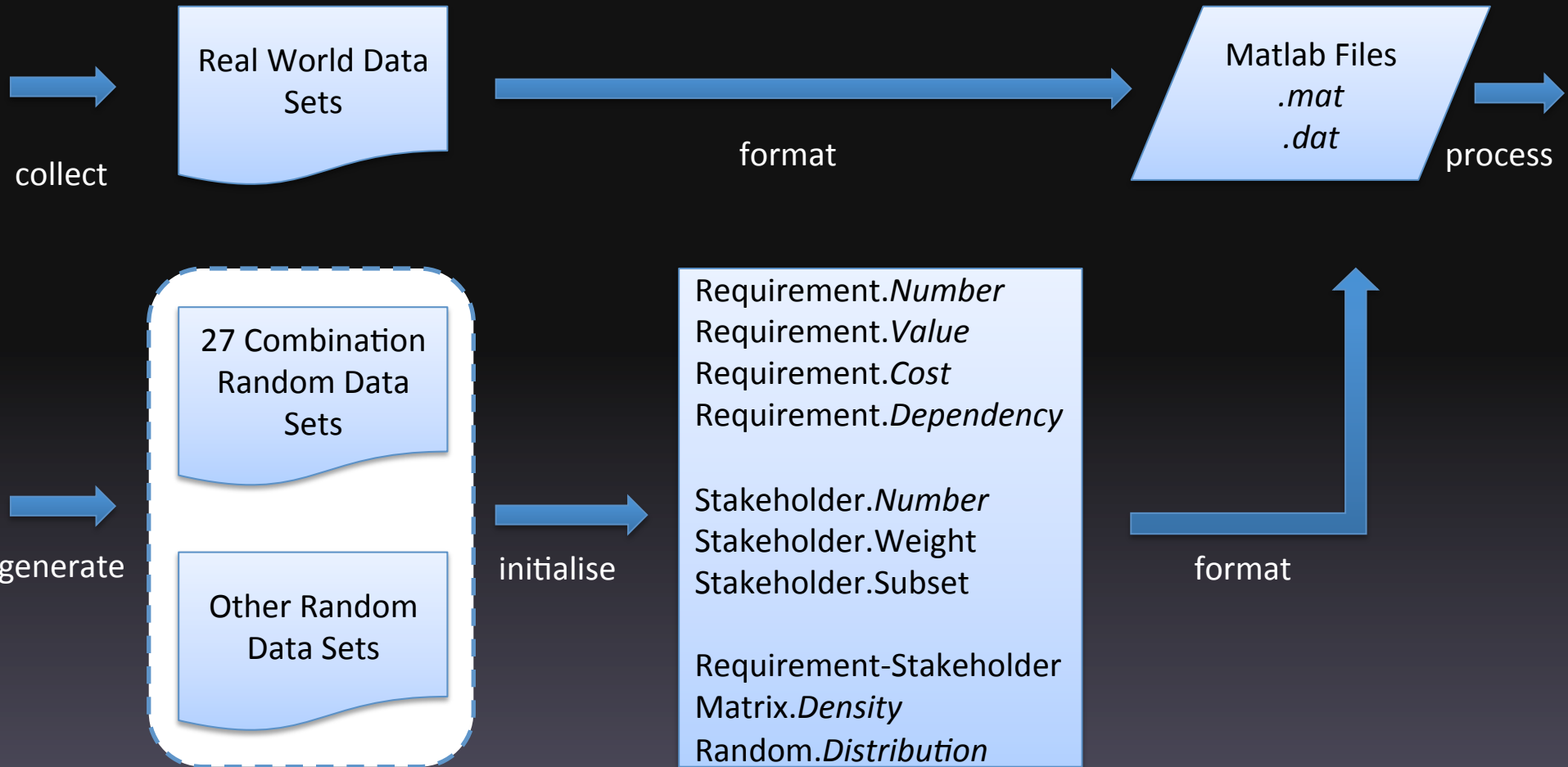
- Each stakeholder c_j has a subset of requirements that expect to be fulfilled denoted by R_j

$$R_j \subseteq R, \quad \forall r \in R_j \quad value(r, c_j) > 0$$

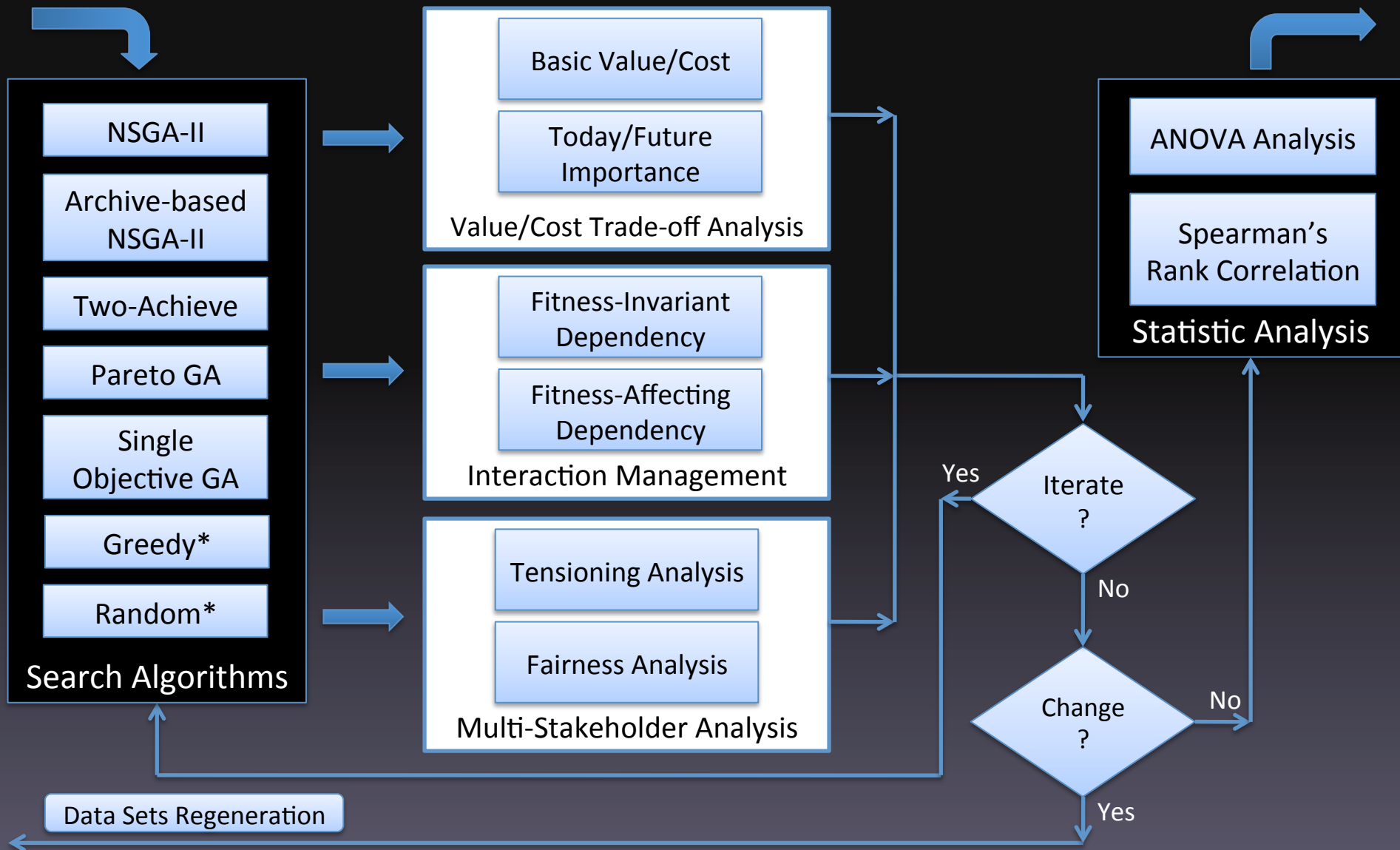
- The overall *score* of a given requirement r_i can be calculated by:

