



Decision Making Methods in Space Economics and Systems Engineering

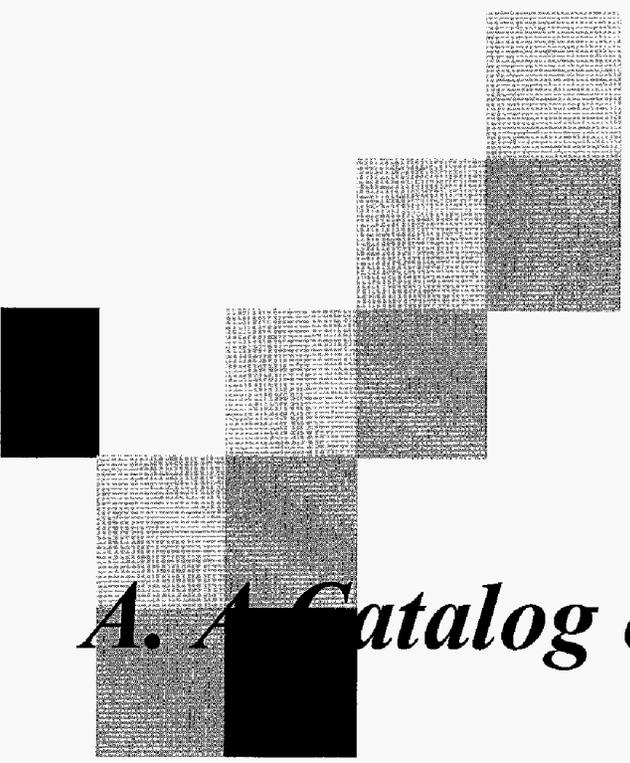
Robert Shishko, Jet Propulsion Laboratory,
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Decision Methods Outline

A. A Catalog of DM Methods

- Business Case Analysis / Financial Techniques
- Systems Engineering / Trade Studies Techniques

B. Summary and Observations



A. A Catalog of Decision Making Methods

A Catalog of Decision Making Methods in Space Economics and Systems Engineering

- Present Value and IRR
 - Cost-Benefit Analysis
 - Real Options
- } Business Case Analysis/Financial Techniques
-
- Cost-Effectiveness Analysis
 - Cost-Utility Analysis
 - Multi-Attribute Utility Theory (MAUT)
 - Analytic Hierarchy Process (AHP)
- } SE/Trade Study Techniques

Note: Decision processes may use decision analysis (trees), Monte Carlo simulation techniques, and game theory (e.g., Nash equilibrium) as an overlaid framework for handling uncertainty.

Present Value and IRR

- Primary application in business case analysis for commercial and some government investments
- Requires estimation of the stream of costs, C , and revenues, R (or savings)

□ PV calculation
$$PV(r, T) = \sum_{t=0}^{t=T} (R_t - C_t)(1+r)^{-t}$$

□ IRR calculation
$$IRR = r^* \text{ such that } PV(r^*, T) = 0$$

- Treatment of uncertainty?

Cost-Benefit Analysis

- Given a market failure situation, should a particular (public) investment be made?
- The foundation of cost-benefit analysis is the *Kaldor-Hicks compensation principle*.
- Distributional (equity) issues are not addressed, and so-called “secondary” or “intangible” benefits are typically ignored.
- Benefit measures for public investment projects designed to ultimately benefit consumers are generally based on the concept of *willingness-to-pay*.
- Cost-benefit analysis is appropriate when there are market prices that can be used to infer the values of the goods or services produced by the public investment.

Real Options Analysis

- Real options valuation is a financial technique for evaluating investments under conditions of uncertainty, particularly uncertainty associated with market variables such as future product demand or the future value of an asset.
- Option pricing is a well-developed area of financial engineering, where it deals with the valuation of puts, calls, and more complex derivatives, but when applied to non-financial assets, the term “real options” is used.
- In real options valuation, the general ideas from financial options pricing theory are used along with some of the mathematics.

Basic Concepts of Real-Options Valuation

“The future is uncertain . . . and in an uncertain environment, having the flexibility to decide what to do after some of that uncertainty is resolved definitely has value. Option-pricing theory provides the means for assessing that value”

Robert C. Merton

Nobel Lecture, 1998

Would You Have Invested?



Microsoft Corporation, 1978

Basic Concepts of Real-Options Valuation

- The real-options approach captures the additional value of managerial flexibility inherent in any phased investment, which makes it more suitable than older methods.
 - Mathematical representation of flexibility to alter course
 - Bring market information to bear when its available
 - Powerful analytic techniques based on stochastic processes
- The value v of a real (non-income producing) option that pays off $W(T)$ at future time T is given by the general formula:

$$v(t,T) = \exp(-r(T-t)) E[\max(0, W(T))]$$

where t is current time, E denotes the risk-neutral expected value, and r is the riskless discount rate.

Space Applications of Real-Options Techniques

- Making commercial satellites compatible with on-orbit servicing

Lamassoure, E., Saleh, J., and Hastings, D., “Space Systems Flexibility Provided by On-Orbit Servicing, Parts 1 and 2”, *Journal of Spacecraft and Rockets*, Vol. 39, No. 4, pp. 551-570, July-August 2002

- NASA technology evaluation

Shishko, R. Ebbeler, D., and Fox, G. “NASA Technology Assessment Using Real Options Valuation”, *Systems Engineering*, Vol. 7, No. 1, pp. 1-12, February 2004

Cost-Effectiveness Analysis

- A general approach to the analysis of alternatives (trade studies) that attempts to quantify selected MoEs and LCC
 - Focus on choosing the best alternative from among competing alternatives, not on whether the project is worth doing
 - Should not be confused with Cost-Benefit Analysis

- Heavy reliance on engineering, operations, and cost models/simulations

- Permits exploration of the tradespace for design optimization (batch mode)

Measures of Effectiveness

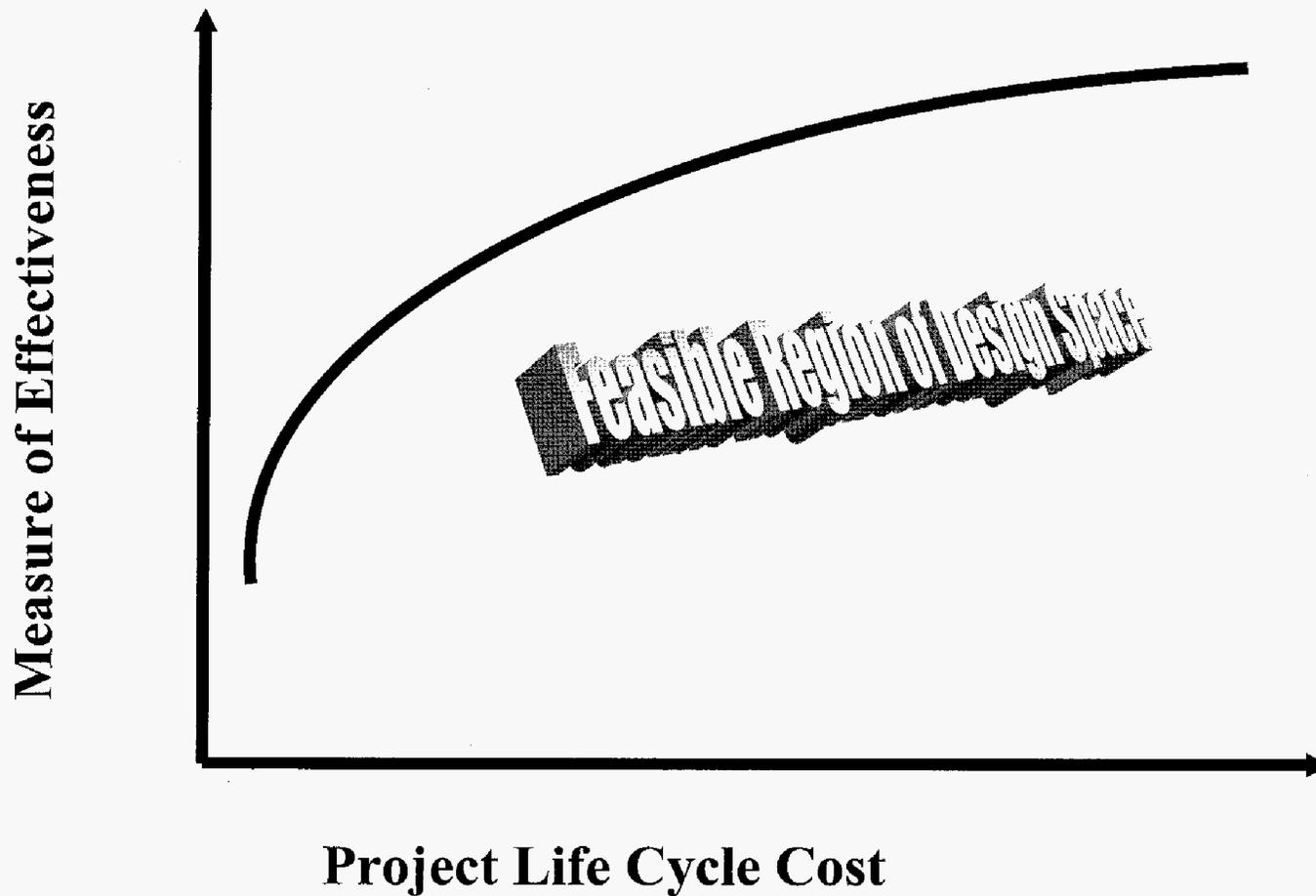
- A mission Measure of Effectiveness (MoE) is defined as a quantitative measure of the degree to which the mission's purpose is achieved.
 - For example, how much the mission returns in science data, and what the resolution and coverage are.

- It often depends on measures of performance (MoP), or size and quantity measures of several inputs. (A "Design Vector")
 - Launch vehicle probability of success and injected mass
 - EDL reliability
 - Instrument performance
 - Spacecraft subsystem (power, C&DH, telecomm, etc.) performance.

- Cost is NOT a Measure of Effectiveness; cost is a measure of the resources foregone.

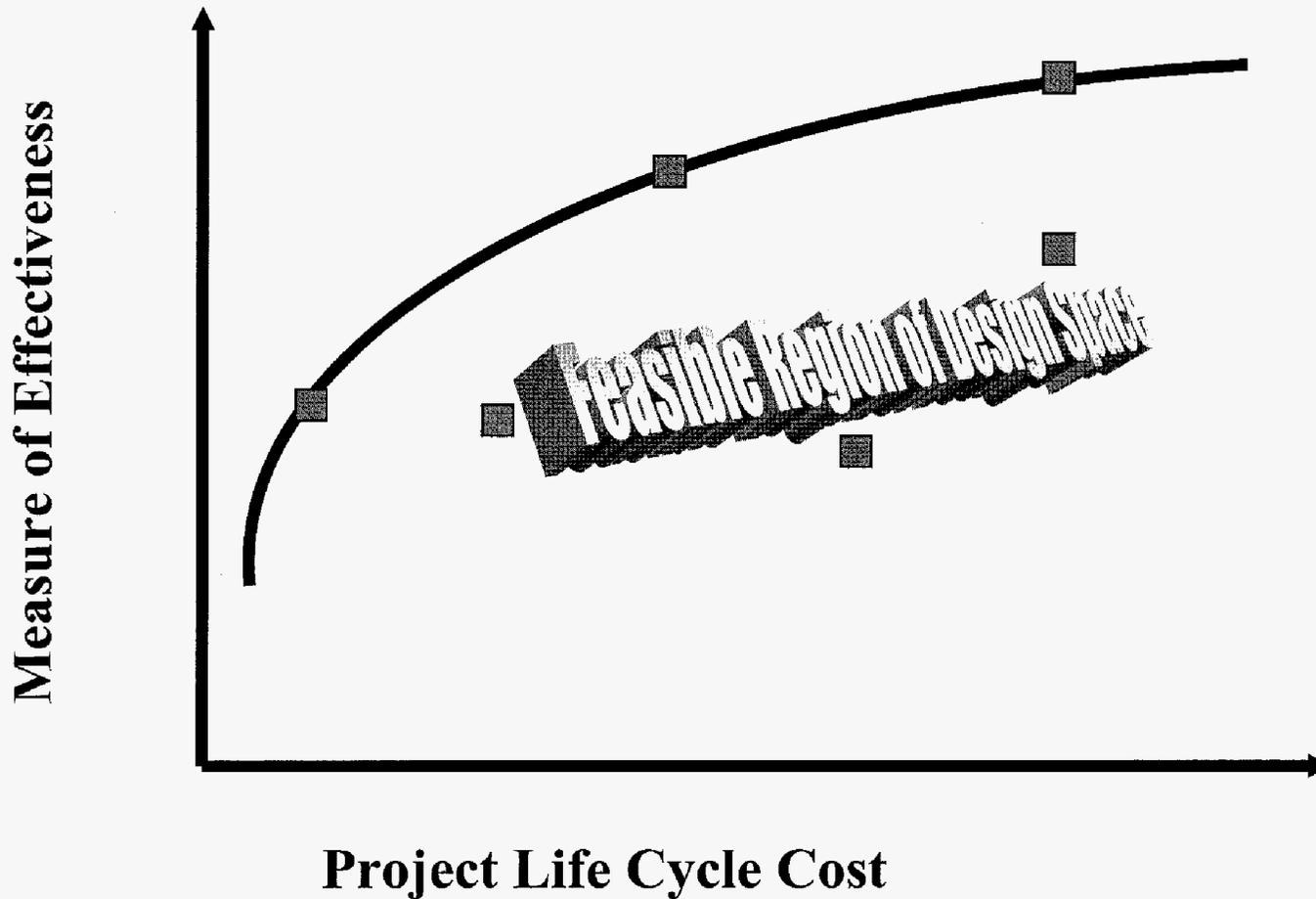
- Cost-effectiveness is represented by a combination of both -- preferably as the whole Pareto frontier, but sometimes as points on that frontier.

Cost-Effectiveness Frontier (Pareto Optimal Frontier)



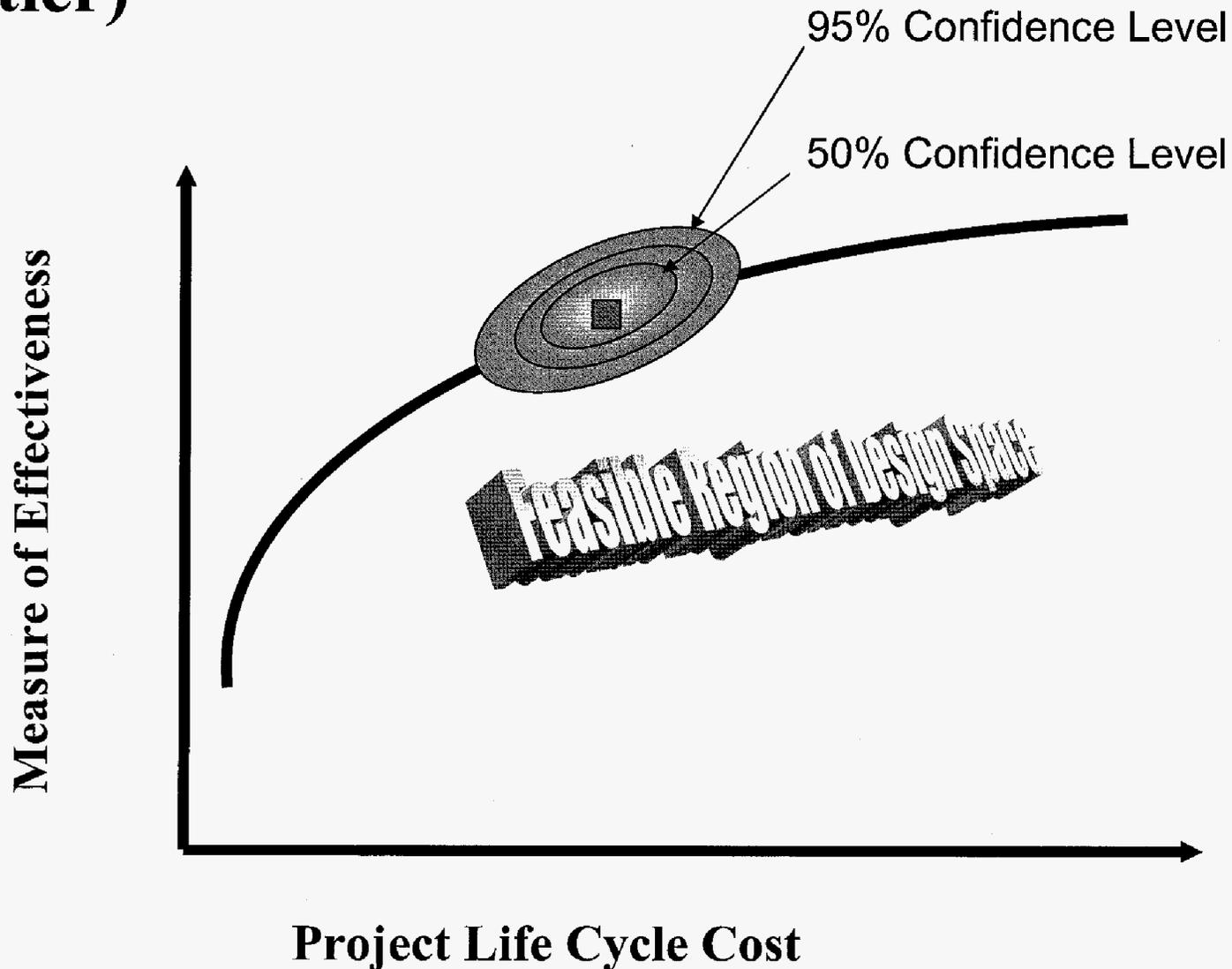
Level of effectiveness or level of cost usually is an external decision.

Cost-Effectiveness Frontier (Pareto Optimal Frontier)



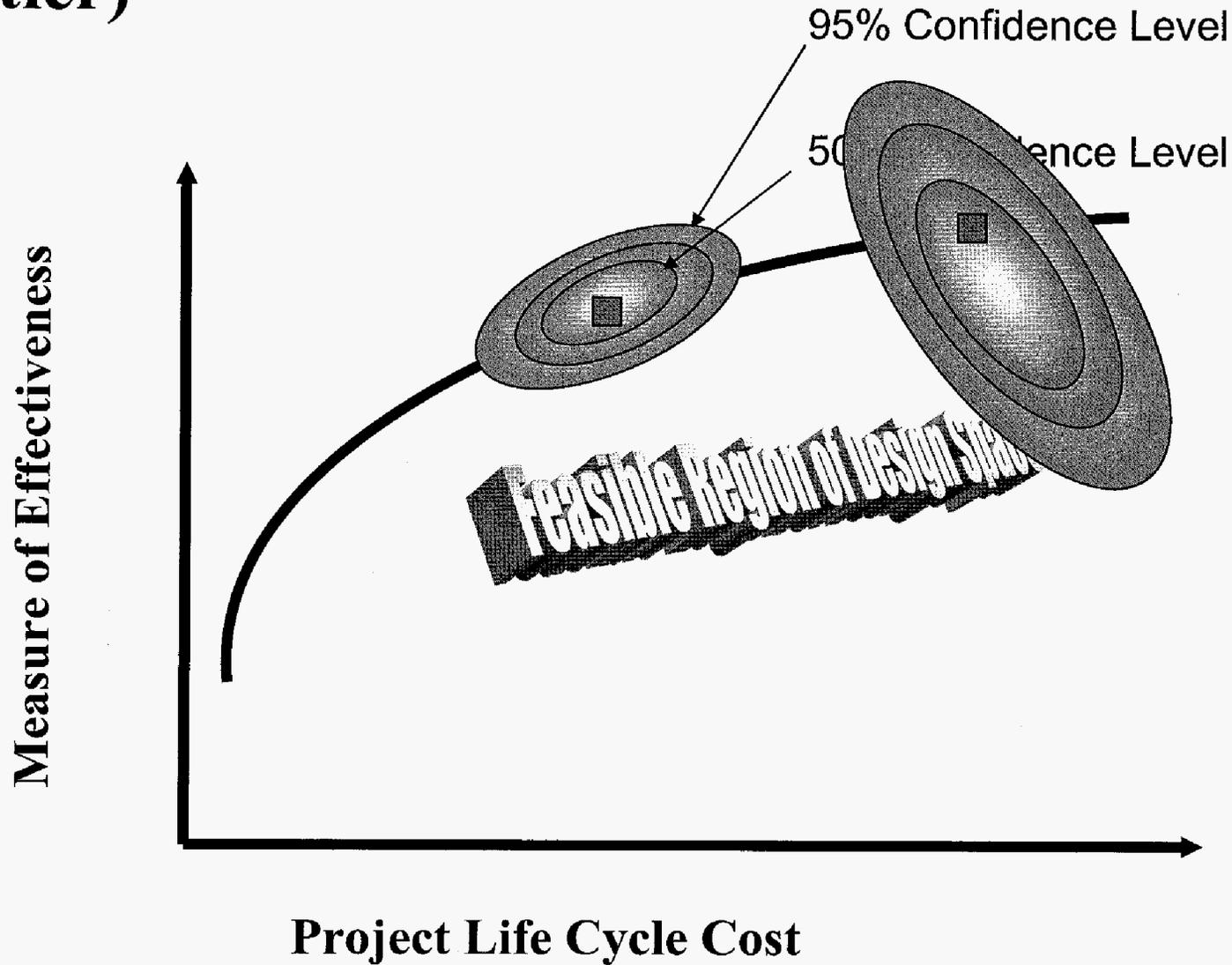
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Cost-Effectiveness Frontier (Pareto Optimal Frontier)



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Cost-Utility Analysis

- Similar in intention to cost-effectiveness, but uses a utility function to translate outcomes into a single decision variable
- Uses basic decision analysis concepts (vN-M utility theory) for decisions under risk.
- Decision analysis rests on 4 (sometimes 5) axioms requiring rationality when making decisions under risk.
- Useful when outcomes are uncertain and decision maker is not risk neutral

Cost-Utility Analysis: Technical Aspects

- Alternative with the highest expected utility is preferred

$$E(U) = \sum_i p_i U(X_i)$$

$$CE = X \text{ such that } U(X) = E(U)$$

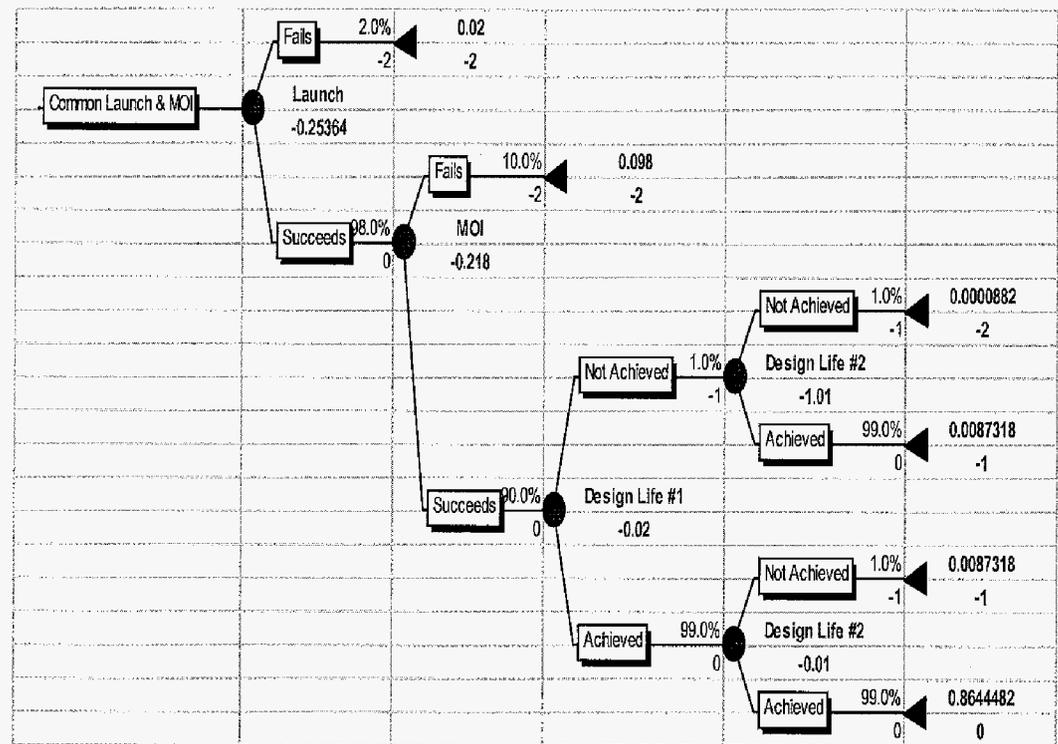
- Typically, decision trees are used to represent alternatives and probabilities of risk events
- Functional form selected for convenience
- Exponential form:

$$U(X) = a - be^{-rX} \quad r, b > 0$$

- Parameters are determined by asking decision maker to compare various “lotteries”

