



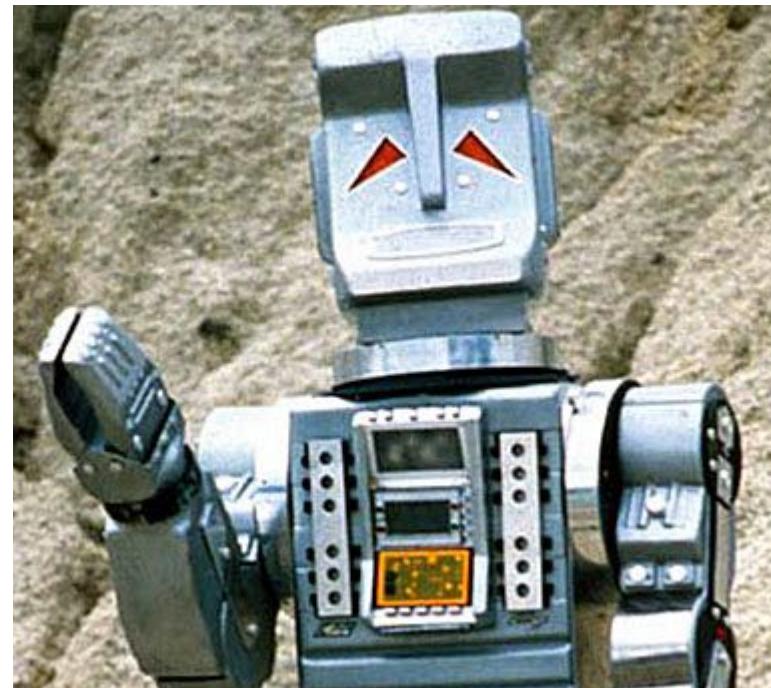
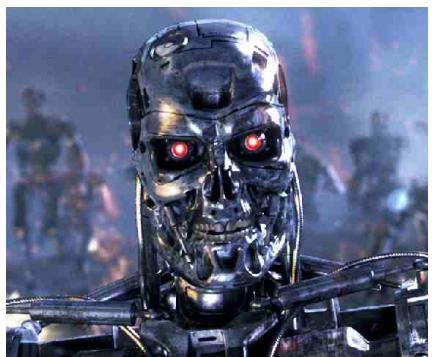
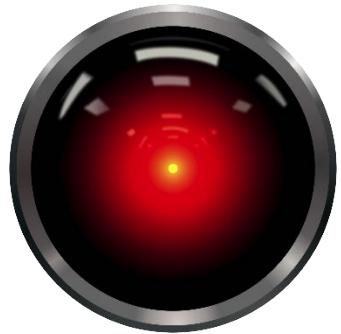
29th Annual **INCOSE**
international symposium

Orlando, FL, USA
July 20 - 25, 2019

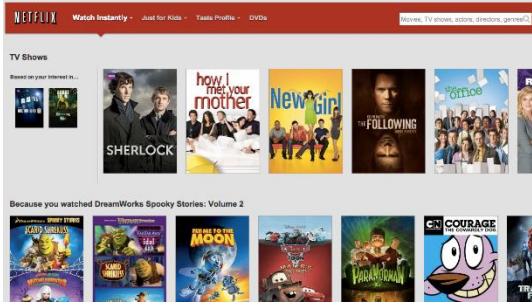
**Augmented Intelligence: Combining Model Based Systems Engineering with
AI & Machine Learning**

Augmented Intelligence for Systems Engineering

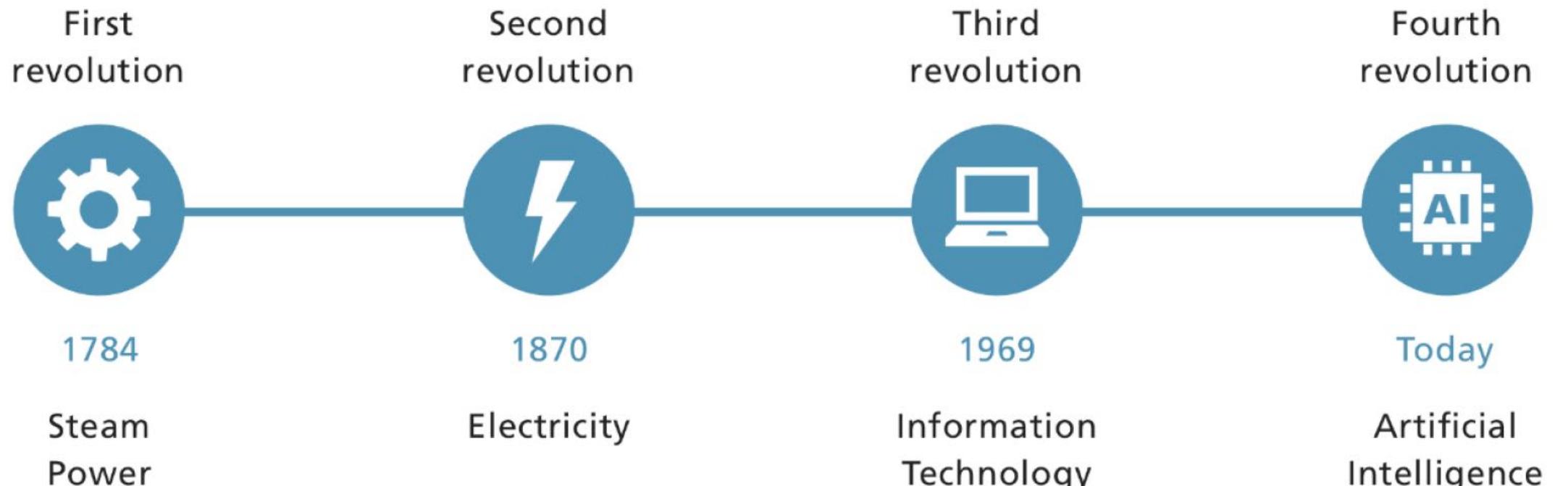
AI We Expected



AI We Got



AI and the Fourth Industrial Revolution



+ robotics, nanotechnology,
quantum computing, biotechnology,
IoT, 3D printing and autonomous
vehicles.

But wait...



Augmented Intelligence in SE: Beginnings



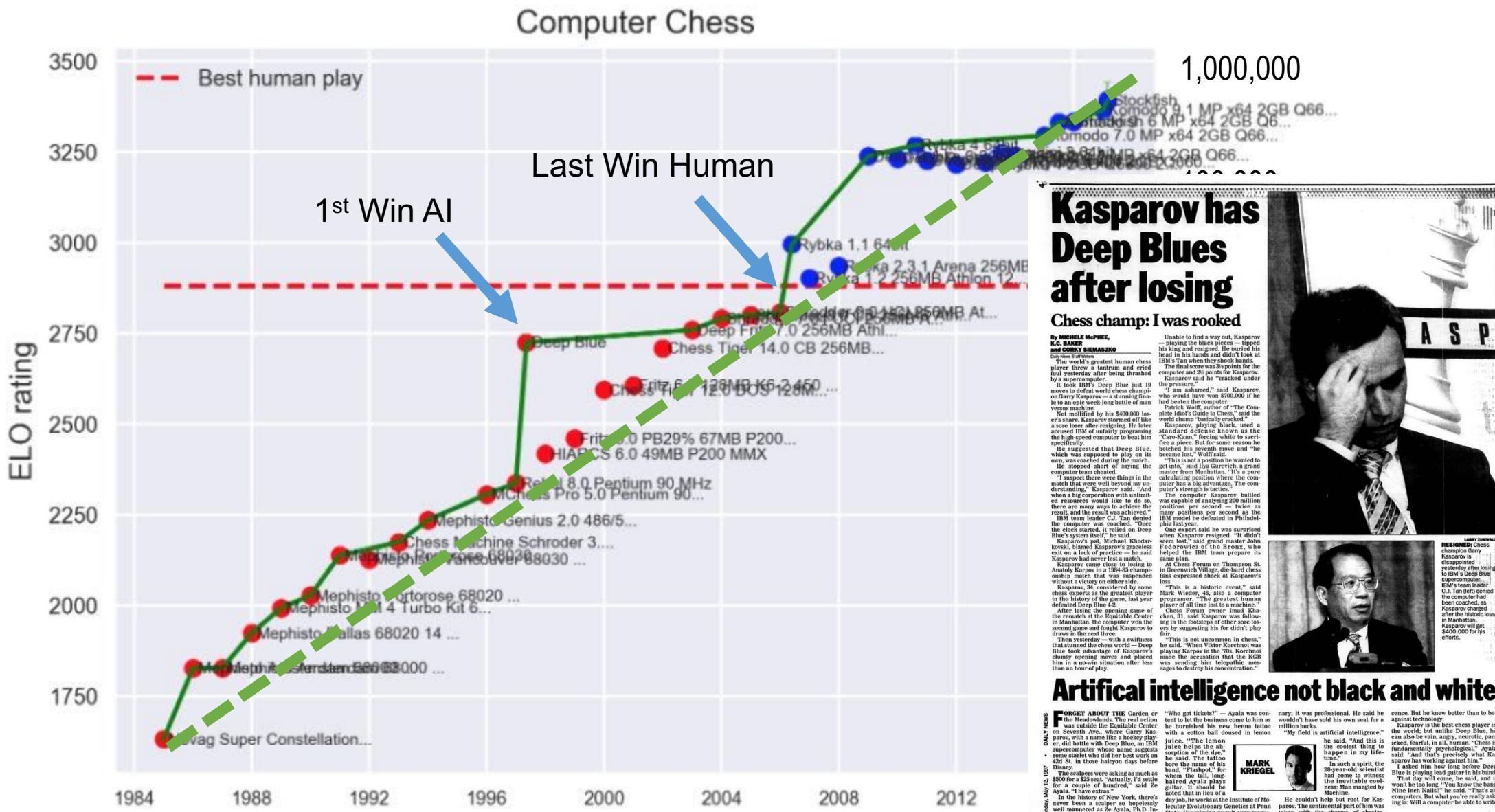
- INCOSE Challenge Team: Augmented Intelligence for Systems Engineering
- Chartered 2018:

...further the understanding of how computational approaches, such as artificial intelligence, machine learning, and data science, can collaborate with human systems engineers to measurably improve the system engineering effort. The challenge team will seek out approaches that enhance human capabilities in systems engineering.

