

# Identifying Collaboration and Communication Needs for Complex Systems Design Using Digital Engineering

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**By**

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Motivation

Problem Description

Trade Space Evaluation in Complex System Design

A Mathematical Surrogate Model for Studying Behaviors

Evaluation Metrics for Validating the Model

Future Work

**Special thanks to my advisor, Dr. Paul T. Grogan**

- DoD Digital Engineering Strategy
  - Published June 2018
  - Modernize design, development, operation and sustainment
  - Transform acquisition and implementation
  - Improve speed for critical capability delivery to the warfighter
  - Connected data in a digital environment

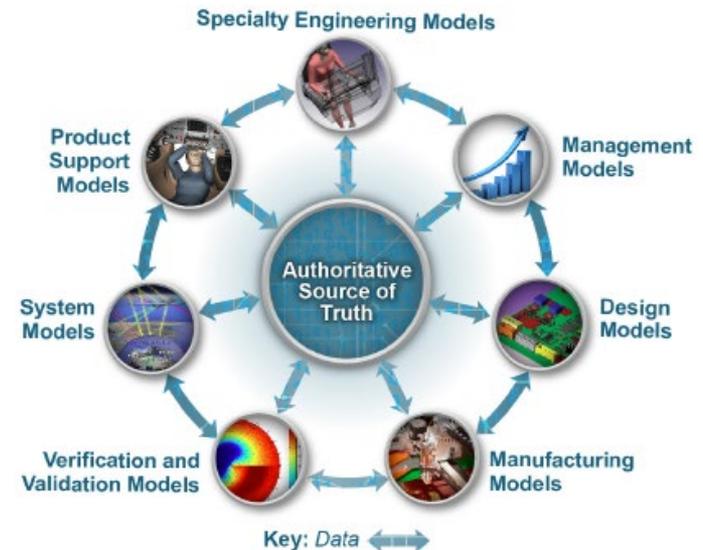
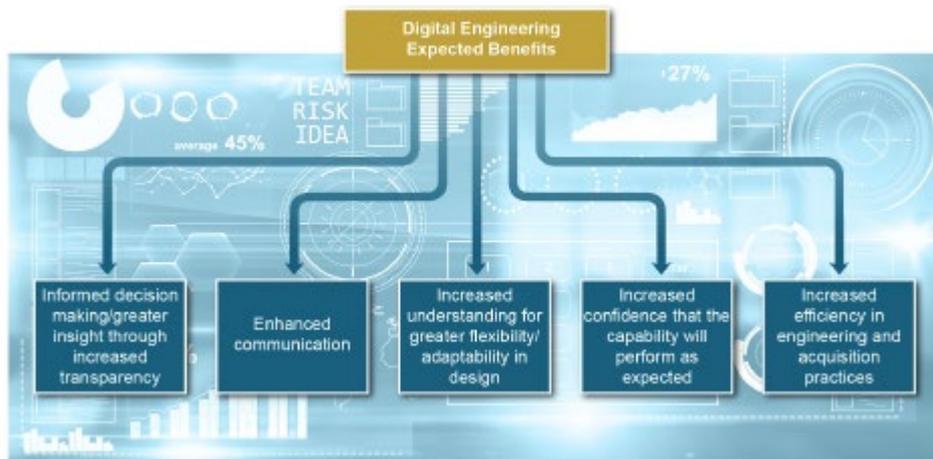
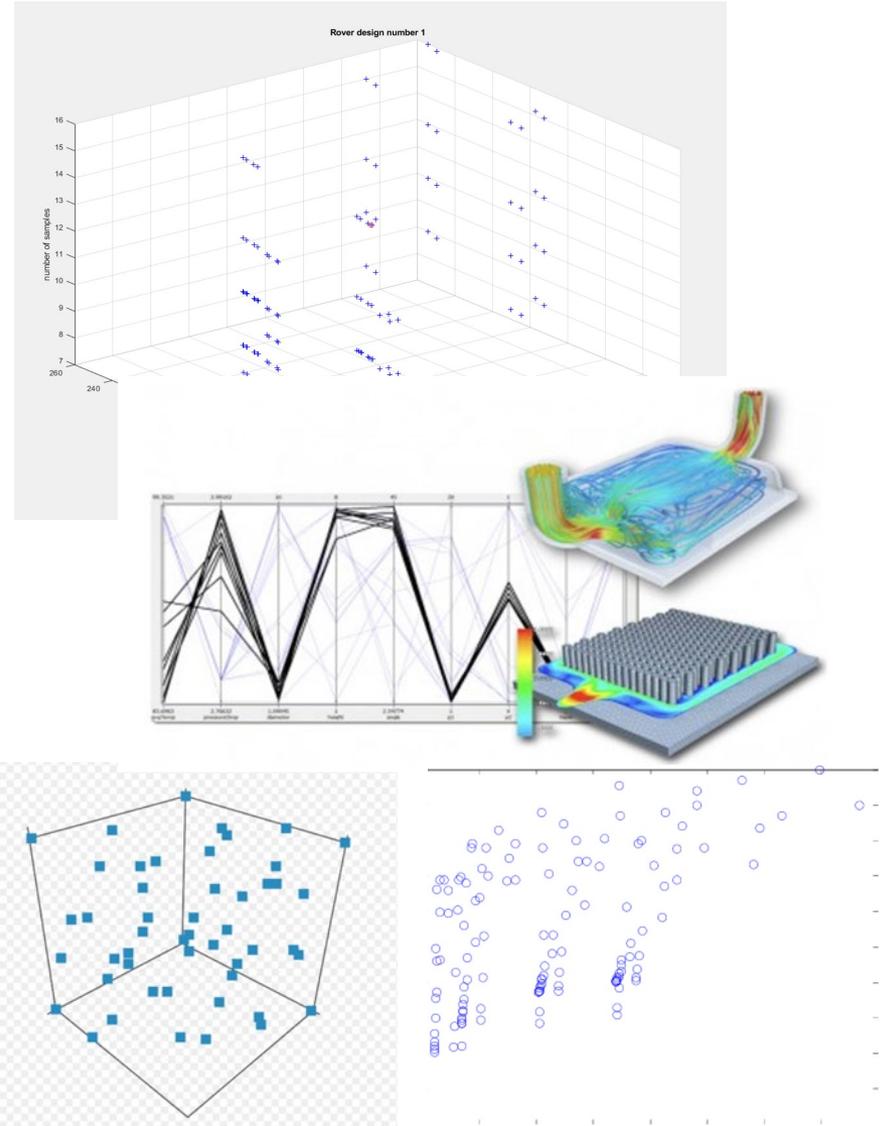