$$score_i = \sum_{j=1}^m w_j \cdot value(r_i, c_j)$$

Data Set Collection & Initialisation

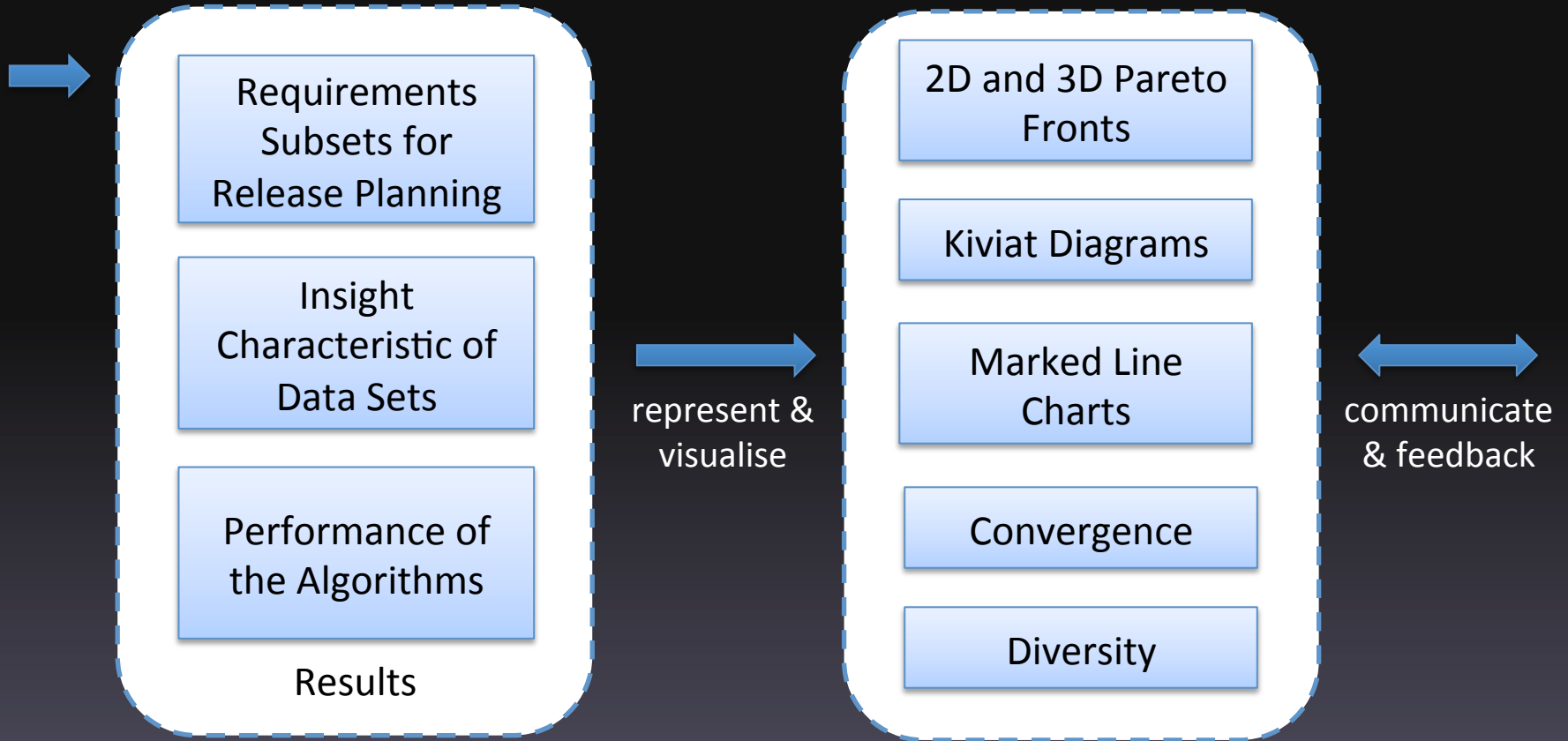


Requirements Selection Process



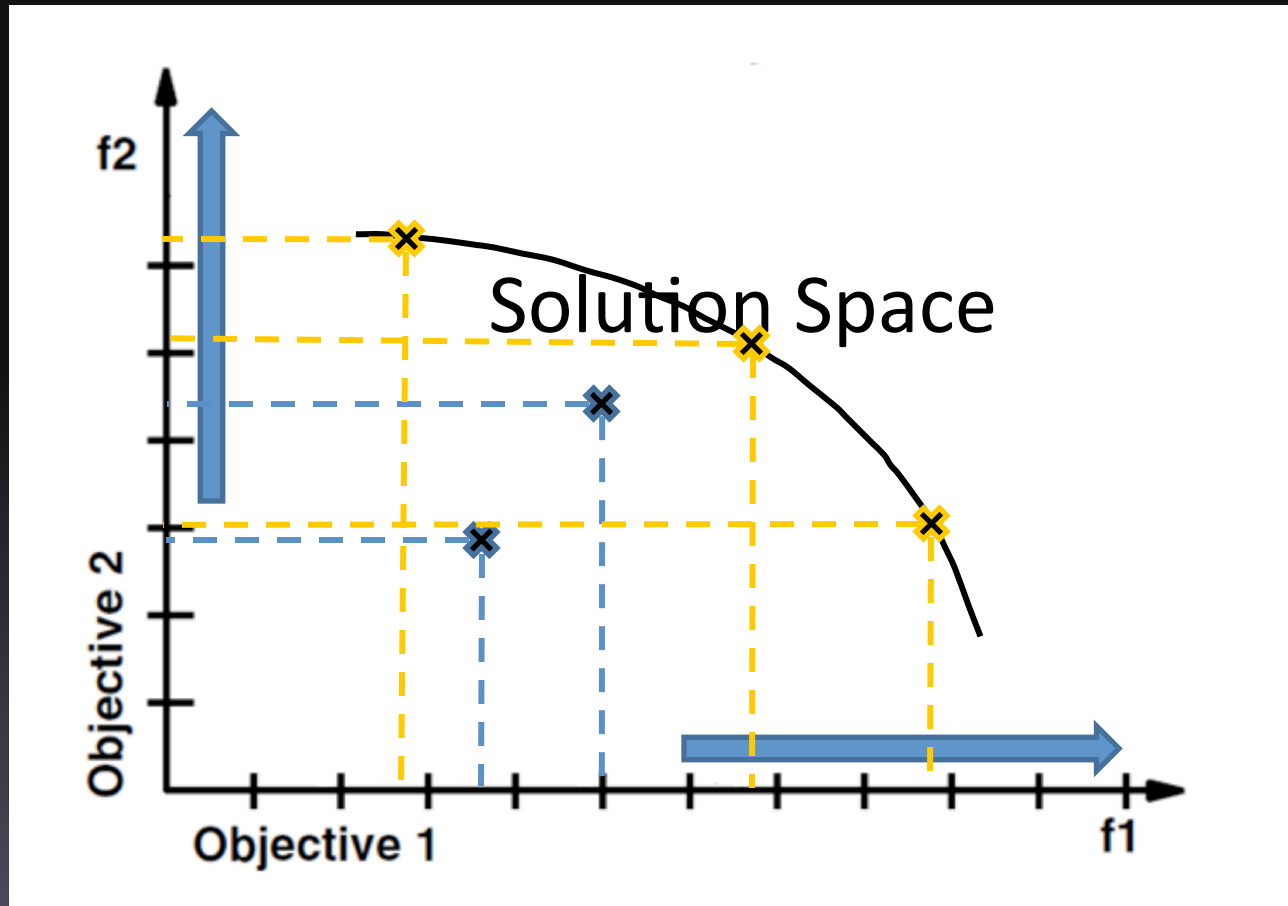
* Strictly speaking, these are not search algorithms.

Result Representation and Visualisation



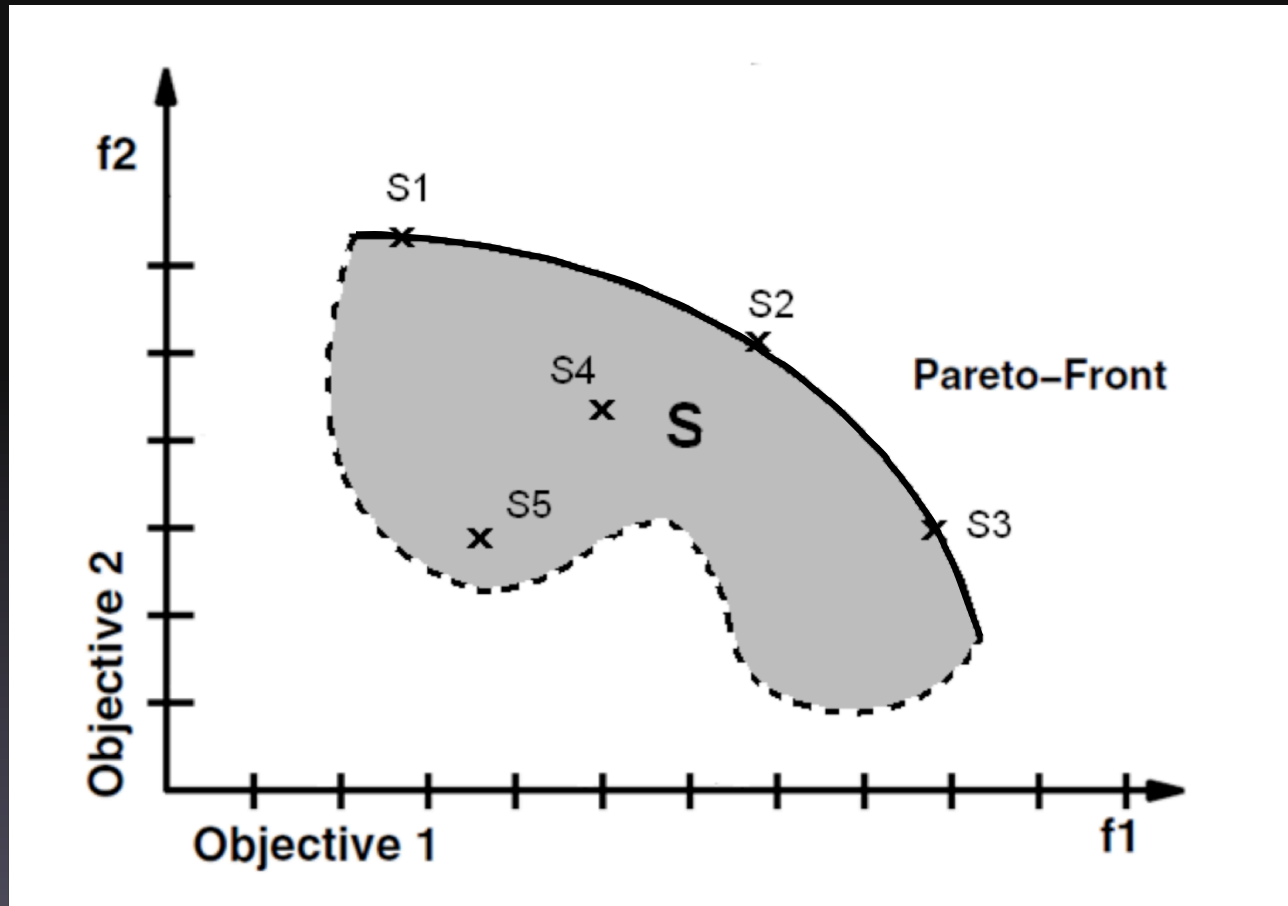
Visualisation

Pareto Optimal Front



Visualisation

Pareto Optimal Front



1. Basic Value/Cost Trade-off

The problem is to select a set of requirements that maximise customers' satisfaction (total value) and minimise required cost.

The model of fitness functions represented as:

Maximise

$$f_1 \left(\begin{matrix} \mathbf{r} \\ x \end{matrix} \right) = \sum_{i=1}^n score_i \cdot x_i$$

Minimise

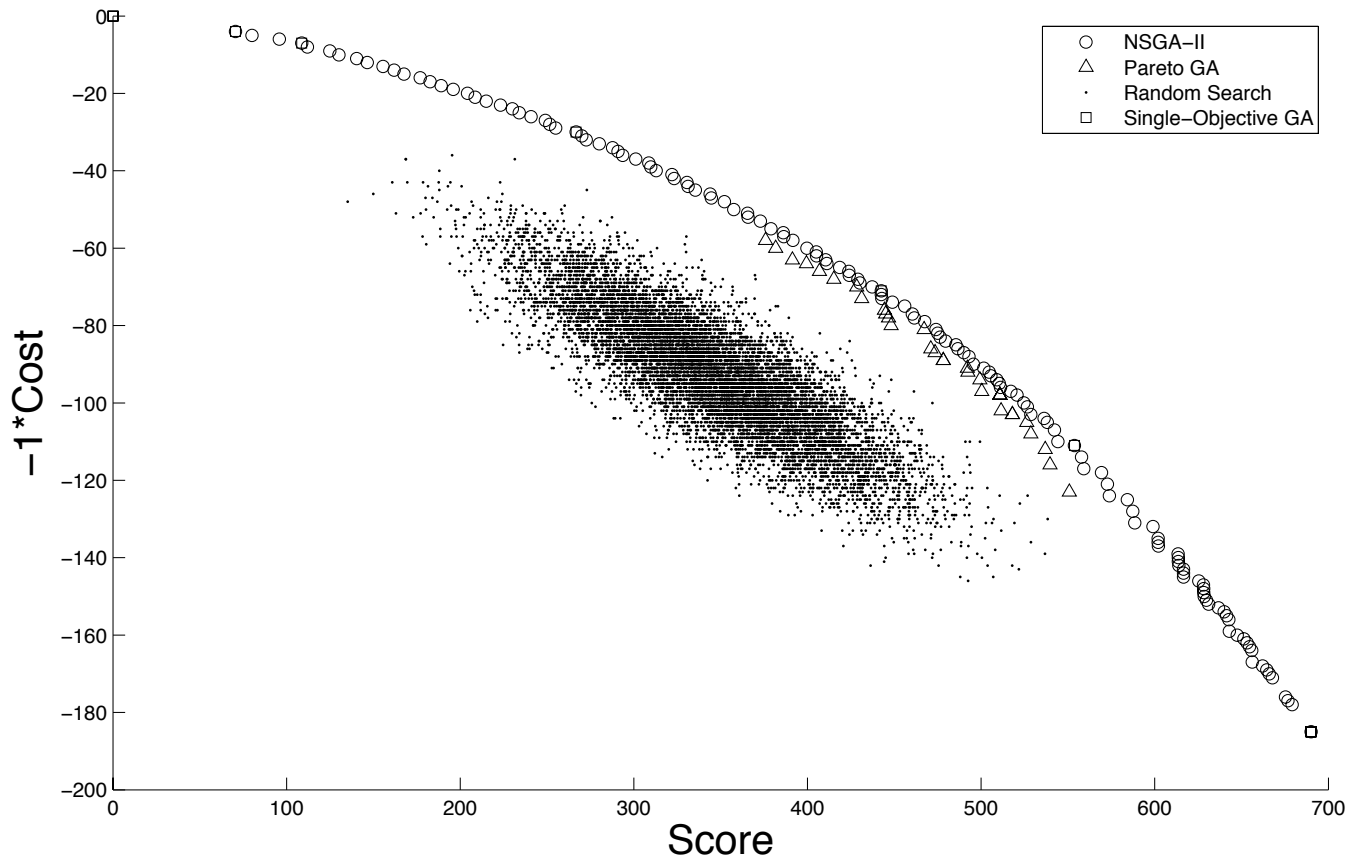
$$f_2 \left(\begin{matrix} \mathbf{r} \\ x \end{matrix} \right) = \sum_{i=1}^n cost_i \cdot x_i$$

1. Basic Value/Cost Trade-off

Scale Problem

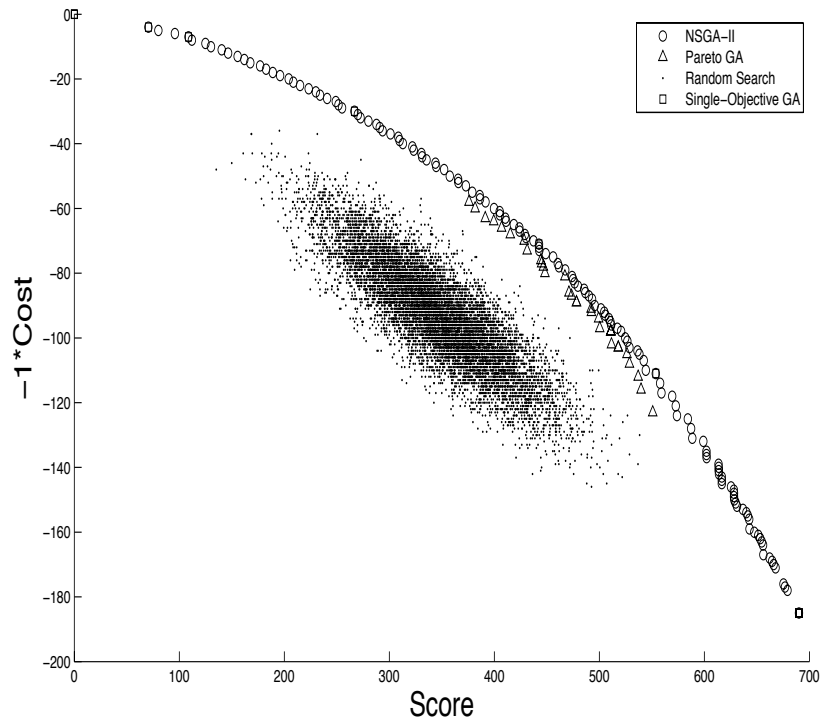
- consider three typical 'scales' cases of problem, with the number of customers ranging from 15 to 100 and the number of requirements ranging from 40 to 140.
- Investigate the relative performance of the approaches for cases.

