

Decision Management

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Many systems engineering decisions are difficult because they include numerous stakeholders, multiple competing objectives, substantial uncertainty, and significant consequences. In these cases, good decision making requires a formal decision management process. The purpose of the decision management process is:

“...to provide a structured, analytical framework for objectively identifying, characterizing and evaluating a set of alternatives for a decision at any point in the life cycle and select the most beneficial course of action.”(ISO/IEC/IEEE 15288)

Decision situations (opportunities) are commonly encountered throughout a system’s lifecycle. The decision management method most commonly employed by systems engineers is the trade study. Trade studies aim to define, measure, and assess shareholder and stakeholder value to facilitate the decision maker’s search for an alternative that represents the best balance of competing objectives. By providing techniques for decomposing a trade decision into logical segments and then synthesizing the parts into a coherent whole, a decision management process allows the decision maker to work within human cognitive limits without oversimplifying the problem. Furthermore, by decomposing the overall decision problem, experts can provide assessments of alternatives in their area of expertise.



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Decision Management Process

The decision analysis process is depicted in Figure 1 below. The decision management process is based on several best practices, including:

- Utilizing sound mathematical technique of decision analysis for trade studies. Parnell (2009) provided a list of decision analysis concepts and techniques.
- Developing one master decision model, followed by its refinement, update, and use, as required for trade studies throughout the system life cycle.
- Using Value-Focused Thinking (Keeney 1992) to create better alternatives.
- Identifying uncertainty and assessing risks for each decision.

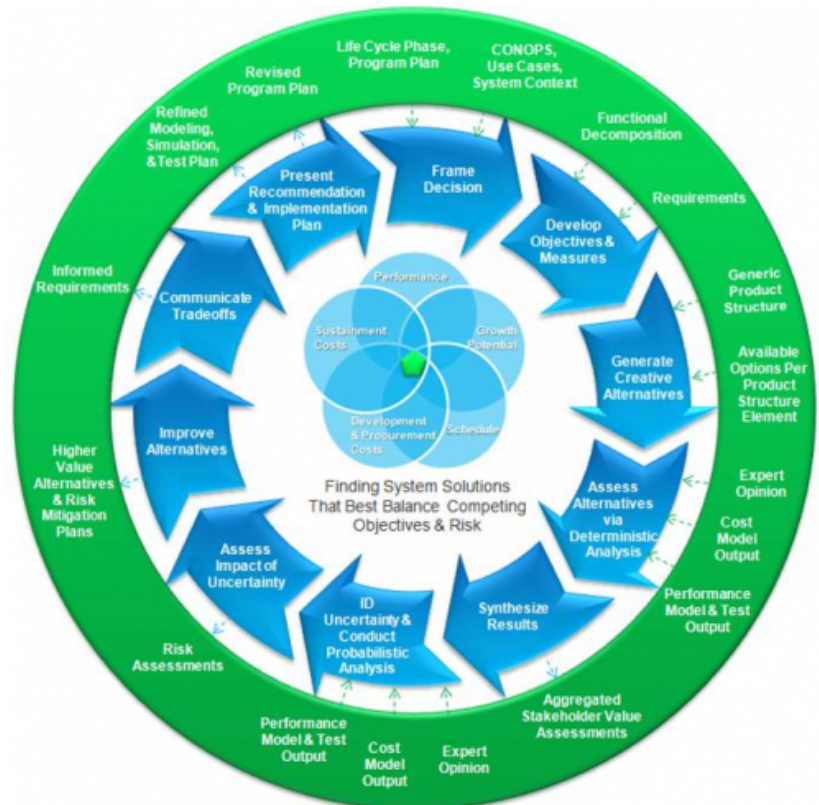


Figure 1. Decision Management Process (INCOSE DAWG 2013). Permission granted by Matthew Cilli who prepared image for the INCOSE Decision Analysis Working Group (DAWG). All other rights are reserved by the copyright owner.

The center of the diagram shows the five trade space objectives (listed clockwise): Performance, Growth Potential, Schedule, Development & Procurement Costs, and Sustainment Costs. The ten blue arrows represent the decision management process activities and the white text within the green ring represents SE process elements. Interactions are represented by the small, dotted green or blue arrows. The decision analysis process is an iterative process. A hypothetical UAV decision problem is used to illustrate each of the activities in the following sections.