<http://www.omgwiki.org/MBSE/doku.php?id=mbse:augmented>

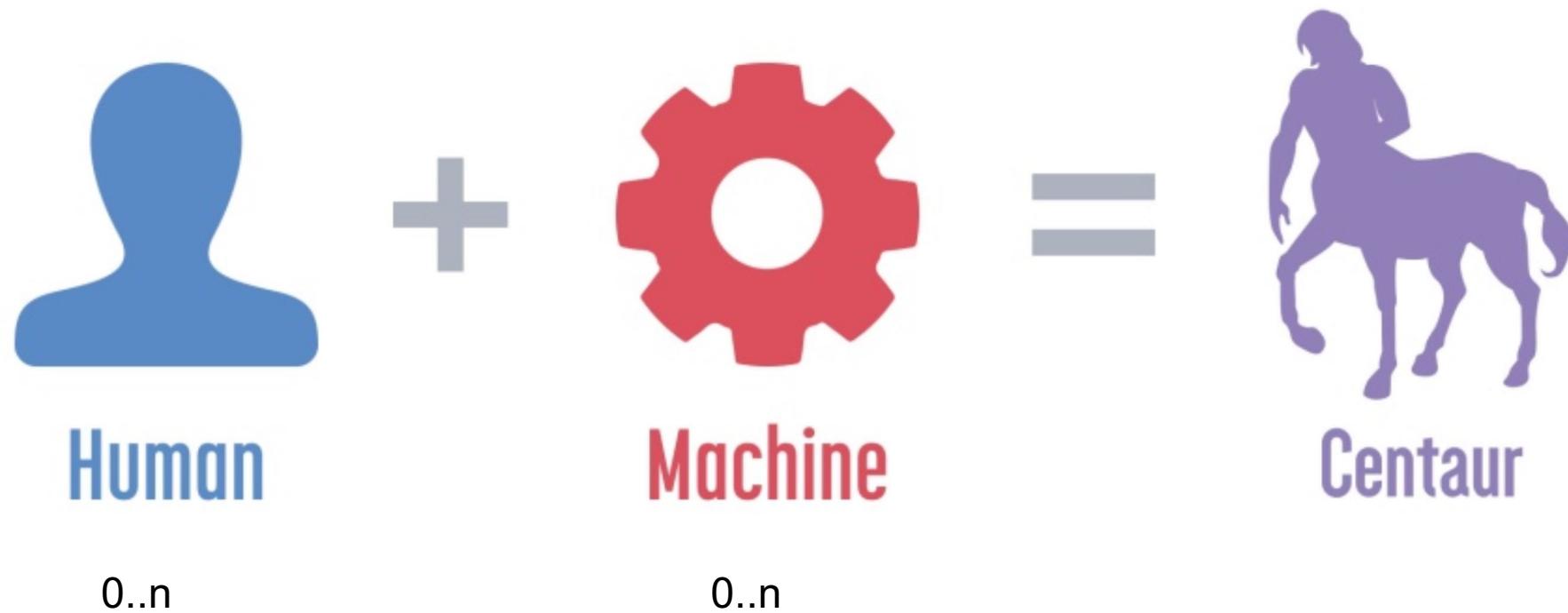
- *How can Humans and AI Form Effective Engineering Teams?*

Chess: Human vs AI



Freestyle / “Centaur” Chess

Rules: {}



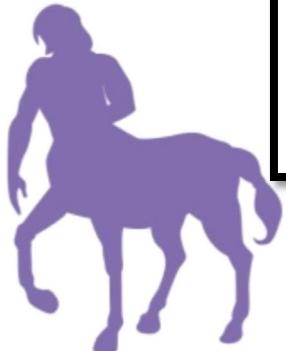
2005 Freestyle Tournament



Dark horse ZackS wins Freestyle Chess Tournament

 [I like it!](#) | [0 Comments](#)

6/19/2005 – The computer-assisted PAL/CSS Freestyle Chess Tournament, staged on Playchess.com, ended with a shock win by two amateurs: Steven Cramton, 1685 USCF and Zackary Stephen, 1398 USCF, using three computers for analysis, defeated teams of strong grandmasters all the way to victory in the finals. We bring you a first flash report with games and results.



Centaur



Centaur



Centaur



Centaur

GM + 

GM + 

GM + 

“ZackS”



Q: What makes for a great freestyle operator?

KASPAROV: Someone who can work out the most effective combination, bringing together human and machine skills. I reached the formulation that a weak human player plus machine plus a better process is superior, not only to a very powerful machine, but most remarkably, to a strong human player plus machine plus an inferior process.

Kasperov's Law:

Weak Human + Machine + Good Process
Outperforms

Strong Human + Machine + Inferior Process



Augmented Intelligence



What is Augmented Intelligence?

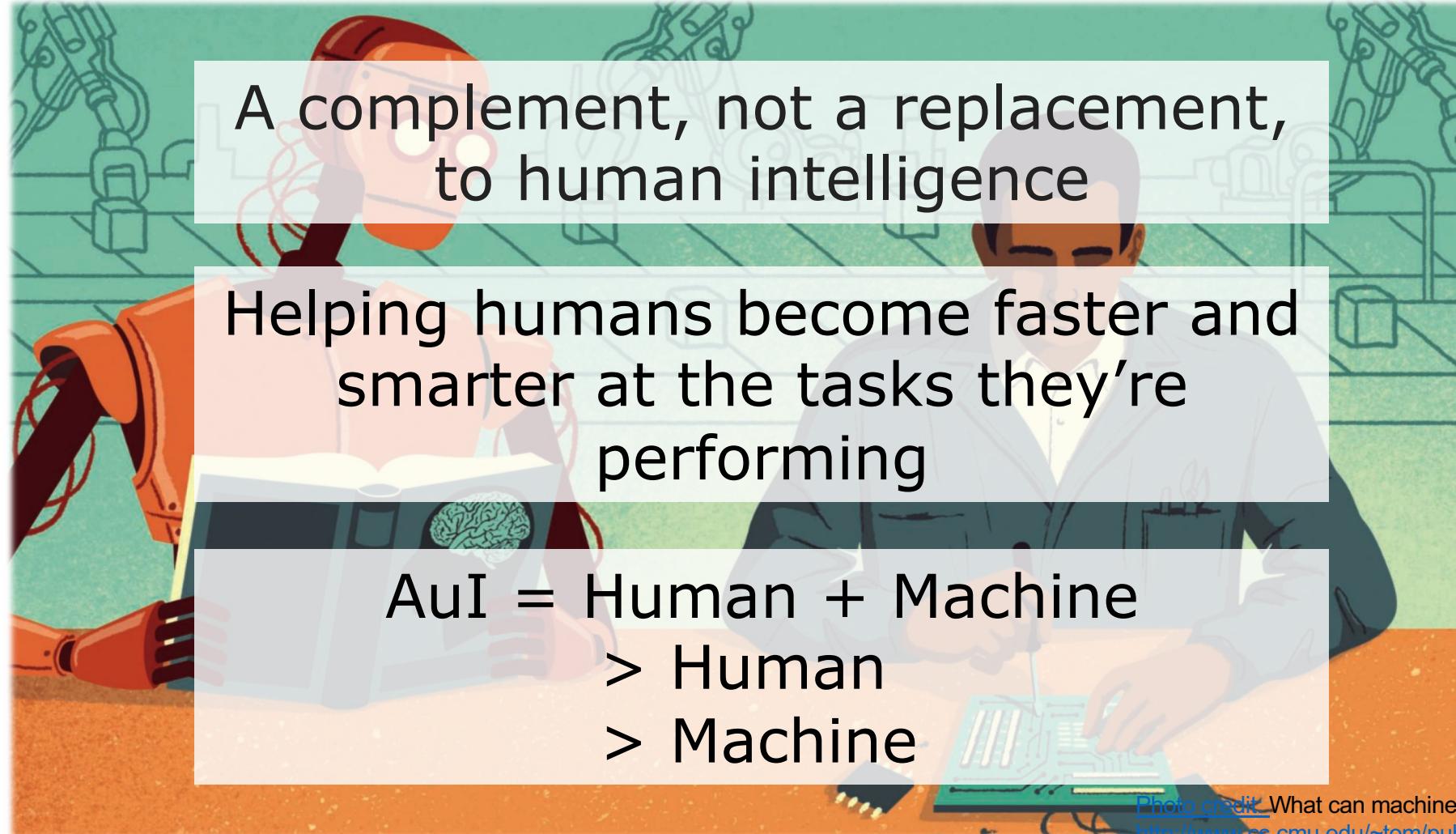


Photo credit: What can machine learning do? Workforce implications
http://www.cs.cmu.edu/~tom/pubs/Science_WorkforceDec2017.pdf