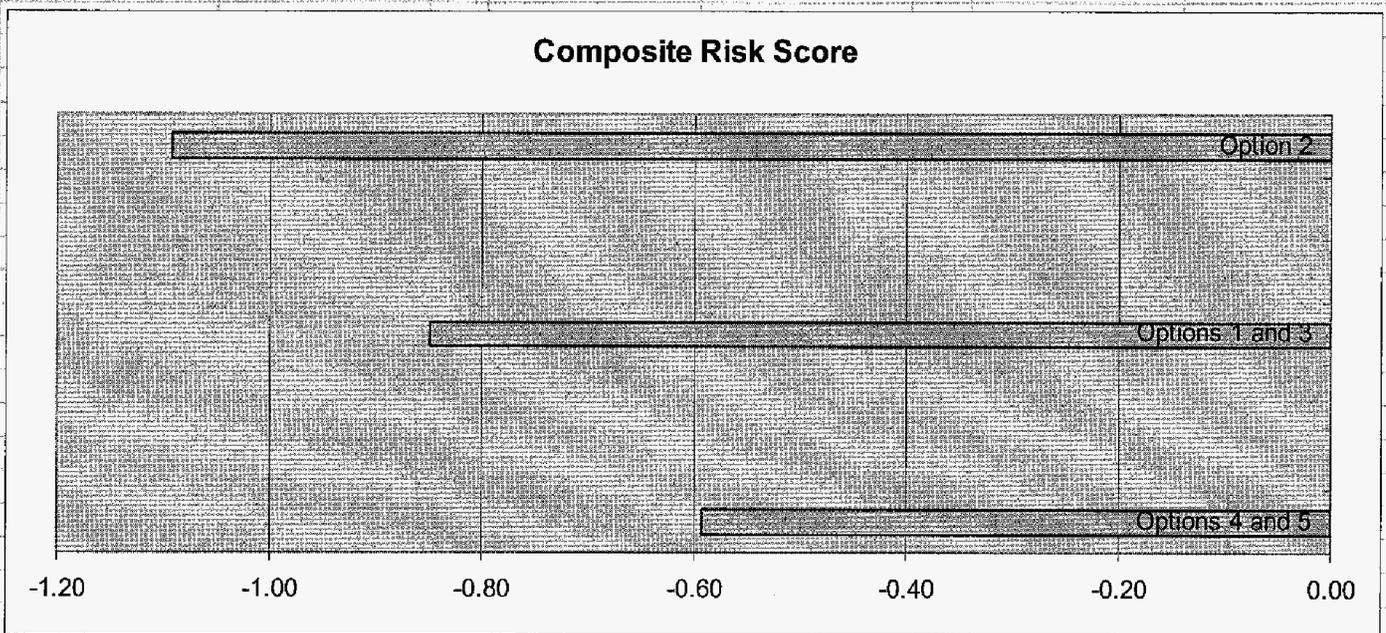
Mars Risk Model: Decision Tree

- Risk model building process involves (for each option):
 - Identify key risk events
 - Develop a decision tree
 - Solve tree
 - Calculate certainty equivalent (CE) as risk metric



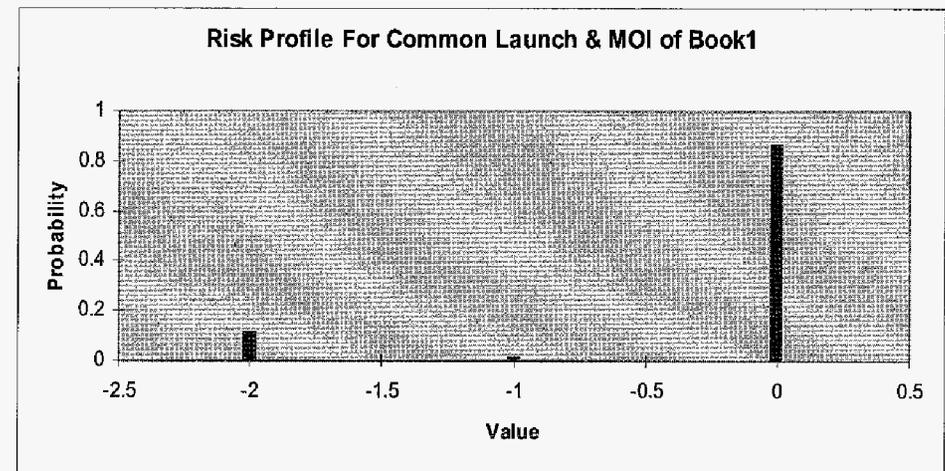
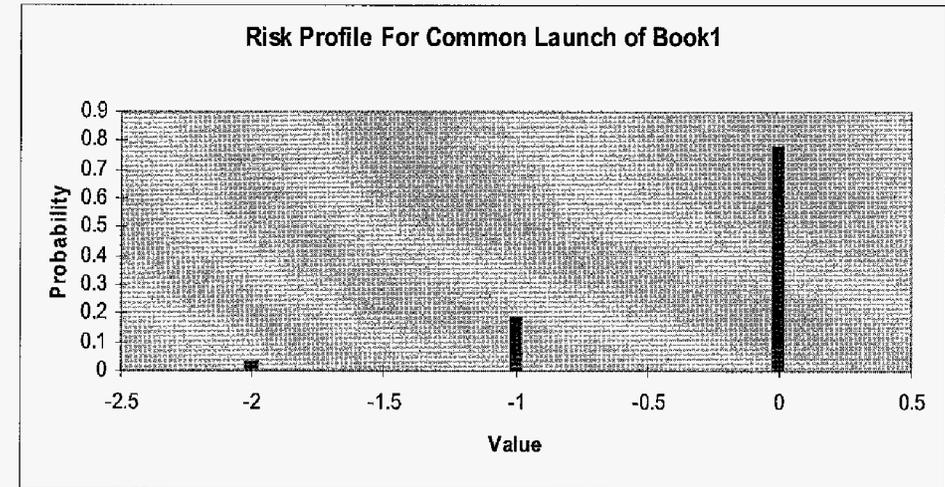
Mars Risk Model: Spreadsheet Version

Project:	Mars '03		Risk Assessment		Risk Aversion	2.00
Option Name:	Options 4 and 5		Options 1 and 3		Option 2	
Option Description:	2 Delta launches; 2 separate S/C		1 launch; 2 separate S/C		1 launch; 1 S/C thru MOI	
Events	Likelihood of Occurrence	Severity of Consequences	Likelihood of Occurrence	Severity of Consequences	Likelihood of Occurrence	Severity of Consequences
S/C #1 Launch Failure	2.00%	Lose one S/C	5.00%	Lose two S/C	5.00%	Lose two S/C
S/C #1 MOI Failure	10.00%	Lose one S/C	10.00%	Lose one S/C	10.00%	Lose two S/C
S/C #1 Design Life Failure	1.00%	Lose one S/C	1.00%	Lose one S/C	1.00%	Lose one S/C
S/C #2 Launch Failure	2.00%	Lose one S/C				
S/C #2 MOI Failure	10.00%	Lose one S/C	10.00%	Lose one S/C		
S/C #2 Design Life Failure	1.00%	Lose one S/C	1.00%	Lose one S/C	1.00%	Lose one S/C
	Options 4 and 5		Options 1 and 3		Option 2	
Composite Risk Score	-0.69		-0.85		-1.09	



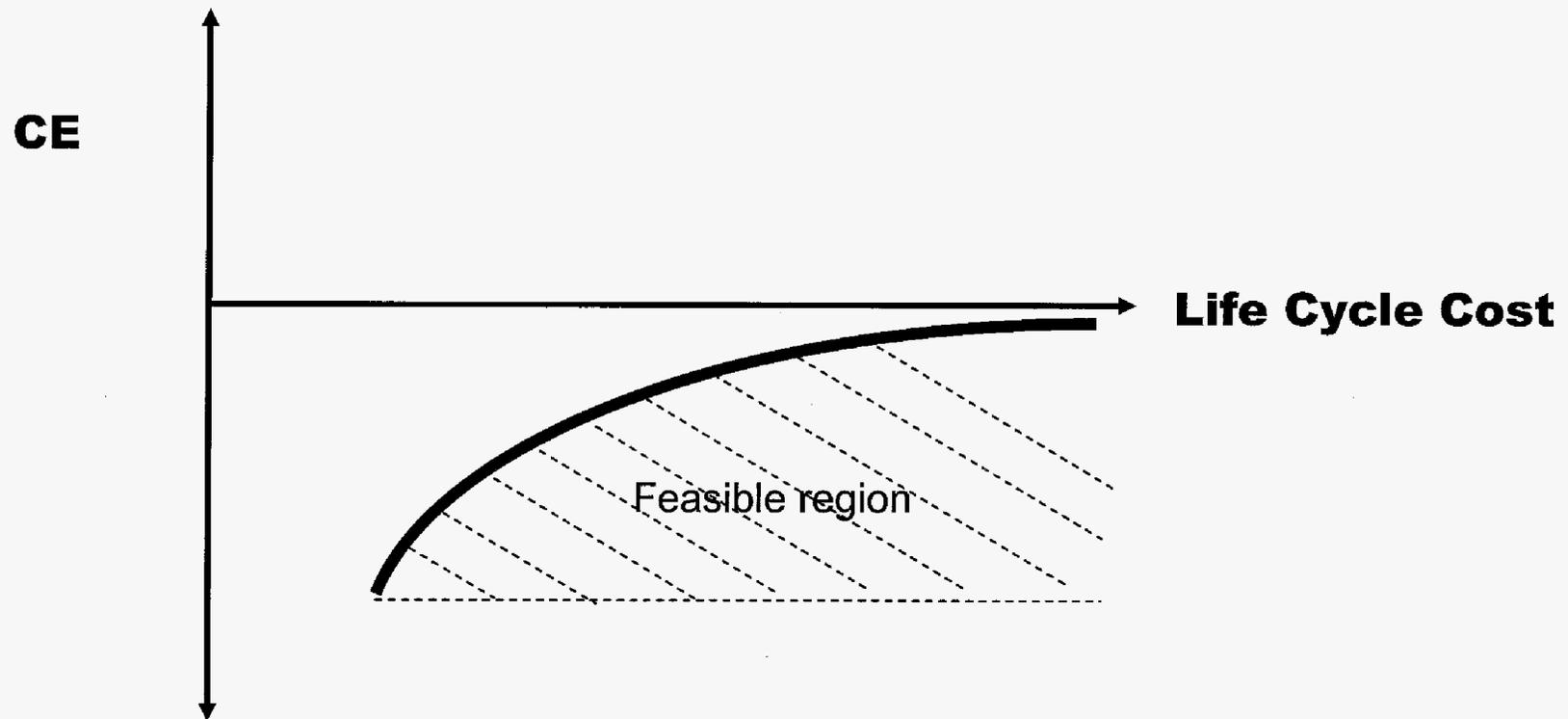
Mars Risk Model: What is the Risk Metric?

- Risk metric is “certainty equivalent” (CE)
 - Amount (outcome) that makes the decision-maker indifferent between accepting the “gamble” (shown at right) or taking with certainty
 - Comparing CEs provides a rational way of selecting alternatives; the higher CE option is always preferred
 - Mathematically related to the concepts of “risk aversion” and “risk premium”
- In Mars Risk Model spreadsheet, CE units are “spacecraft lost”, hence a less negative number is preferred



Mars Risk Model: Adding Costs

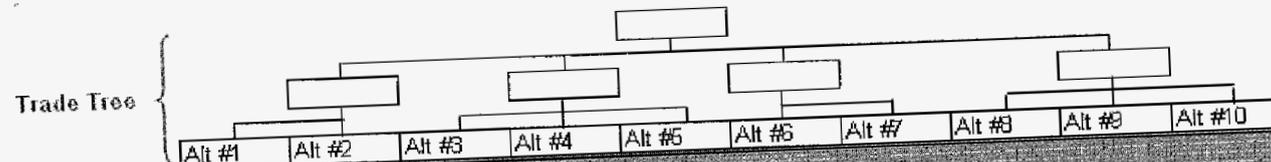
Add costs to develop the Pareto optimal frontier



Multi-Attribute Utility Theory

- Extends utility concept to situations with multiple, usually conflicting, decision attributes
- Select a set of decision attributes (criteria)
 - Recommend between 3 and 10
 - Attributes can be quantitative or qualitative
 - Getting the right ones can be difficult; need to distinguish among alternatives and should be orthogonal
- Create weights; different groups might disagree on weights
- Develop individual utility functions for each attribute
 - $u(\text{best}) = 1.0$; $u(\text{worst}) = 0.0$
- Calculate total rating (score) for each alternative

Multi-Attribute Utility Theory: Mars Exploration Rovers Example



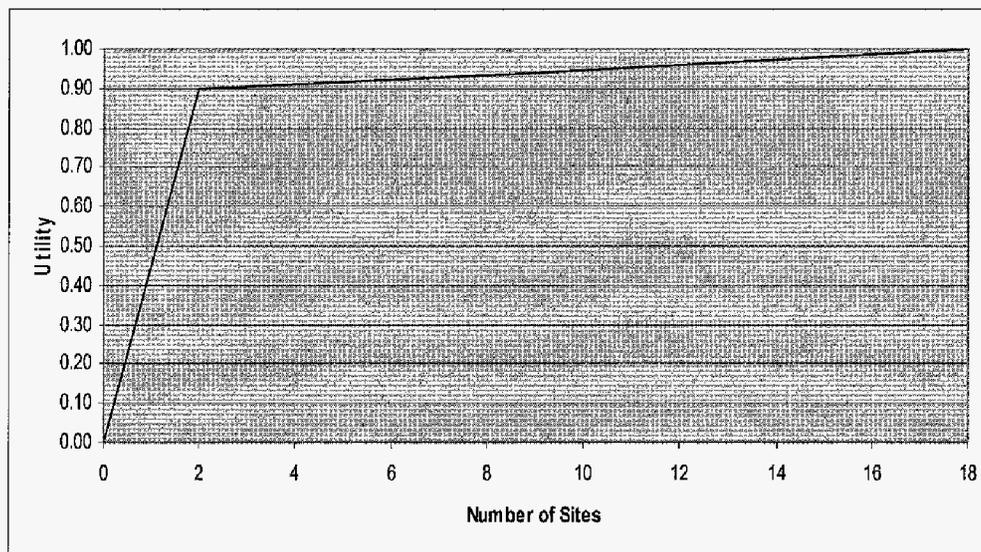
Decision Attributes

Major Criteria	Weight	Sub-criteria	Weight	Alt #1	Alt #2	Alt #3	Alt #4	Alt #5	Alt #6	Alt #7	Alt #8	Alt #9	Alt #10
Mission Reliability/Survivability	40%	Launch Reliability	30%										
		EDL Survivability	35%										
		Surface Reliability	35%										
Science Diversity	30%	E[Mission Duration]	20%										
		Num of Sites	50%										
		Mission Distance	30%										
Affordability	10%		100%										
Operability	20%	Downlink Thruput @3db	50%										
		Decision Cycle Time	25%										
		Command Complexity	25%										
Rating													

$$Rating_j = \sum_i w_i u_{i,j}$$

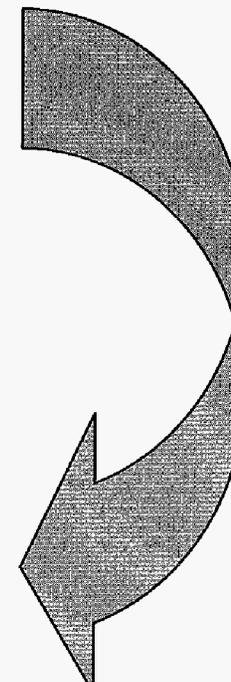
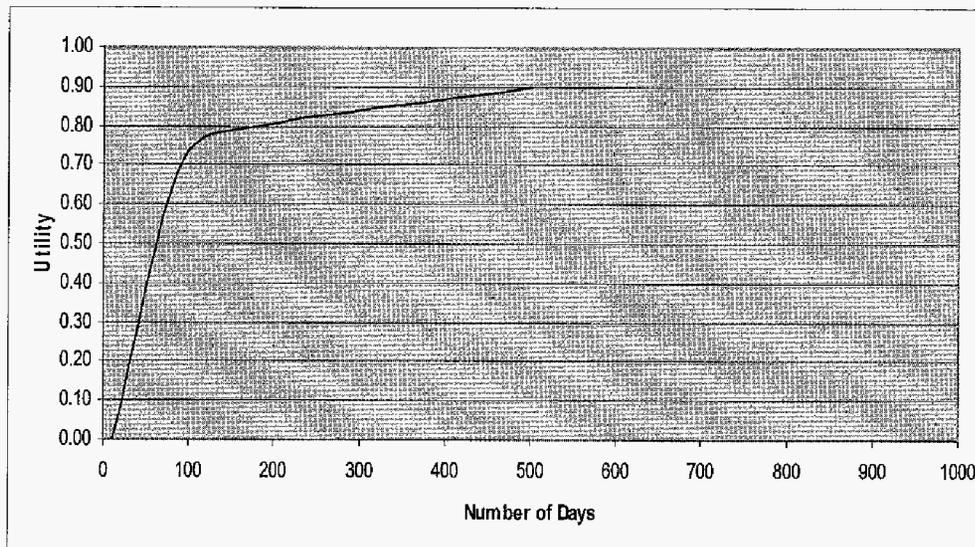
Mars Exploration Rovers Example

Attribute	Max	Min	Value	Utility
<i>Experiment 1</i>				
Number of Sites	18	0	0	0.00
			2	0.90
			9	0.94
			15	0.98
			18	1.00
<i>Experiment 3</i>				
Number of Sols	1000	10	10	0.00
			90	0.70
			180	0.80
			500	0.90
<i>Experiment 4</i>				
Distance Traversed	80	0	0	0.00
			10	0.80
			40	0.96
			60	0.98
			80	1.00



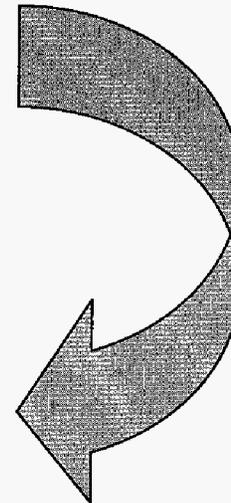
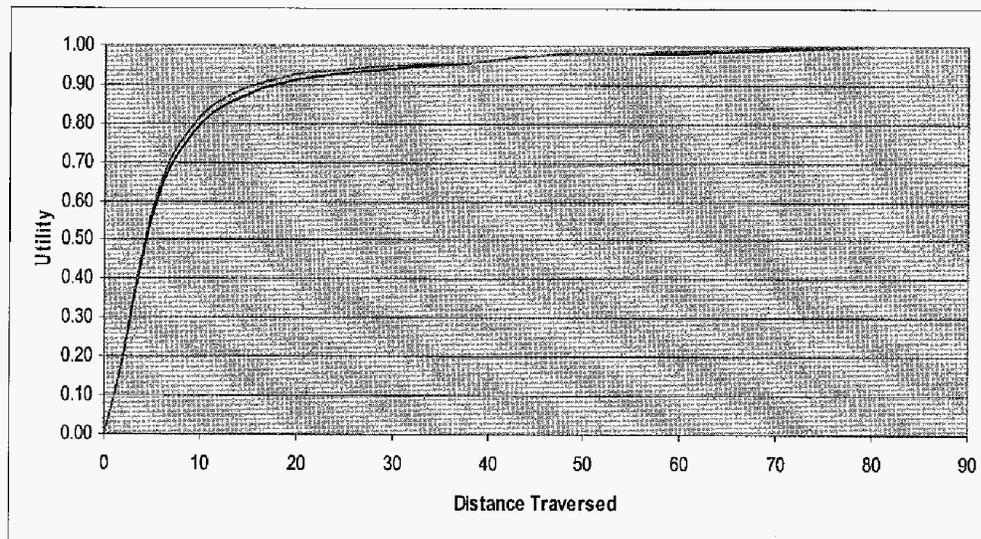
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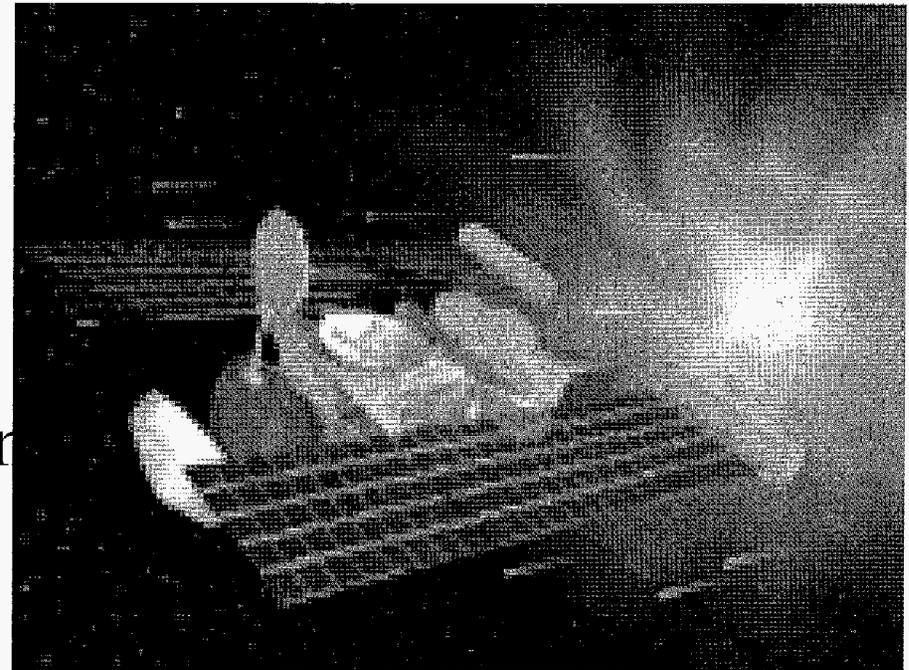
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			40	0.96
			60	0.98
			80	1.00



Comet Flyby Example

- Scientists want to get close to obtain better samples, but are aware of the risks.
- Engineers are concerned about losing the s/c, so they want to stand-off.
- Assume a flyby closest approach between 100 km and 1,000 km of the nucleus are acceptable to both Project Engineers and Project Scientists.



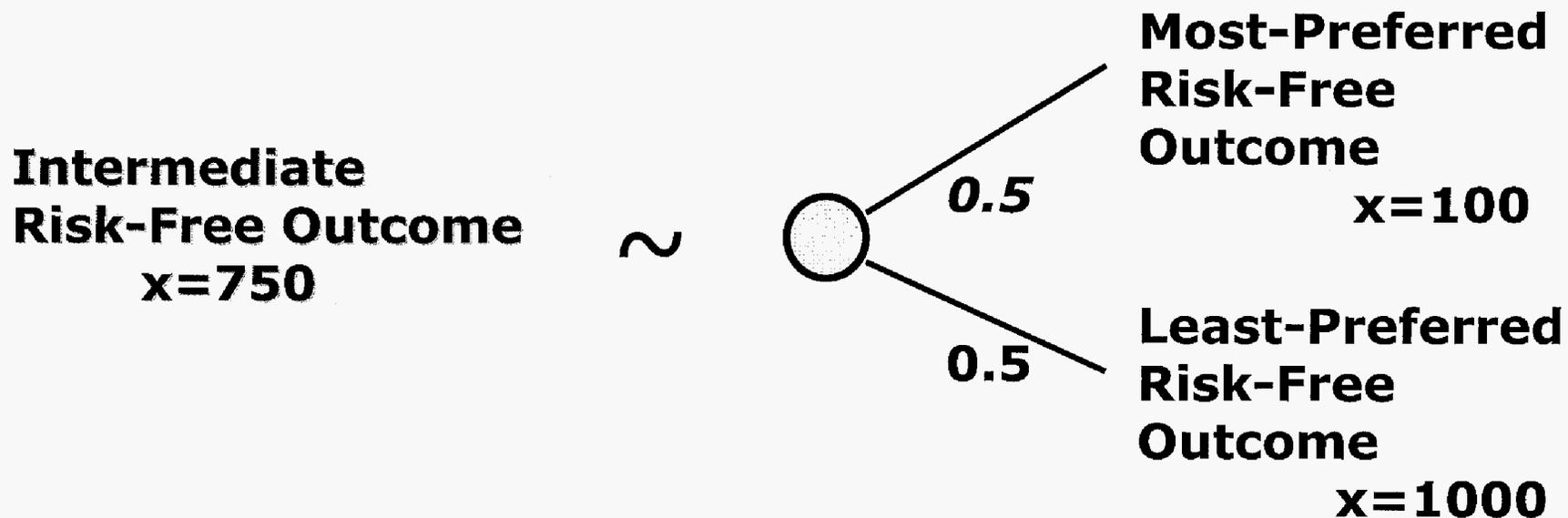
Project Manager's Decision Dilemma

- Different decisions favor different groups
- Individual utility functions need to be quantified
- Need to combine the individual utility functions into a single (SW) function to be “optimized.”
- Largely a “political” decision

Comet Flyby Example: Scientist's Utility Function

Hypothetical Example to Illustrate the Theory

$$u(x=1000) = 0.0 \quad u(x=100) = 1.0$$



Then $u(x=750) = 0.5$, from above indifference relationship

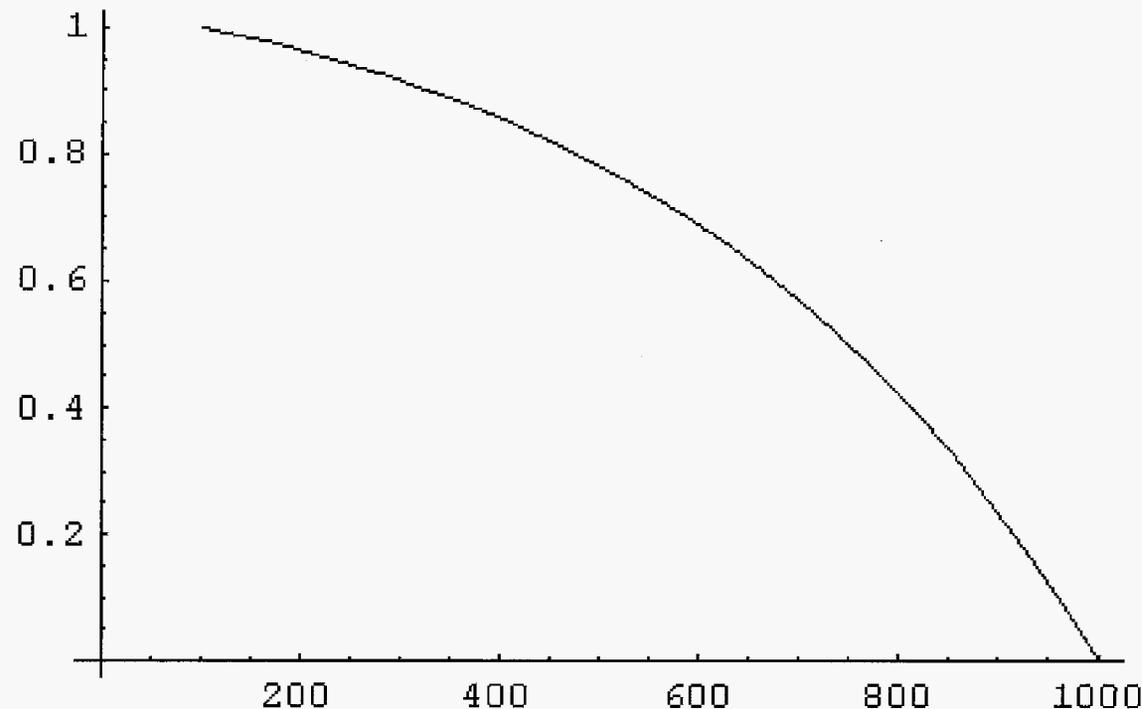
Using the constant risk aversion utility function:

$$u(x) = 1.145(1 - e^{-0.0023(1000-x)})$$

Science Value for Comet Flyby

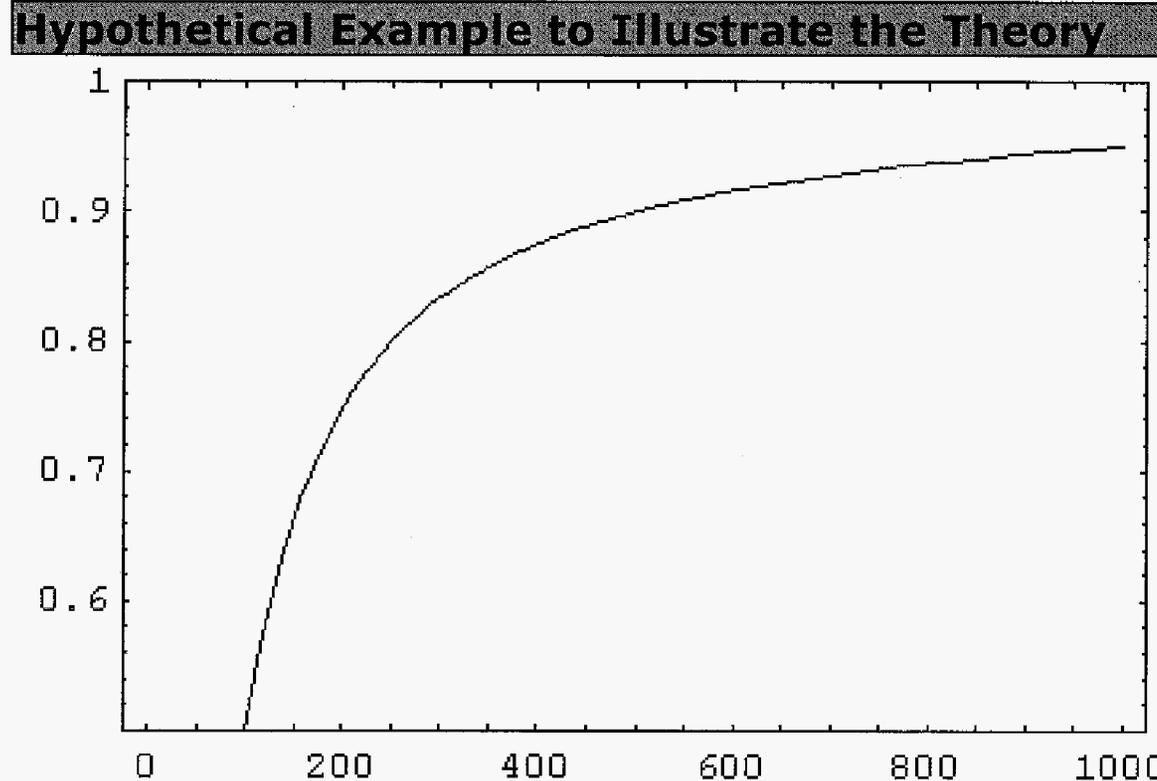
Assume indifference between science at 750 km and a 50/50 gamble between 100 km and 1,000 km. This yields the curve for $v(x)$.