Framing and Tailoring the Decision

To ensure the decision team fully understands the decision context, the analyst should describe the system baseline, boundaries and interfaces. The decision context includes: the system definition, the life cycle stage, decision milestones, a list of decision makers and stakeholders, and available resources. The best practice is to identify a decision problem statement that defines the decision in terms of the system life cycle.

Developing Objectives and Measures

Defining how an important decision will be made is difficult. As Keeney (2002) puts it:

Most important decisions involve multiple objectives, and usually with multiple-objective decisions, you can't have it all. You will have to accept less achievement in terms of some objectives in order to achieve more on other objectives. But how much less would you accept to achieve how much more?

The first step is to develop objectives and measures using interviews and focus groups with subject matter experts (SMEs) and stakeholders. For systems engineering trade-off analyses, stakeholder value often includes competing objectives of performance, development schedule, unit cost, support costs, and growth potential. For corporate decisions, shareholder value would also be added to this list. For performance, a functional decomposition can help generate a thorough set of potential objectives. Test this initial list of fundamental objectives by checking that each fundamental objective is essential and controllable and that the set of objectives is complete, non-redundant, concise, specific, and understandable (Edwards et al. 2007). Figure 2 provides an example of an objectives hierarchy.

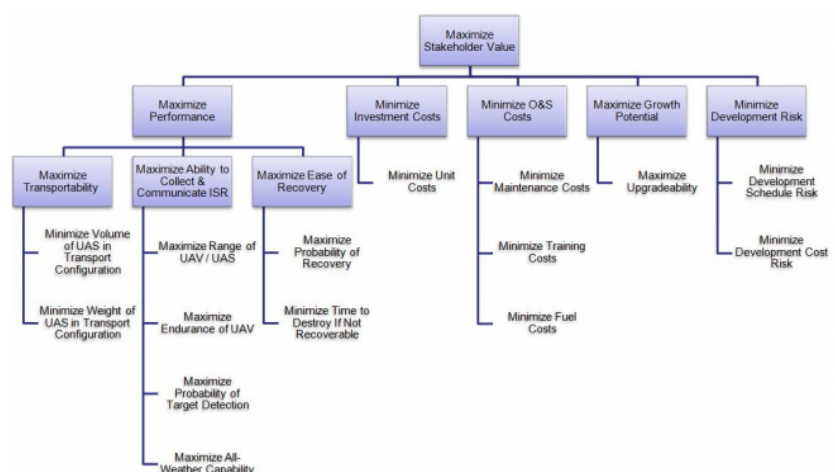


Figure 2. Fundamental Objectives Hierarchy (INCOSE DAWG 2013). Permission granted by Matthew Cilli who prepared image for the INCOSE Decision Analysis Working Group (DAWG). All other rights are reserved by the copyright owner.

For each objective, a measure must be defined to assess the value of each alternative for that objective. A

measure (attribute, criterion, and metric) must be unambiguous, comprehensive, direct, operational, and understandable (Keeney & Gregory 2005). A defining feature of multi-objective decision analysis is the transformation from measure space to value space. This transformation is performed by a value function which shows returns to scale on the measure range. When creating a value function, the walk-away point on the measure scale (x-axis) must be ascertained and mapped to a 0 value on the value scale (y-axis). A walk-away point is the measure score where regardless of how well an alternative performs in other measures, the decision maker will walk away from the alternative. He or she does this through working with the user, finding the measure score beyond, at which point an alternative provides no additional value, and labeling it "stretch goal" (ideal) and then mapping it to 100 (or 1 and 10) on the value scale (y-axis). Figure 3 provides the most common value curve shapes. The rationale for the shape of the value functions should be documented for traceability and defensibility (Parnell et al. 2011).