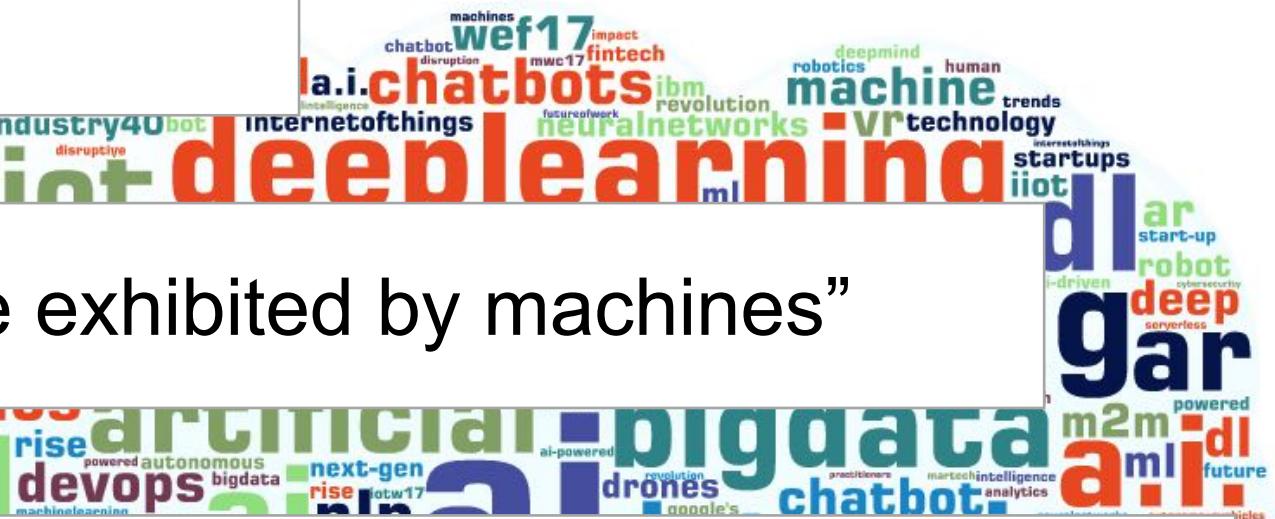
What is Artificial Intelligence?



Using computers to solve problems

“Human intelligence exhibited by machines”

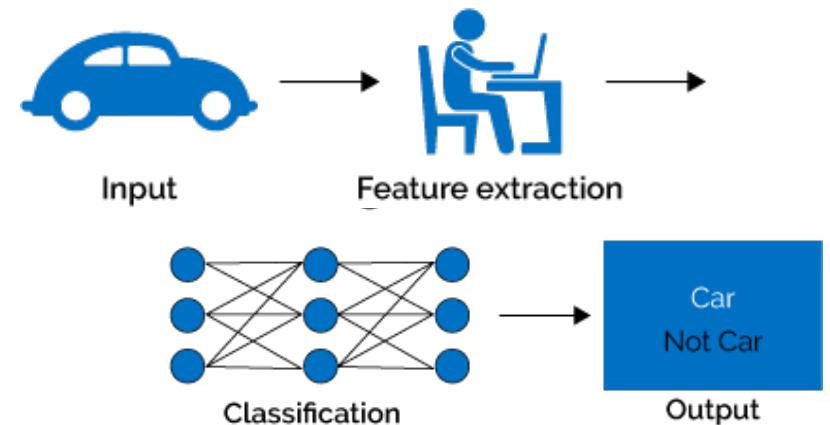
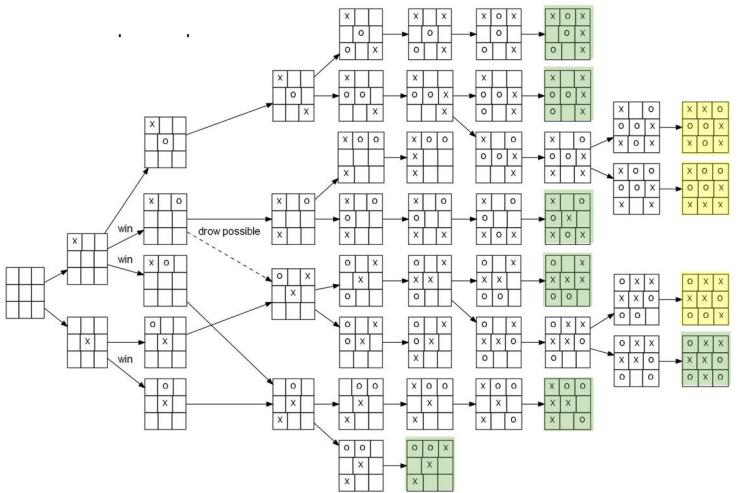
“Any program can be considered AI if it does something that we would normally think of as intelligent in humans”



Two Categories of AI

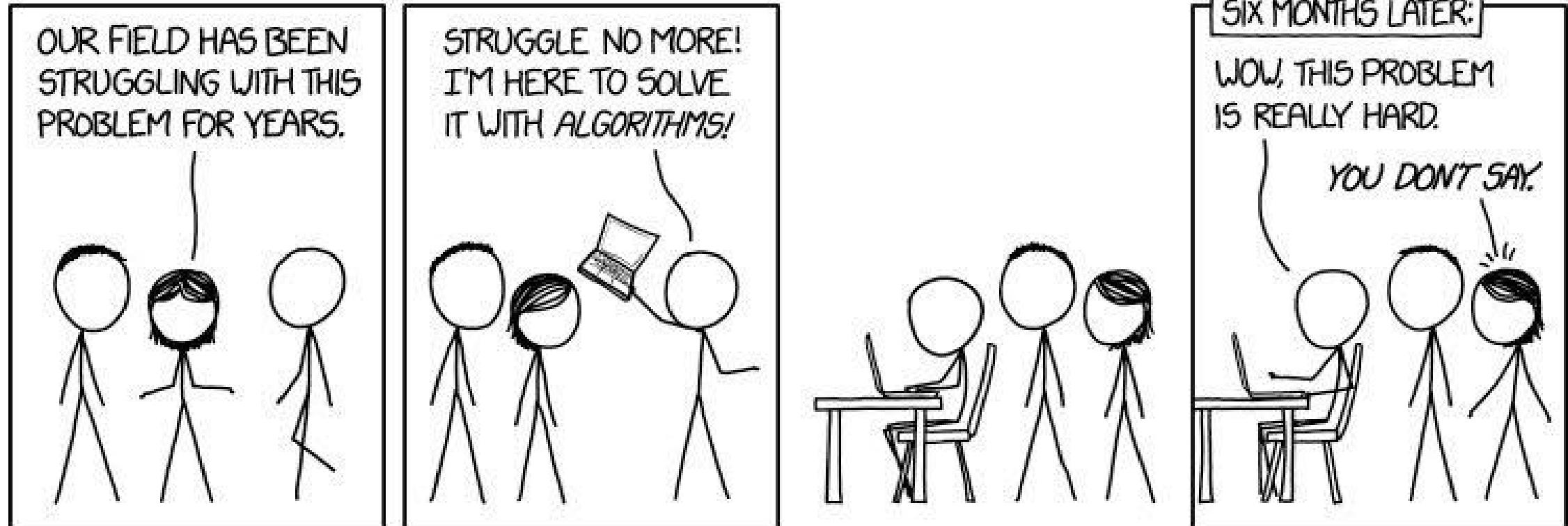


Categories	Example Techniques
<p>“Rules Based” Explicit Knowledge</p>  	<ul style="list-style-type: none">Expert SystemsNLPGenetic AlgorithmsOptimization Solvers
<p>“Patterns Based” Tacit Knowledge</p>  	<ul style="list-style-type: none">Machine LearningNeural networksDeep Learning