Hypothetical Example to Illustrate the Theory



Mission Failure Risk for Comet Flyby

Assume risk increases as $(a + b/x)$, where x is the flyby closest-approach distance.

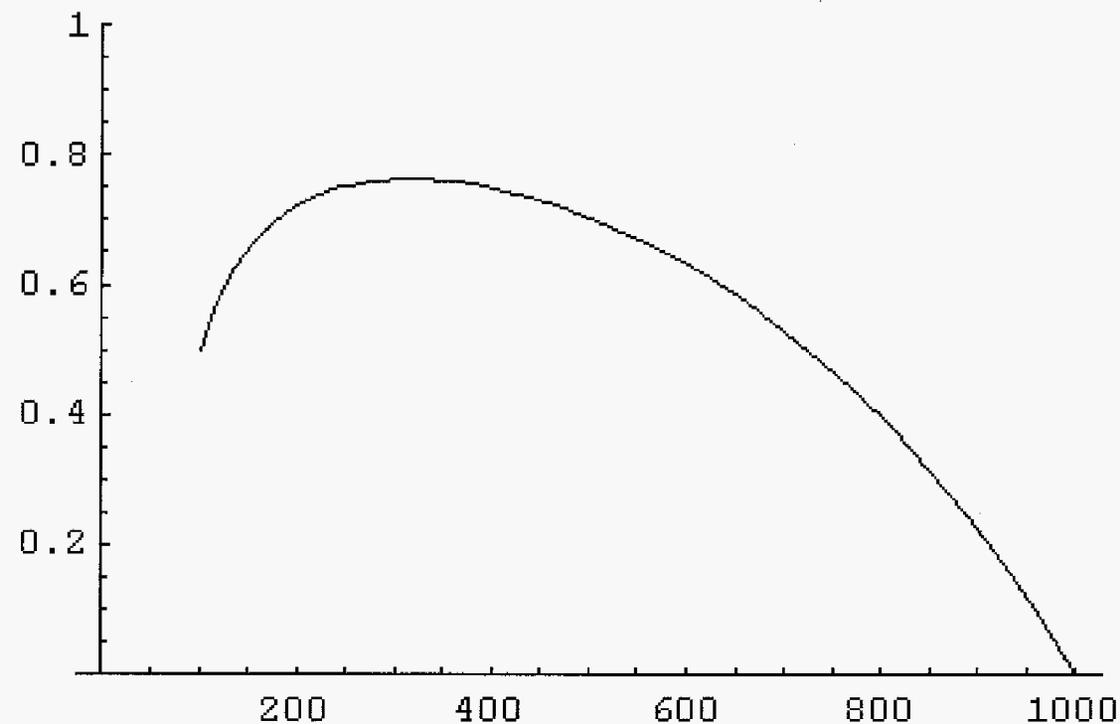


In this figure, $p(x) = 0.5$ @ 100 km and 0.95 @ 1,000 km.

Utility Function for Comet Scientists

Expected Science Value = $p(x) v(x)$

Hypothetical Example to Illustrate the Theory



The maximum Expected Science Value is at 315 km.

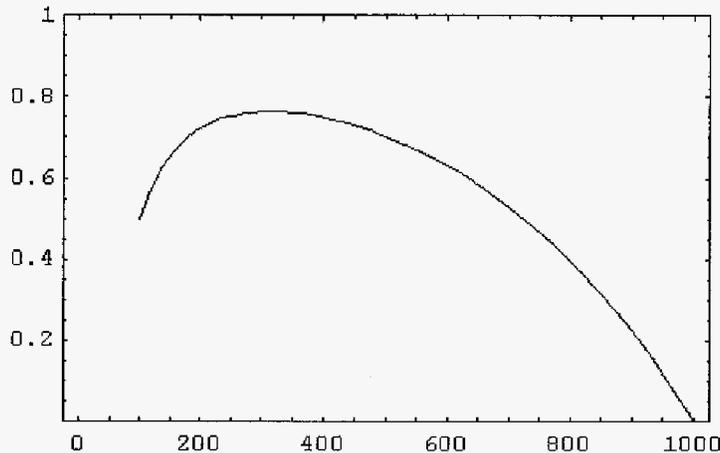
Assume the Following Social Welfare Function

$$\begin{aligned} SW(U_{\text{Scientists}}, U_{\text{Engineers}}) &= kU_{\text{Scientists}} + (1-k)U_{\text{Engineers}} \\ &= kp(x)v(x) + (1-k)p(x) \\ 0 &\leq k \leq 1 \end{aligned}$$

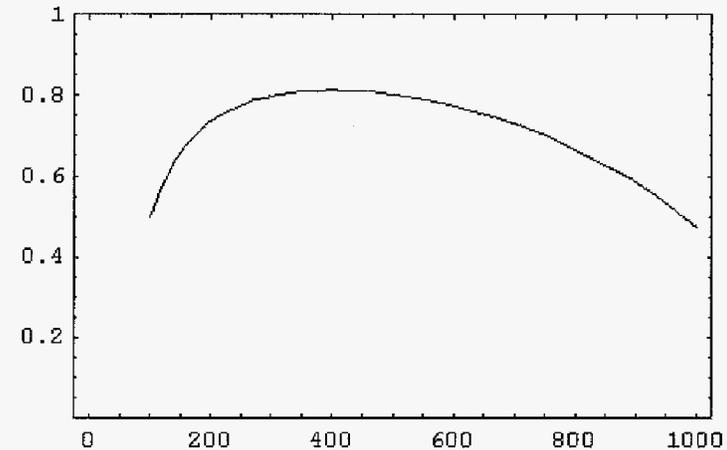
A high value of k favors the project scientists; a low value favors the project engineers in the social welfare function

Different Social Welfare Curves for Comet Flyby

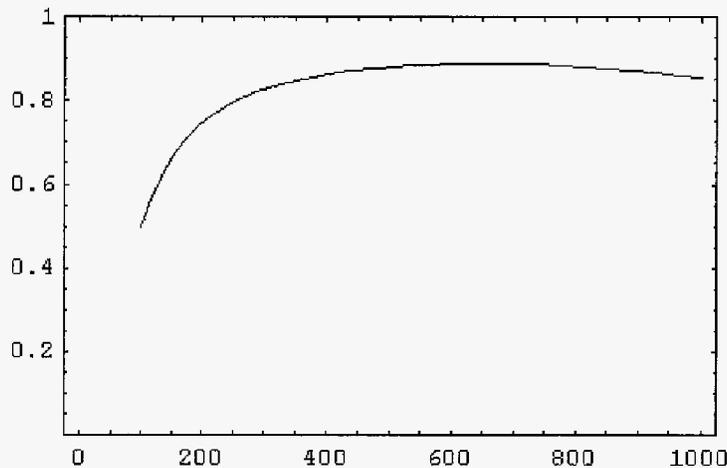
Hypothetical Example to Illustrate the Theory



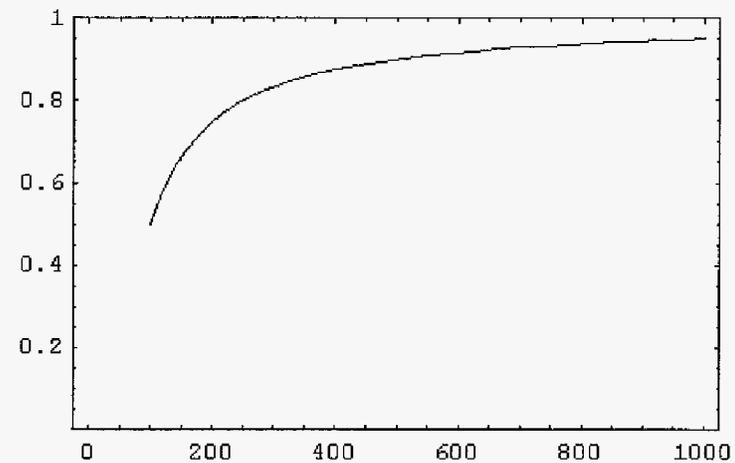
For $k = 0.0$, SW Max @ 315 km. Risk = 0.16



For $k = 0.5$, SW Max @ 400 km. Risk = 0.13



For $k = 0.9$, SW Max @ 658 km. Risk = 0.08

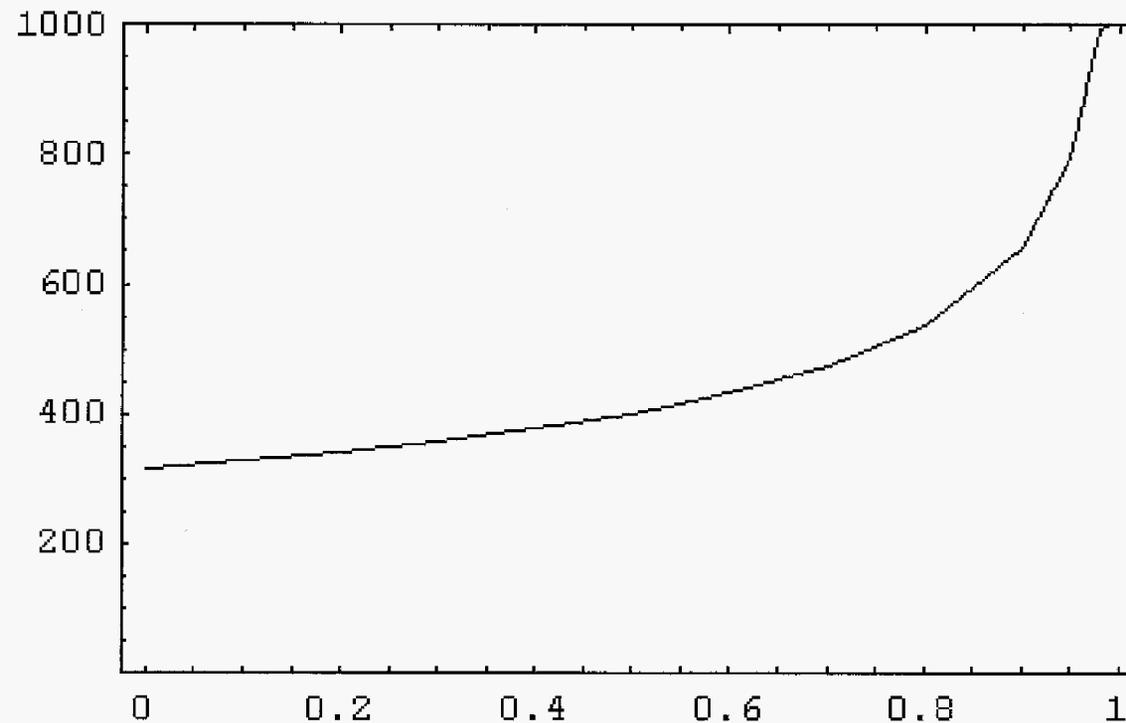


For $k = 1.0$, SW Max @ 1,000 km. Risk = 0.05

Optimum Flyby Altitude vs. Risk Factor r

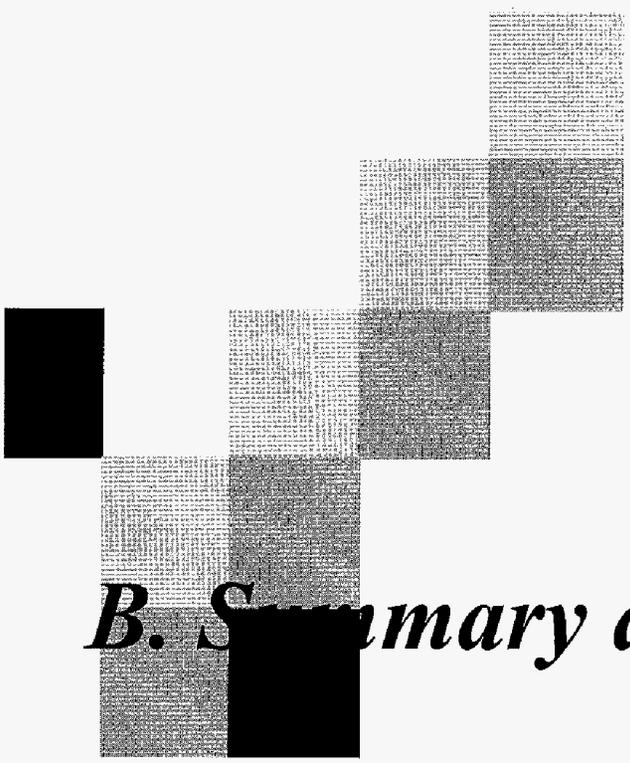
As the Risk Factor k varies from 0 to 1, the optimum flyby altitude varies from 315 km to 1,000 km.

Hypothetical Example to Illustrate the Theory



Analytic Hierarchy Process (AHP)

- In AHP as in MAUT, a rating (score) is determined for each alternative.
- As in MAUT, a hierarchy of decision attributes is developed.
 - Decision attributes are evaluated in a series of pair-wise comparisons
 - Comparisons are usually on a qualitative scale, which is then converted to a quantitative one
- Multiple experts make separate pair-wise comparisons of the alternatives with respect to each decision attribute.
- An AHP software program then computes the rating.
- Consensus may take several iterations, or may not achieve consistency.



B. Summary and Observations

Summary and Observations

- Variety of DM available; apply the one that best fits the decision problem
- Have the right mix of skills as part of the analysis team
- Physics-based models are fairly reliable; costs of alternatives are the hardest to estimate
- Treatment of uncertainty is an important part of most decision problems in space SE and BCA
- Document analysis assumptions, alternatives, data, methods, models, and decision results to avoid unnecessary replication



Tradespace Exploration: Some Fundamental Concepts and Models

Robert Shishko, Jet Propulsion Laboratory,
California Institute of Technology

Tradespace Exploration Outline

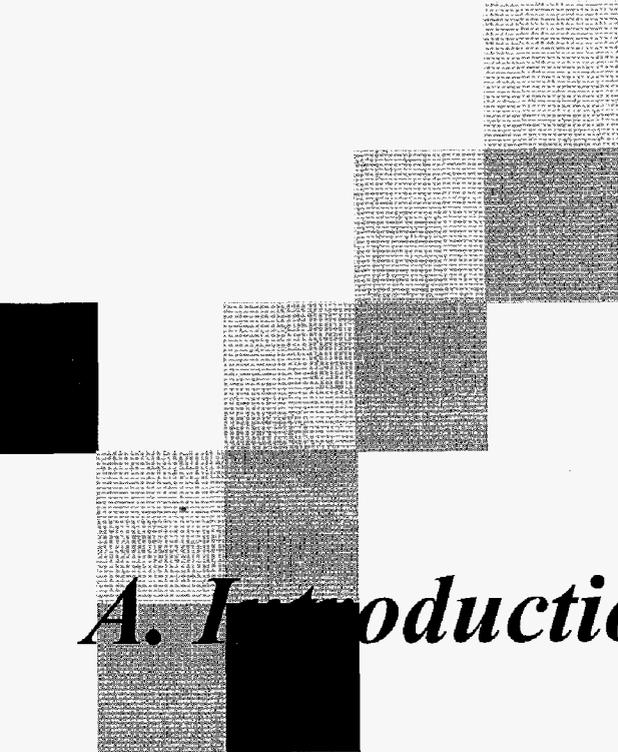
A. Introduction and Fundamental Concepts

- What is a Tradespace?
- Project tradespace modeling
- Pareto optimality and Pareto frontiers
- Iso-effectiveness curves
- Issues in building a Project Tradespace Model

B. Building a Project Tradespace Model for Mars Science Laboratory

C. Tradespace Visualization

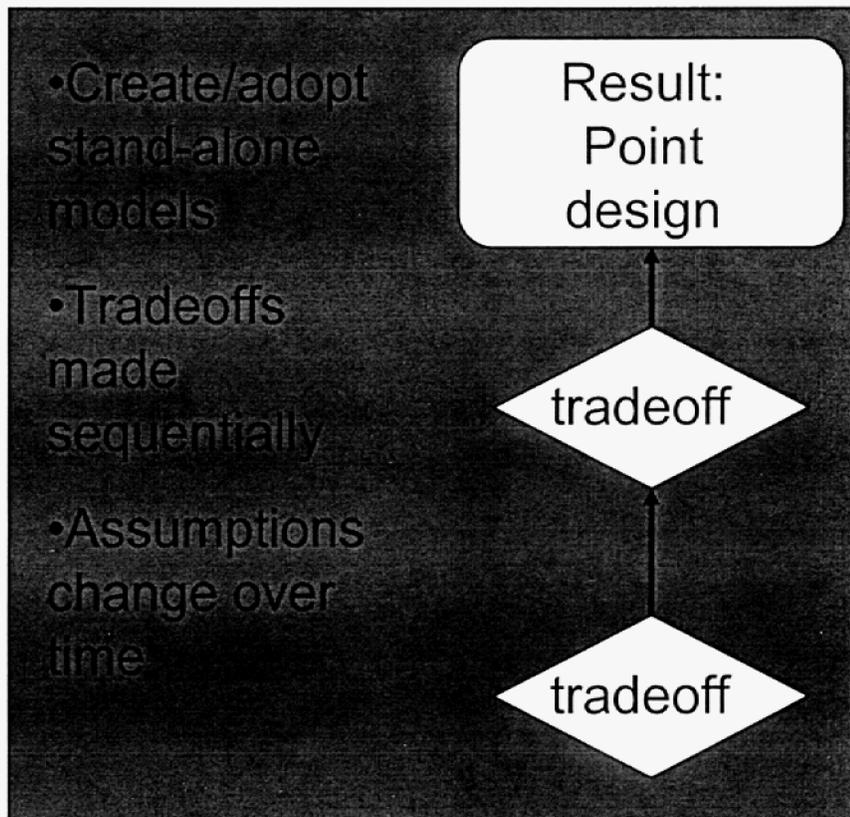
D. Tradespace Mini-Workshop



A. Introduction and Fundamental Concepts

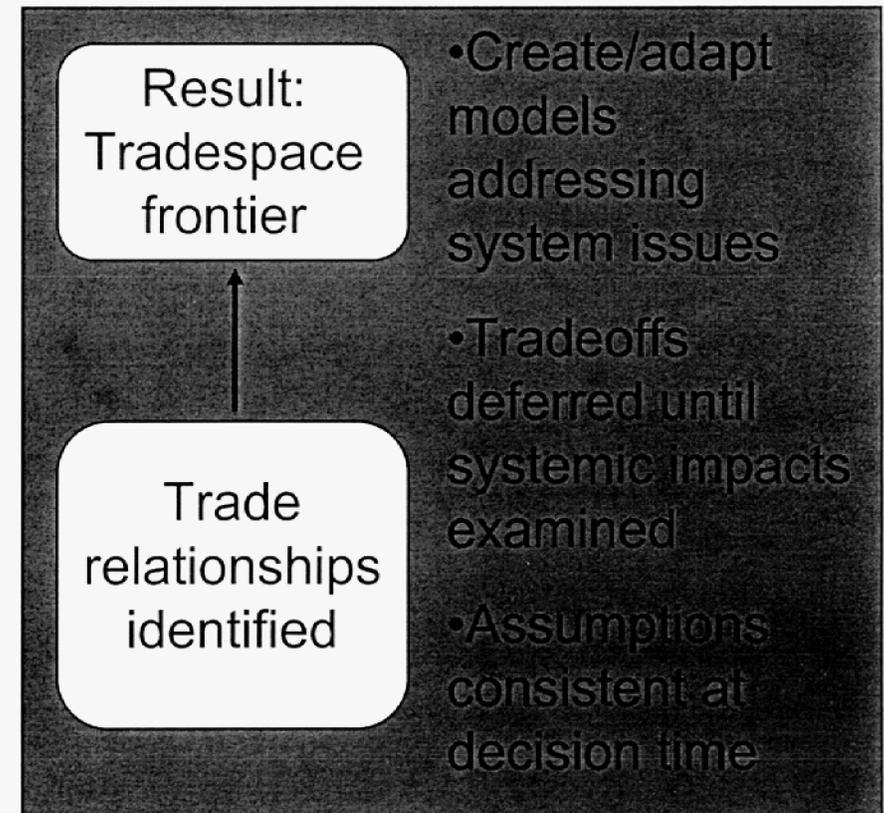
Old Problem, New Solution

Traditional Approach



< 5 designs

Tradespace Modeling Approach



100-1000 designs

time ↑

What is a Tradespace?

- A *project tradespace* is a complete, quantitative description of the alternatives available to a project based on potentially available technologies, in terms of design and operational inputs, and project outcomes.
 - **Inputs:**
 - System and subsystem (e.g., spacecraft and instrument) performance ratings, sizes, quantities, etc.
 - Operational scenario parameters, orbits, launch dates, etc.
 - **Outputs:**
 - Project/mission Measures of Effectiveness, including mission risks (chances of mission success),
 - Life-cycle costs, including cost risk

- A project tradespace model (PTM) has been used in previous projects/programs to integrate these considerations.
 - Space Station Freedom
 - Pluto Flyby
 - Europa Orbiter
 - Mars Science Laboratory

Project Tradespace Model (PTM)

- Captures project/system design relationships and associated *life-cycle cost* information
- Calculates project-level implications of design changes for trade studies (sensitivities)
- Permits rapid exploration of the tradespace neighborhood for design optimization (batch mode)

Measures of Effectiveness

- A mission Measure of Effectiveness (MoE) is defined as a quantitative measure of the degree to which the mission's purpose is achieved.
 - For example, how much the mission returns in science data, and what the resolution, and coverage are.

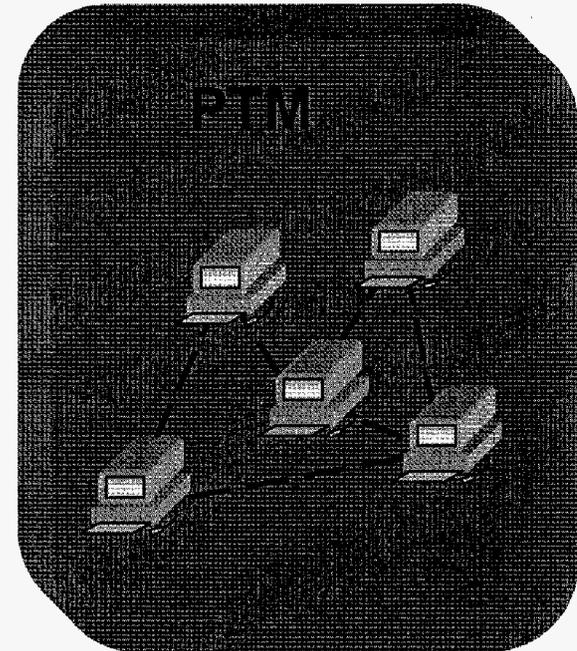
- It often depends on measures of performance (MoP), or size and quantity measures of several inputs. (A "Design Vector")
 - Launch vehicle probability of success
 - Injection, cruise, and EDL reliability
 - Navigation accuracy
 - Instrument performance
 - Spacecraft subsystem (power, C&DH, telecomm, etc.) performance.