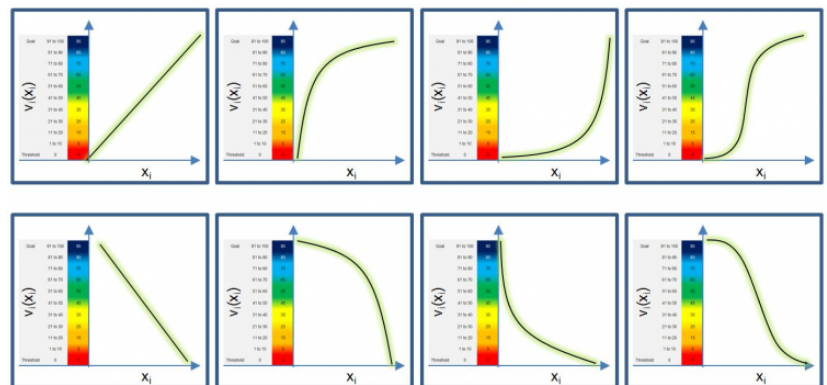


Figure 3. Value Function Examples (INCOSE DAWG 2013).

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The mathematics of multiple objective decision analysis (MODA) requires that the weights depend on importance of the measure and the range of the measure (walk away to stretch goal). A useful tool for determining priority weighting is the swing weight matrix (Parnell et al. 2011). For each measure, consider its importance through determining whether the measure corresponds to a defining, critical, or enabling function and consider the gap between the current capability and the desired capability; finally, put the name of the measure in the appropriate cell of the matrix (Figure 4). The highest priority weighting is placed in the upper-left corner and assigned an unnormalized weight of 100. The unnormalized weights are monotonically decreasing to the right and down the matrix. Swing weights are then

assessed by comparing them to the most important value measure or another assessed measure. The swing weights are normalized to sum to one for the additive value model used to calculate value in a subsequent section.

Highest Weight		Level of importance of the value measure								
		Mission Critical	Swt	Mwt	Mission Effectiveness	Swt	Mwt	Mission Enhancing	Swt	Mwt
Capability gap	Large	UAV range in nm	100	0.285	All weather capability	50	0.142	UAV P[R]	10	0.028
	Medium	UAV endurance in minutes	80	0.228	UAS volume in ft3	40	0.114			
	Small	UAV P[D]	50	0.142	UAS weight in pounds	20	0.057	UAV destruct time in seconds	1	0.003
Total swing weight			351	f _i = swing weight			w _i = measure weight			$w_i = \frac{f_i}{\sum_{i=1}^n f_i}$

Figure 4. Swing Weight Matrix (INCOSE DAWG 2013).

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Generating Creative Alternatives

To help generate a creative and comprehensive set of alternatives that span the decision space, consider developing an alternative generation table (also called a morphological box) (Buede, 2009; Parnell et al. 2011). It is a best practice to establish a meaningful product structure for the system and to be reported in all decision presentations (Figure 5).







	Cardinal	Buzzard	Crow	Pigeon	Robin	Dove
						
Subsystem	Design Choice	Design Choice	Design Choice	Design Choice	Design Choice	Design Choice
Propulsion System	Electric 300W & Li P	Electric 300W w/ Li Ion	Electric 600W w/ Solar	Elect 600W w Fuel Cell	Piston Engine 2.5 HP	Piston Engine 4.0 HP
Fuel	NA	NA	NA	NA	JP-8	JP-8
Fuel Tank Capacity	NA	NA	NA	NA	5 liter	7 liter
Propeller	18" Rear	20" rear	22" rear	24" Front	26" Front	28" Front
Wing Configuration	5 ft, Conventional	6 ft, Canard	6 ft, Tandem Wing	7 ft, Three Surface	8 ft, Conventional	9 ft, Conventional
Fin Configuration	Twin Boom Conv.	Inverted V	V Tail	Conventional	H Tail	Cruciform
Actuators	Electromagnetic	Hydraulic	MEMS	Hydraulic	Hydraulic	Hydraulic
Fuselage X Section	12" Diameter	14" Diameter	16" Diameter	18" Diameter	20" Diameter	22" Diameter
Airframe Material	Graphite Epoxy	Graphite Epoxy	Aramid-epoxy	Boron-epoxy	Fiberglass-epoxy	Fiberglass-epoxy
Avionics Arch.	Simplex	Simplex	Triplex	Triplex	Triplex	Triplex
Navigation Sensor	MEMS GPS / INS	MEMS GPS / INS	MEMS GPS / INS	MEMS GPS / INS	MEMS GPS / INS	MEMS GPS / INS
External Comms	LOS COMM Link	LOS COMM Link	LOS + SATCOM Link	LOS + SATCOM Link	LOS + SATCOM Link	LOS + SATCOM Link
Internal Comms	MIL-STD-1553B	MIL-STD-1553B	MIL-STD-1553B	MIL-STD-1553B	MIL-STD-1553B	MIL-STD-1553B
Autopilot	Pre-Programmed, Auto	Semi-Autonomous	Remotely Piloted	Pre-Programmed, Auto	Pre-Programmed, Auto	Pre-Programmed, Auto
Launch / Recovery	Hand / Belly	Hand / Belly	Hand / Belly	Hand / Belly	Hand / Belly	Hand / Belly
Acquisition Sensor	Un-cooled IR	Day Video	Day Video, Cooled IR	Day Video, Cooled IR	Day Video	SAR, Acoustic, Day, IR
Sensor Actuation	Pan-tilt	Pan-tilt-roll	Roll-tilt	Pan-tilt	Pan tilt	Pan tilt
Characteristics	Measurement	Measurement	Measurement	Measurement	Measurement	Measurement
Weight	5 lbs	10 lbs	10 lbs	15 lbs	30 lbs	40 lbs
Max Airspeed	60 kph	50 kph	80 kph	70 kph	60 kph	80 kph
Climb Rate	200 m / minute	150 m / minute	250 m / minute	200 m / minute	200 m / minute	250 m / minute