1. Basic Value/Cost Trade-off

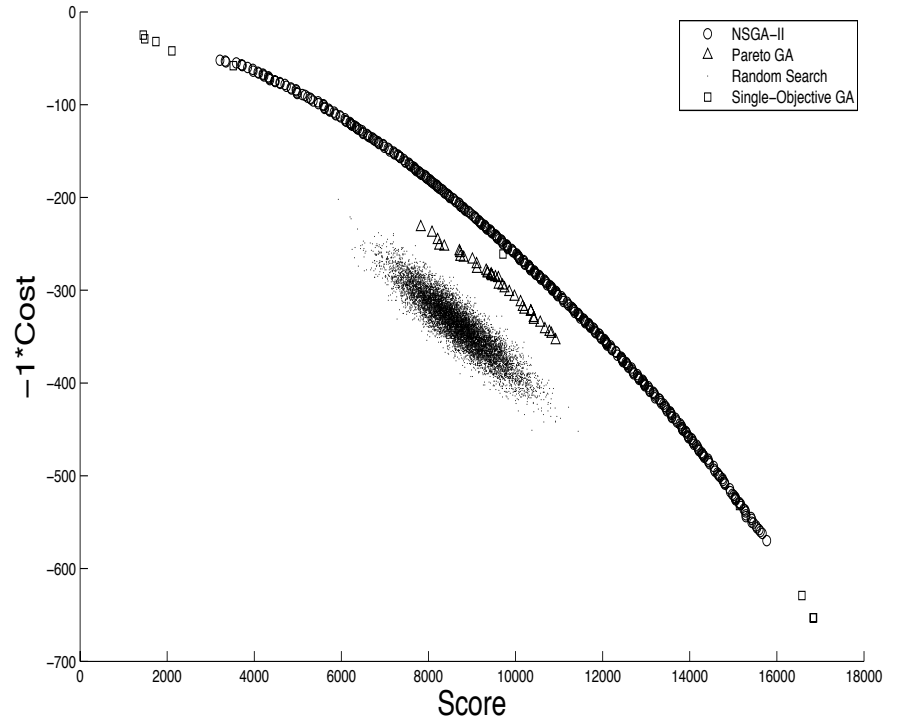


Synthetic data set: 15 stakeholders; 40 requirements

1. Basic Value/Cost Trade-off



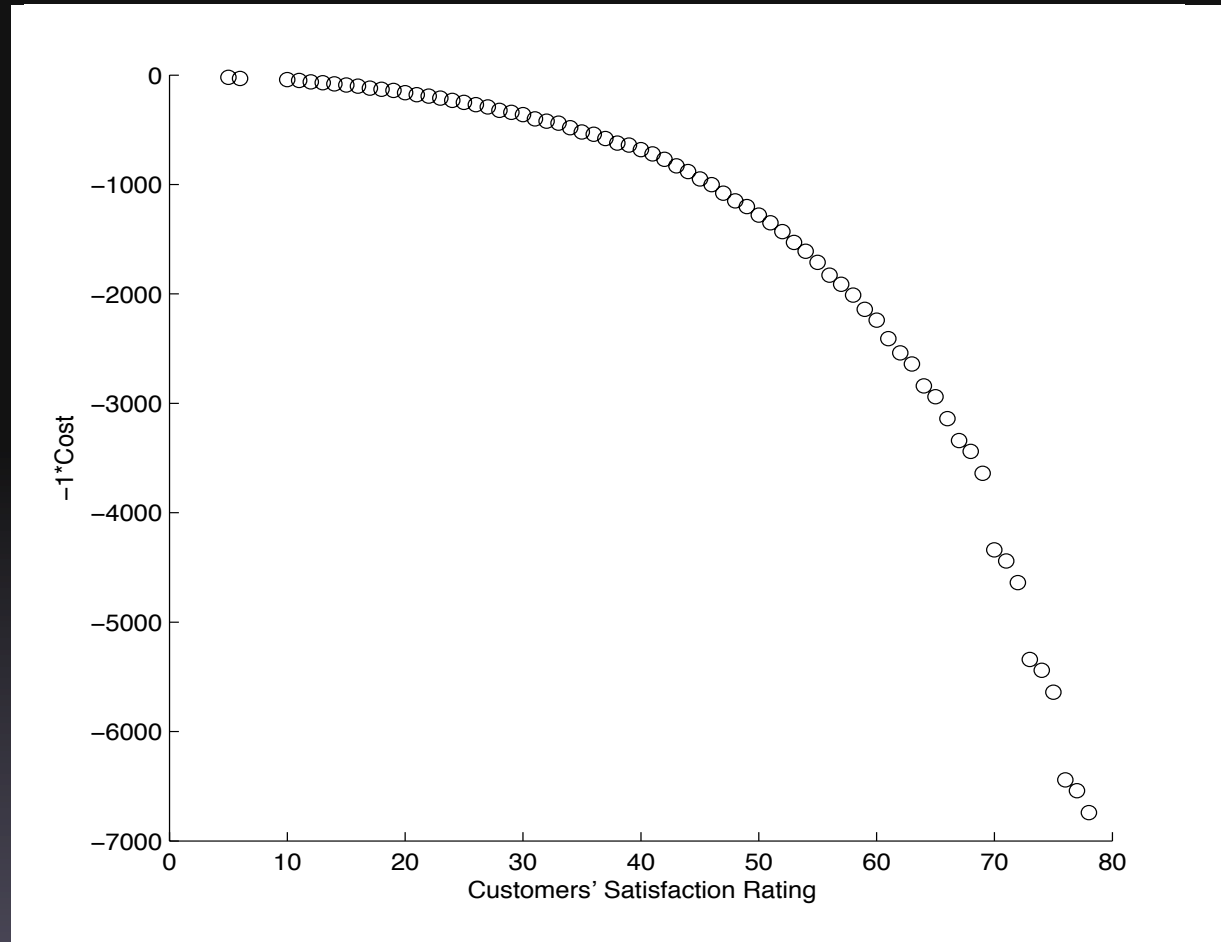
15 Stakeholders
40 Requirements



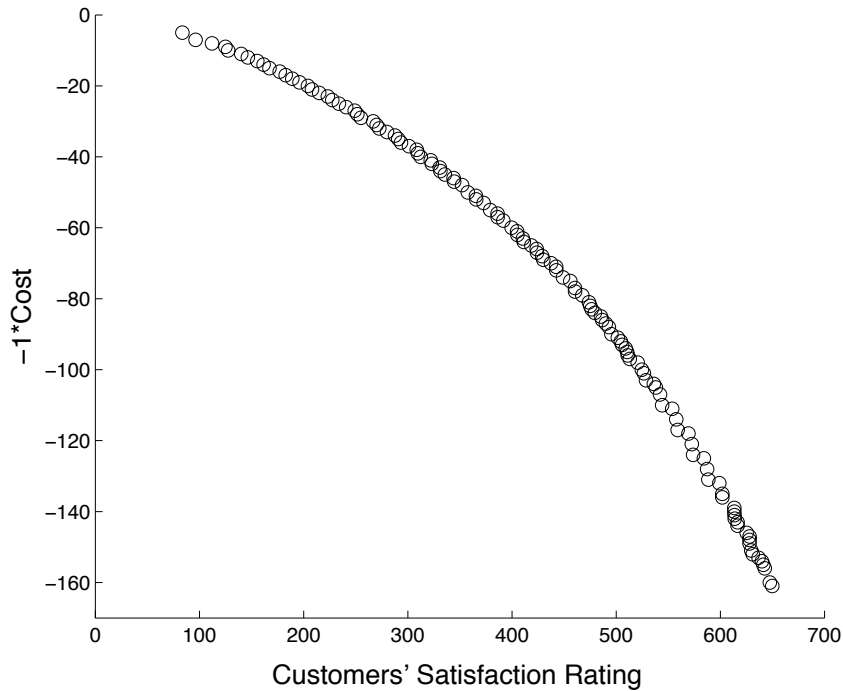
100 Stakeholders
140 Requirements

1. Basic Value/Cost Trade-off

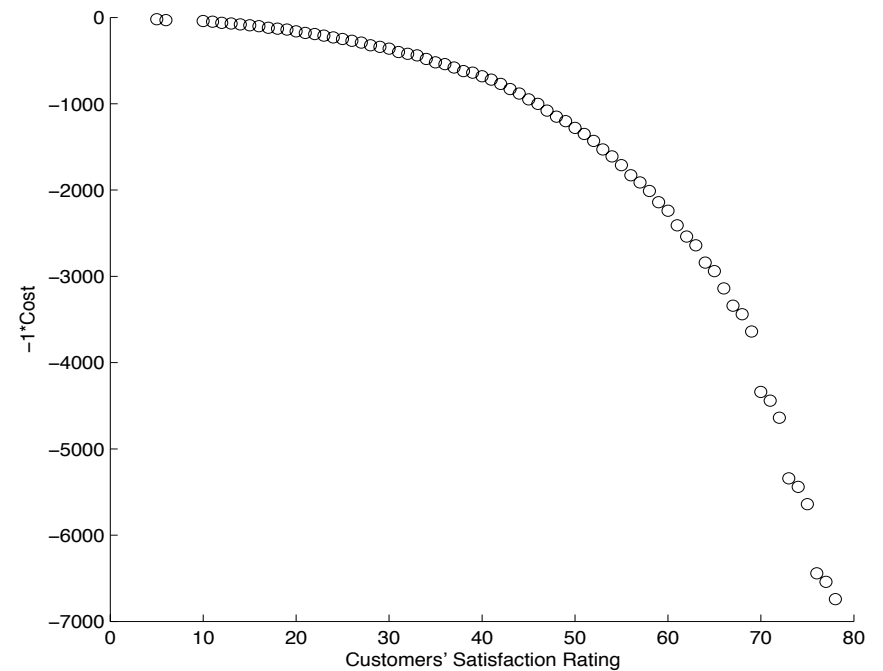
Motorola Data set
4 Stakeholders
35 Requirements