Augmented Systems Engineering

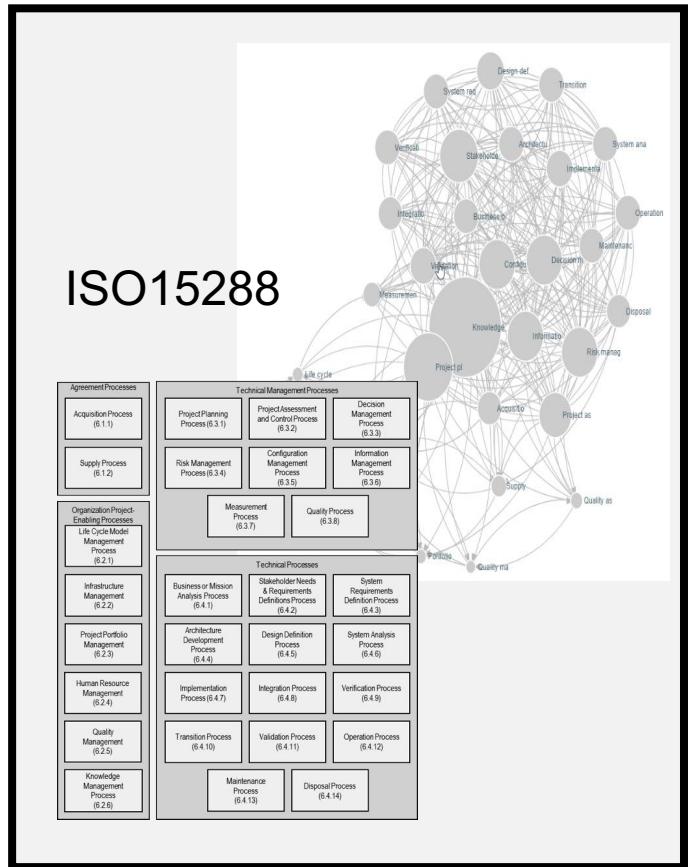


XKCD: Here to Help, <https://xkcd.com/1831/>

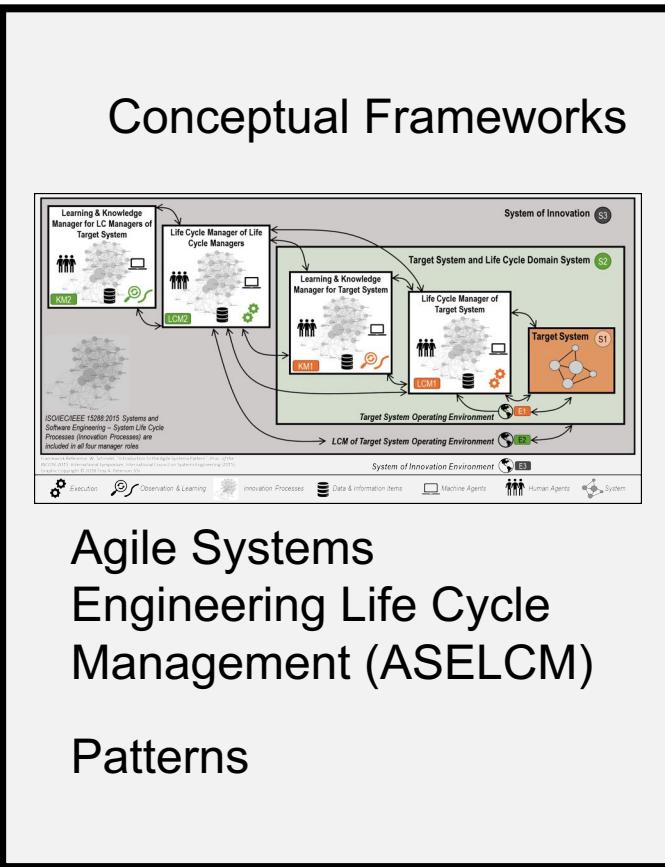
Piecing Together Augmented Intelligence in SE



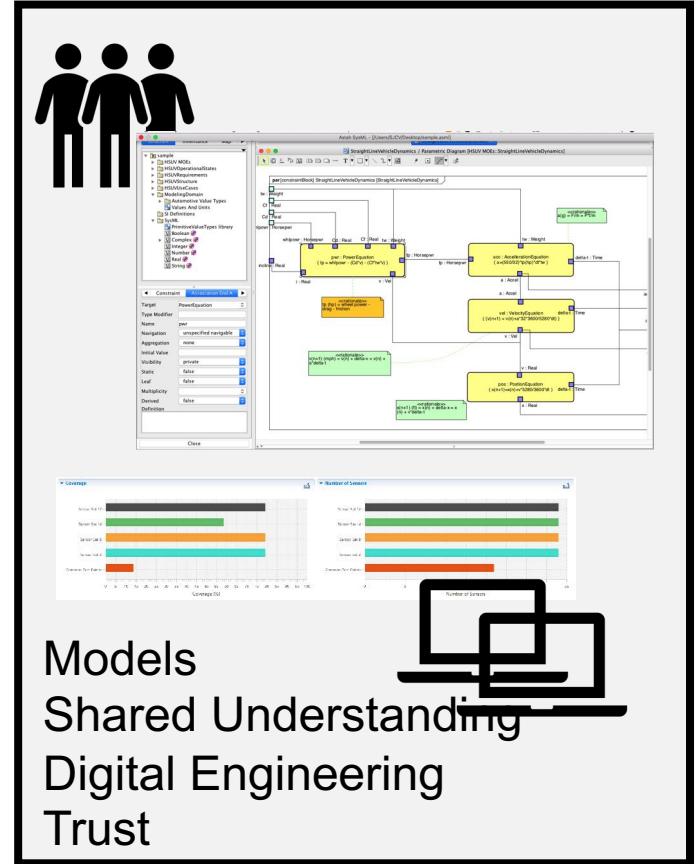
Processes



Framework

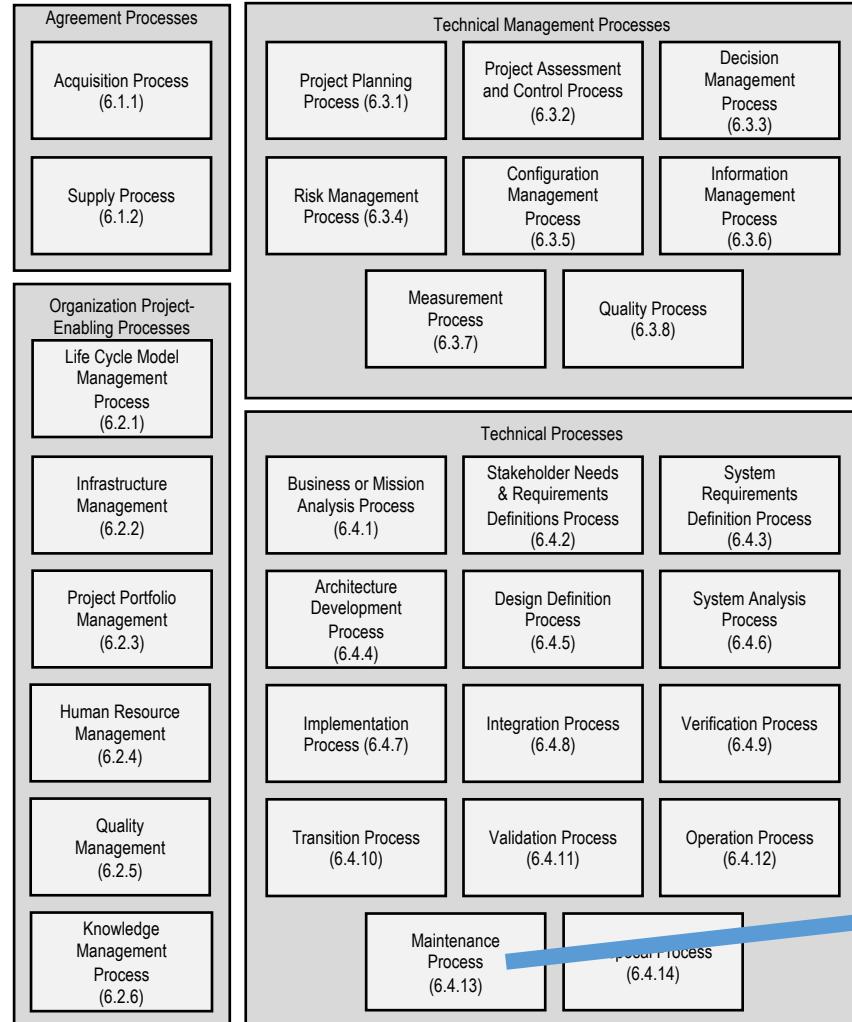


Knowledge & Decision Making

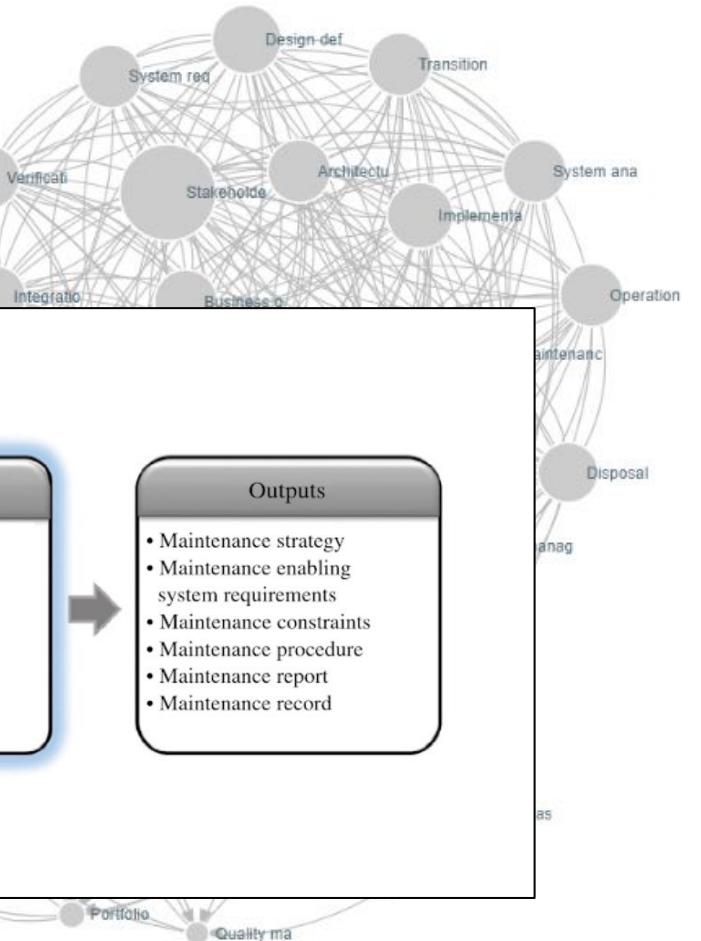


How can Humans and AI Form Effective Engineering Teams?