Figure 5. Descriptions of Alternatives (INCOSE DAWG 2013).

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Assessing Alternatives via Deterministic Analysis

With objectives and measures established and alternatives having been defined, the decision team should engage SMEs, equipped with operational data, test data, simulations, models, and expert knowledge. Scores are best captured on scoring sheets for each alternative/measure combination which document the source and rationale. Figure 6 provides a summary of the scores.







Raw Scorecard (expected performance values)			RELOCATE UAV			EMPLOY UAV			RECOVER UAV		GROWTH	UNIT	DEVELOPMENT RISK		OPERATION & SUPPORT COST		
			Minimize UAV weight	Minimize UAV volume	Maximize all weather capability	Maximize UAV range	Maximize UAV probability of detection	Maximize UAV endurance	Maximize probability of recovery	Minimize time to destroy if not recoverable	Maximize Lag reliability	Minimize unit cost	Minimize development schedule risk	Minimize test & development cost	Minimize training cost	Minimize maintenance cost	Minimize fuel cost
ID	Name	Image	(lbs)	ft ³	index	km	P[D]	hours	P[R]	sec	index	FY13\$K	index	index	FY13\$	FY13\$	FY13\$
1	Cardinal		5	12	3	10	0.92	0.5	0.6	1	0.3	250	0.9	0.9	300	300	0
2	Buzzard		10	15	1	10	0.9	1	0.7	2	0.6	300	0.8	0.8	300	300	0
3	Crow		10	20	3	70	0.92	1	0.8	2	0.6	350	0.7	0.7	300	300	0
4	Pigeon		15	30	3	80	0.92	1.5	0.9	2	0.6	400	0.6	0.6	300	300	0
5	Robin		30	40	1	90	0.9	2	0.9	2	0.6	500	0.5	0.5	500	500	300
6	Dove		40	50	5	100	0.94	2	0.9	3	0.9	700	0.4	0.4	500	500	500
7	Ideal		5	10	5	100	1	2	0.9	1	1	200	1	1	250	0	0

Figure 6. Alternative Scores (INCOSE DAWG 2013).

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Note that in addition to identified alternatives, the score matrix includes a row for the ideal alternative. The ideal is a tool for value-focused thinking, which will be covered later.

Synthesizing Results

Next, one can transform the scores into a value table, by using the value functions developed previously. A color heat map can be useful to visualize value tradeoffs between alternatives and identify where alternatives need improvement (Figure 7).