Results comparison



Synthetic



Motorola

Background

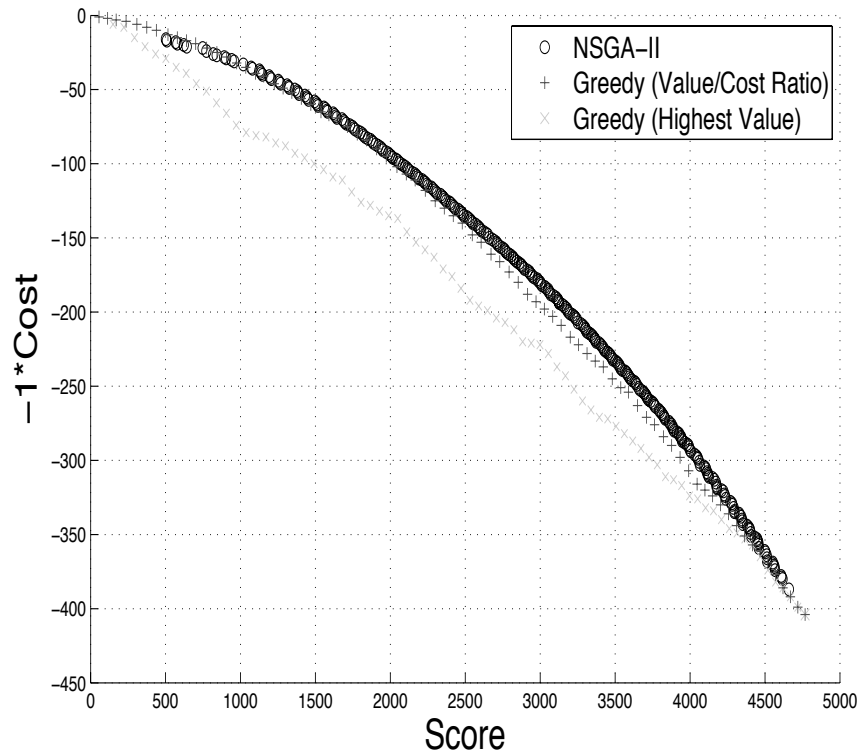
Problem

Solution

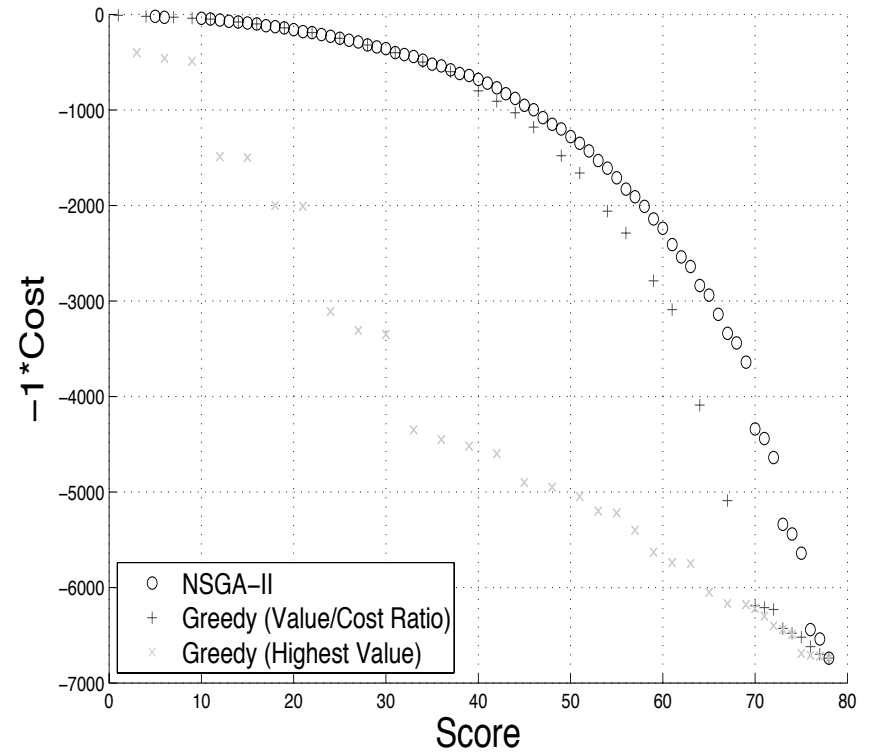
Empirical Study

Conclusion

Search vs. Greedy



Synthetic



Motorola

Background

Problem

Solution

Empirical Study

Future Work

2. Today/Future Importance Analysis

To provide robust solutions not only in the context of present conditions but also in response to those future changes that can be anticipated

Maximise

$$f_1 \left(\begin{matrix} \mathbf{r} \\ x \end{matrix} \right) = \sum_{i=1}^n score_{i, today} \cdot x_i$$

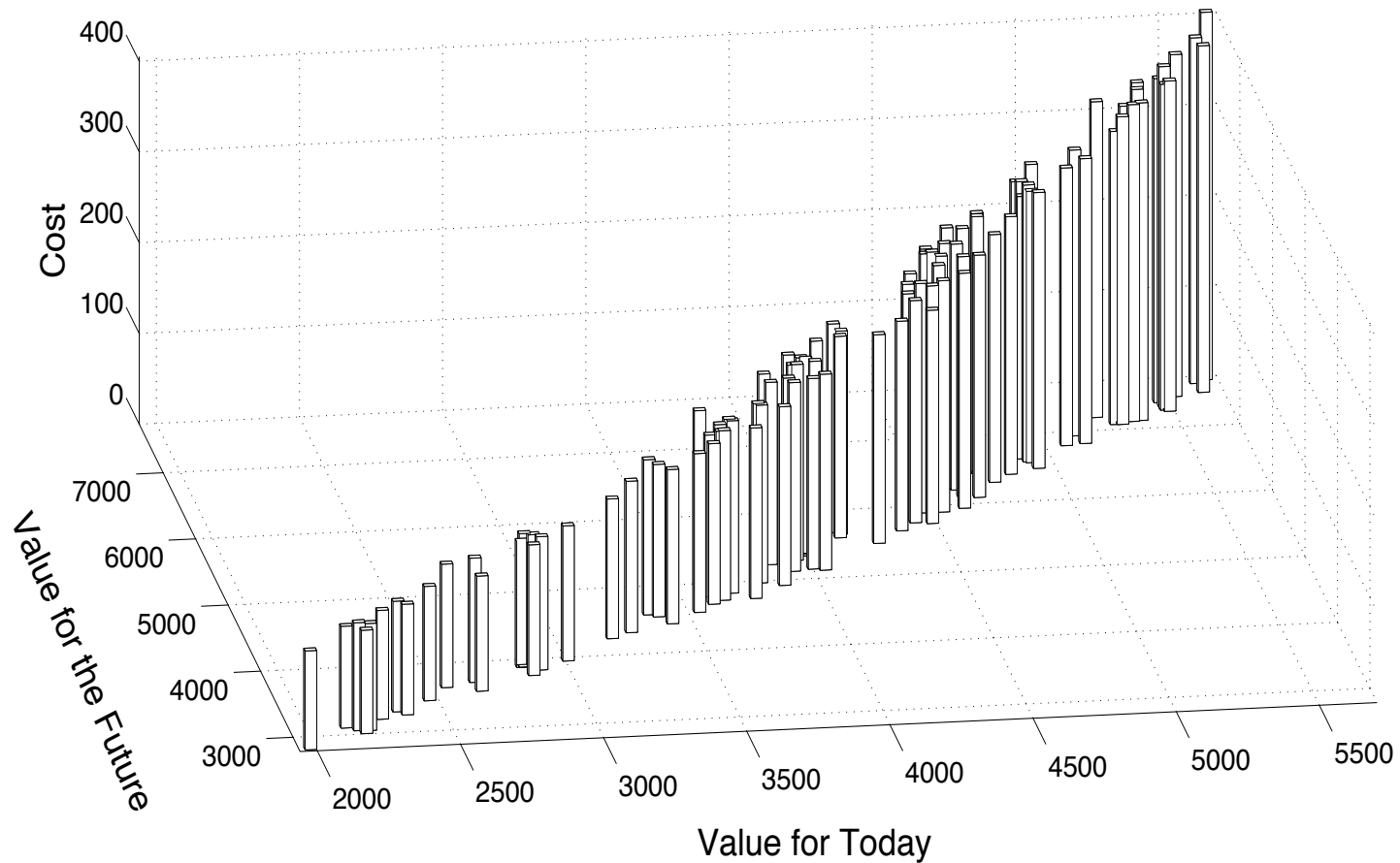
Maximise

$$f_1 \left(\begin{matrix} \mathbf{r} \\ x \end{matrix} \right) = \sum_{i=1}^n score_{i, future} \cdot x_i$$

Minimise

$$f_3 \left(\begin{matrix} \mathbf{r} \\ x \end{matrix} \right) = \sum_{i=1}^n cost_i \cdot x_i$$

2. Today/Future Importance Analysis



Results from Ericsson Data Sets: 124 Requirements, 14 Customers

Background

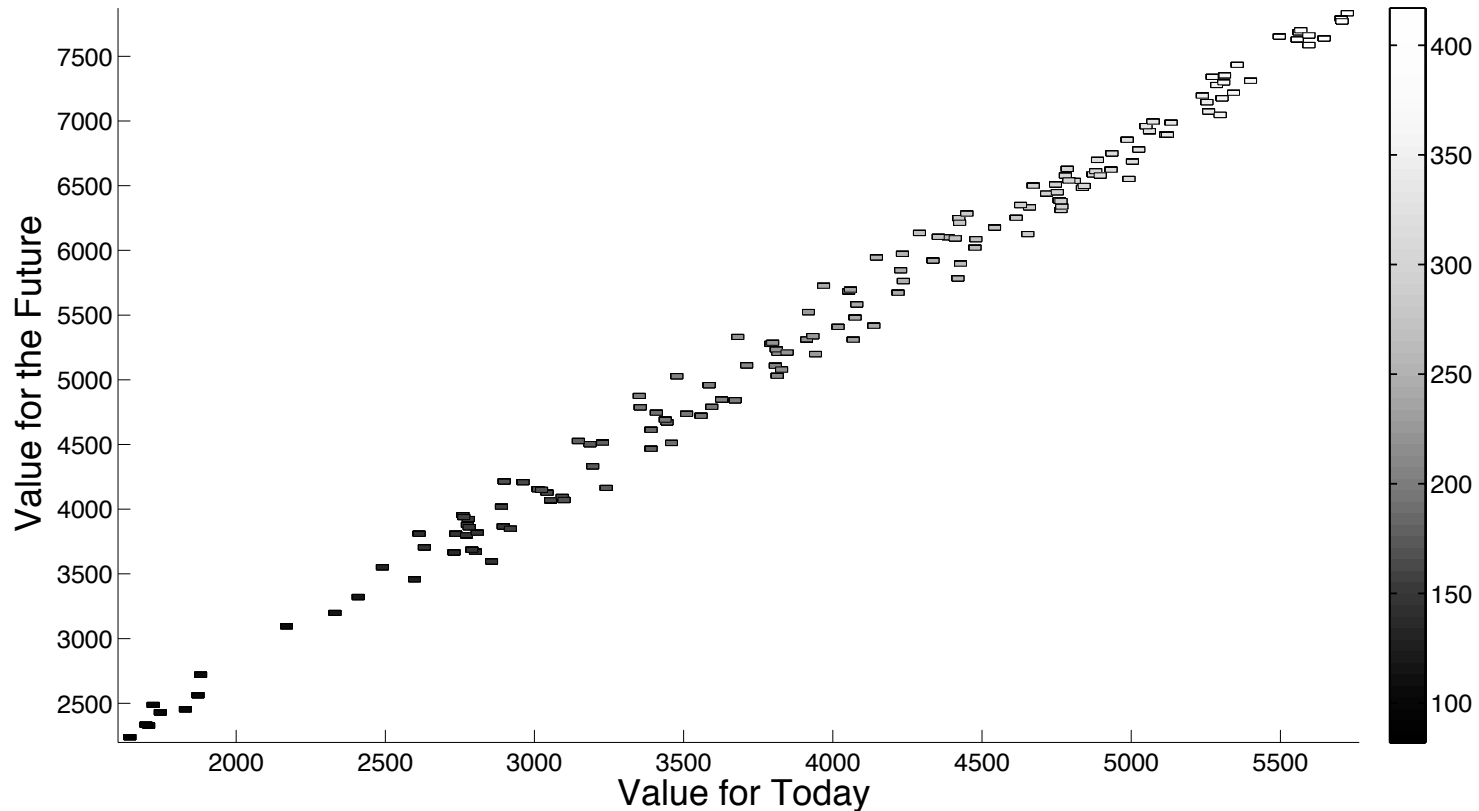
Problem

Solution

Empirical Study

Conclusion

2. Today/Future Importance Analysis



Projection onto the X-Y Plane

Background

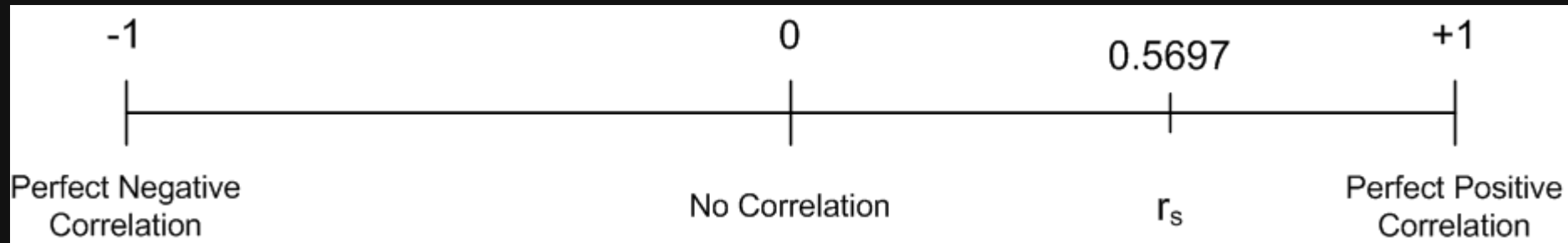
Problem

Solution

Empirical Study

Conclusion

2. Today/Future Importance Analysis



Spearman's Rank Correlation Coefficient

This indicates a positive correlation between the value for today and value for the future.