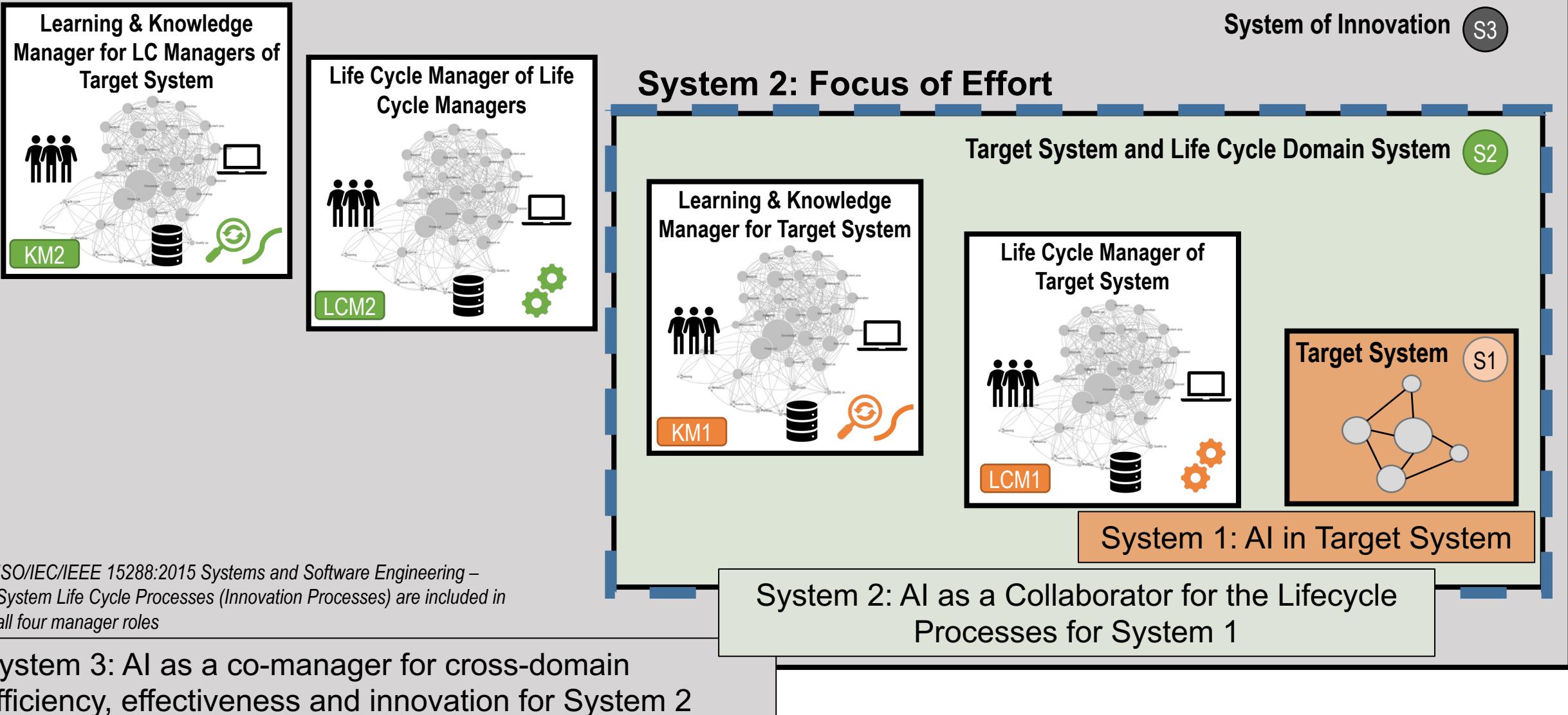
Breadth of Systems Engineering – ISO 15288 Process Areas



- Each process is defined by a purpose, outcomes, and activities
- Comprises 25 processes which have 123 outcomes derived from 403 activities