Heat-Indexed Value Scorecard			RELOCATE UAV		EMPLOY UAV				RECOVER UAV		GROWTH POT.	UNIT COST	DEVELOPMENT RISK		OPERATION & SUPPORT COST		
			Minimize UAV weight 0.06	Minimize UAV volume 0.11	Maximize all weather capability 0.14	Maximize UAV range 0.28	Maximize UAV probability of detection 0.14	Maximize UAV endurance 0.23	Maximize probability of recovery 0.03	Minimize time to destroy if not recoverable 0.01	Maximize upgradability 1	Minimize unit cost 1	Minimize development schedule risk 0.5	Minimize development cost risk 0.5	Minimize training cost 0.33	Minimize maintenance cost 0.33	Minimize fuel cost 0.33
ID	Name	Image															
1	Cardinal		100	97	90	13	59	4	47	100	31	94	83	88	83	85	100
2	Buzzard		86	92	1	13	42	66	77	90	61	88	67	67	89	85	100
3	Crow		85	85	90	83	59	60	80	75	61	82	50	50	83	85	100
4	Pigeon		72	87	86	86	59	80	100	75	81	78	34	34	89	85	100
5	Robin		29	29	1	95	42	100	100	60	61	96	18	18	17	67	51
6	Dove		1	1	100	100	75	100	100	1	91	1	1	1	17	67	18
7	Ideal		100	100	100	100	100	100	100	100	100	100	100	100	100	100	100

Figure 7. Value Scorecard with Heat Map (INCOSE DAWG 2013). Permission granted by Richard Swanson who prepared image for the INCOSE Decision Analysis Working Group (DAWG). All other rights are reserved by the copyright owner.

The additive value model uses the following equation to calculate each alternative's value:

$$v(x) = \sum_{i=1}^n w_i v_i(x_i)$$

where

$v(x)$ is the alternative's value

$i = 1$ to n is the number of the measure

x_i is the alternative's score on the i^{th} measure

$v_i(x_i)$ is the single dimensional value of a score of x_i

w_i is the weight of the i^{th} measure

and $\sum_{i=1}^n w_i = 1$ (all weights sum to one)

The value component chart (Figure 8) shows the total value and the weighted value measure contribution of each alternative (Parnell et al. 2011).

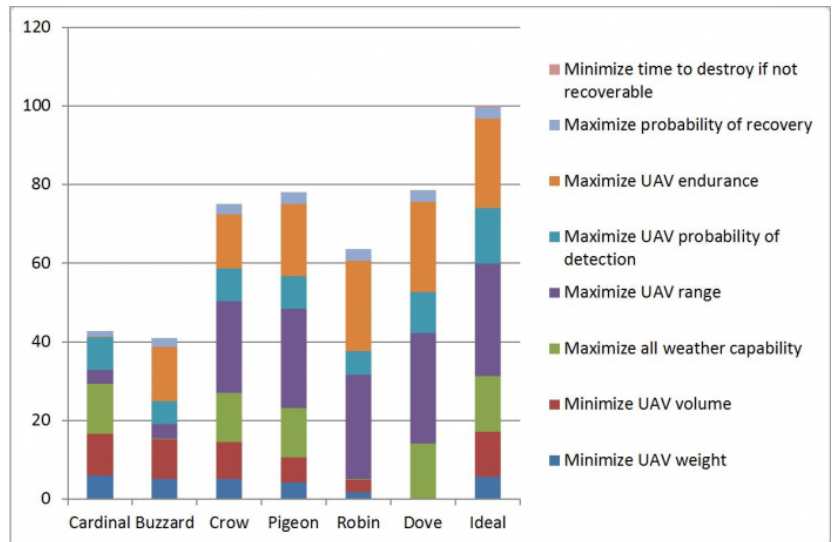


Figure 8. Value Component Graph (INCOSE DAWG 2013).

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The heart of a decision management process for system engineering trade off analysis is the ability to assess all dimensions of shareholder and stakeholder value. The stakeholder value scatter plot in Figure 9 shows five dimensions: unit cost, performance, development risk, growth potential, and operation and support costs for all alternatives.