3. Multi-Stakeholder Tensioning Analysis

The problem is to select a set of requirements that maximise the total value to each stakeholder, which is expressed as a percentage.

The model of fitness functions represented as:

$$\begin{aligned} &\text{Maximise} && \frac{\sum_{i=1}^n \text{value}(r_i, c_j) \cdot x_i}{\sum_{r \in R_j} \text{value}(r, c_j)} \\ &\text{subject to} && \sum_{i=1}^n \text{cost}_i \leq B, \quad B > 0 \end{aligned}$$

Data Sets Used

1. Motorola Data Set:

35 Requirements and 4 Stakeholders

2. Greer and Ruhe Data Set:

20 Requirements and 5 Stakeholders

Data Sets Used

3. 27 Combination Levels of Random Data Sets:

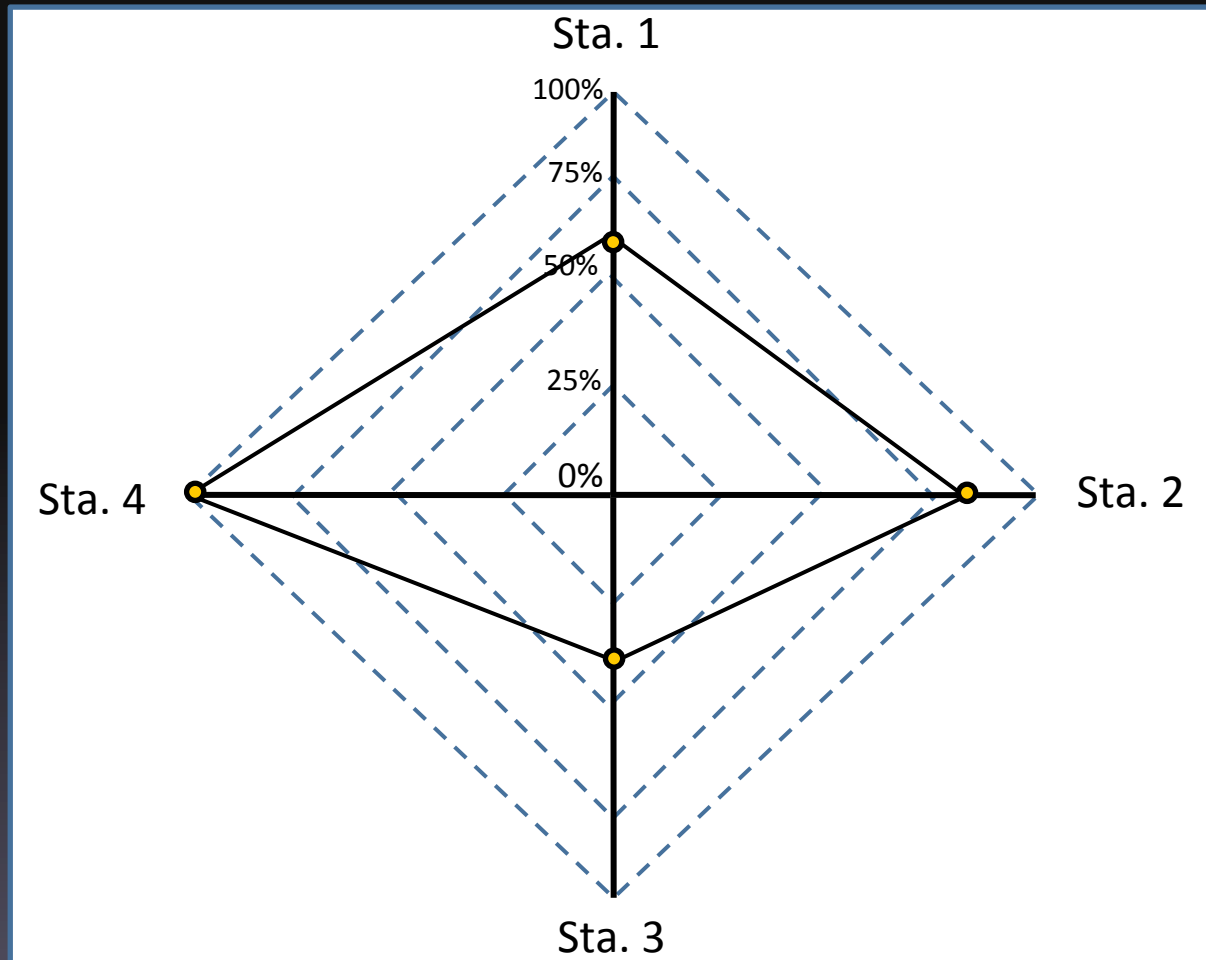
the No. of requirements

the No. of stakeholders

the density of the
stakeholder-requirement
matrix

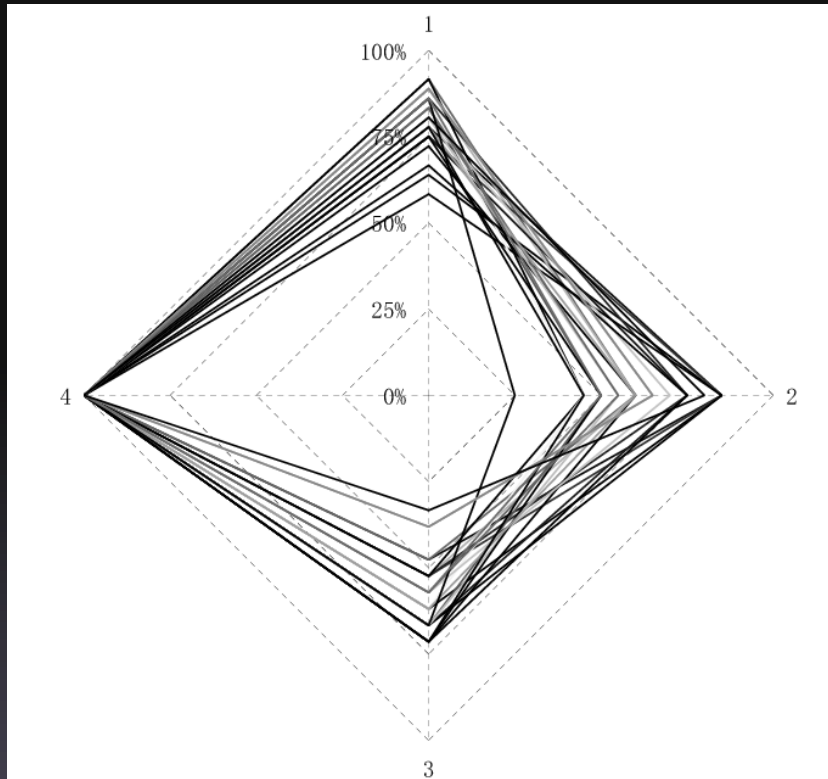
	R_{small}	R_{medium}	R_{large}
C_{small}	$C_s R_s D_{low}$	$C_s R_m D_{low}$	$C_s R_l D_{low}$
	$C_s R_s D_m$	$C_s R_m D_m$	$C_s R_l D_m$
	$C_s R_s D_h$	$C_s R_m D_h$	$C_s R_l D_h$
C_{medium}	$C_m R_s D_{low}$	$C_m R_m D_{low}$	$C_m R_l D_{low}$
	$C_m R_s D_m$	$C_m R_m D_m$	$C_m R_l D_m$
	$C_m R_s D_h$	$C_m R_m D_h$	$C_m R_l D_h$
C_{large}	$C_l R_s D_{low}$	$C_l R_m D_{low}$	$C_l R_l D_{low}$
	$C_l R_s D_m$	$C_l R_m D_m$	$C_l R_l D_m$
	$C_l R_s D_h$	$C_l R_m D_h$	$C_l R_l D_h$