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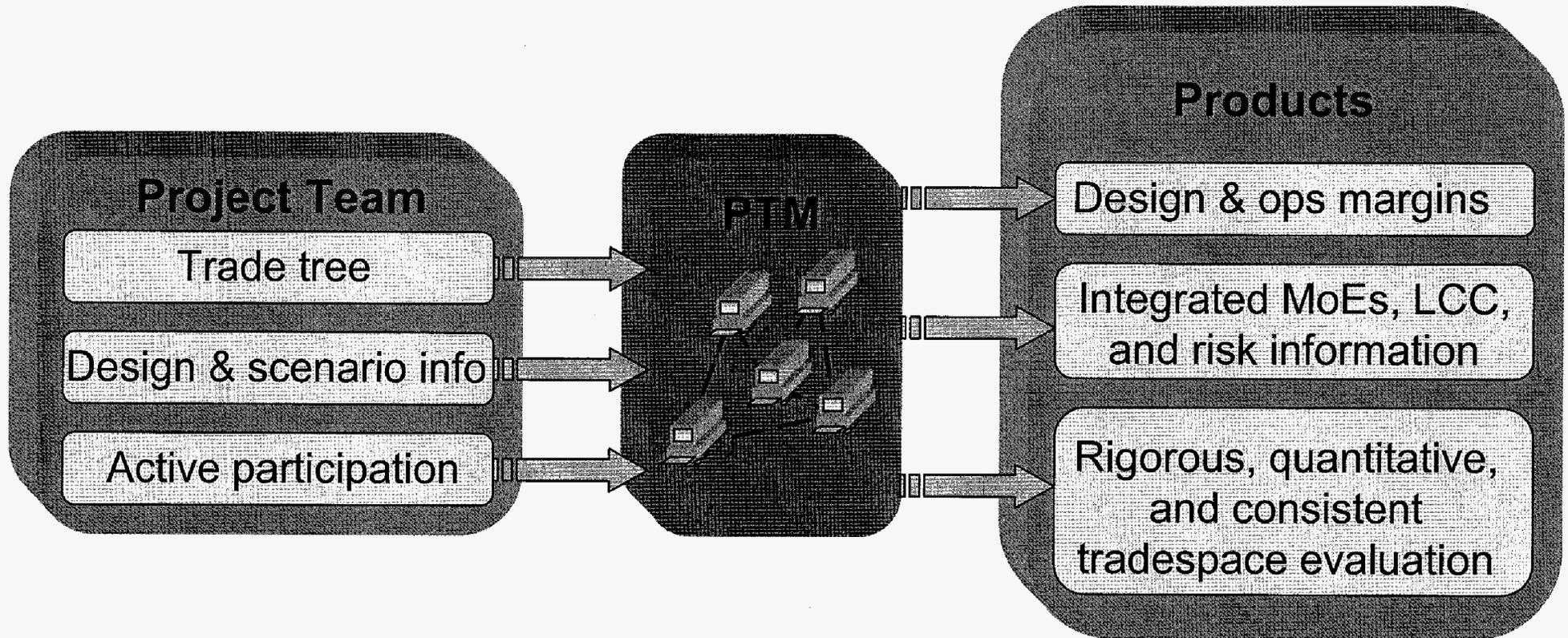
- Cost-effectiveness is represented by a combination of both -- preferably as the whole Pareto frontier, but sometimes as a point on that frontier.

Building a Project Tradespace Model (PTM)

- Identify the project/system MoEs and MoPs
- Identify Life-Cycle Cost (LCC) drivers
- Determine the “threads of calculation” needed to quantify these
- Implement the calculations by integrating appropriate space system design, cost, and operations models / simulations

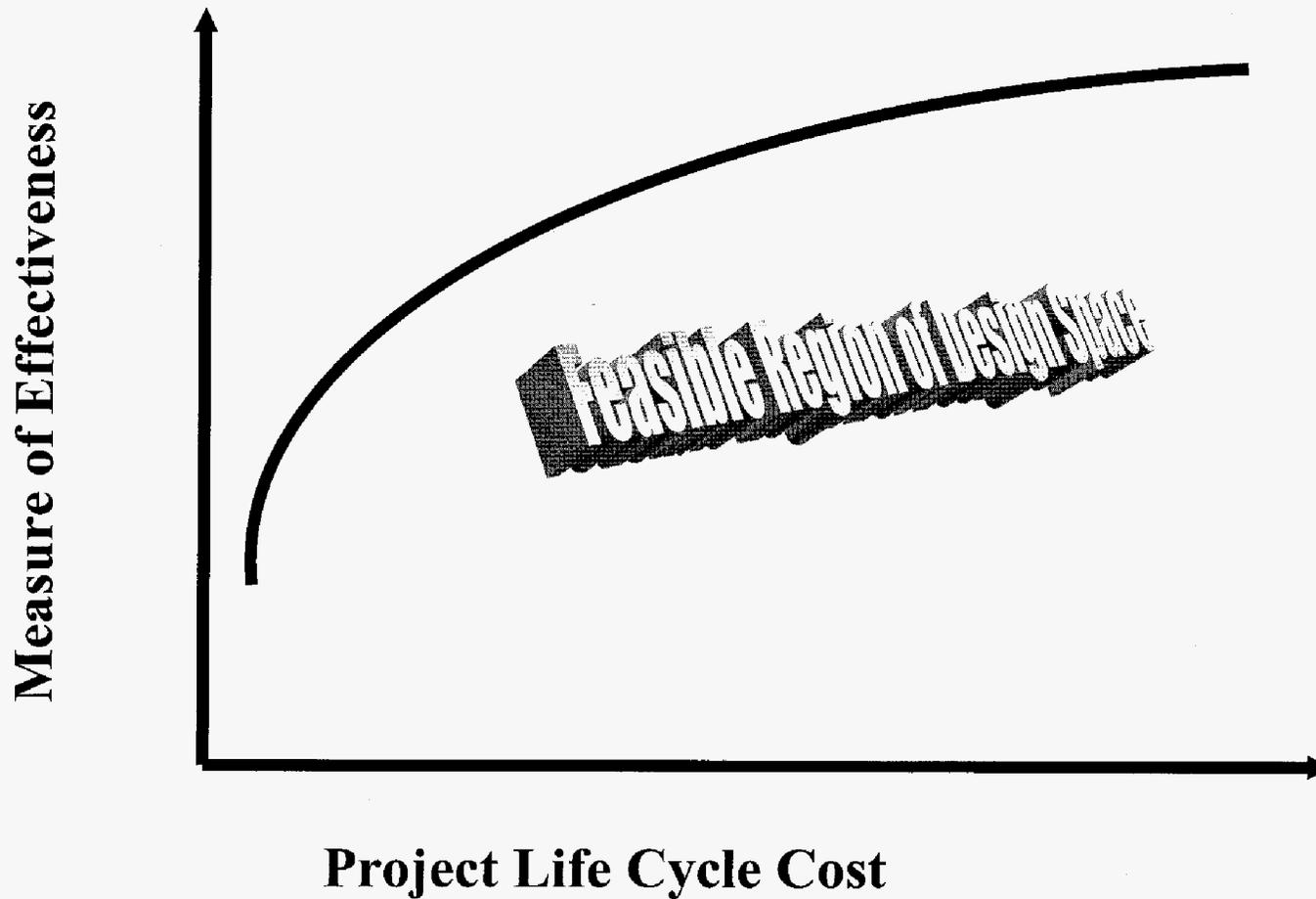


PTM: Inputs and Outputs



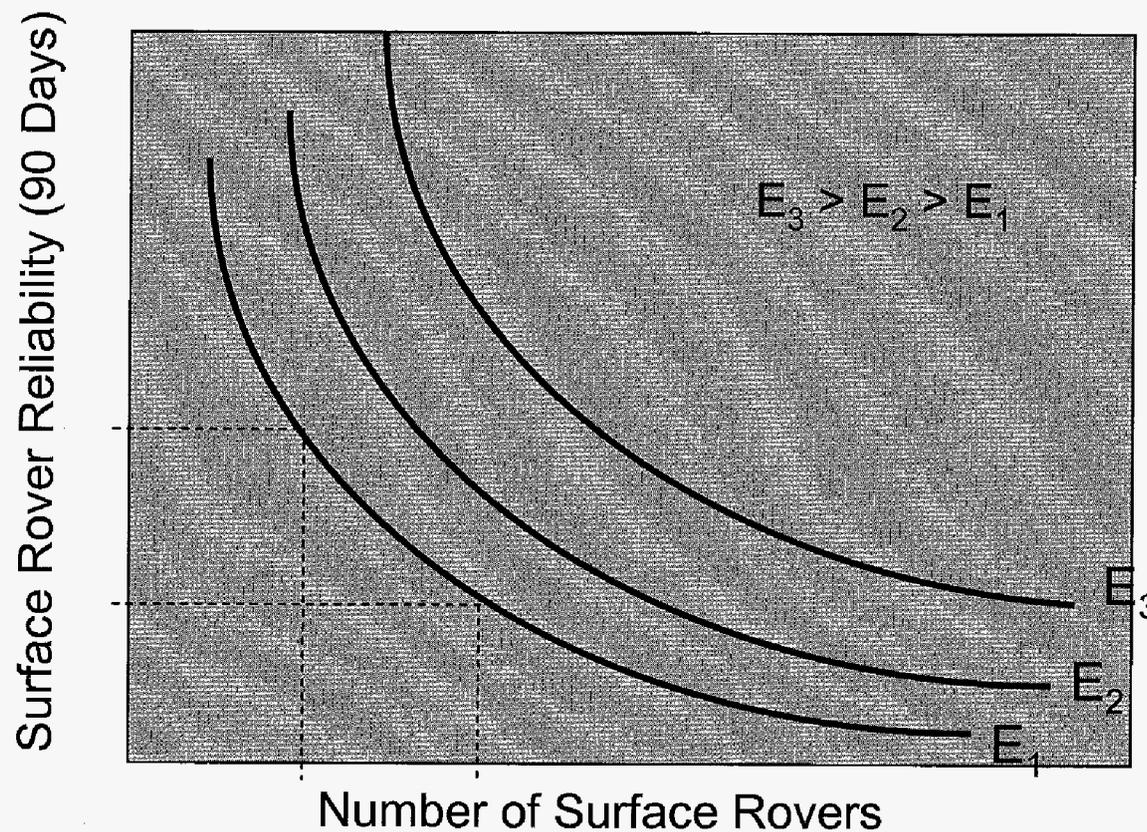
Each PTM is custom-built to a project, but contains many reusable elements

Cost-Effectiveness Frontier (Pareto Optimal Frontier)



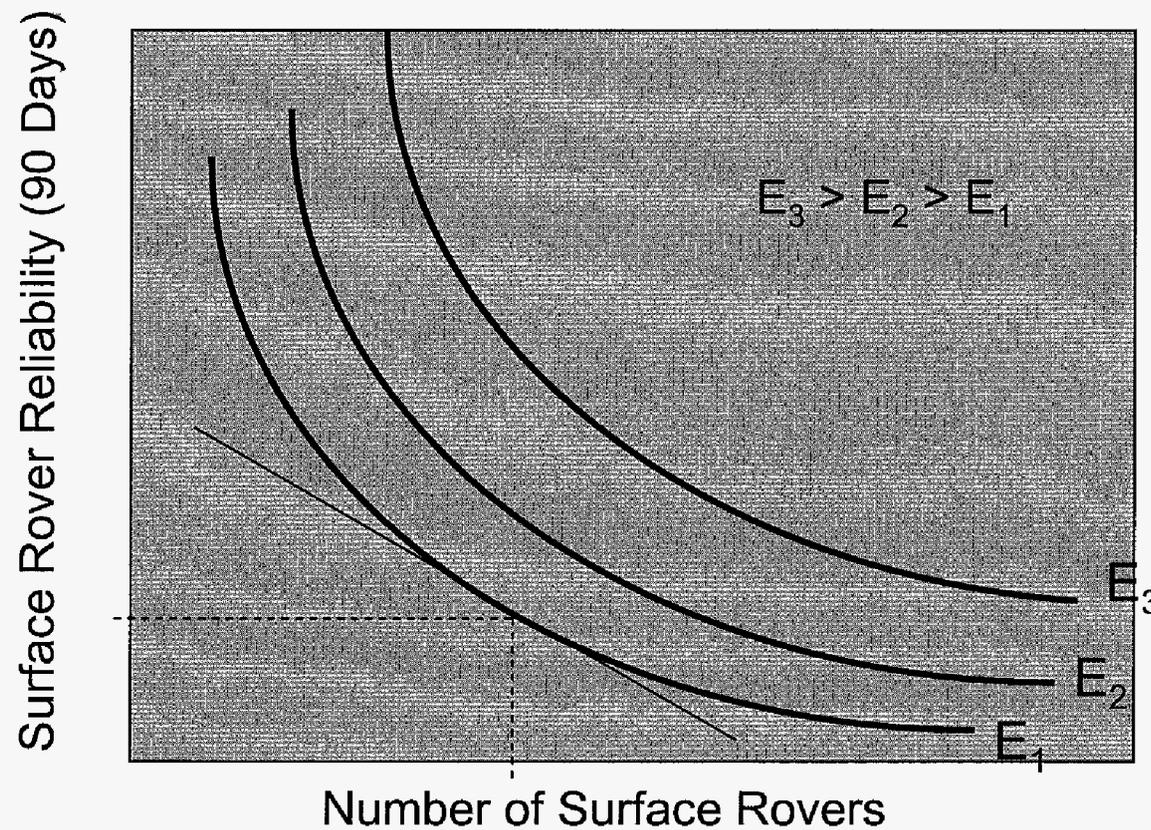
Iso-Effectiveness Curves

Iso-effectiveness curves represent combinations of the input variables that yield the identical value for a Measure of Effectiveness. For illustration:



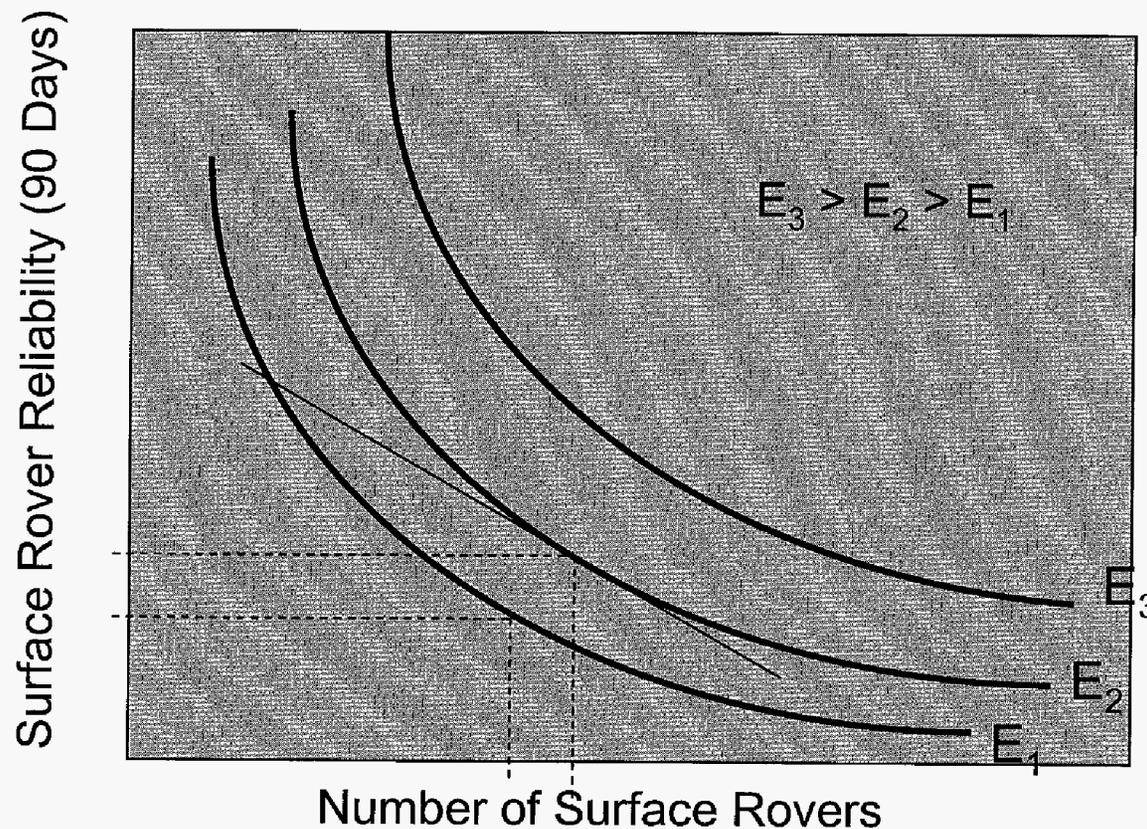
Iso-Effectiveness Curves

Generally each iso-effectiveness curve has only one point that's Pareto optimal.



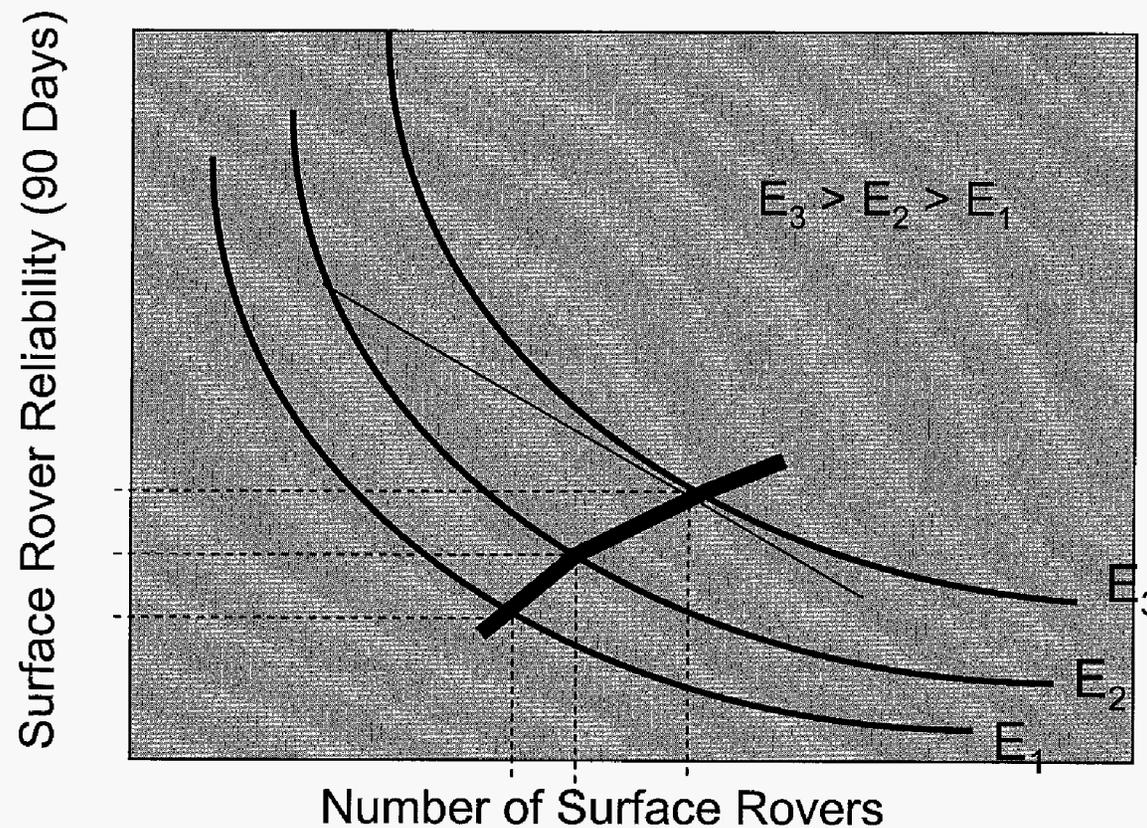
Iso-Effectiveness Curves

Generally each iso-effectiveness curve has only one point that's Pareto optimal.

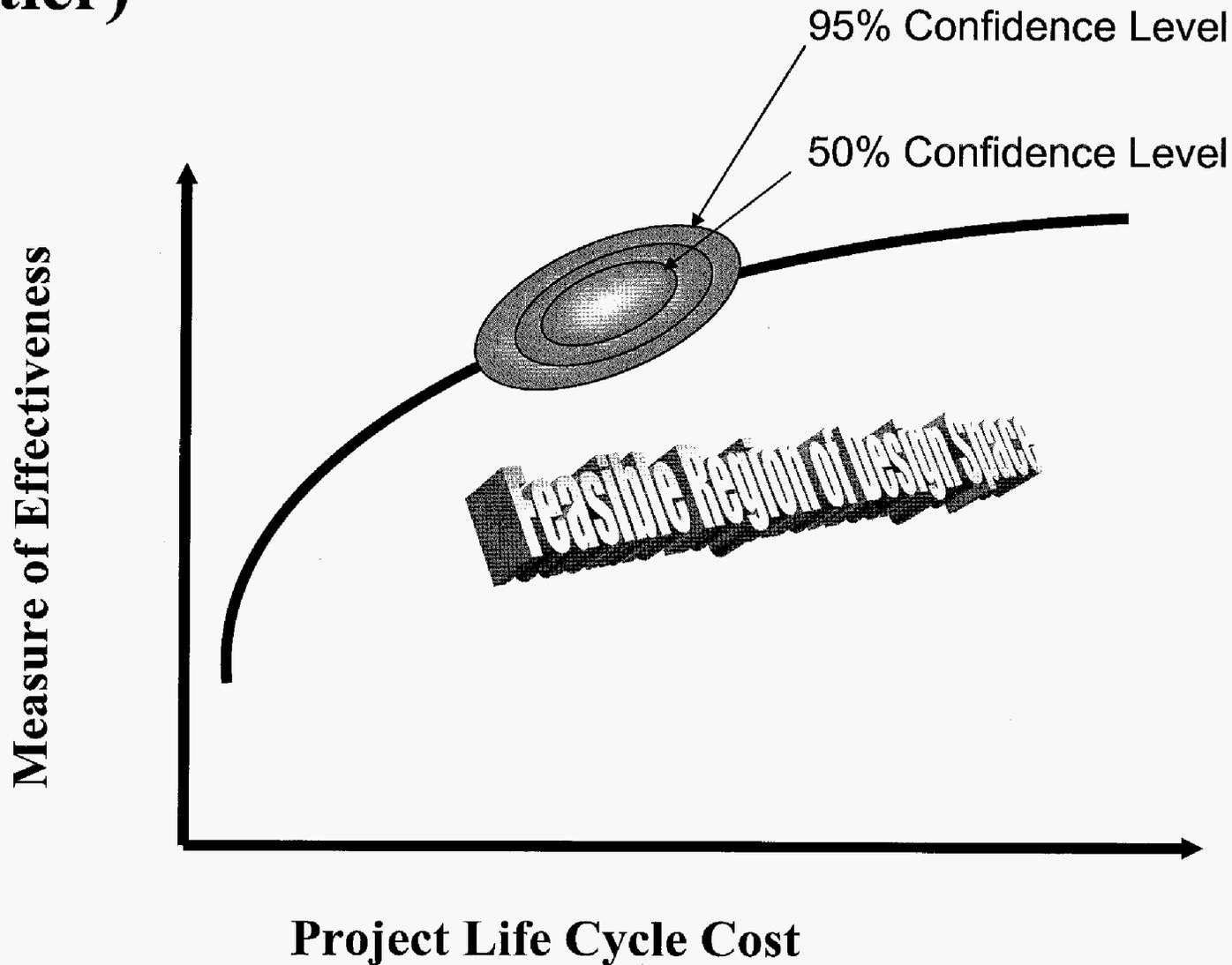


Iso-Effectiveness Curves

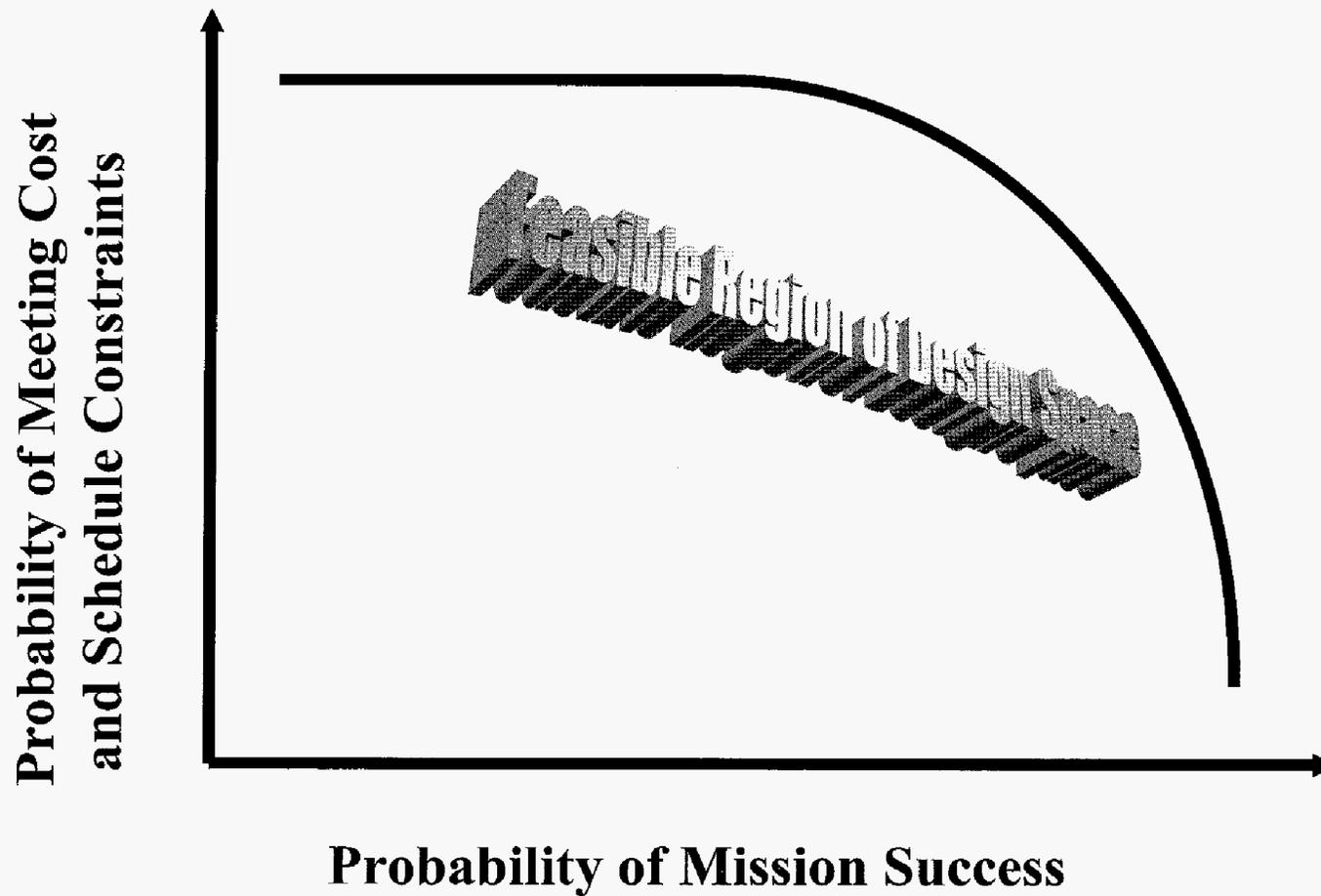
Generally each iso-effectiveness curve has only one point that's Pareto optimal. Connecting these points produces the optimal expansion path