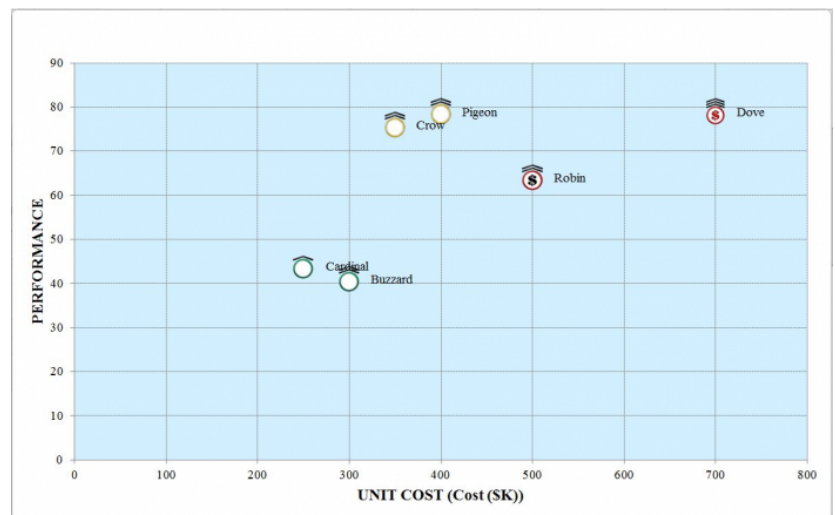


Figure 9. Example of a Stakeholder Value Scatterplot (INCOSE DAWG 2013).

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Each system alternative is represented by a scatter plot marker (Figure 9). An alternative's unit cost and performance value are indicated by x and y positions respectively. An alternative's development risk is indicated by the color of the marker (green = low,

yellow= medium, red = high), while the growth potential is shown as the number of hats above the circular marker (1 hat = low, 2 hats = moderate, 3 hats = high).

Identifying Uncertainty and Conducting Probabilistic Analysis

As part of the assessment, the SME should discuss the potential uncertainty of the independent variables. The independent variables are the variables that impact one or more scores; the scores that are independent scores. Many times the SME can assess an upper, nominal, and lower bound by assuming low, moderate, and high performance. Using this data, a Monte Carlo Simulation summarizes the impact of the uncertainties and can identify the uncertainties that have the most impact on the decision.

Accessing Impact of Uncertainty - Analyzing Risk and Sensitivity

Decision analysis uses many forms of sensitivity analysis including line diagrams, tornado diagrams, waterfall diagrams and several uncertainty analyses including Monte Carlo Simulation, decision trees, and influence diagrams (Parnell et al. 2013). A line diagram is used to show the sensitivity to the swing weight judgment (Parnell et al. 2011). Figure 10 shows the results of a Monte Carlo Simulation of performance value.

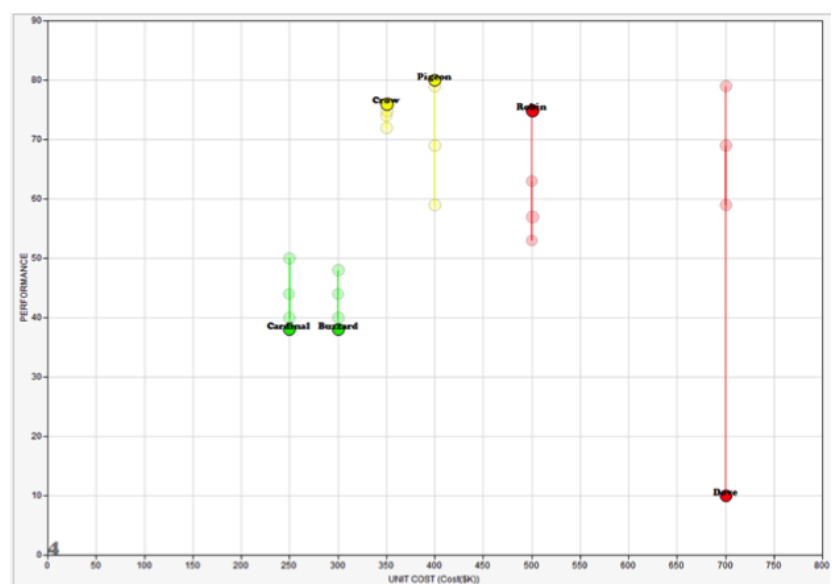


Figure 10. Uncertainty on Performance Value from Monte Carlo Simulation (INCOSE DAWG 2013). Permission granted by Matthew Cilli who prepared image for the INCOSE Decision Analysis Working Group (DAWG). All other rights are reserved by the copyright owner.

Improving Alternatives

Mining the data generated for the alternatives will likely reveal opportunities to modify some design choices to claim untapped value and/or reduce risk. Taking advantage of initial findings to generate new and creative alternatives starts the process of transforming the decision process from "alternative-focused thinking" to "value-focused thinking" (Keeney 1993).

Communicating Tradeoffs

This is the point in the process where the decision analysis team identifies key observations about tradeoffs and the important uncertainties and risks.