Multi-Stakeholder Kiviat Diagram

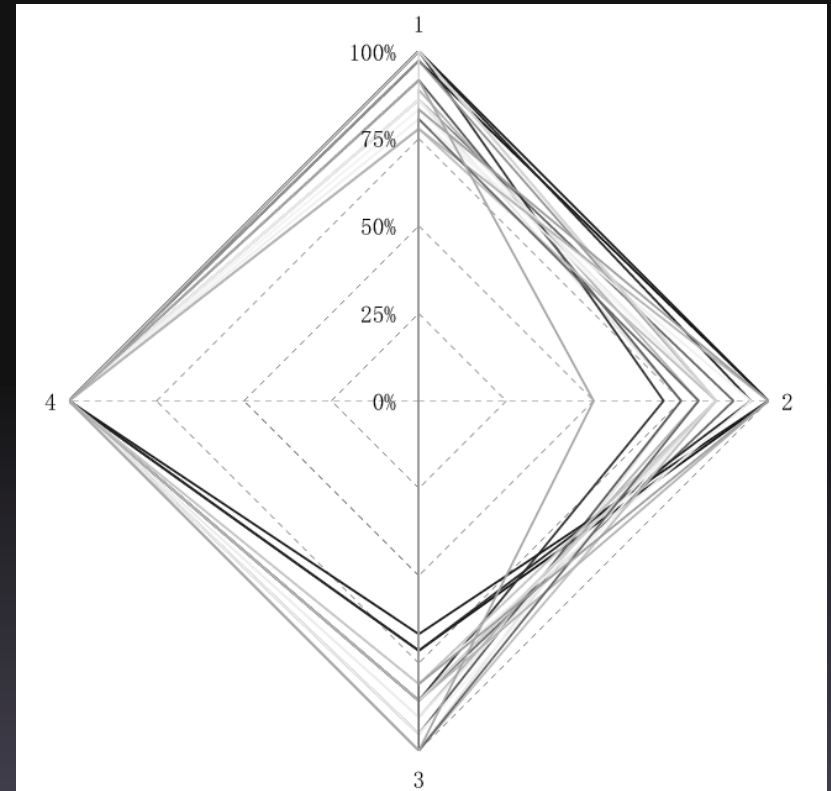


Multi-Stakeholder Tensioning Analysis

Motorola data set

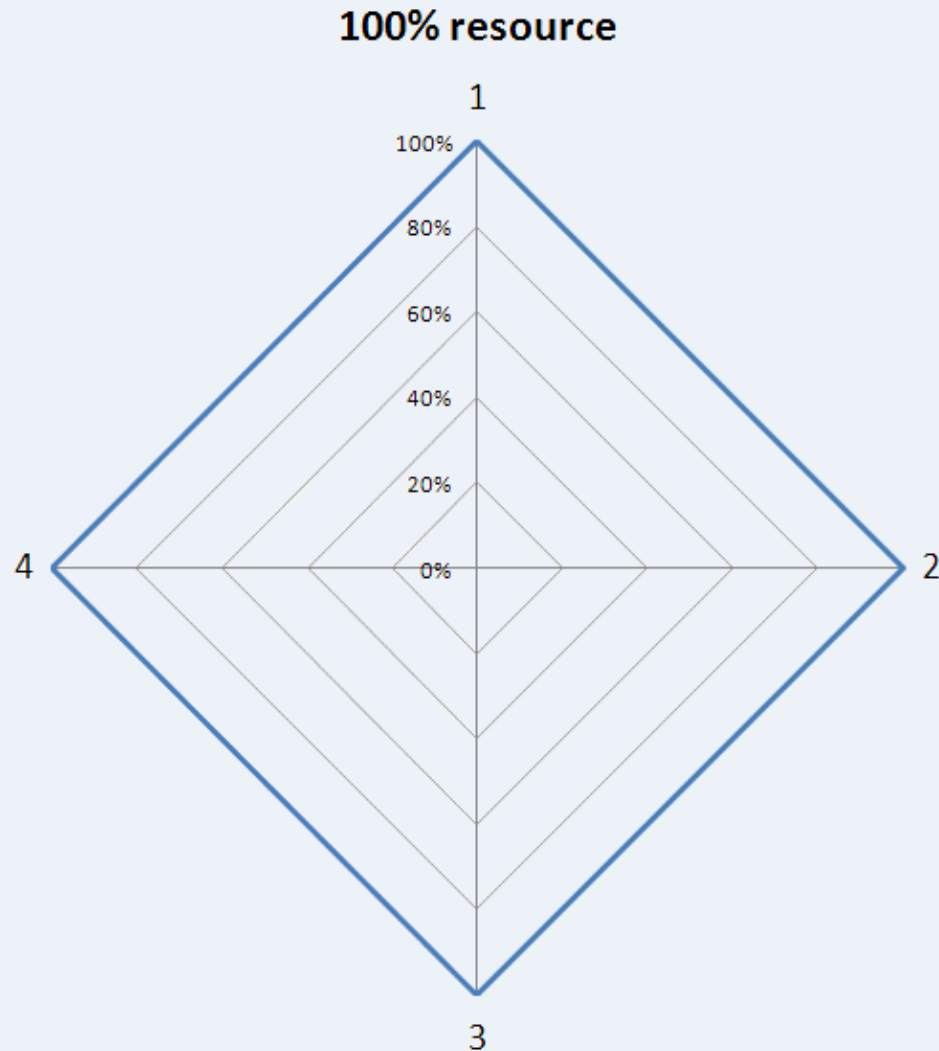


30% Budgetary Resource Constraint

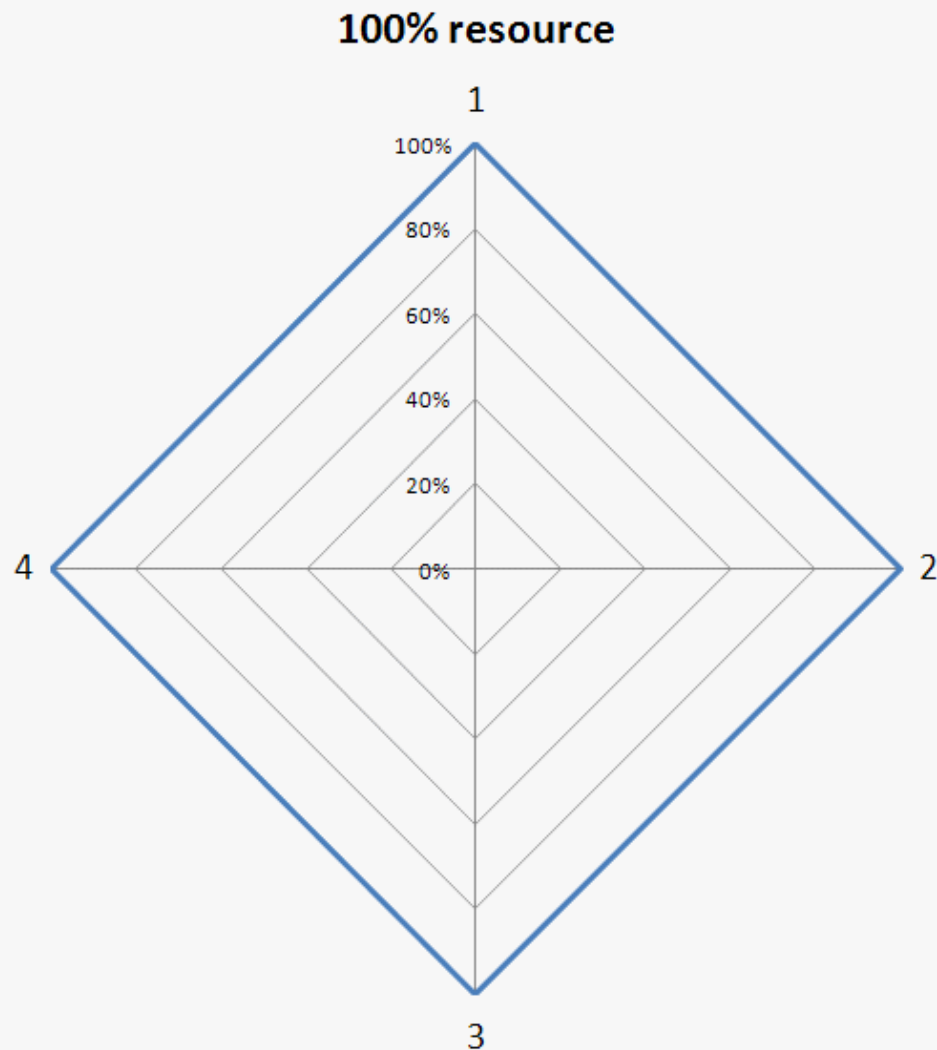


70%

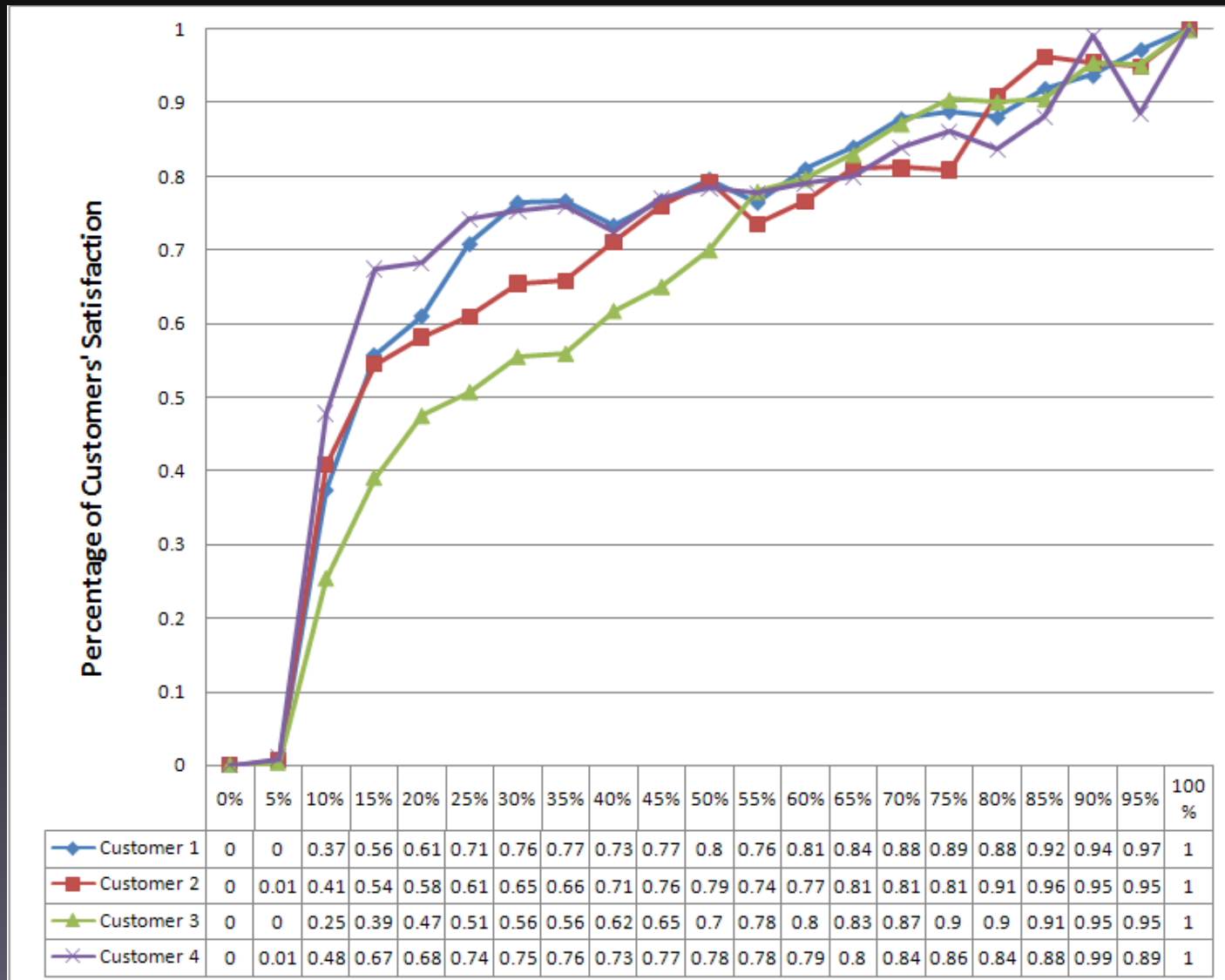
Solutions on the Pareto Front



Average Solutions

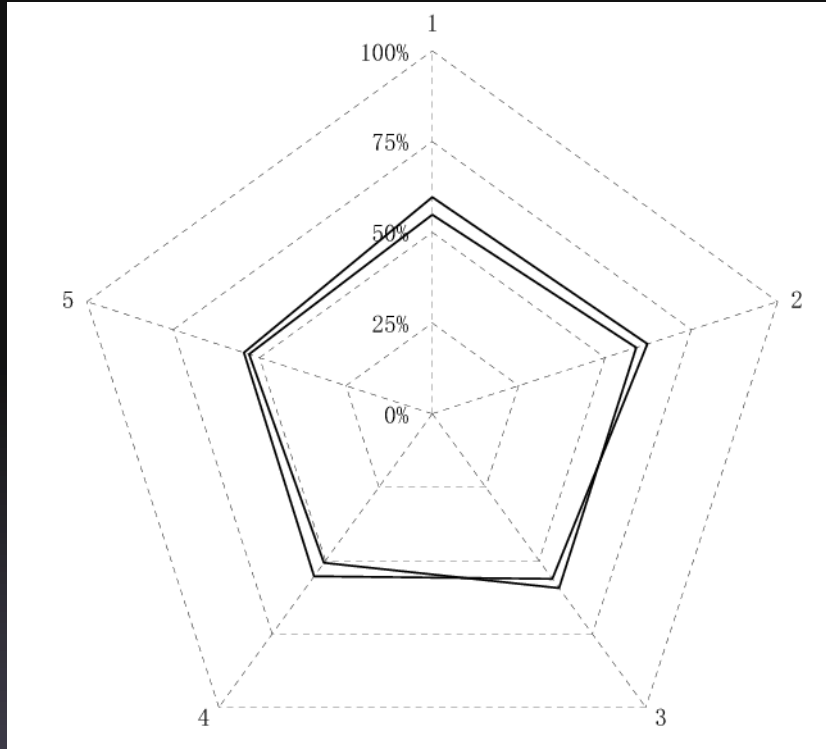


Tensions between the Stakeholders' Satisfaction for Different Budgetary Resource Constraints

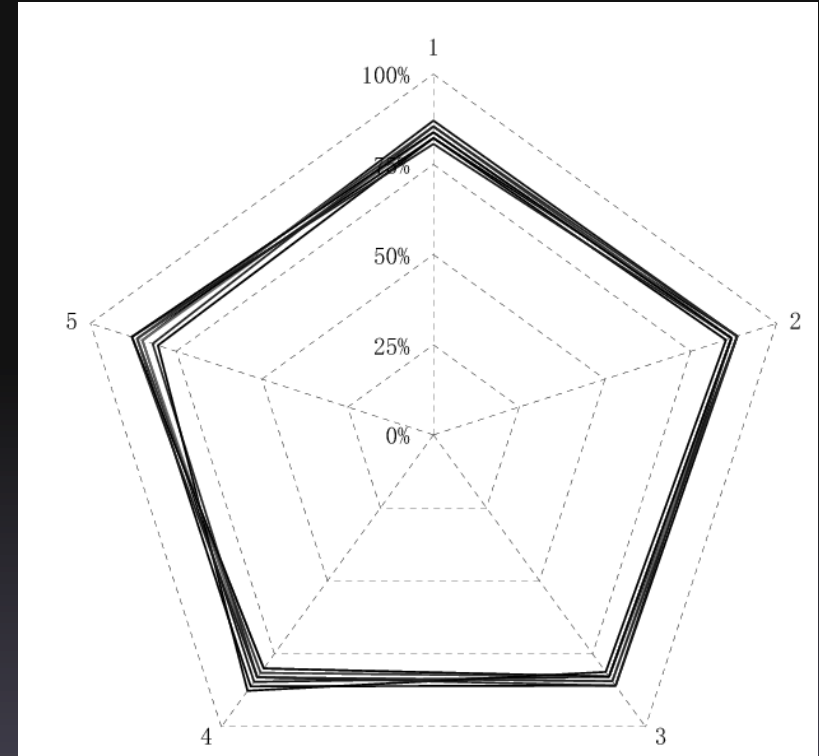


Multi-Stakeholder Tensioning Analysis

Greer and Ruhe data set



30% Budgetary Resource Constraint



70%

Background

Problem

Solution

Empirical Study

Conclusion

Algorithms' Performance

$$C = \frac{\sum_{i=1}^N d_i}{N}$$

$$P = \frac{num}{NUM}$$

Rank Order for Convergence

	winner	runner up	loser
Random Search	0%	0%	100%
Two-Archive	95.19%	4.81%	0%
NSGA-II	7.04%	92.96%	0%

Solutions on the Reference front

Random Search	Two-Archive	NSGA-II
2.68%	94.57%	38.25%

Algorithms' Performance

- 1 . The diversity of the Two-archive algorithm is significant in most cases
2. The Two-archive and NSGA-II algorithms always have a better convergence than the Random Search
3. The Two-Archive algorithm outperforms NSGA-II and Random Search in terms of convergence in some case