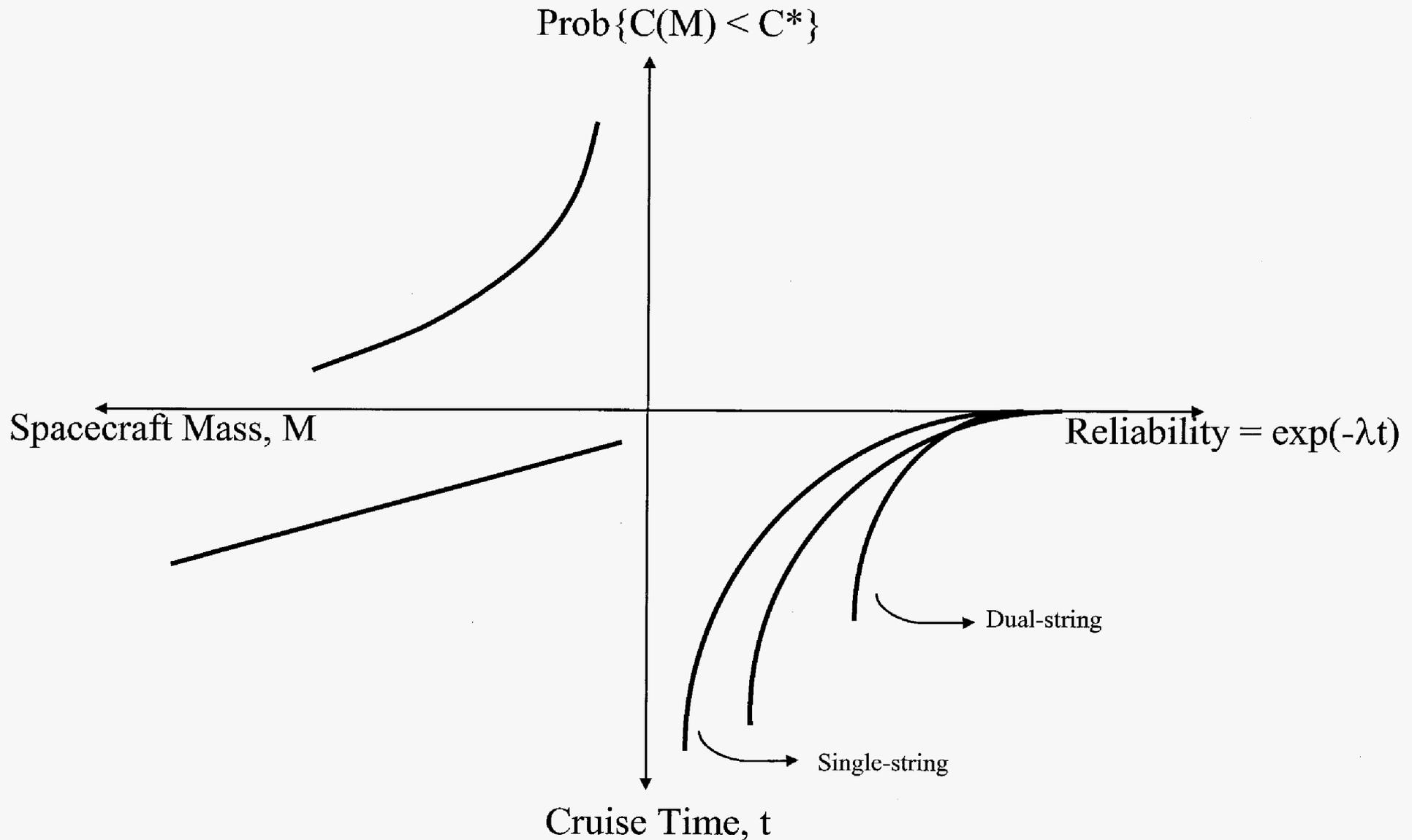
Cost-Effectiveness Frontier (Pareto Optimal Frontier)



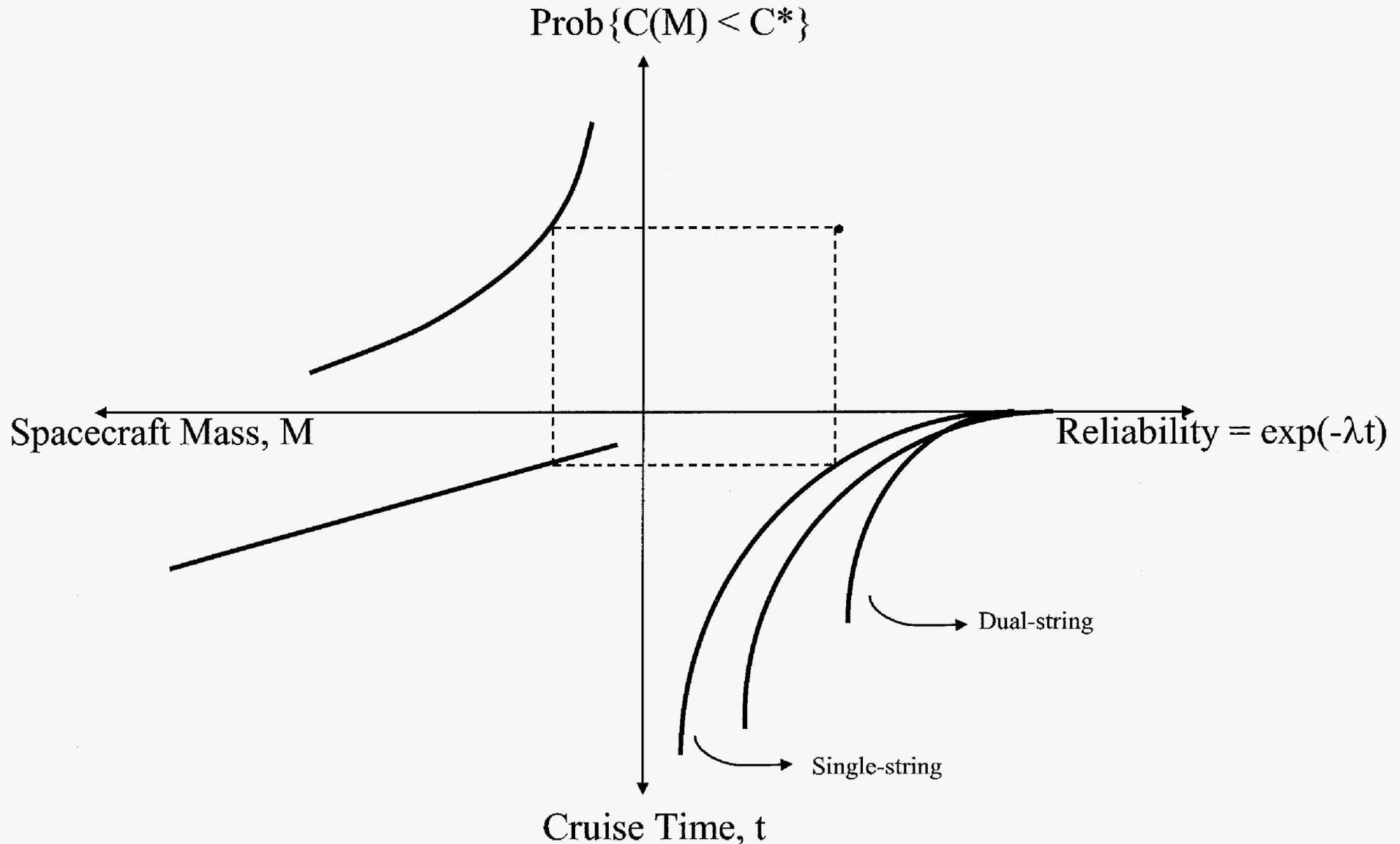
Cost-Effectiveness Frontier (Pareto Optimal Frontier) in Terms of Probabilities



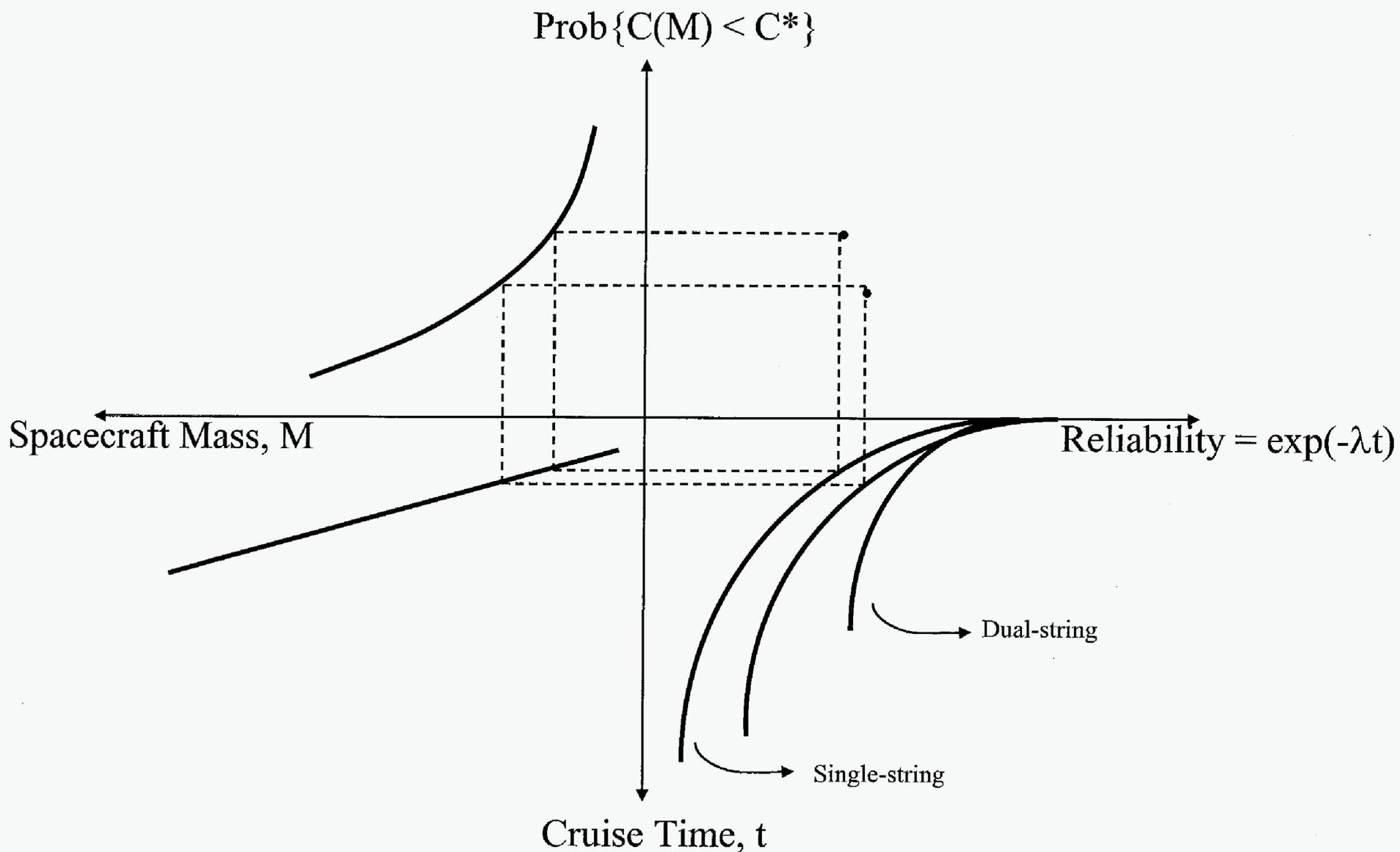
Deriving the Cost-Effectiveness Frontier (Pareto Optimal Frontier)



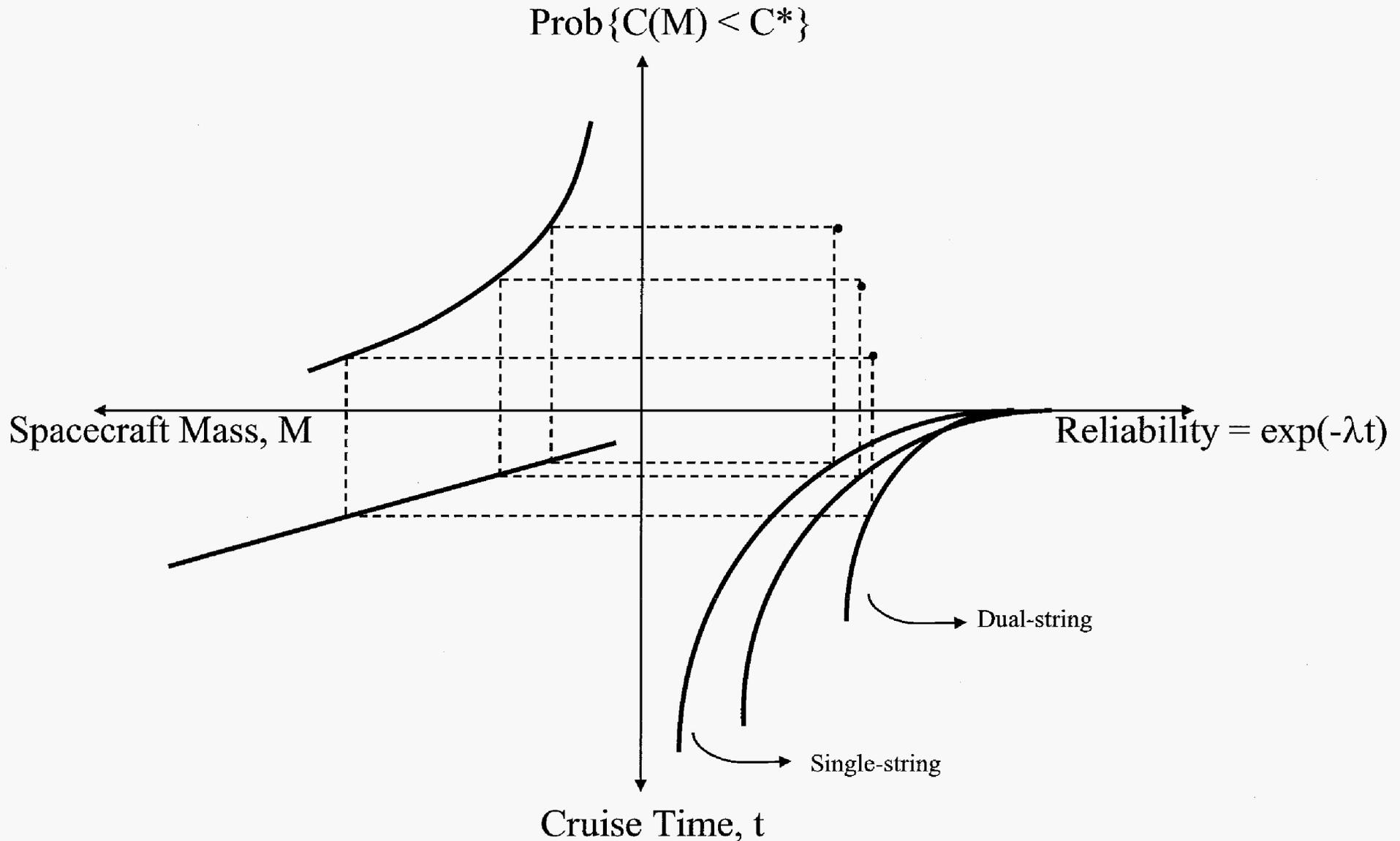
Deriving the Cost-Effectiveness Frontier (Pareto Optimal Frontier)



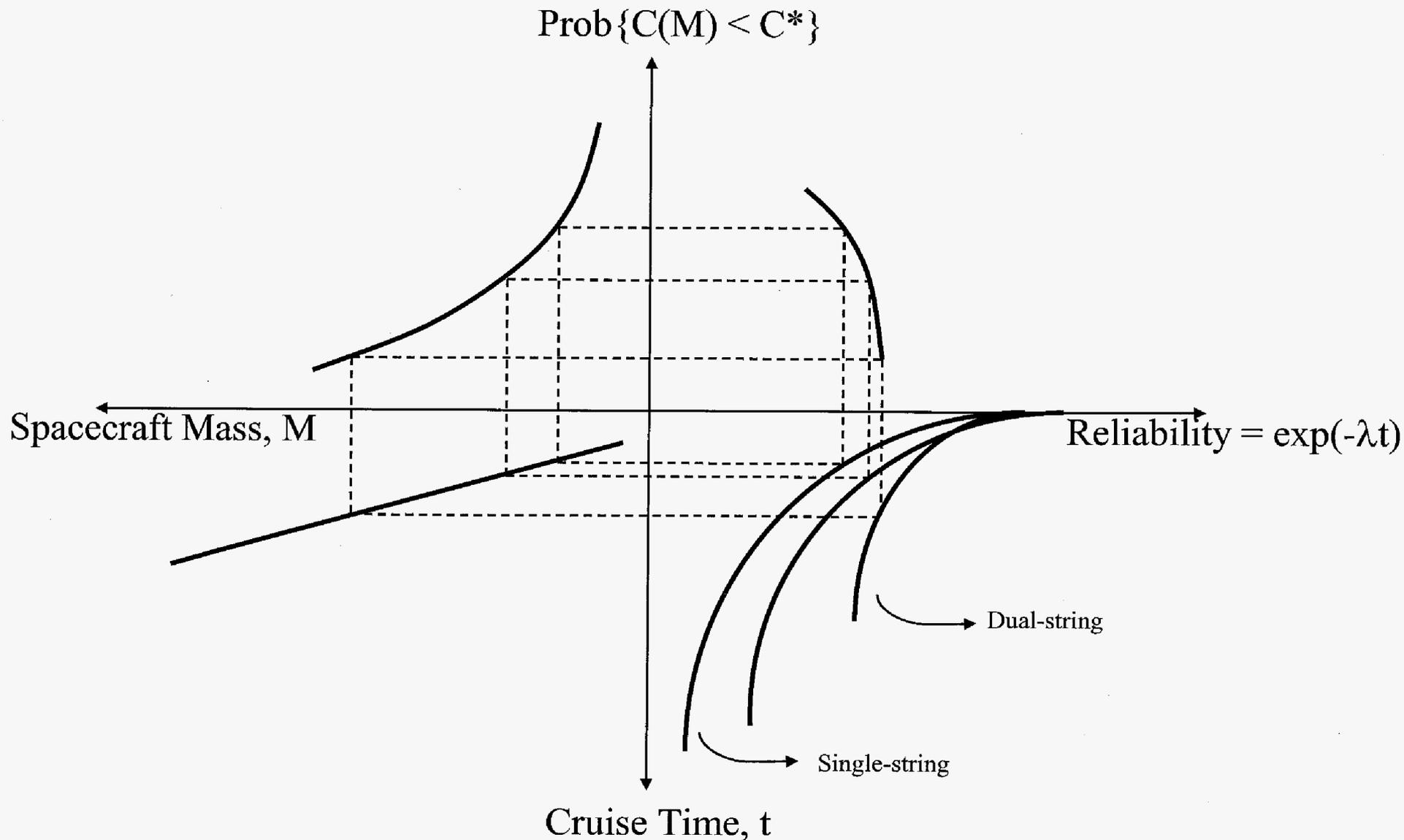
Deriving the Cost-Effectiveness Frontier (Pareto Optimal Frontier)



Deriving the Cost-Effectiveness Frontier (Pareto Optimal Frontier)



Deriving the Cost-Effectiveness Frontier (Pareto Optimal Frontier)

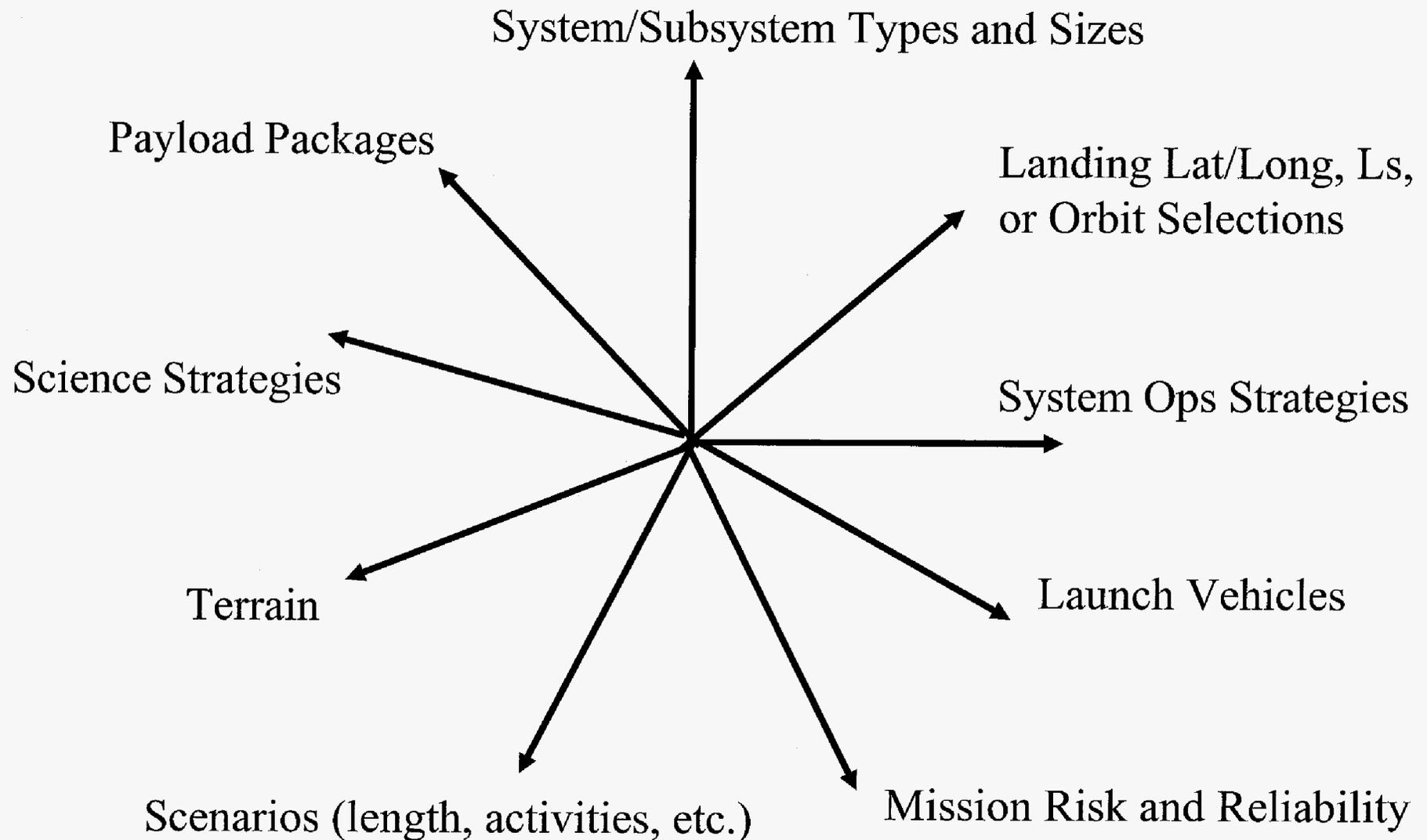


Tradespace Exploration

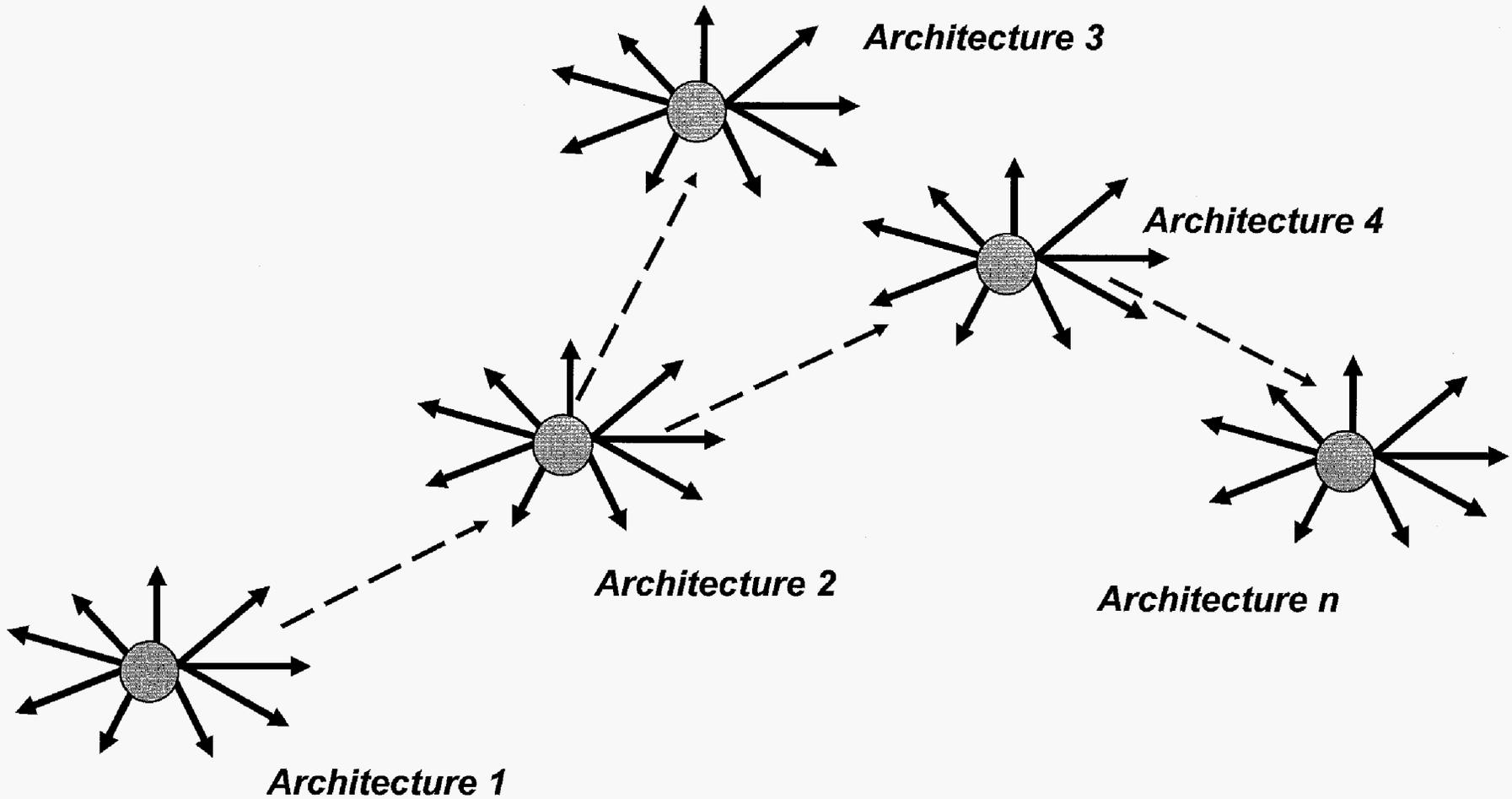
Issues in Building a PTM

- The PTM can involve analytic models, geometry-based models, Monte Carlo, and deterministic simulations.
 - Models are abstractions or representations of real-world processes or structures. In systems analysis, finding reliable relationships between the system's inputs (design vector) and effectiveness (or cost) is never easy.
 - Useful models may not exist, or if they do (from previous work), they come in various states of integration readiness
- Identifying appropriate Measure of Effectiveness can be difficult
- Tradespace dimensionality challenge