Conceptual Framework for Human/Machine Teamwork



Execution



Observation & Learning



Innovation Processes



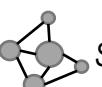
Data & Information Items



Machine Agents

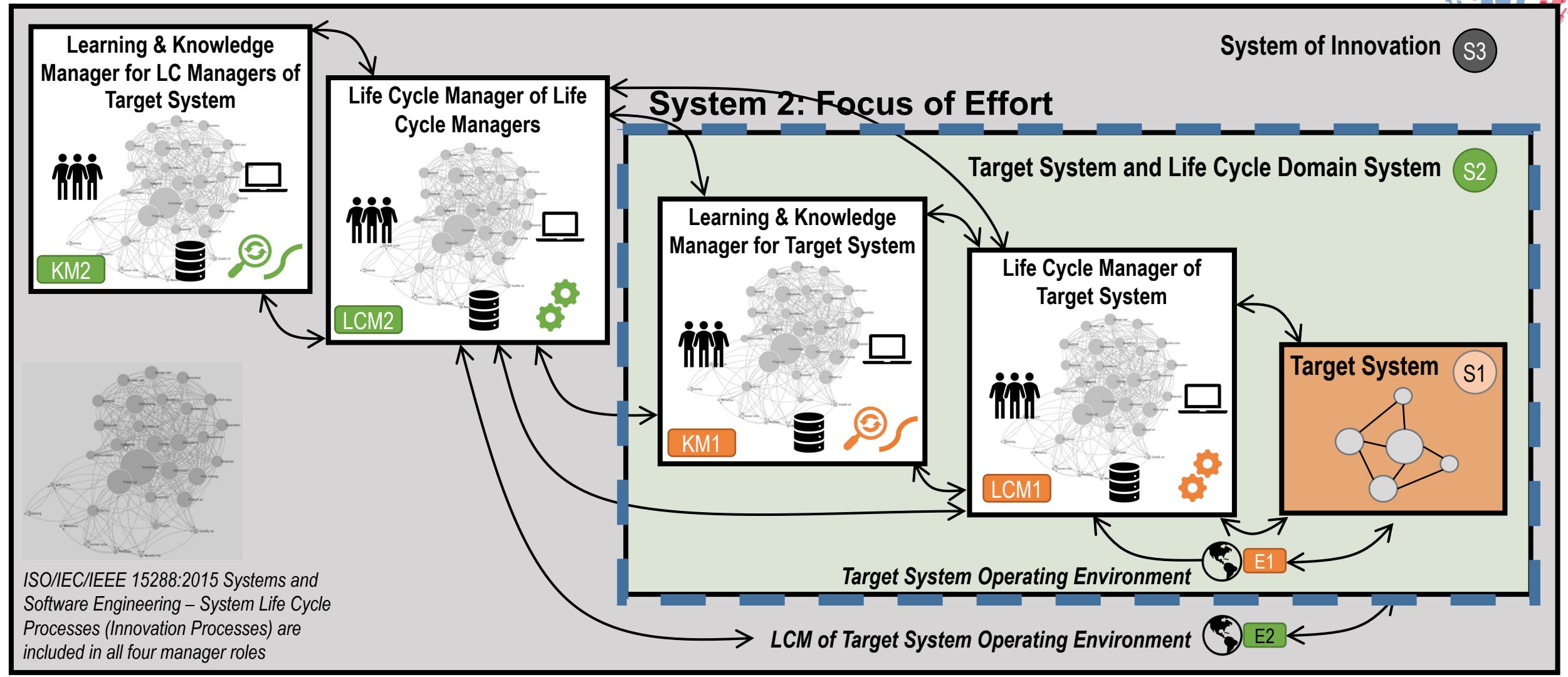


Human Agents



System

Conceptual Framework for Human/Machine Teamwork



Execution



Observation & Learning



Innovation Processes



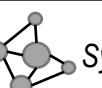
Data & Information Items



Machine Agents



Human Agents



System

Models for Knowledge and Decision Making

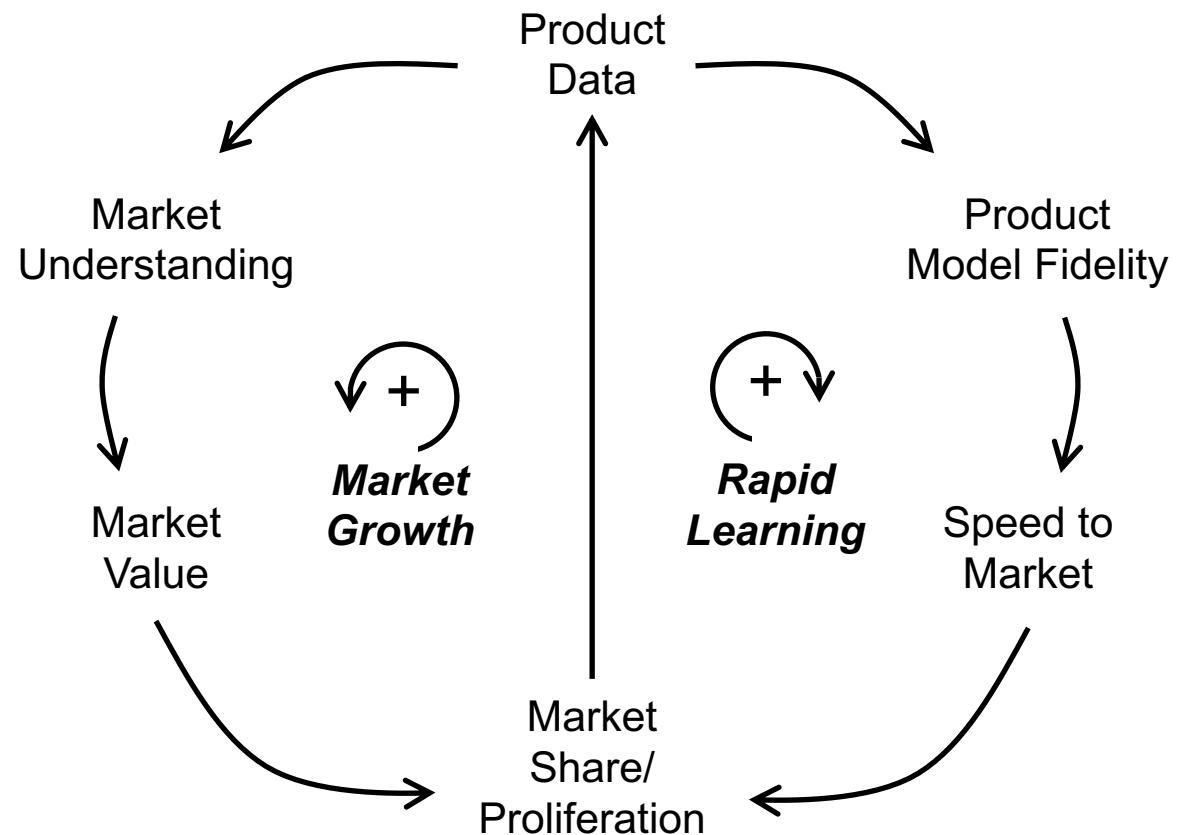


Both AI and MBSE rely on Models

Model based products get better, allowing them to collect more data, which allows them to build better models, making their products better...



What happens when our models are right?



What happens when our models are wrong?

Better Decisions with Human Centered Algorithms



Agency

Reflect the information, goals, and constraints that the decision-maker tends to weigh when arriving at a decision

Perspective

Analyze from a position of domain and institutional knowledge, and an understanding of the process that generated it

Relevancy

Anticipate the realities of the environment in which it is to be used

Objectivity

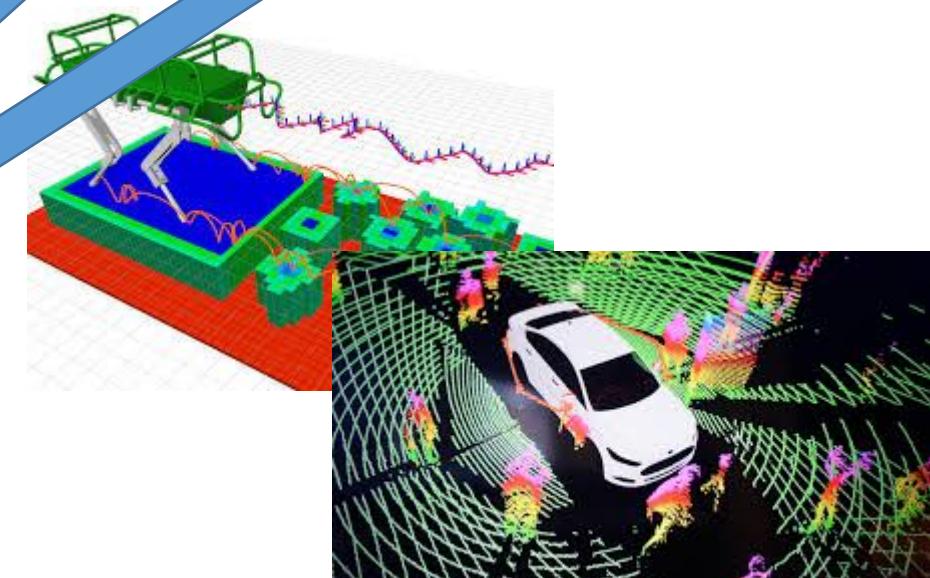
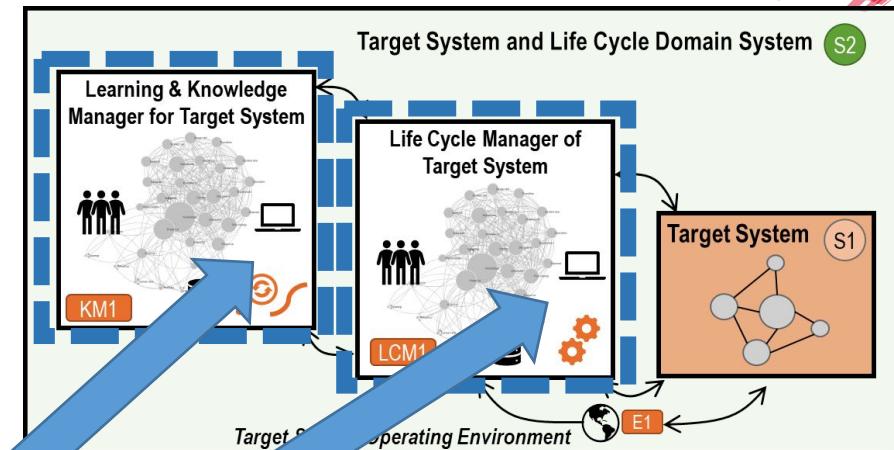
Avoid biased predictors

Transparency

Be transparent, peer-reviewed or audited to ensure that unwanted biases have not inadvertently crept in

Candor

Effectively present measures of confidence and “why” messages (ideally expressed in intuitive language) explaining why a certain algorithmic indication is what it is

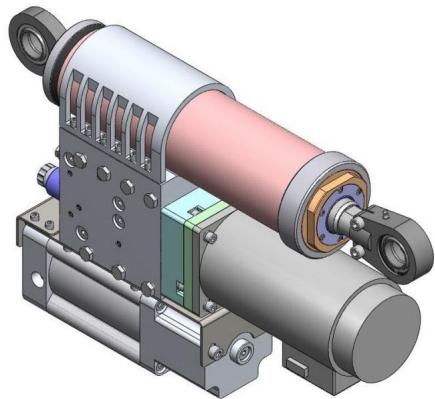
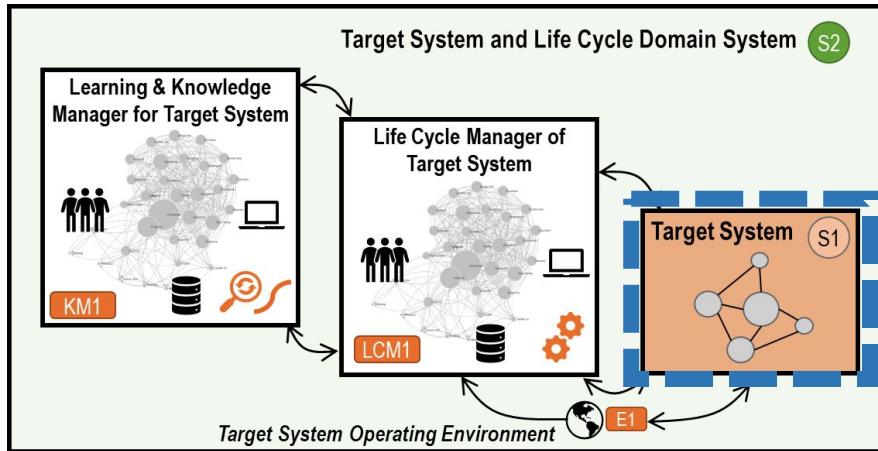


Why artificial intelligence needs human-centered design - Deloitte Review, issue 22
© 2017 SSI - UNCLASSIFIED

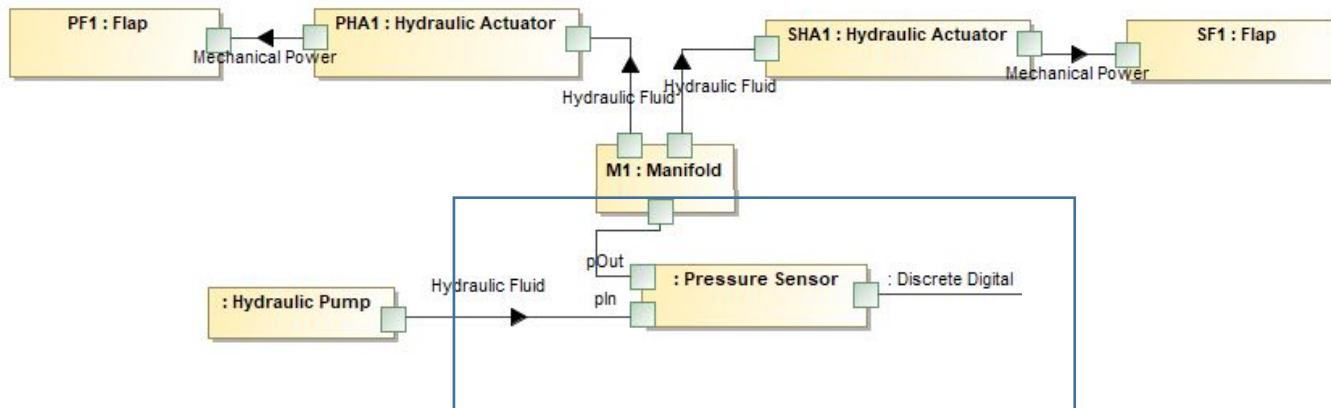
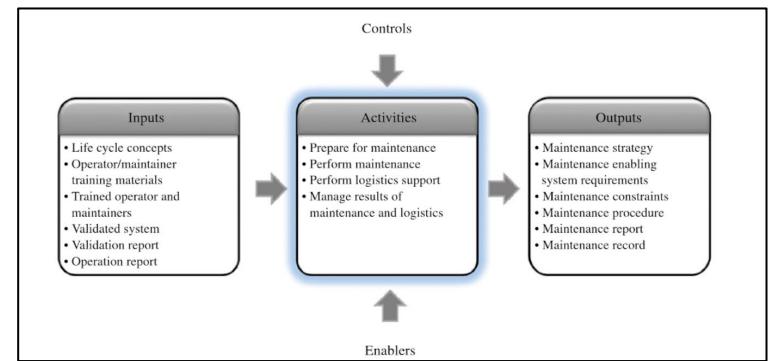