4. Multi-Stakeholder Fairness Analysis

Fairness on Absolute *number* of fulfilled requirements:

Maximise \overline{NA}

Minimise $\sigma(NA)$

Fairness on absolute *value* of fulfilled requirements:

Maximise \overline{VA}

Minimise $\sigma(VA)$

where

$$VA_j = \sum_{i=1}^n value(r_i, c_j) \cdot x_i$$

4. Multi-Stakeholder Fairness Analysis

Fairness on the percentage of value and cost of fulfilled requirements:

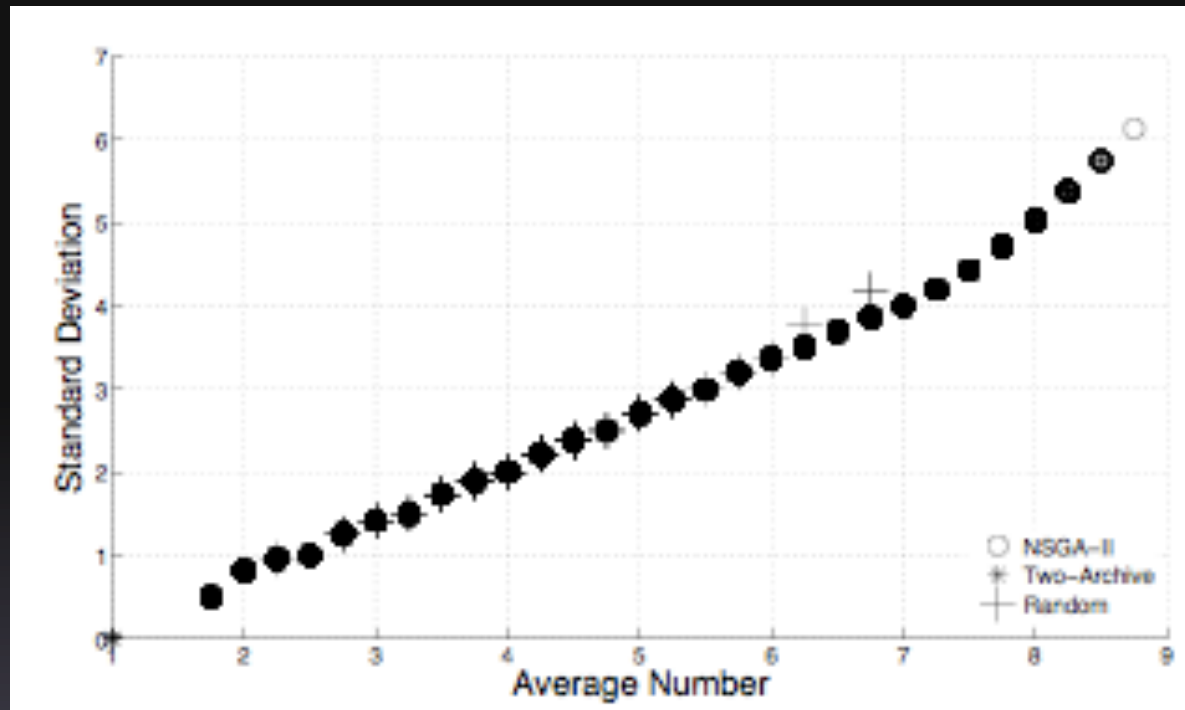
Minimise $\sigma(\text{Cost_}C)$

Maximise \overline{VP}

Minimise $\sigma(VP)$

Minimise $\sum_{i=1}^n \text{cost}_i \cdot x_i$

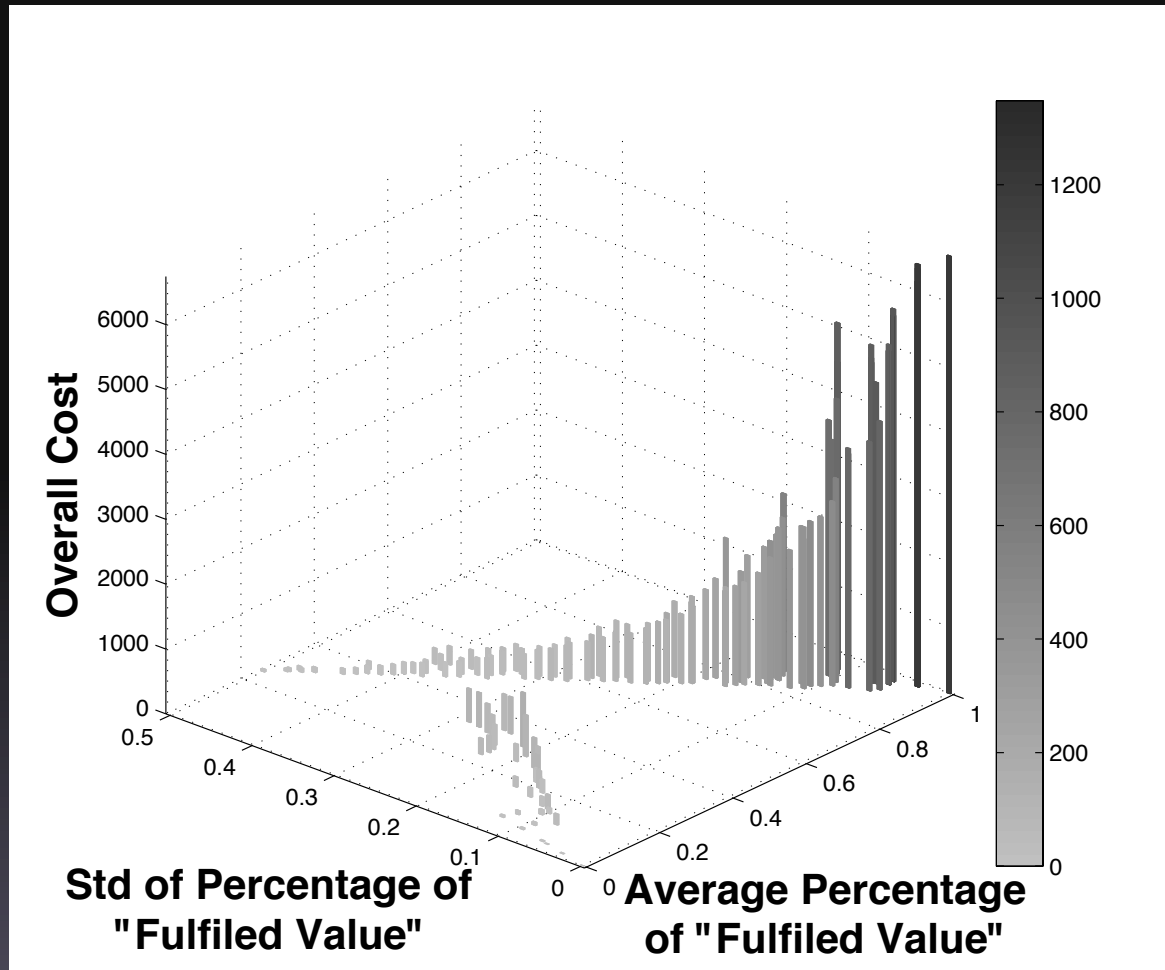
4. Multi-Stakeholder Fairness Analysis



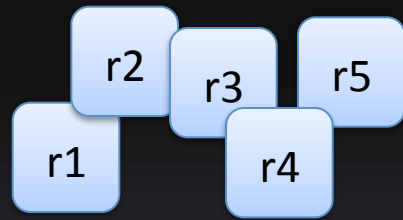
4. Multi-Stakeholder Fairness Analysis

Motorola Data Set

Fairness on
Percentage of
Fulfilled Value



5. Requirements Interaction Management (RIM)



Requirements

Release
Next

And
Or
Precedence
Value-
related
Cost-related

And

Given requirement r_i is selected,
then requirement r_j has to be chosen.

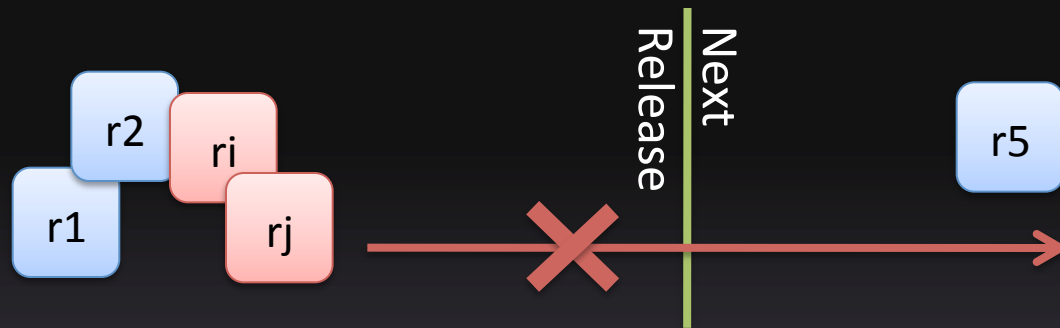
r_i



Define an equivalence relation ξ on the
requirements array R such that $r(i, j) \in \xi$

Or

Requirements r_i and r_j are conflicting to each other, only one of r_i, r_j can be selected.



Define an equivalence relation φ on the requirements array R such that $r(i, j) \in \varphi$

Precedence

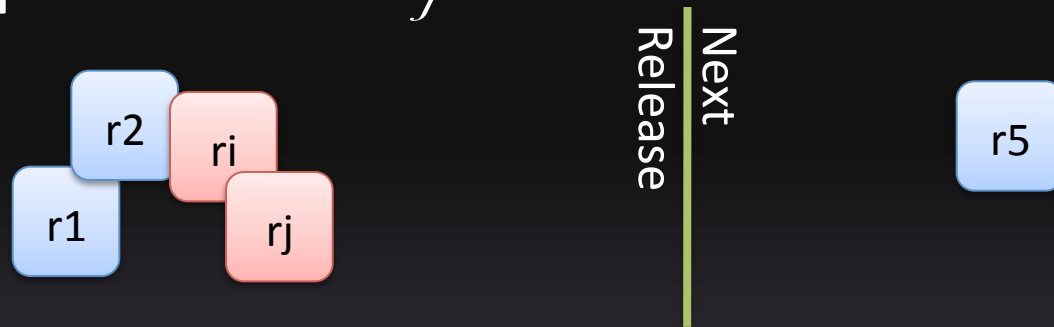
Given requirement r_i has to be implemented before requirement r_j



Define an partial order χ on the requirements array R such that $r(i, j) \in \chi$

Cost-related

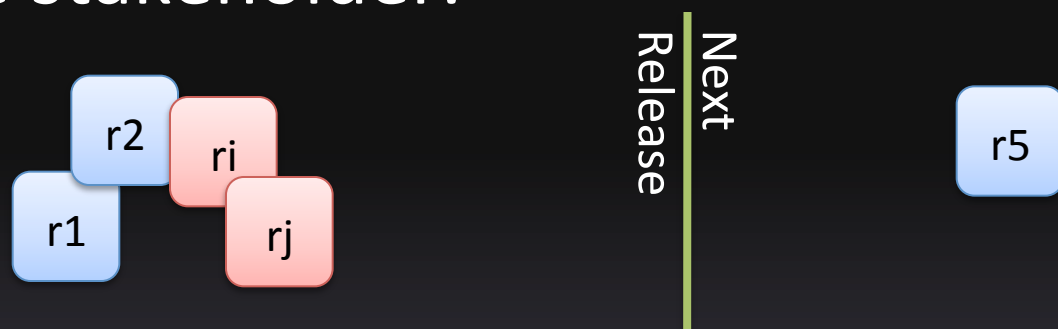
Given requirement r_i is selected, then this selection affects the cost of implementing requirement r_j .



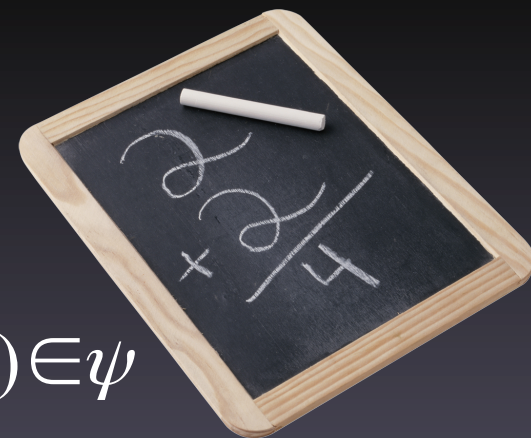
Define an partial order ω on the requirements array R such that $r(i, j) \in \omega$

Value-related

Given requirement r_i is selected, then this selection affects the value of requirement r_j for the stakeholder.