Tradespace Dimensionality Challenge

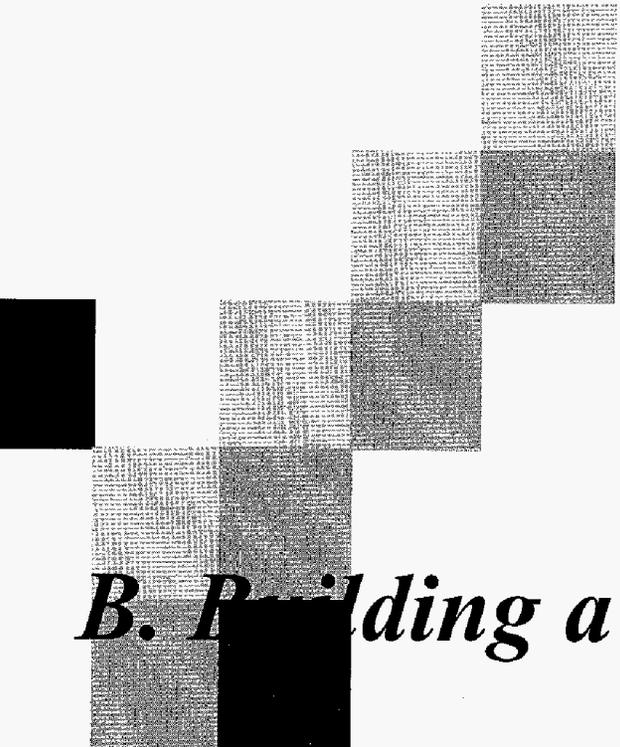


Explorations made possible



Project Tradespace Model Summary

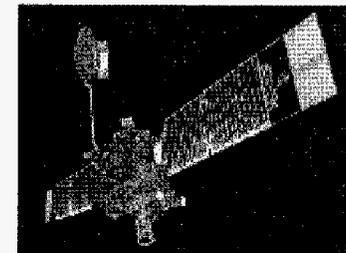
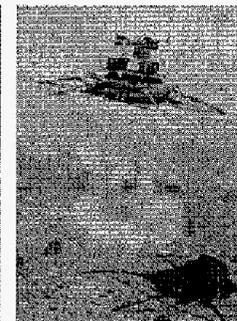
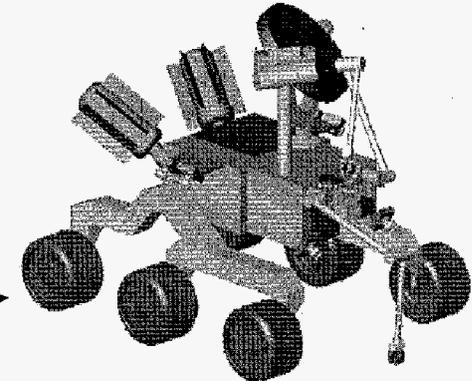
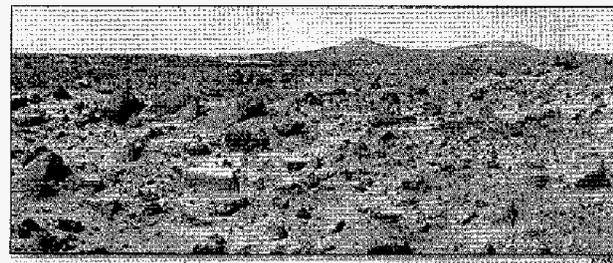
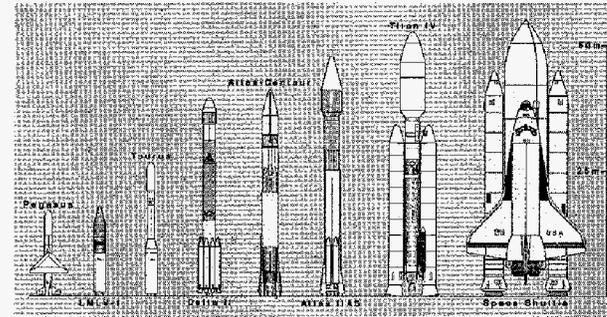
- The PTM is a package of linked models used for analyzing trades among different project/system architectures/designs and operations concepts/scenarios, i.e., exploring the tradespace.
- The PTM captures the relationships between the project's "design vector" (inputs) and the MoEs, life-cycle cost, and risk-metrics (outputs).
- Helps to define the Pareto frontier.



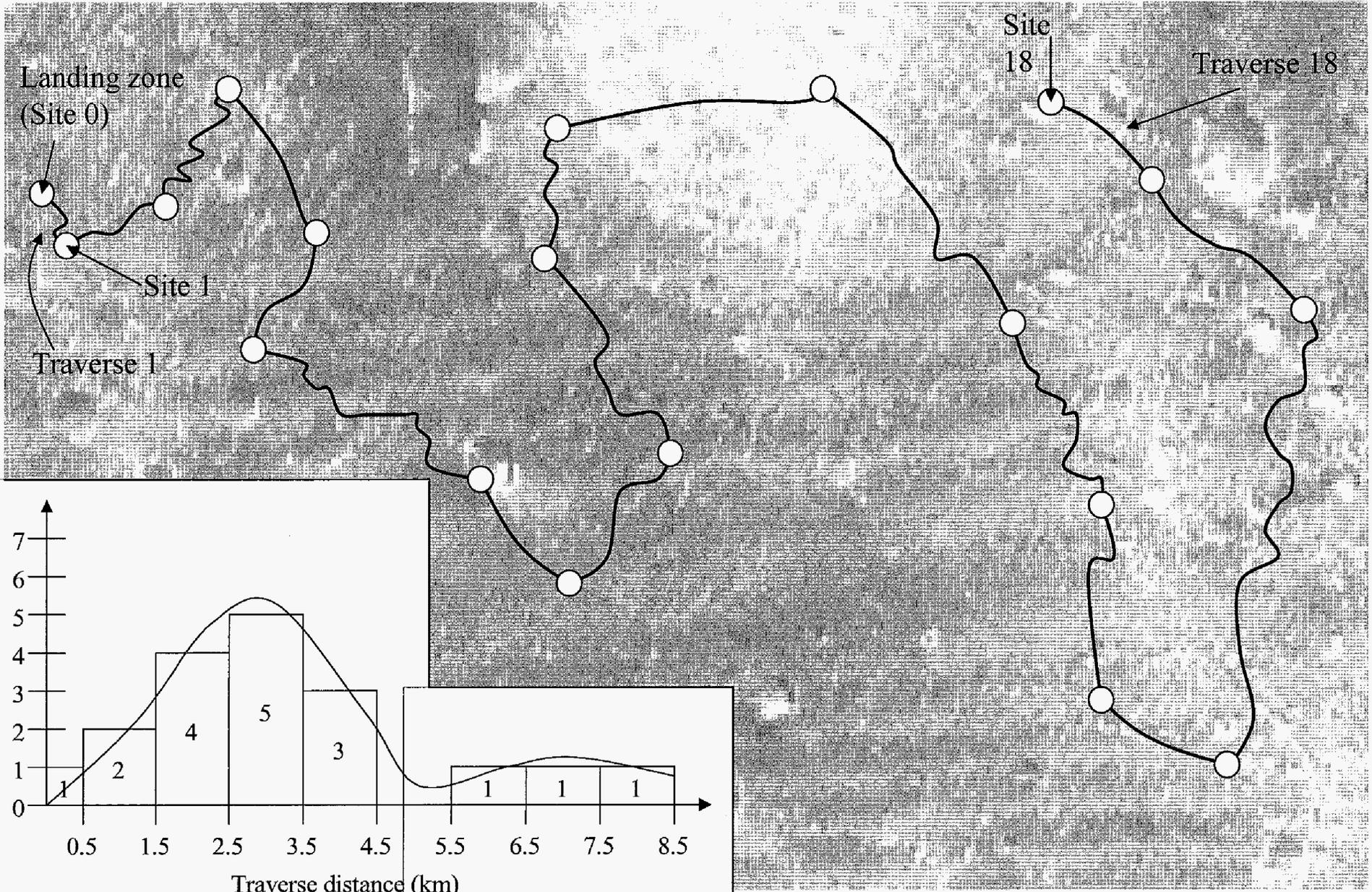
B. Building a Project Tradespace Model: MSL

Mars Science Laboratory (MSL) Tradespace Model Design Vector

- Arrival date (L_s)
- Launch vehicle
- Mars location
- Inter-site distances
- Level of autonomy
- Power system type
- Telecomm strategy
- Reliability requirement
- Instrument suite
- EDL system type



Traverse Lengths



Mars Science Laboratory (MSL) Tradespace Model Output Vector

- Achieved science data return (Mb/sol)
 - Achieved site investigations
 - Traverse distance margin (km)
 - Flight system margin (kg)
 - Surface system margin (kg)
 - Launch vehicle excess capacity (kg)
 - Power and energy margins (kW and kWh)
 - EDL feasibility
 - DSN availability (%)
 - Estimated life cycle cost (\$)
 - Cost confidence (%)
- } Measures of Effectiveness
 } Measures of Performance
 } Measures of Cost

Linking Design and Output Vectors Requires Connecting Islands of Analytical Capability

APGEN

ROAMS

Terrain (Maker2)

Mobility & Structure

Launch Vehicle Database

Mars Data

MEL/PEL

Instrument Database

MER-based Phase E Workforce Estimator

Satellite Orbit Analysis Program (SOAP)

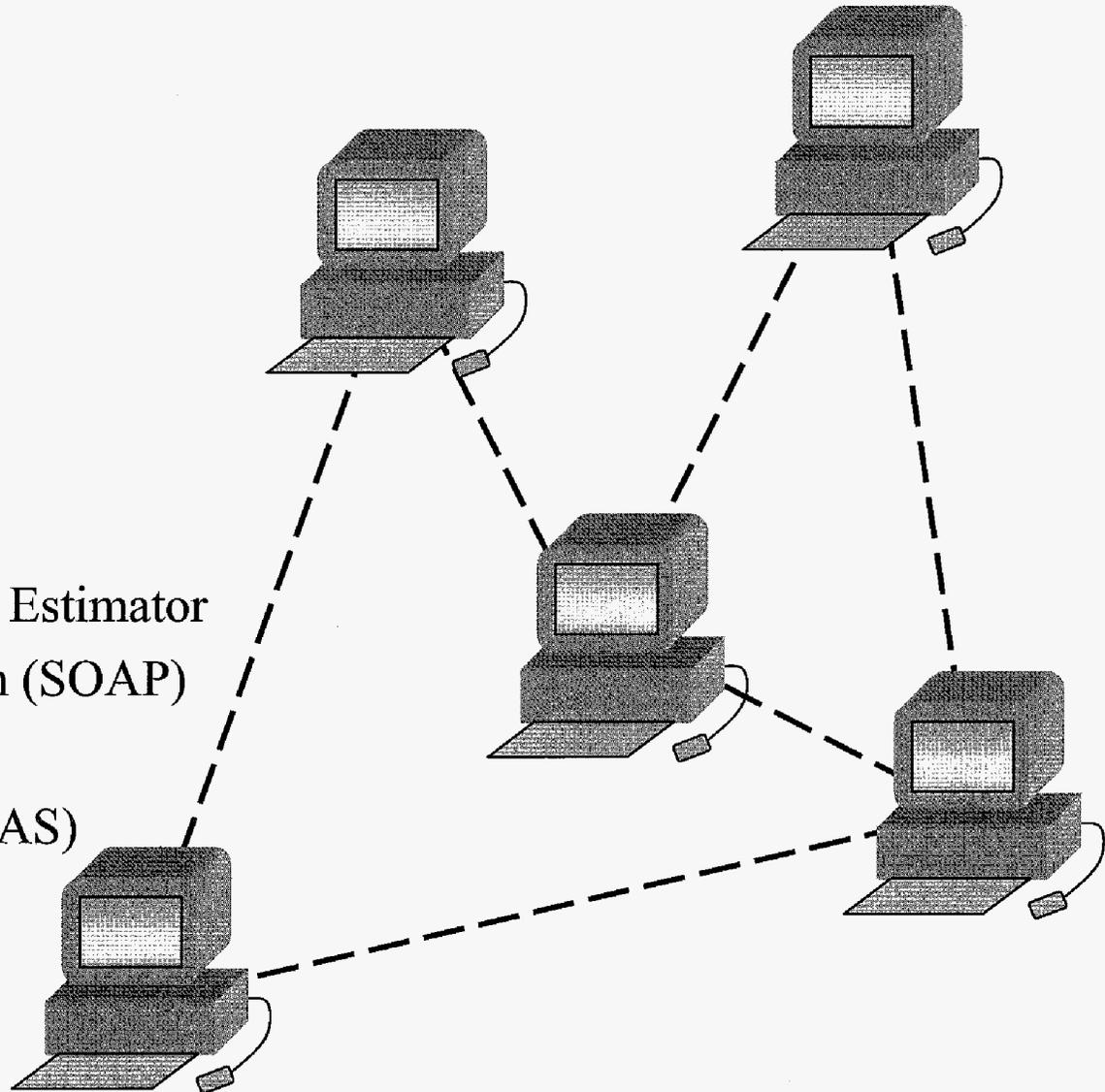
Telecom Link Analysis Tool

DSN Availability Model (TIGRAS)