Case Studies of Augmented Intelligence

Case Study: Design for CBM



ISO15288/6.4.13

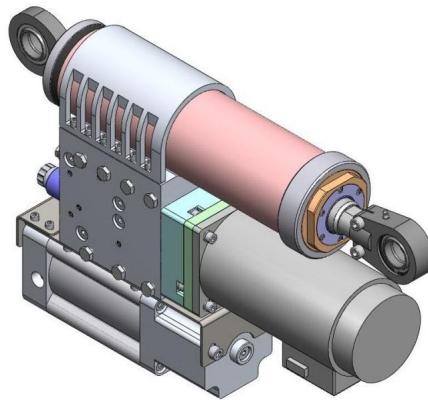
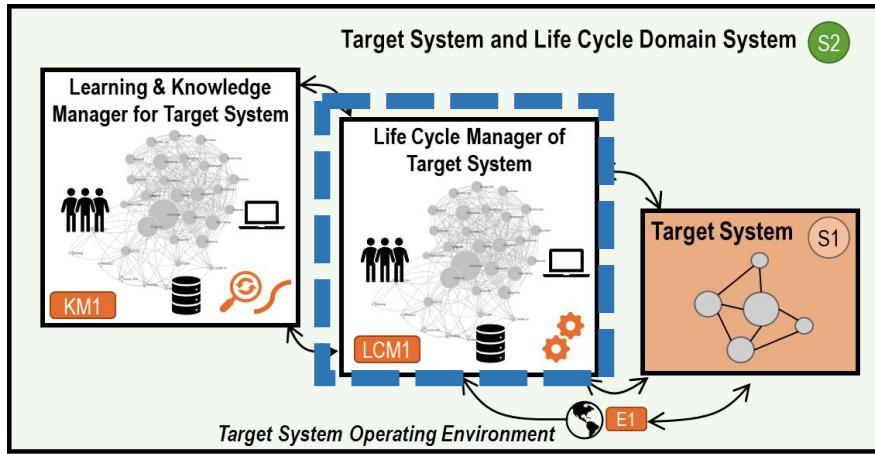


Performance requirement:
Achieve > 90% fault isolation

Constraints:
Sensors can be placed on
hydraulic or mechanical
connectors

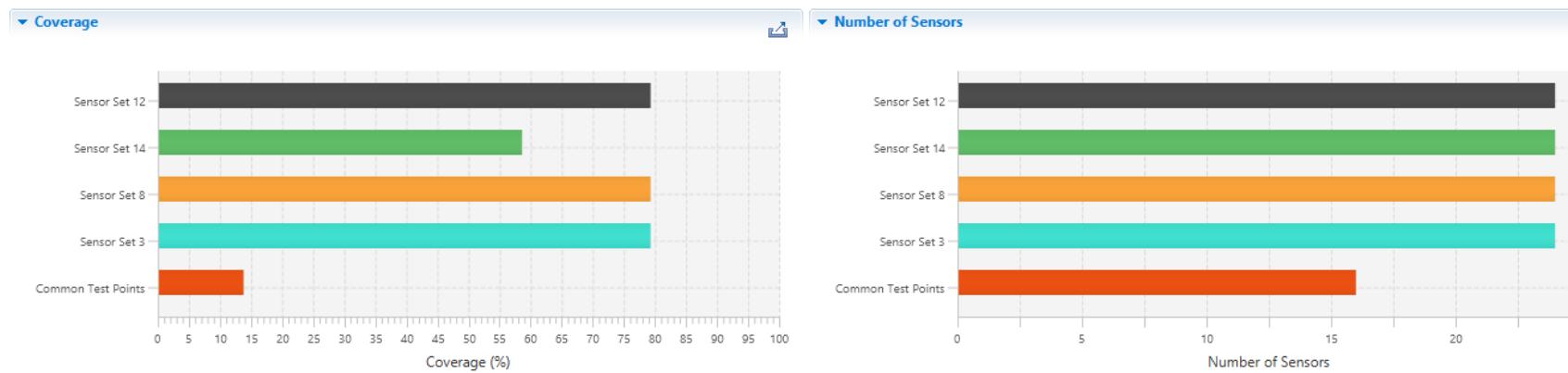
Initial Conditions: 14% fault isolation

Case Study: Design for CBM



Fault Isolation

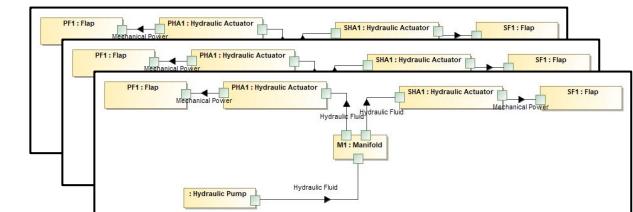
User: compare sensor sets parameters (coverage, size, weight, cost)



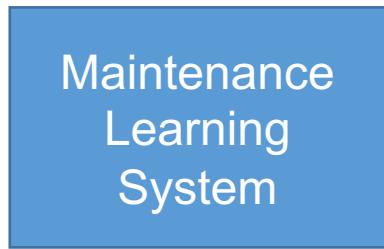
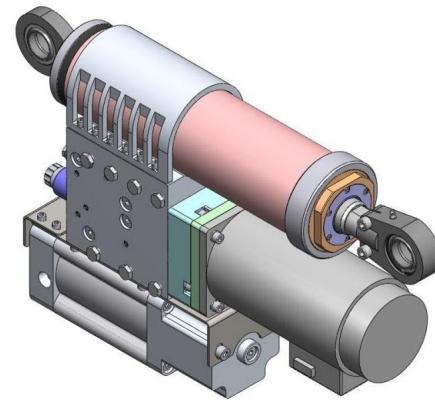
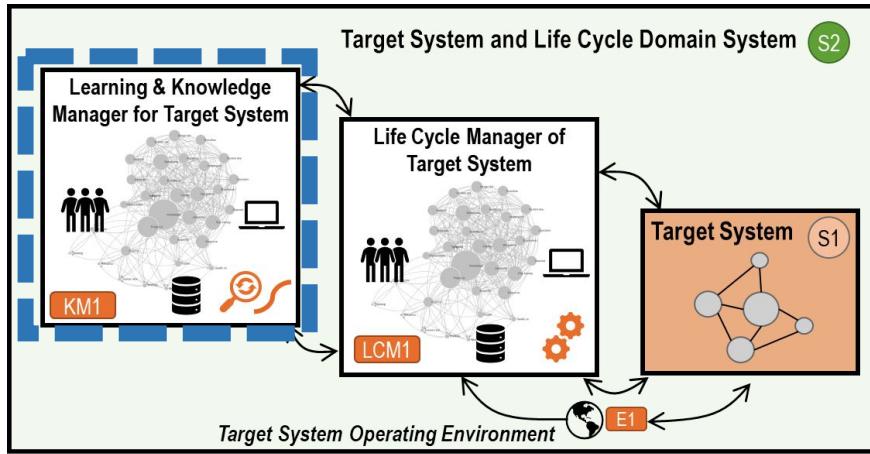
Method: Genetic Algorithm

Apply
User: compare sensor sets parameters (coverage, size, weight, cost, **reliability**) → isolation %

Apply Maintenance Design

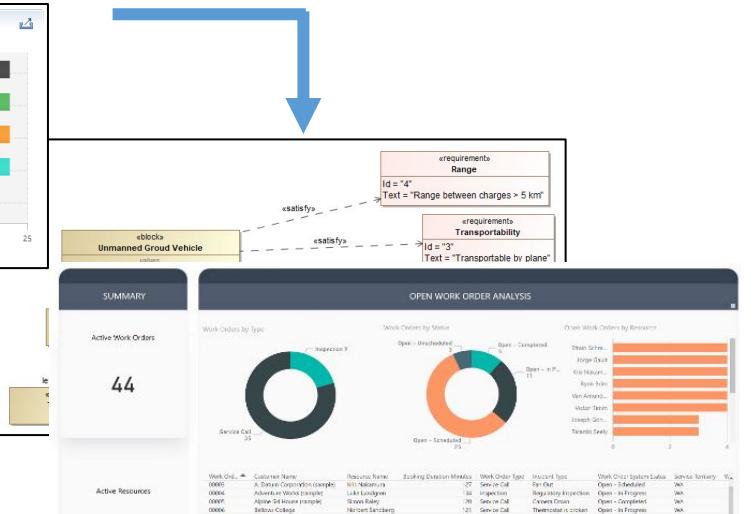


Case Study: Design for CBM



Learn: Sensor count → Reliability

Observe: Increased failure modes from sensors



Service Reports

How Can SE Form Effective Augmented Teams?