Presenting Recommendations and Implementing Action Plan

It is often helpful to describe the recommendation(s) in the form of a clearly-worded, actionable task-list in order to increase the likelihood of the decision implementation. Reports are important for historical traceability and future decisions. Take the time and effort to create a comprehensive, high-quality report detailing study findings and supporting rationale. Consider static paper reports augmented with dynamic hyper-linked e-reports.

The Cognitive Bias Effect on Decisions

Research by (Kahneman 2011) and (Thaler and Sunstein 2008) has concluded that cognitive bias can seriously distort decisions made by any decision maker. Both Kahneman and Thaler were awarded the Nobel prize for their work. The cause of this distortion is called the cognitive bias. These distorted decisions have contributed to major catastrophes, such as Challenger and Columbia. Other sources attributing major catastrophes are (Murata, Nakamura, and Karwowski 2015) and (Murata 2017).

(Kahneman 2011) and (Thaler and Sunstein 2008) have identified a large number of individual biases, the most well-known of which is the confirmation bias. This bias states that humans have a tendency to interpret new evidence as confirmation of one's existing beliefs or

theories. Regarding mitigation of these biases, there is general agreement that self-mitigation by the decision-maker is not feasible for most biases. (Thaler and Sunstein 2008) provide methods to influence the mitigation of most biases. They refer to these influences as “nudges”.

Considering cognitive biases in a systems engineering is discussed by (Jackson 2017, Jackson and Harel 2017), and (Jackson 2018). The primary theme of these references is that rational decisions are rarely possible and that cognitive bias must be taken into account.

Decisions with Cognitive Bias

According to (INCOSE 2015) ideal decisions are made while “objectively identifying, characterizing, and evaluating a set of alternatives for a decision...” Research in the field of behavioral economics has shown that these decisions can be distorted by a phenomenon known as cognitive bias. Furthermore, most decision makers are unaware of these biases. The literature also provides methods for mitigating these biases.

According to (Haselton, Nettle, and Andrews 2005, p. 2) a cognitive bias represents a situation in which “human cognition reliably produces representations that are systematically distorted compared to some aspect of objective reality.” Cognitive biases are typically stimulated by emotion and prior belief. The literature reveals large numbers of cognitive biases of which the following three are typical:

1. The rankism bias. According to (Fuller 2011), rankism is simply the idea that persons of higher rank in an organization are better able to assert their authority over persons of lower rank regardless of the decision involved. Rankism frequently occurs in aircraft cockpits. According to (McCreary et al. 1998), rankism was a factor in the famous Tenerife disaster.
2. The complacency bias. According to (Leveson 1995, pp. 54-55), complacency is the disregard for safety and the belief that current safety measures are adequate. According to (Leveson 1995, pp. 54-55), complacency played a role in the Three Mile Island and Bhopal disasters.
3. The optimism bias. According to (Leveson 1995, pp. 54-55), famous physicist Richard Feynman states that NASA “exaggerates the reliability of the system.” This

is an example of the optimism bias.

Mitigation of Cognitive Bias

Various sources have suggested methods to mitigate the effects of cognitive bias. Following are some of the major ones.

1. Independent Review. The idea of independent review is that advice on decisions should come from an outside body, called by the Columbia Accident Investigation Board (CAIB) (NASA 2003, 227) as the Independent Technical Authority (ITA). This authority must be both organizationally and financially independent of the program in question. That is, the ITA cannot be subordinate to the program manager.
2. Crew Resource Management. Following a period of high accident rate, several airlines have adopted the crew resource management (CRM) method. The primary purposes of this method are first to assure that all crew members do their job properly and secondly that they communicate with the pilot effectively when they have a concern. The impetus for this method was the judgment that many pilots were experiencing the rankism bias or were preoccupied with other tasks and simply did not understand the concerns of the other crew members. The result is that this strategy has been successful, and that the accident rate has fallen.
3. The Premortem. (Kahneman 2011) (pp. 264-265) suggests this method of nudging in an organizational context. This method, like others, requires a certain amount of willingness on the part of the decision-maker to participate in this process. It calls for decision-makers to surround themselves with trusted experts in advance of major decisions. According to Kahneman the primary job of the experts is to present the negative argument against any decision. For example, the decision-maker should not authorize the launch now, perhaps later.

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