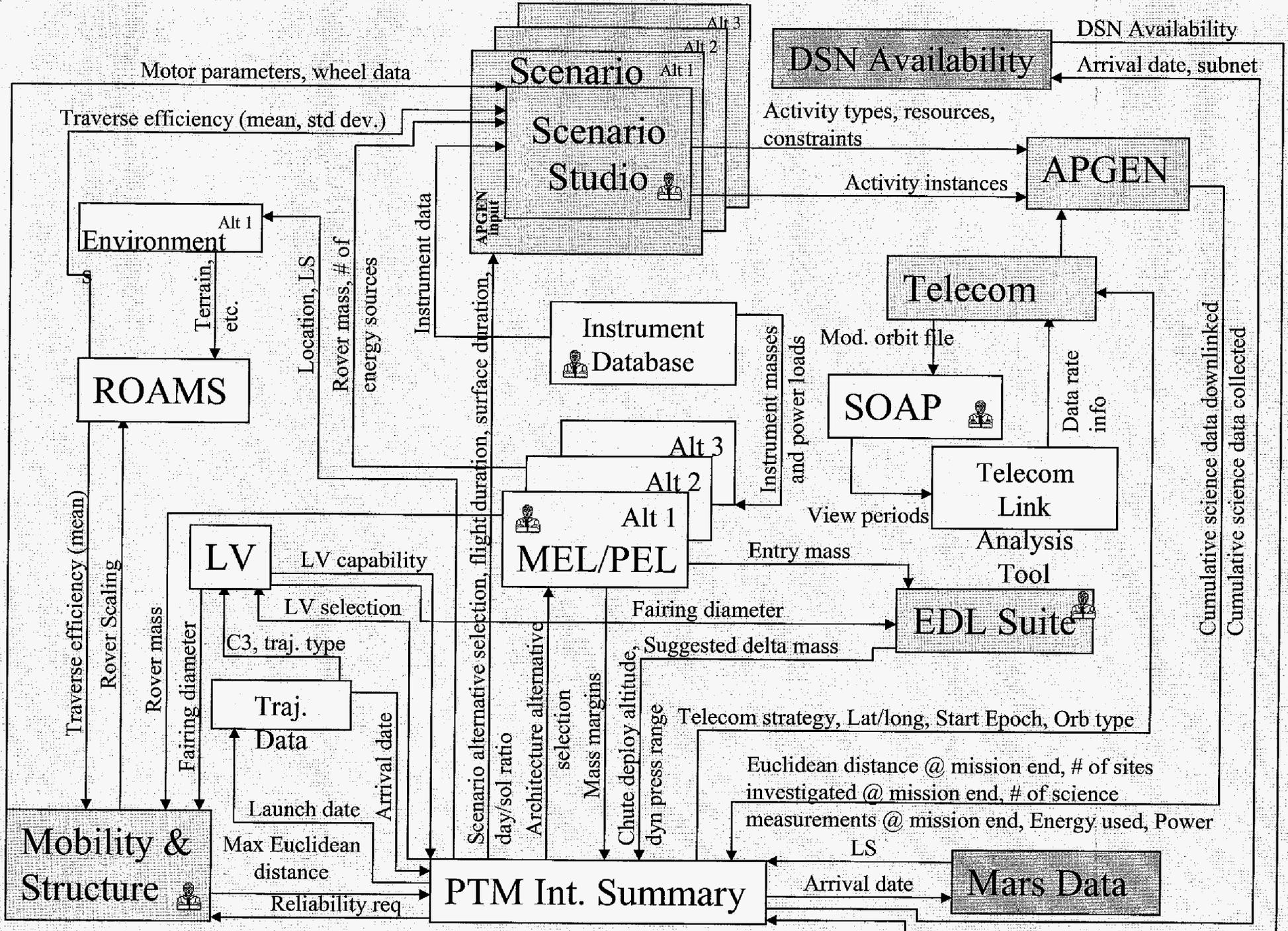
Upper EDL Model

Developmental Cost Models

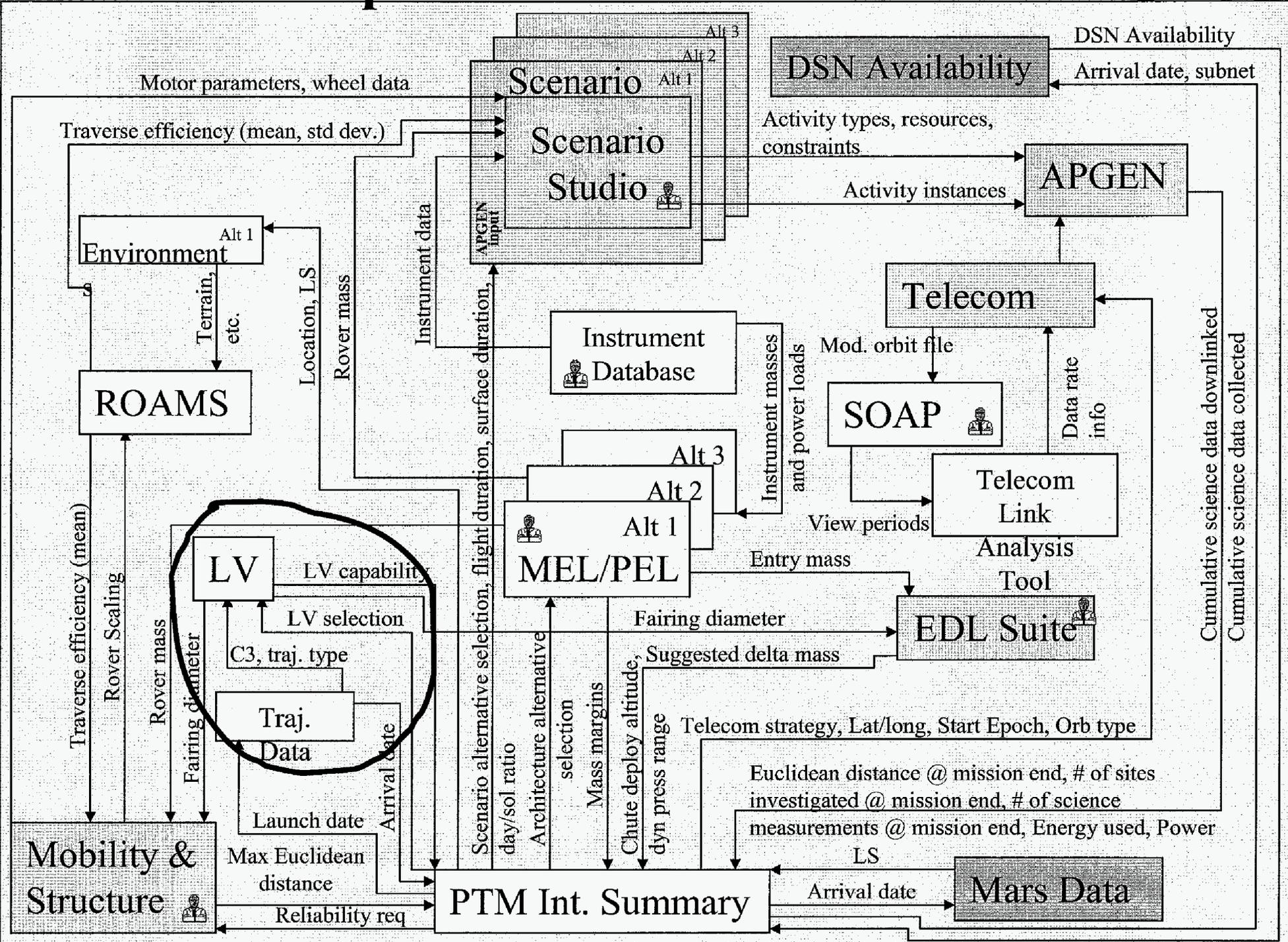
Team X Instrument Cost Model



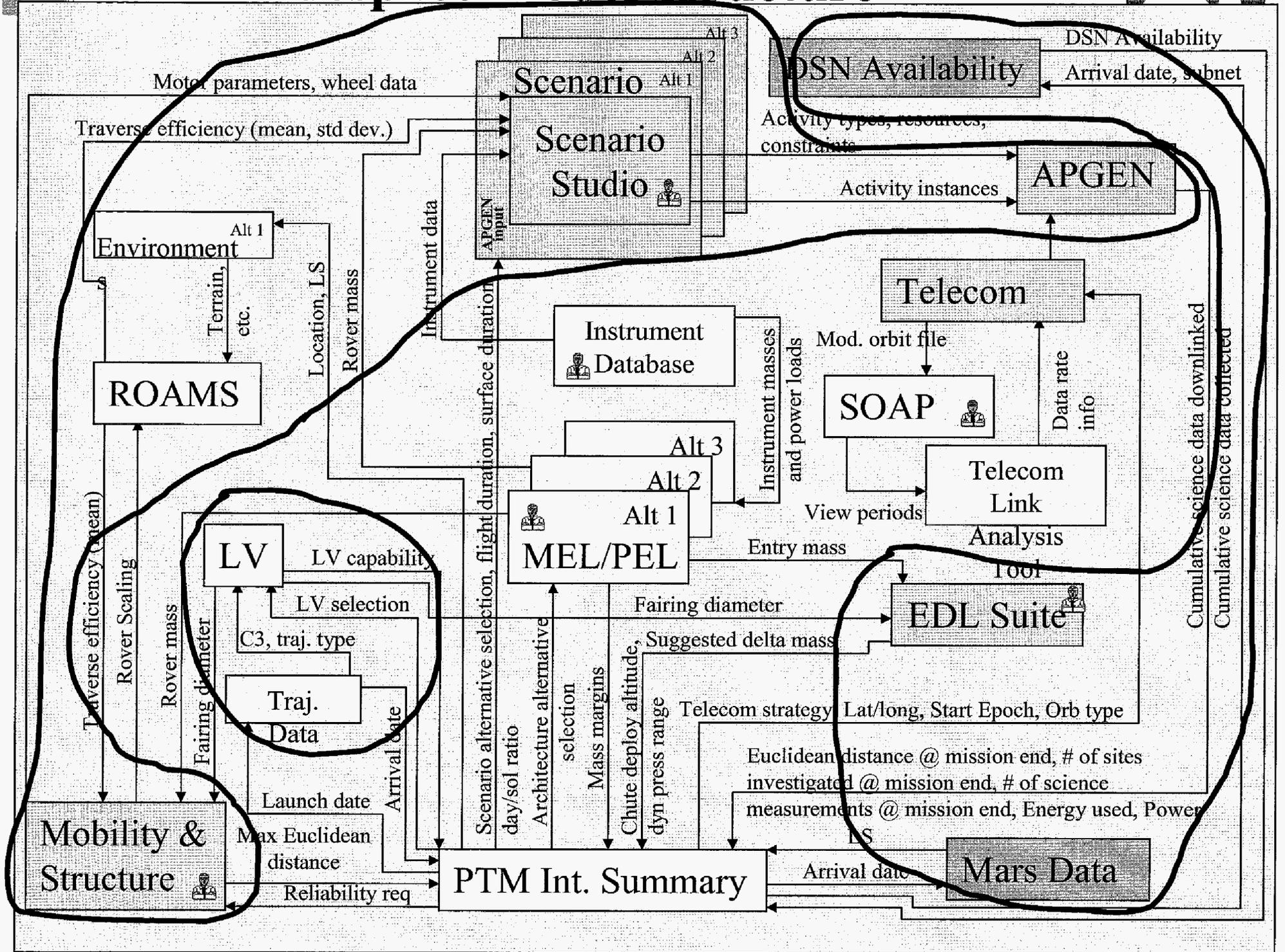
Mars Tradespace Model Structure



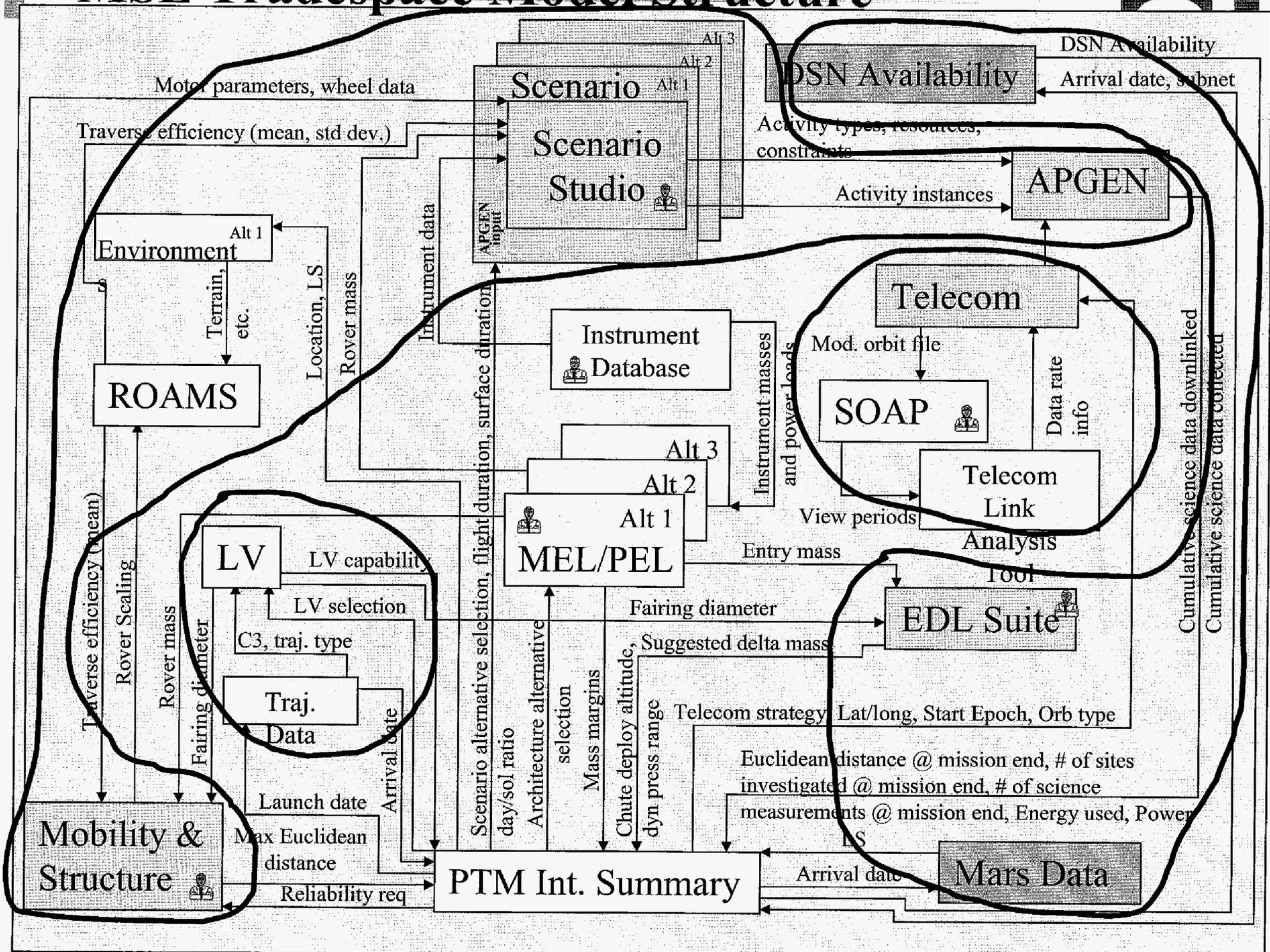
Mars Tradespace Model Structure



MSE Tradespace Model Structure

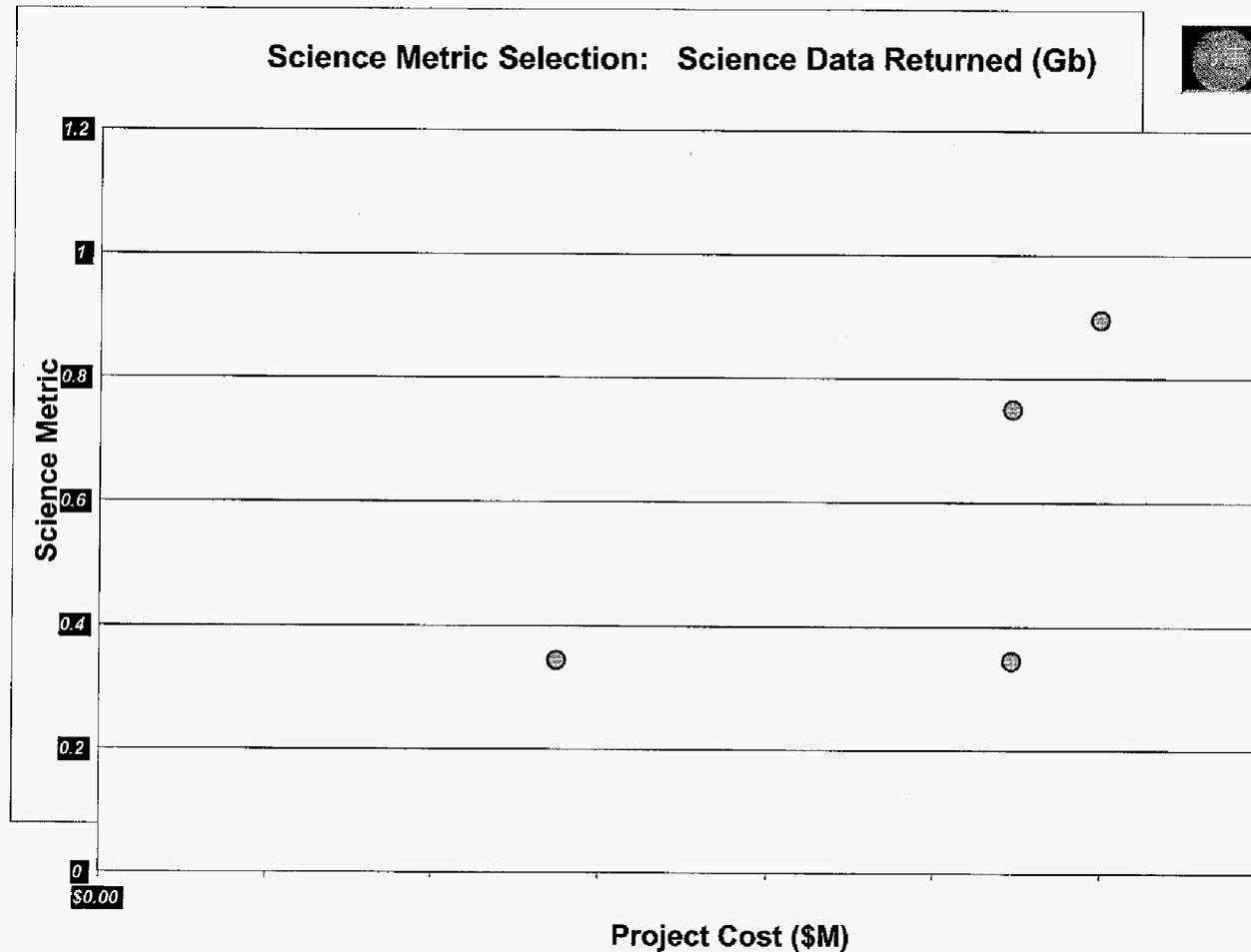


Mars Tradespace Model Structure



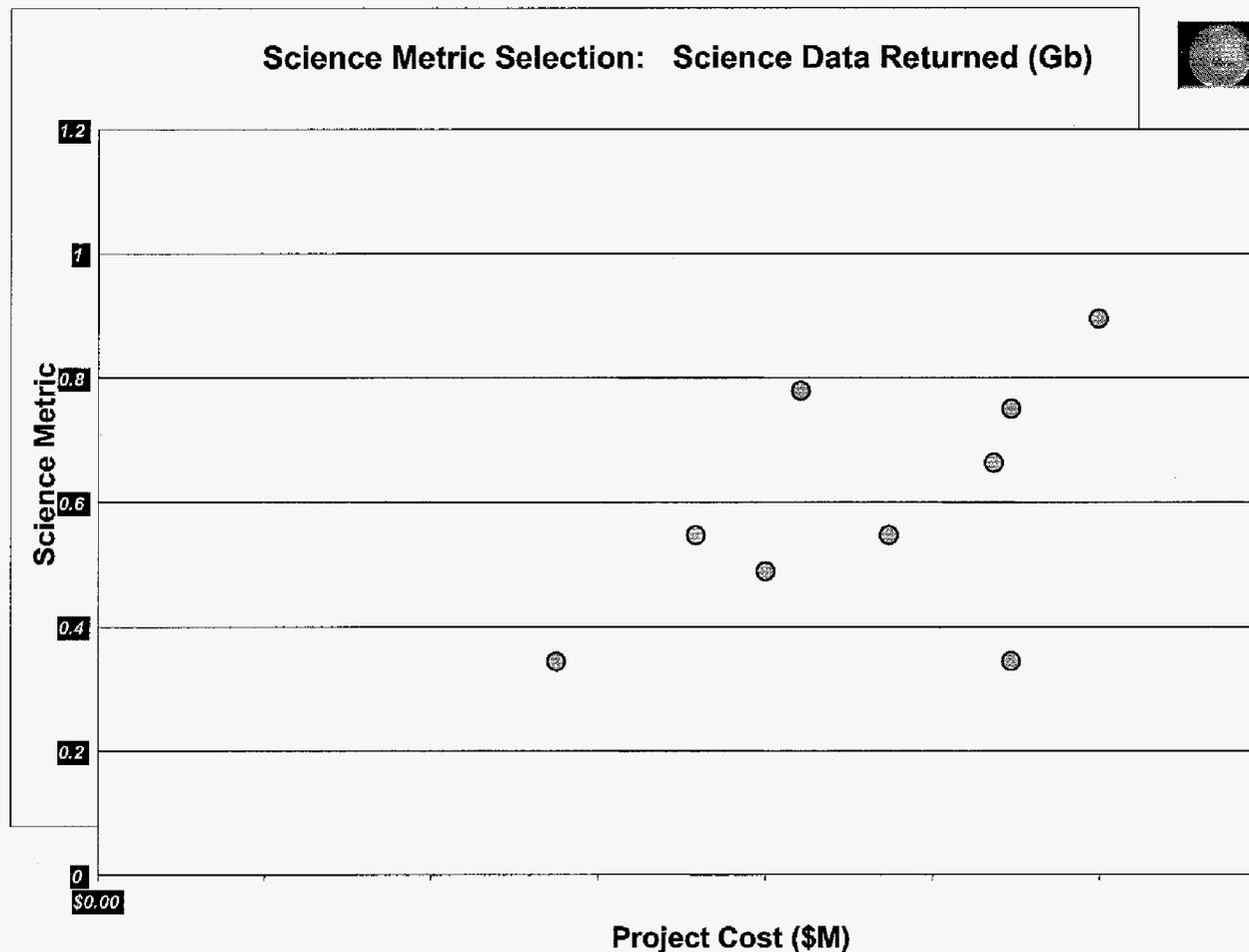
MSL Tradespace Exploration

- Enables team to more effectively explore the tradespace to find the best combination of life-cycle cost and mission effectiveness
- Shows HQ what is feasible with different funding levels



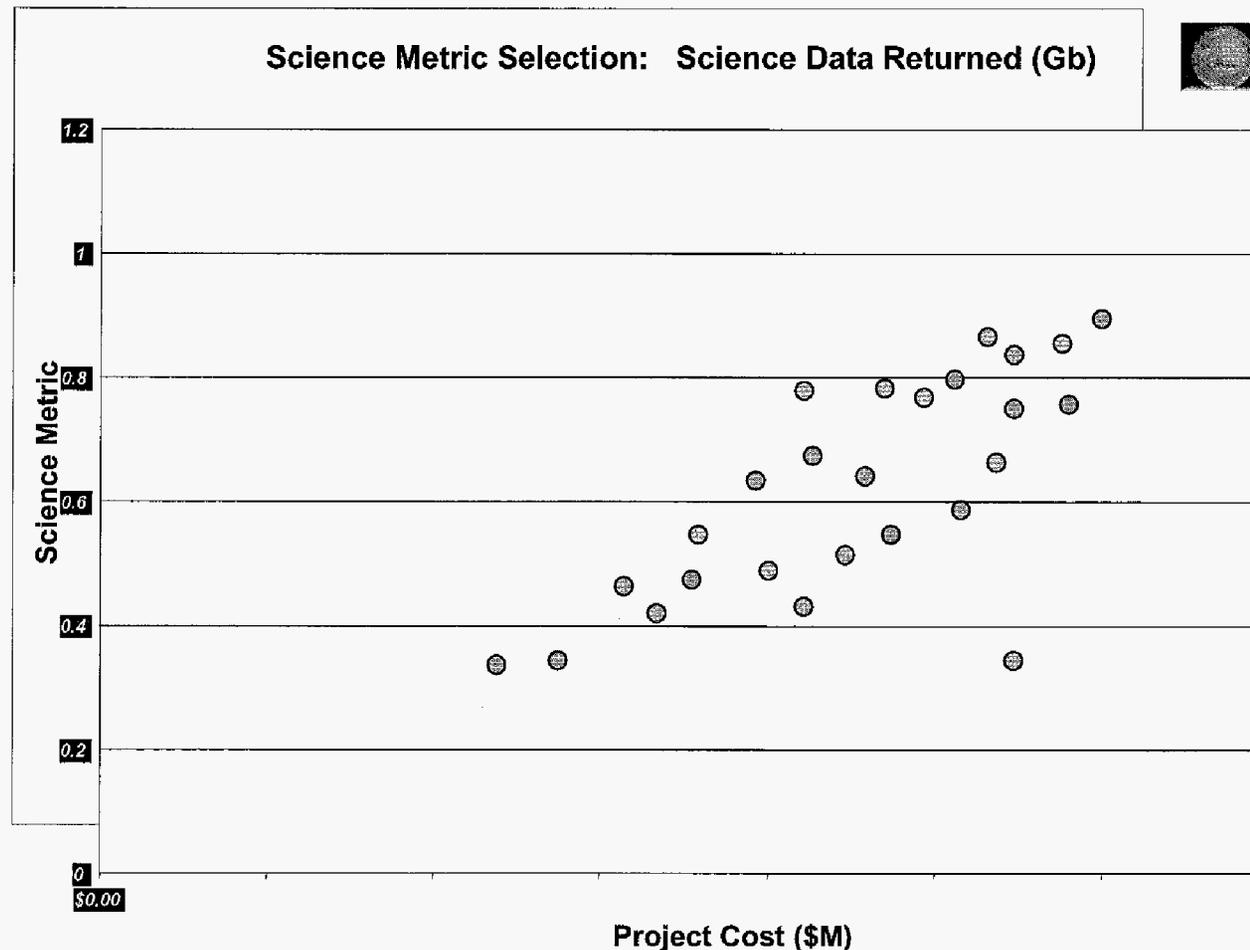
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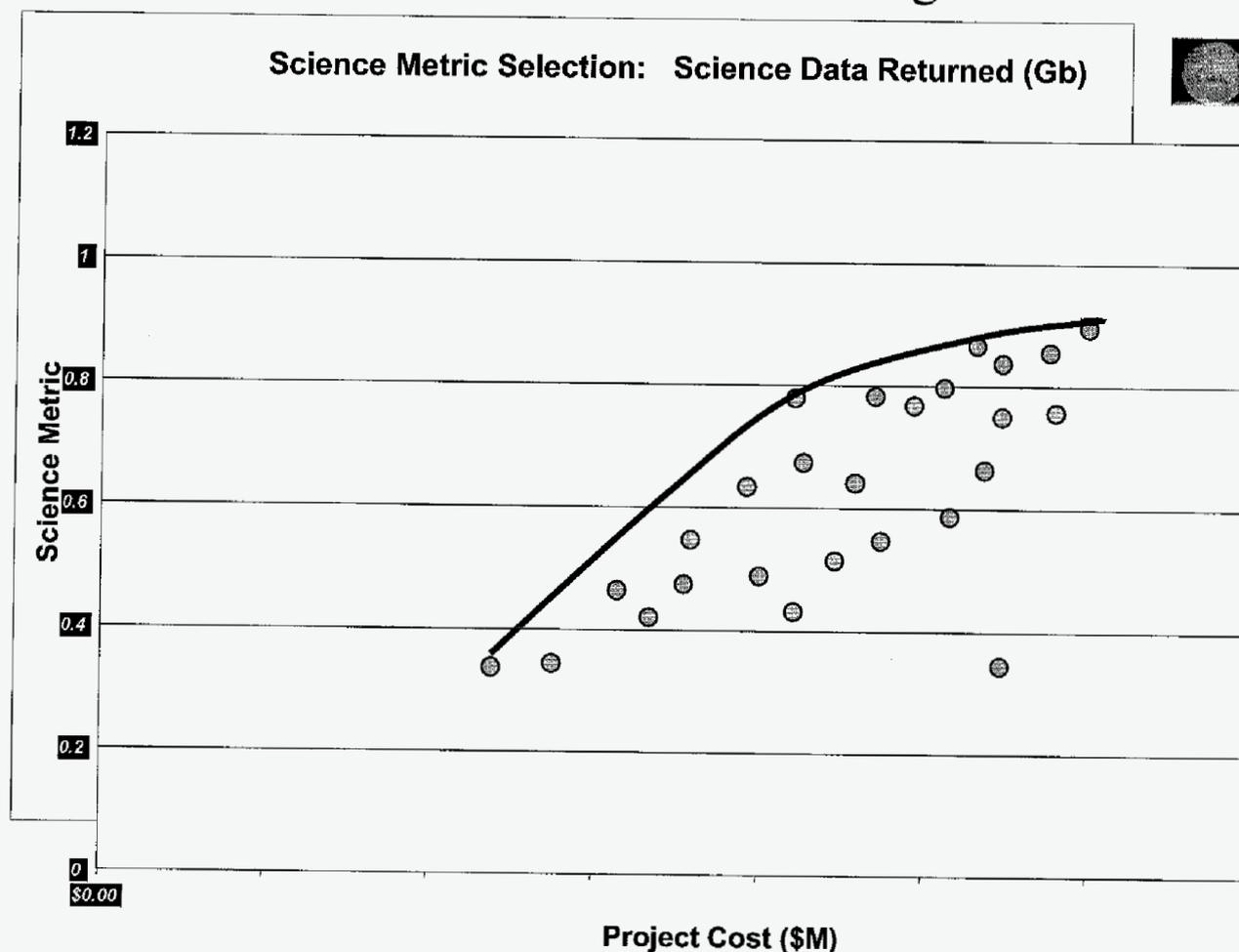
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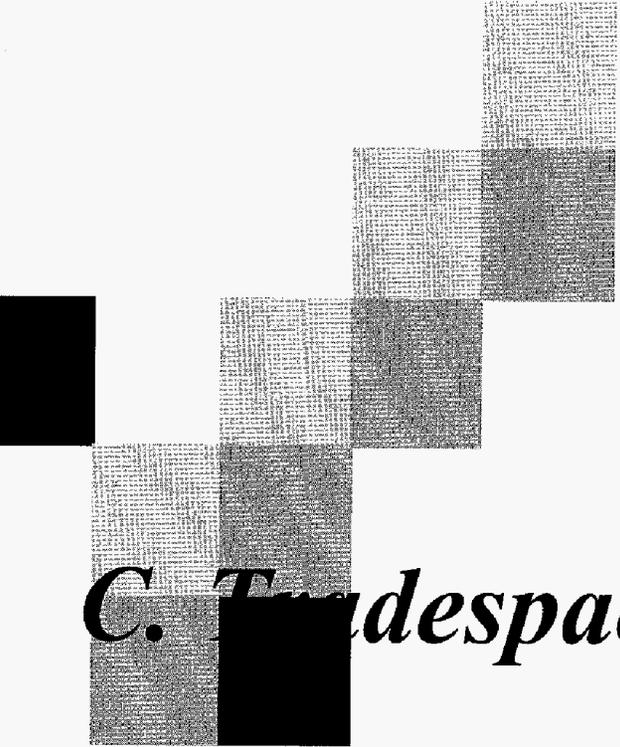
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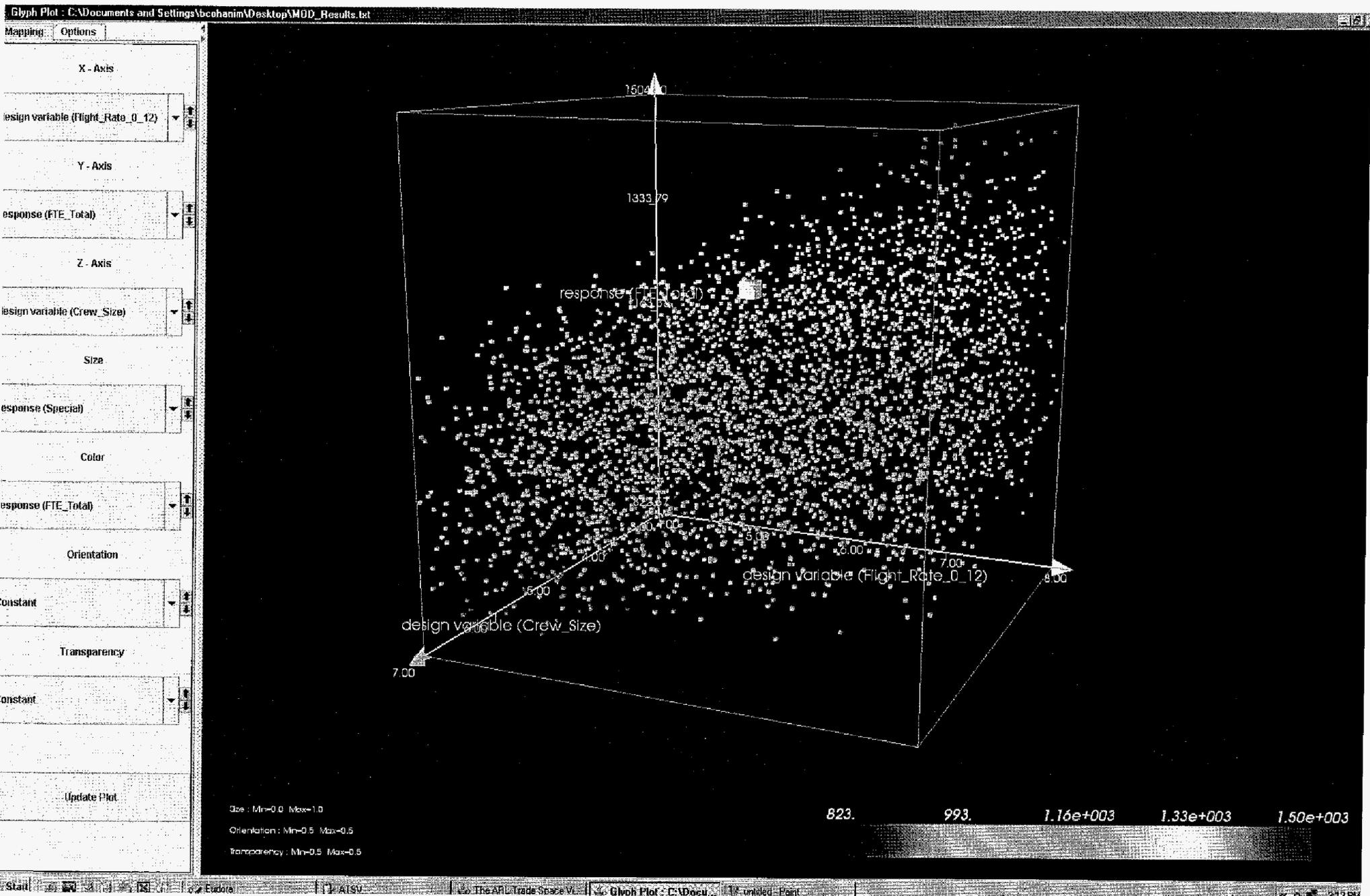
Observations on Building the MSL PTM

- PTM “components” may not be available inside or outside the organization
 - Develop, adapt, and standardize
 - Legacy models typically not “integration-ready”
 - Excellent cooperation from tool owners/originators allows fast progress
- Most successful application occurs when project tradespace modeling is a leadership priority
 - Design vectors are reasonably well-defined
 - Decision criteria (MoEs, LCC) are identified early
- Development increases system thinking and raises understanding of interrelationships
- Archives inputs/outputs, assumptions, and captures rationale
- Capability to pursue uncustomary trades