Image credit: DoD Digital Engineering Strategy, June 2018

- Lack of theory to suggest how Digital Engineering will change:
  - Engineering practice
  - Engineering lifecycle processes
  - Effect on system performance
- Implementing Digital Engineering goals without this understanding could limit desired results
- Need to study how complex communications and decision-making by individuals and teams impact system performance to identify how to best implement Digital Engineering
  - To adopt new technologies more rapidly
  - To design higher performing systems
  - To address workforce challenges

- Multidimensional and Multidisciplinary problem spaces
- Requires trade space to evaluate and determine best solution
- Can be very expensive and lengthy to create the models for the trades to be evaluated
- Solution choice still subject to stakeholder viewpoints
  - Priorities of budget, schedule, performance
  - Quantification of utility



- Evaluation techniques
  - Pareto front comparisons
  - Detailed performance models
  - Utility models
  
- Limitations of these techniques
  - Dependent model variables limit coupled or emergent behavior analysis
  - Can be impacted by:
    - Team or Contract organization
    - Task structure
    - Data accessibility
    - Subject Matter Expert availability
  - Decision making authority dependent
  - Time to create versus decision need date

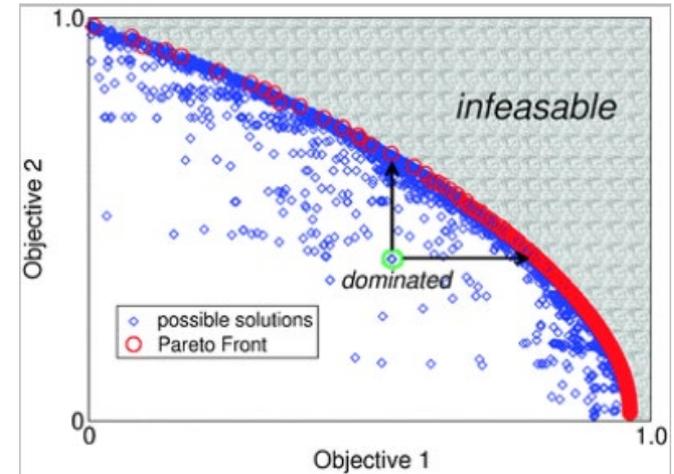
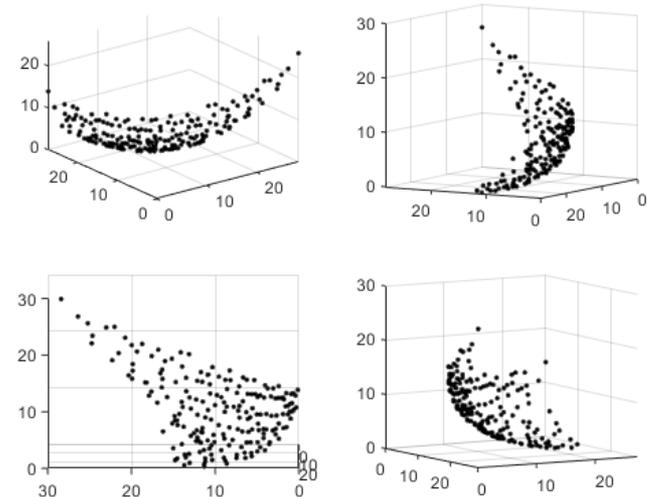


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- Mathematical model as a surrogate for detailed design or utility models
  - Used to study impact of coupled behavior on system performance
  - Have been used to analyze adaptive evolution in immune response and organizational performance
- An enabler to studying the impact of DE on:
  - Decision analysis of individuals, groups, human/machine teaming
  - Without detailed design model development
  - Behavior changes based on change in data and connected information
- Candidate mathematical models exist to evaluate the approach
  - First investigated is the NK model and variants

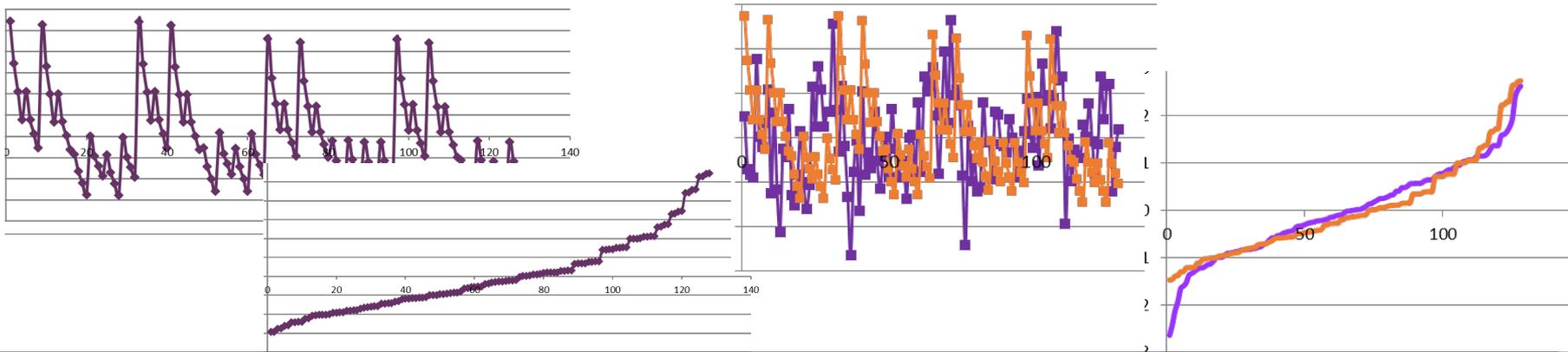
- Basic NK model description
  - A system has  $N$  variables, each variable can take on  $A$  possible values
  - The model assigns a “fitness contribution” to each variable ( $w_i$ )
    - This can be assigned at random from the uniform distribution on  $(0,1)$
  - The total fitness ( $W$ ) of a system is an average of the fitness contributions of each variable
  - $K$  defines the number of coupled variables influencing the fitness value of  $w_i$
- Some parts of the real-world problem easily align to the NK model elements:

Complex System Design Element	NK Model Element
Design variables	$N$
Design variable alternatives	$A$
Coupled behavior	$K$
Performance/Utility Value	$w_i, W$



# First look at alignment of complex system models and NK model - Structure

- Need to determine if the mathematical model can be utilized to perform simulations to study the practices, processes, behaviors and how DE changes them
  - Performance and utility models differ
  - Can the mathematical surrogate be tuned to represent those differences
- What is our evaluation metric for alignment to proceed
  - Previously compared normalized results from detailed performance model to the mathematical model to look at the performance (W) predicted by each



Look at System Performance Evaluation to Correlate Surrogate Model

- Initial abstractions describe the structure of the trade space
- Need abstractions to represent the intricate roles involved in evaluating and decision-making for system performance
  - To evaluate how those behaviors are impacted by DE
- Define agents that represent the behaviors of those roles
  - Utilize different decision-making strategies
  - Represent different collaboration and communication techniques

Complex System Design Element	Agent Element Definition
Design engineer	Individual decision-maker, consumes and produces data
Subject matter expert	Individual decision-maker, set knowledge base, consumes and produces data
Integrated Product Team	Group decision, consume and produce data

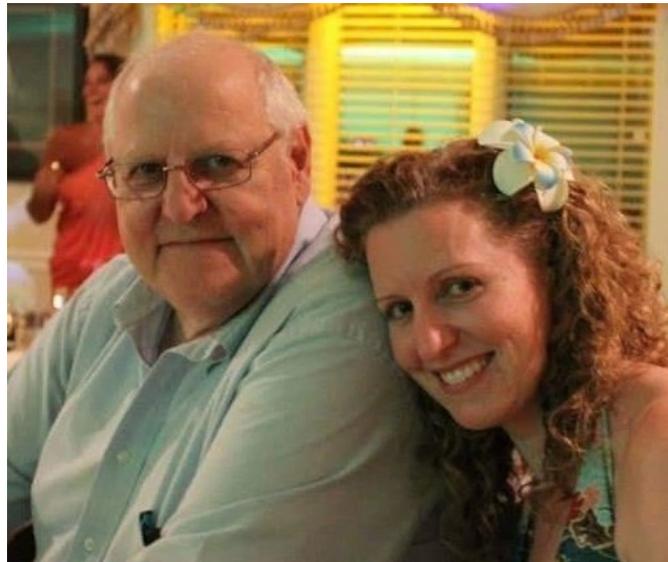


- How do engineers explore the trade space to find better performing solutions
  - Set number of design iterations
  - Set number of dependent variables to be explored
  - Constraints on technical solutions to be selected
  - Limits on design dimensions used to make decisions
- How can the mathematical fitness landscape be explored to imitate the same
  - Local neighborhood search of the solutions
  - Constraints on number of evaluations performed to improve performance
  - Incorporate probability into decision analysis rubric
  - Change dependency structure of the model for different technology infusion or strategy

**Composing the Agents to Represent Engineers, SMEs and Teams**

- Develop agents and behavior evaluation criteria for exercising on surrogate mathematical models
  - Correlate these to real world techniques for selecting system designs of desired performance and utility
  - Correlate to engineering practice, processes and lifecycle stages
- Identify how Digital Engineering could change these agents and behaviors
  - Data access and awareness
  - Design decision influence of data
  - Calibration of engineering decision-making
  - Needed level of expertise to evaluate
- Perform simulations and evaluate against the potential metrics
  - Expand comparison to other detailed design model representations

- Special thanks for continued research guidance and support:
  - Dr. Paul T. Grogan, Stevens Institute of Technology, Assistant Professor, School of Systems and Enterprises
  - Dr. Marshall Mattingly, Raytheon Missile Systems
  - The Collective Design (CoDe) Lab students and researchers, Stevens Institute of Technology, School of Systems and Enterprises
- Dedicated to Charles Andrew Sharo
  - August 18, 1948 – April 18, 2019