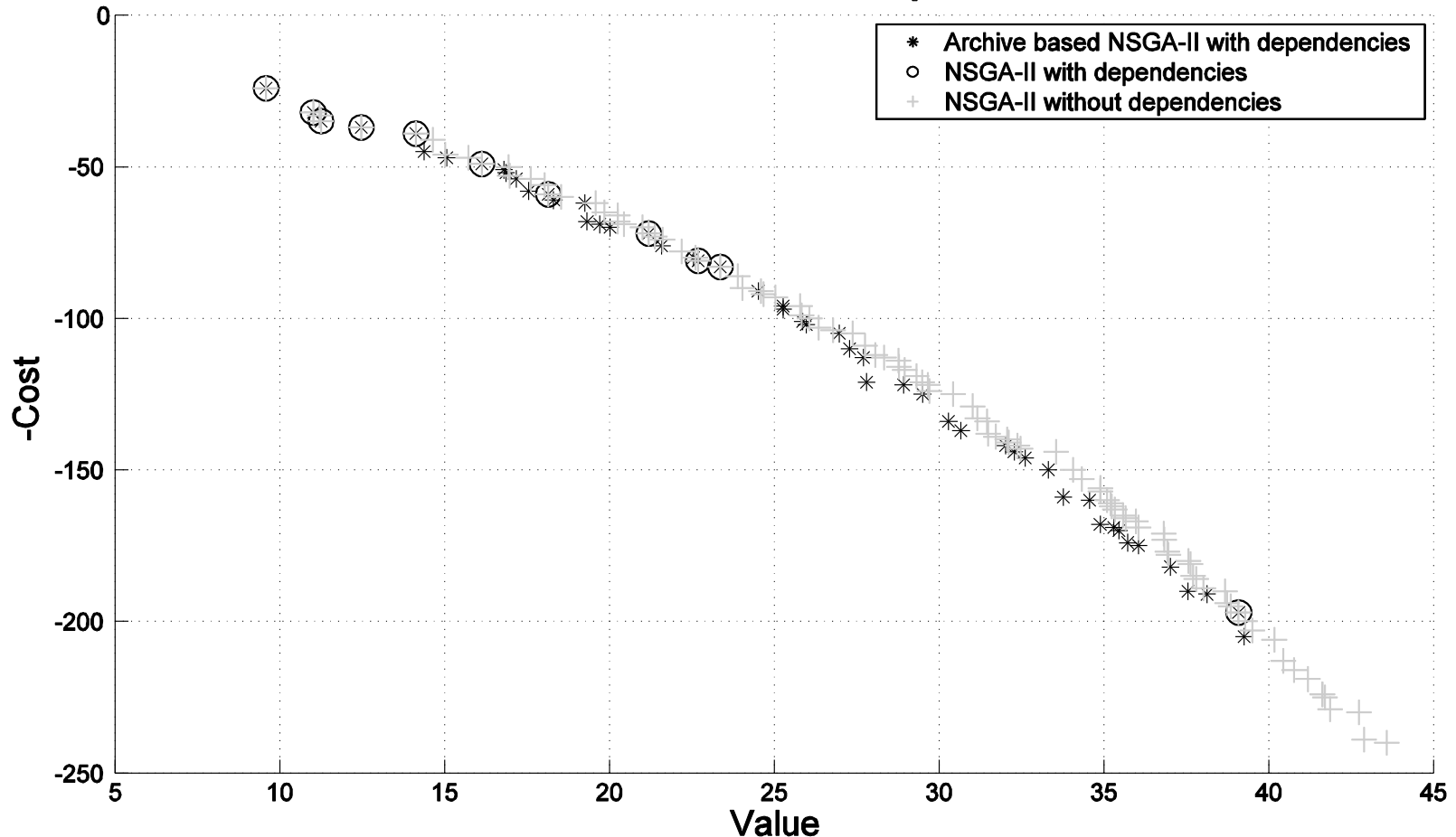


Define an partial order ψ on the requirements array R such that $r(i, j) \in \psi$



Empirical Study 5: RIM

And, Or and Precedence Dependencies



34 Customers, 50 Requirements

Background

Problem

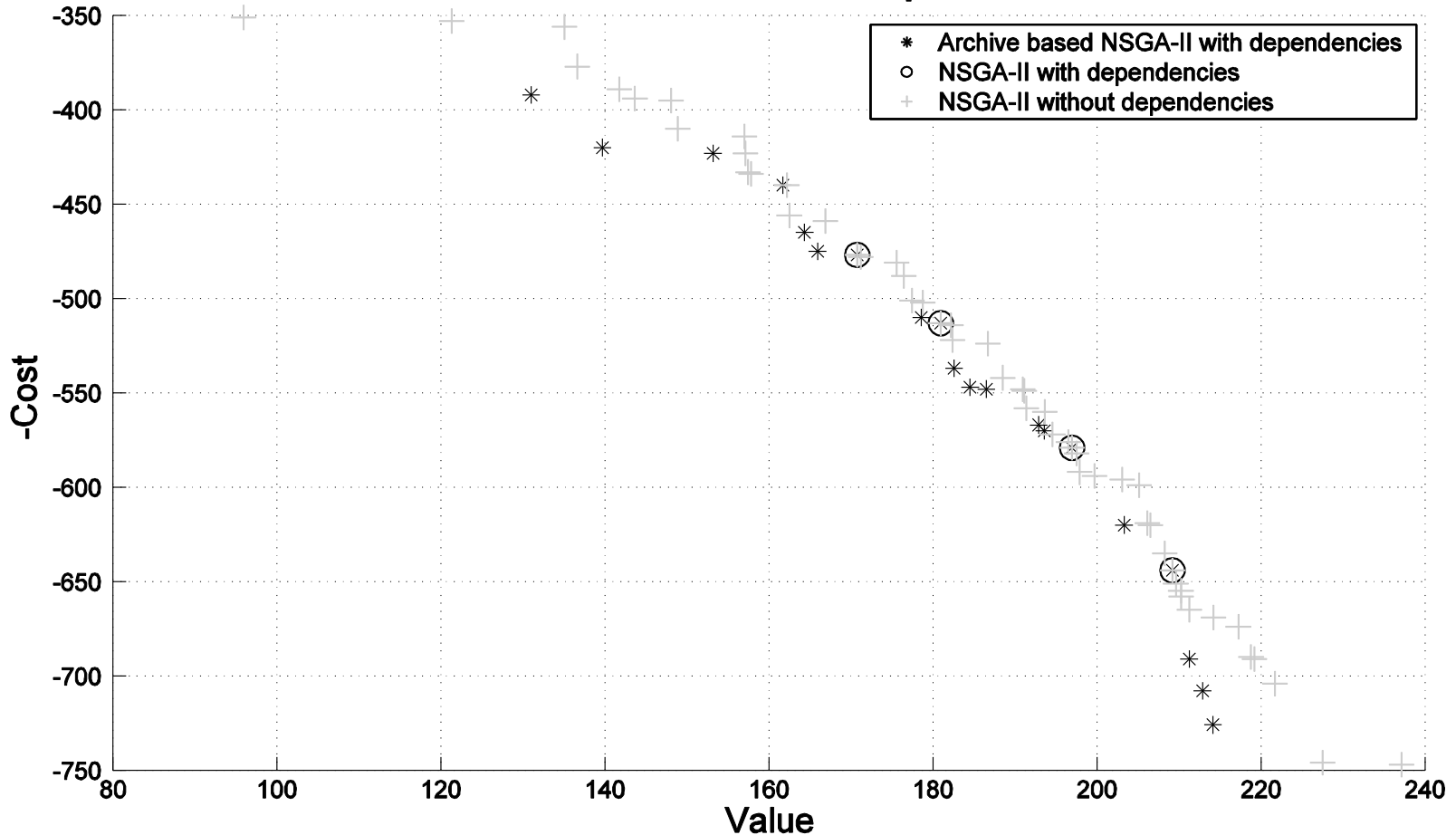
Solution

Empirical Study

Conclusion

Empirical Study 5: RIM

And, Or and Precedence Dependencies



4 Customers, 258 Requirements

Background

Problem

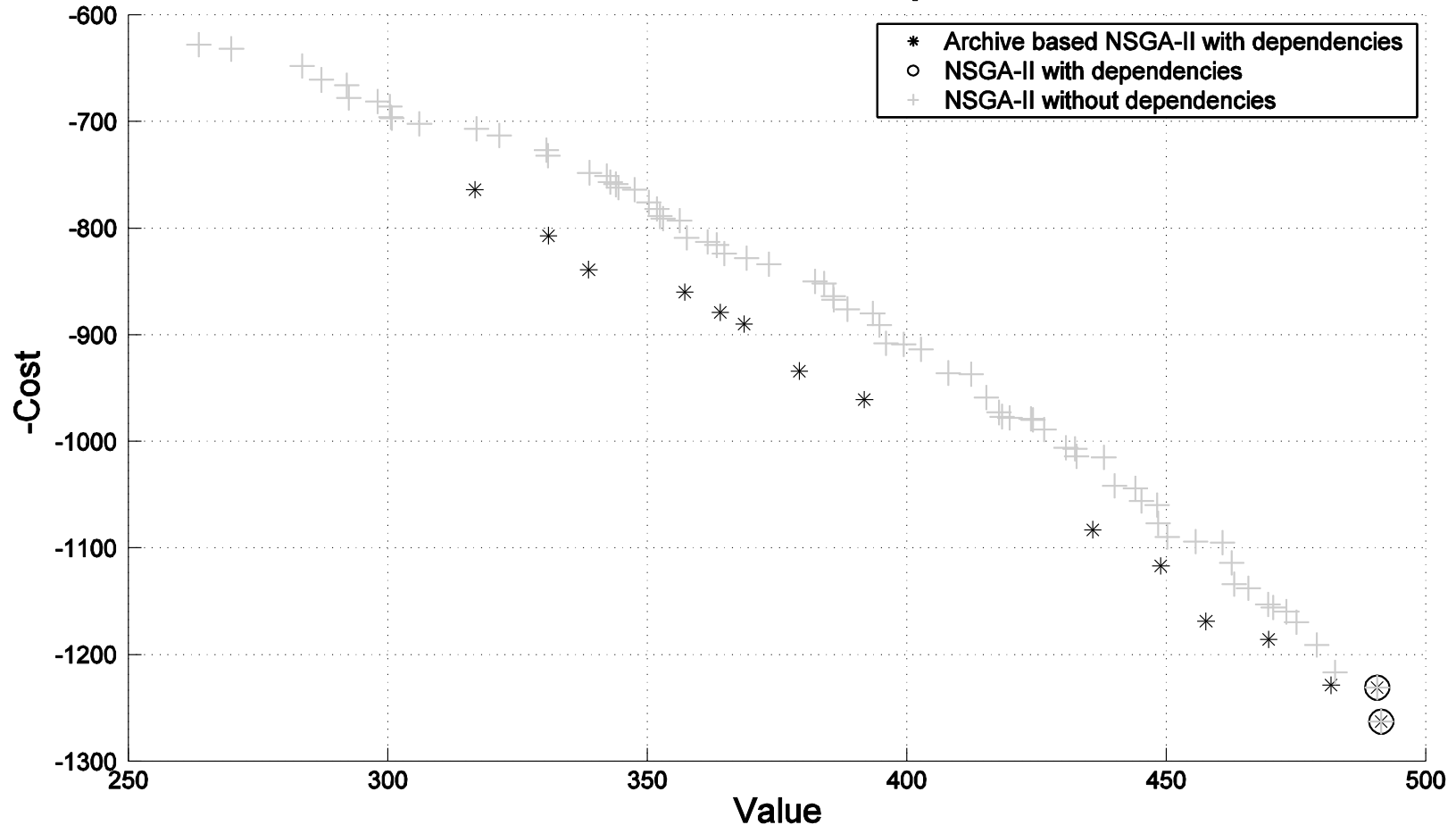
Solution

Empirical Study

Conclusion

Empirical Study 5: RIM

And, Or and Precedence Dependencies



21 Customers, 412 Requirements

Background

Problem

Solution

Empirical Study

Conclusion

Conclusion

- Basic Value/Cost Trade-off analysis
- Today/Future Importance Analysis
- Multi-Stakeholder tension and fairness analysis
- Requirements Interaction Management



[http://crestweb.cs.ucl.ac.uk/resources/
sbse_repository/](http://crestweb.cs.ucl.ac.uk/resources/sbse_repository/)