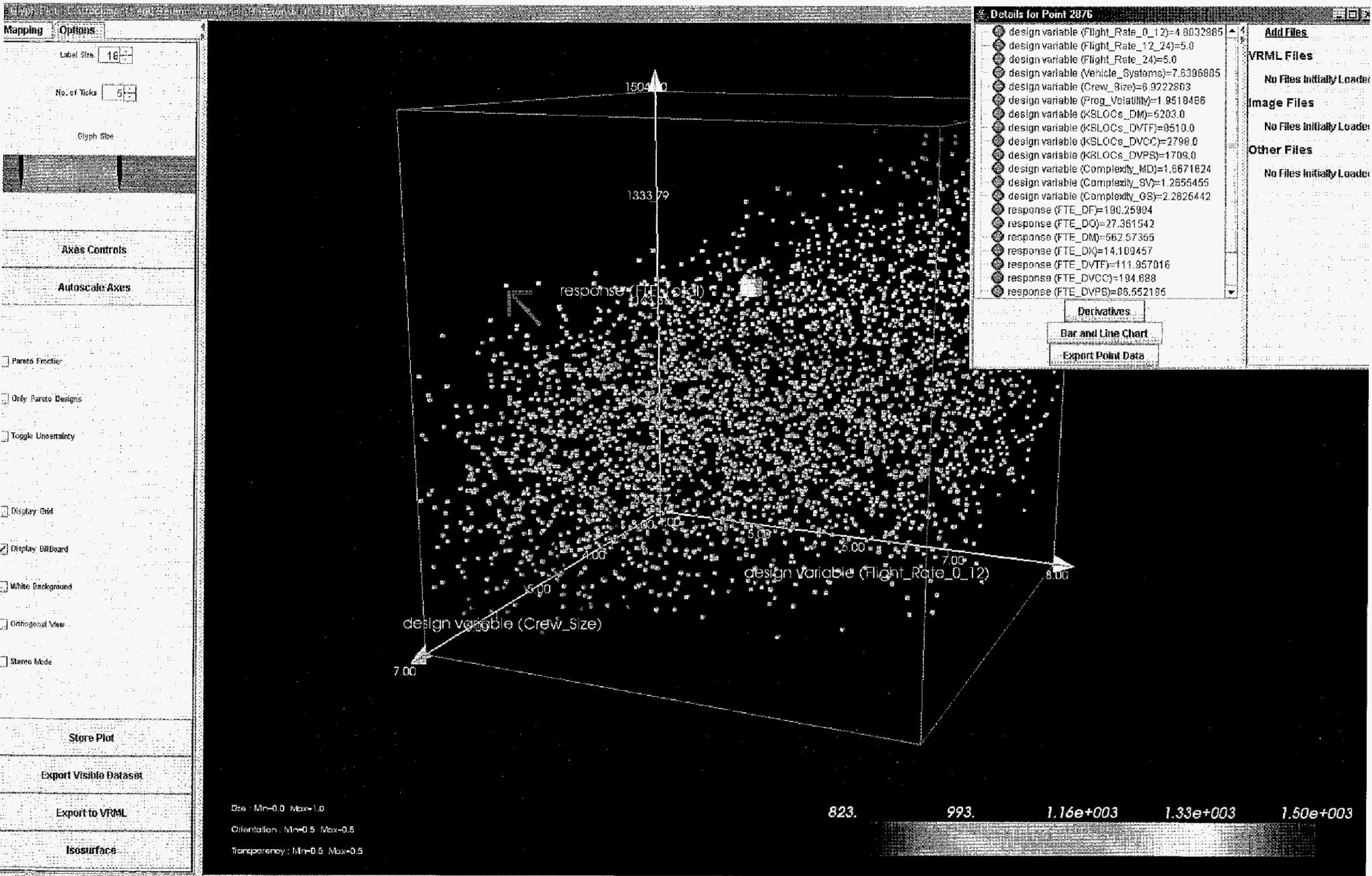


C. Tradespace Visualization

3D Tradespace



View Data on Selected Point



Mapping Options

Label Size: 18

No. of Ticks: 5

Glyph Size:

Axis Controls

Autoscale Axes

Pareto Frontier

Only Pareto Designs

Toggle Uncertainty

Display Grid

Display Billboard

White Background

Orthogonal View

Stereo Mode

Store Plot

Export Visible Dataset

Export to VRML

Isosurface

Details for Point 2876

- design variable (Flight_Rate_0_12)=4.8032985
- design variable (Flight_Rate_12_24)=5.0
- design variable (Flight_Rate_24)=5.0
- design variable (Vehicle_Systems)=7.6396895
- design variable (Crew_Size)=8.0222803
- design variable (Prog_Volatility)=1.9518486
- design variable (KSLOCs_DM)=5203.0
- design variable (KSLOCs_DVTF)=9510.0
- design variable (KSLOCs_DVCC)=2798.0
- design variable (KSLOCs_DVPS)=1709.0
- design variable (Complexity_MD)=1.6671624
- design variable (Complexity_SV)=1.2855455
- design variable (Complexity_OS)=2.2826442
- response (FTE_DF)=190.25894
- response (FTE_DO)=27.361542
- response (FTE_DM)=562.67366
- response (FTE_DX)=14.109457
- response (FTE_DVTF)=111.957016
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- response (FTE_DVPS)=88.552186

Add Files

VRML Files

No Files Initially Loaded

Image Files

No Files Initially Loaded

Other Files

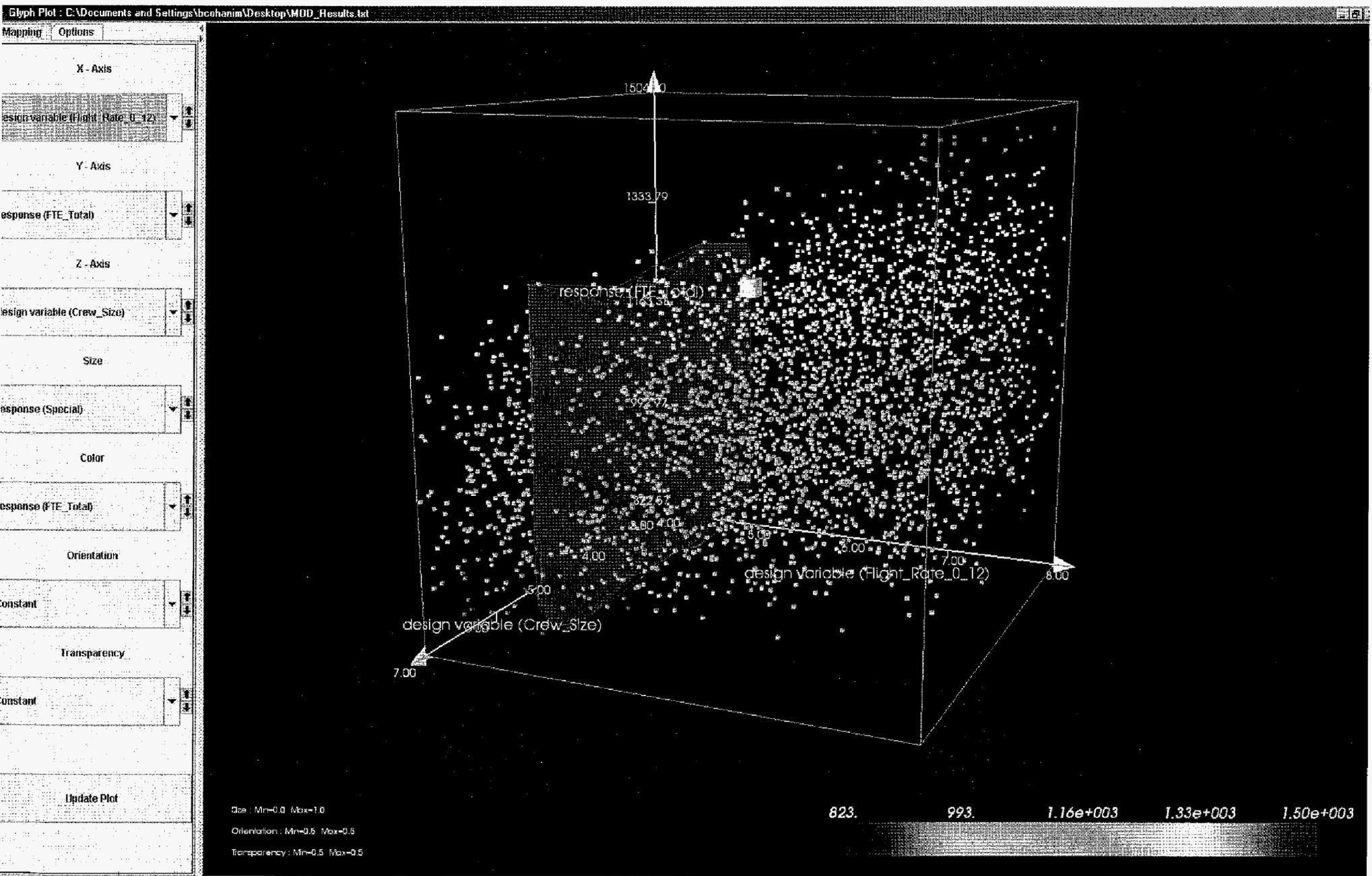
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Derivatives

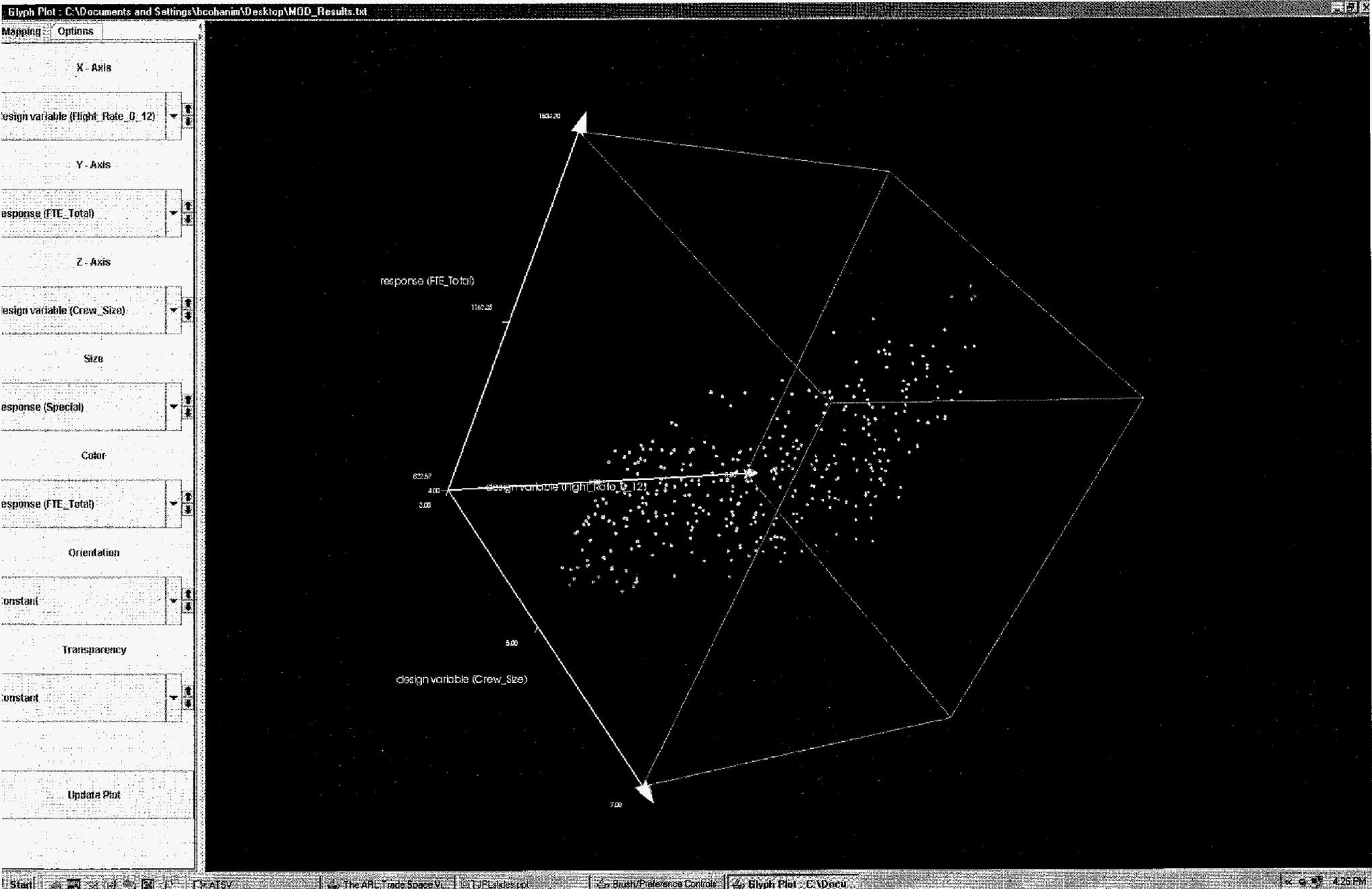
Bar and Line Chart

Export Point Data

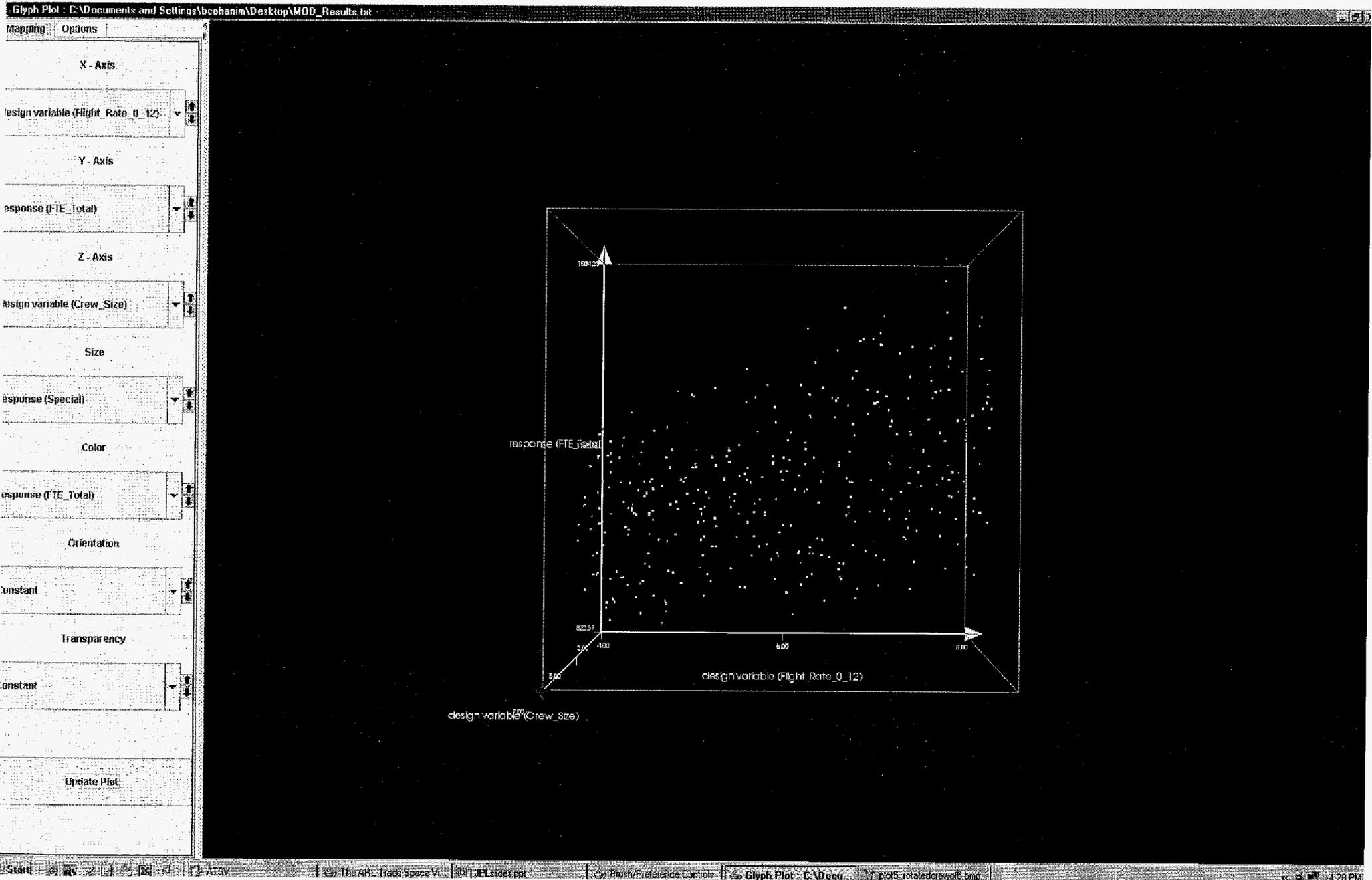
Surface of Constant Value



Filtered Slice



Pareto Surface (Two-Dimensional View)

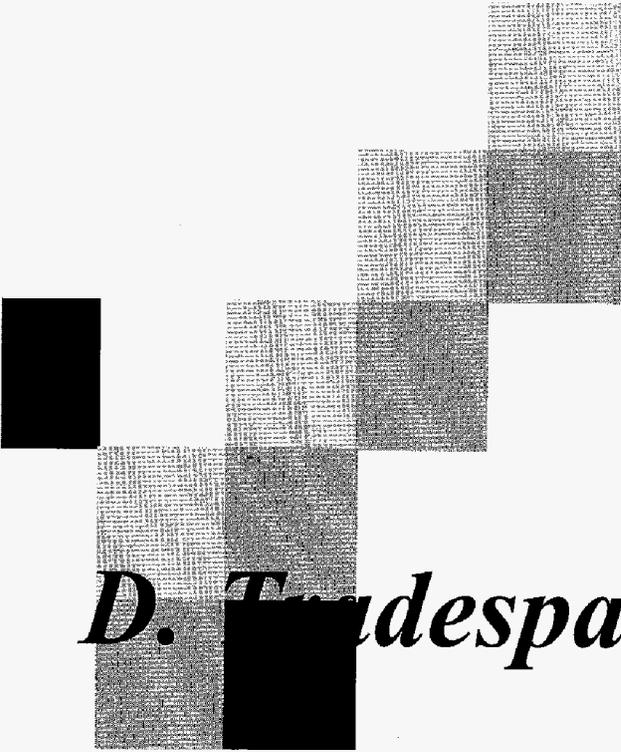


Summary

- Tradespace Exploration is part of the SE Process

- A Project Tradespace Model can help identify the Pareto frontier (i.e., those alternatives that are not dominated)

- Building a Project Tradespace Model requires more up-front effort, but
 - Results in a better understanding of the project-level effects of design changes
 - Helps to identify project disconnects earlier
 - Tends to produce more rigorous and consistent results so as to foster better decisions.



D. Tradespace Mini-Workshop

Tradespace Modeling Challenge for Exploration

- What flight rates would meet minimum lunar outpost operational needs over a variety of scenarios?
- Could lunar outpost concepts be changed to avoid sustainability issues?
- What kind of effort would be necessary to construct and maintain a viable lunar outpost?
- Is the architecture suitable for a given scientific, exploratory, or technology demonstration goal?
- How might Constellation architecture requirements be influenced by the revealed needs of a future outpost?

Lunar Outpost Design and Output Vectors

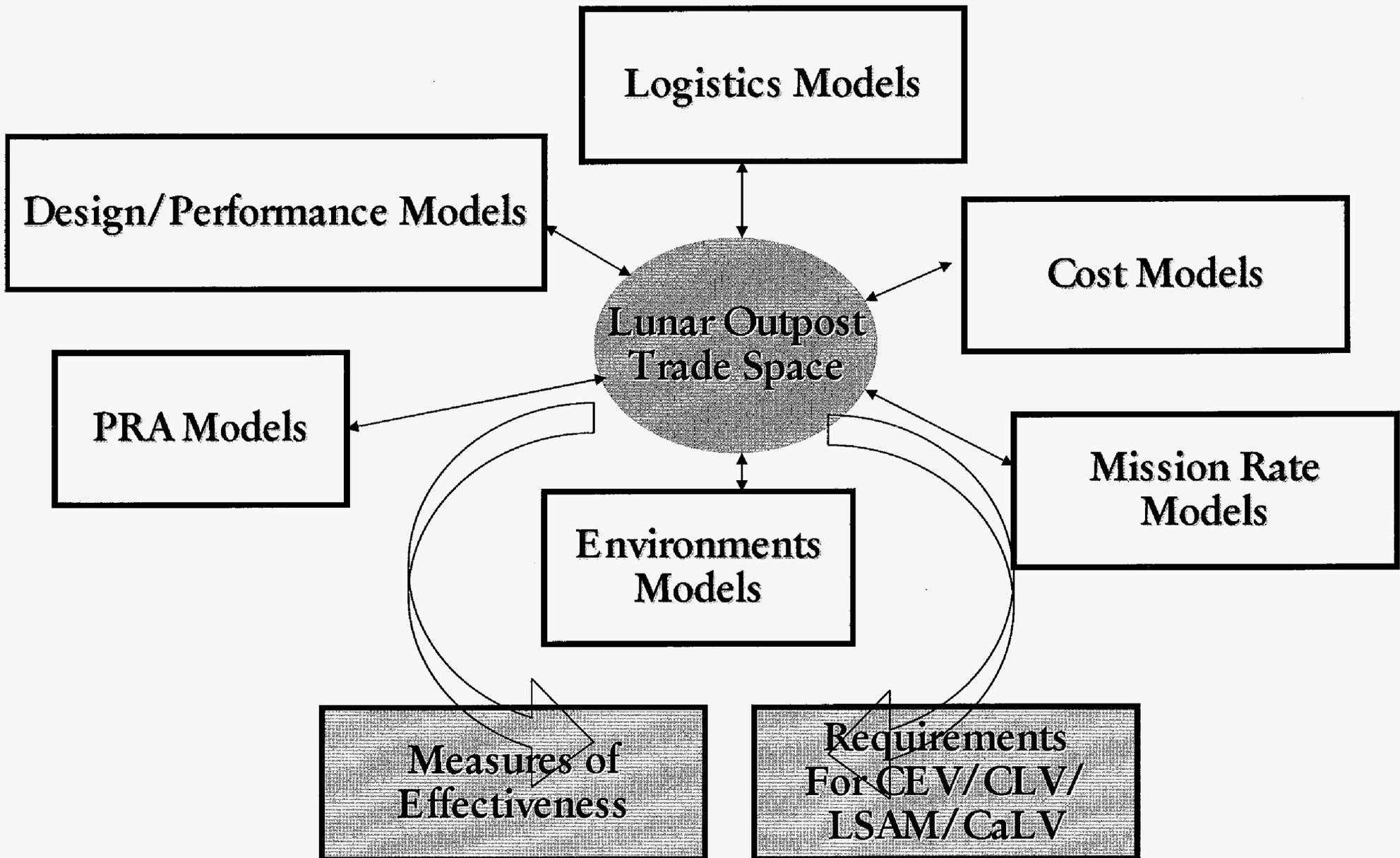
Inputs (Design Vector)

- ECLSS closure (%)
- Outpost location
- Power source type and output
- Level of repair (%)
- Surface crew size
- EVA frequency
- Surface stay duration
- Rover system type and performance
- Crew flight rate
- Cargo flight rate
- Infrastructure flight rate

Outputs (Output Vector)

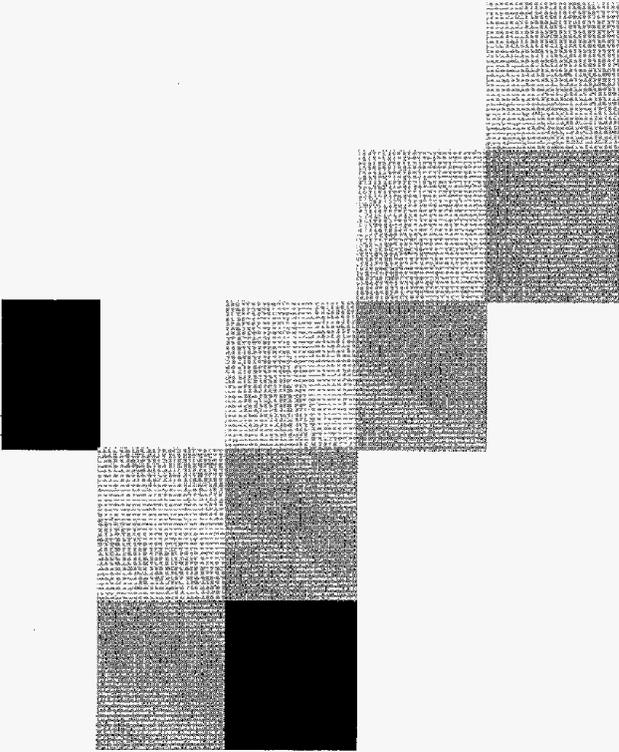
- Lunar surface exploration footprint
- Outpost occupancy (%)
- Outpost EVA time
- Infrastructure mass
- Payload mass
- Logistics mass
- Sample return mass
- Outpost completion date

Lunar Outpost Systems Analysis



Lunar Outpost Trade Space Model

- Measures of Effectiveness
 - Exploration Mass Delivered (Sci/Engr)
 - Exploration Crewtime (EVA)
 - Exploration Footprint
- Major Inputs Affecting MoEs
 - CaLV cargo flights
 - ISRU
 - Surface mobility
 - ECLSS closure



Backup Material

How a Tradespace Model Fits Into the SE Process

- PTM development is a collaborative effort by the project team
 - Step 1: Determine the most important project design issues/trades
 - Step 2: Hold one-on-one meetings with cognizant engineers to develop or identify appropriate models
 - Step 3: Models are linked generally via WBS/PBS elements
 - Step 4: Review by team for completeness and consistency
- PTM development process often illuminates interface issues
- The PTM is not static! Ideally as project progresses:
 - Team augments scope and increases resolution of the models
 - User friendliness increases
 - Databases are improved and expanded
- PTM archives trade study results
- PTM evolves into a capability for future in-situ missions