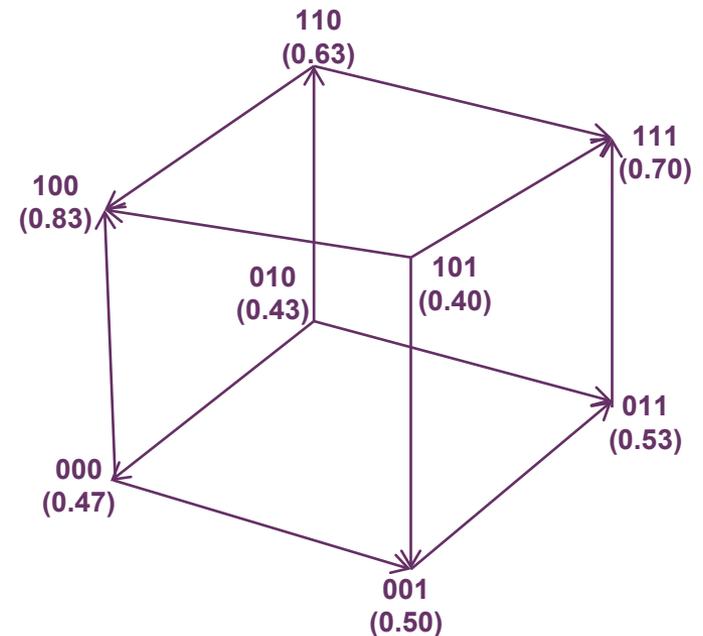
- We need a way to look at the impact of team communication and collaboration on design fitness without relying on a detailed design space model
  - Valid over a range of design problems
  - Before lengthy design and development process to build design models
- Candidate approach is an NK model from a class of mathematical (statistical) models
  - Describe the richness of epistatic interactions
    - The value of a given variable is affected by the values of other variables
  - Have been used to describe adaptive evolution in immune response as well as fitness of organizations
- Can the NK model can be tuned to show that it can be representative of the fitness space defined by complex design models?

- Basic model description

- A system has N variables, each variable can take on A possible values
- The model assigns a “fitness contribution” to each variable ( $w_i$ )
- This can be assigned at random from the uniform distribution on (0,1)
- The total fitness ( $W$ ) of a system is an average of the fitness contributions of each variable

$$W = \frac{1}{N} \sum_{i=1}^N w_i$$

1	2	3	$w_1$	$w_2$	$w_3$	$W$
0	0	0	0.6	0.3	0.5	0.47
0	0	1	0.1	0.5	0.9	0.50
0	1	0	0.4	0.8	0.1	0.43
0	1	1	0.3	0.5	0.8	0.53
1	0	0	0.9	0.9	0.7	0.83
1	0	1	0.7	0.2	0.3	0.40
1	1	0	0.6	0.7	0.6	0.63
1	1	1	0.7	0.9	0.5	0.70



- Contributions to fitness between coupled variables
  - $K$  defines the number of coupled variables influencing the fitness value of  $w_i$
  - $K = 0$  yields a smooth solution fitness landscape with a single peak for the solution with the optimal fitness
    - The contributions of each variable to the system fitness are entirely independent of all other variables
  - As  $K$  increases relative to  $N$ , the fitness landscape becomes rugged with multiple peaks representing local optima
    - For  $K = N-1$  the contributions of each variable are entirely dependent of the values for all other variables in the system
- The statistical model could represent local optima and the distance to reach a local optima

- Created a fitness landscape of potential solutions for Mars rover designs to compare to a randomly generated fitness landscape defined by an NK model
  - K=2 and K=6 have promising potential for representing the design dataset using both the unsorted and sorted fitness plots
- Limitations of this preliminary assessment
  - Single snapshot fitness assessment of the NK model as setup
    - Need to apply Monte Carlo analysis and look at confidence intervals to determine if this could be accepted or rejected as a feasible representation
  - Comparison to a single design fitness model
    - Other design fitness models may have different results in terms of fitness and tuning the NK model to it
  - The evaluation metric needs to be assessed for determining potential of the representation
    - Perhaps sorted fitness is not the best way to evaluate the goodness of fit