- Effective augmented teams will outperform non-augmented teams
- “Go apply the algorithm” is insufficient, even for simple problems
- Experience in other domains indicate systems engineering will change:

Less

- Direct modeling
- Qualitative assessments
- Disconnected data
- Point-based designs

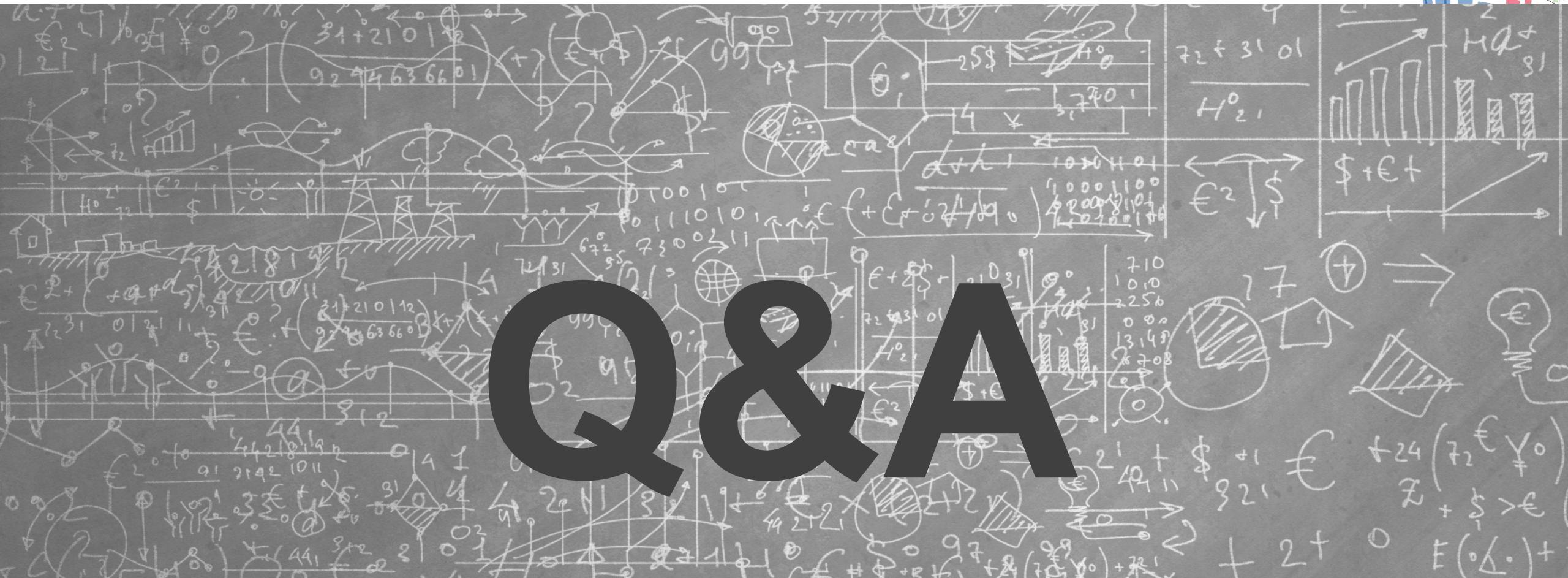
More

- Value of shared understanding
- Validated models & formalized trust
- Problem space & constraint definition
- Encoded domain knowledge
- Use/reuse of patterns
- Visualizations of state space
- Set-based designs
- ...



Q&A

“Many jobs will continue to be lost to intelligent automation. But if you’re looking for a field that will be booming for many years, get into human-machine collaboration.” -Kasperov, Deep Thinking



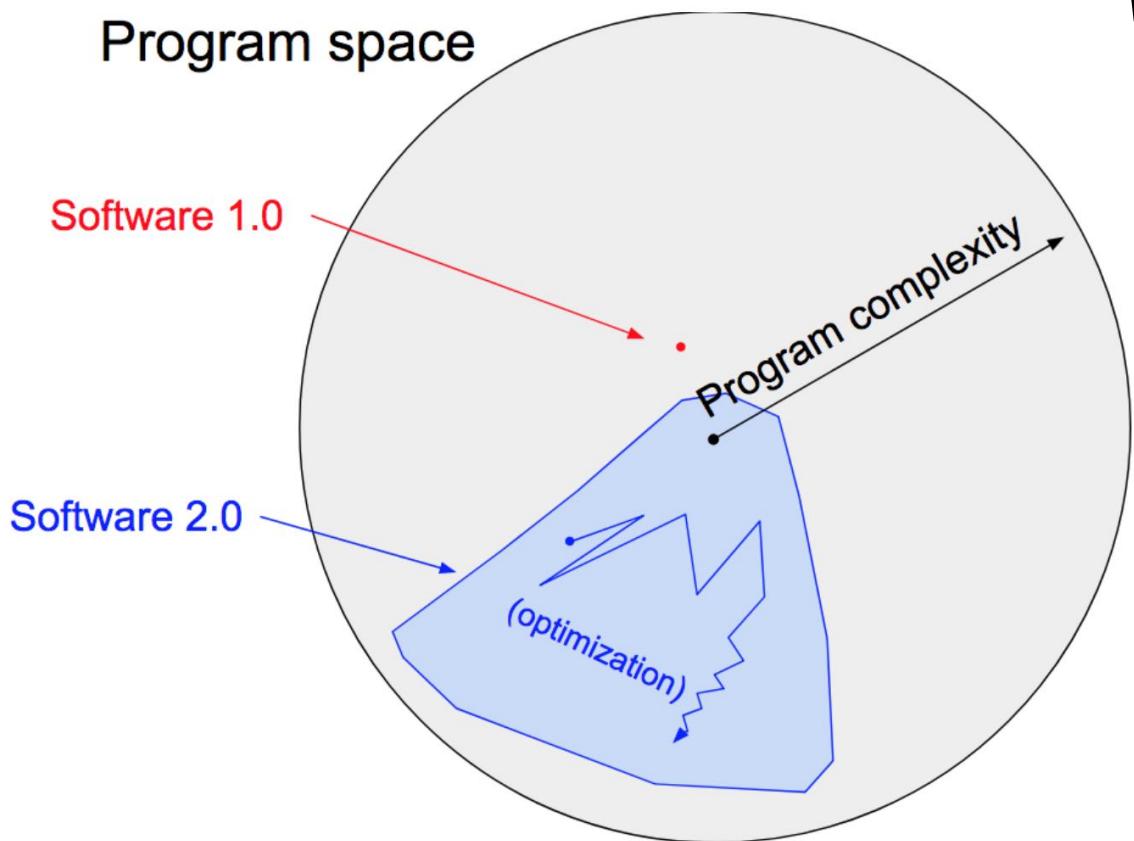


MARK PETROTTA
SYSTEM STRATEGY, INC.
PRINCIPAL
mpetrotta@systemxi.com

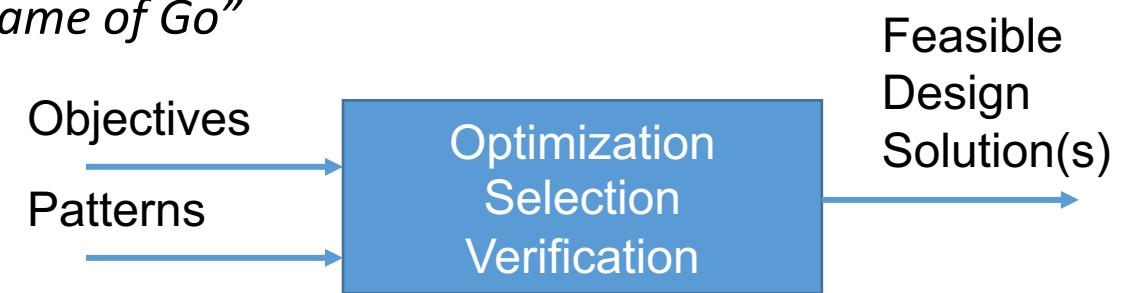
Augmented Intelligence in Software



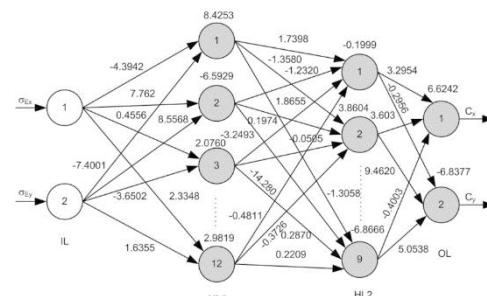
Software 2.0: Identify subset of program space to search, and use the computational resources at our disposal to search this space for a program that works.



"win a game of Go"



Neural net architecture

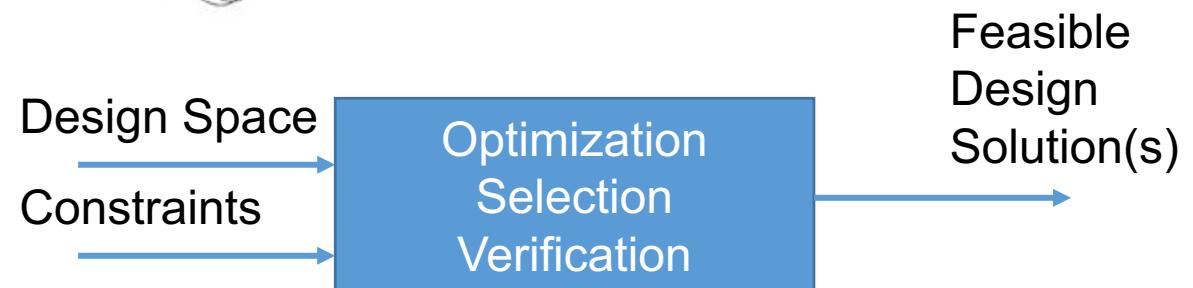
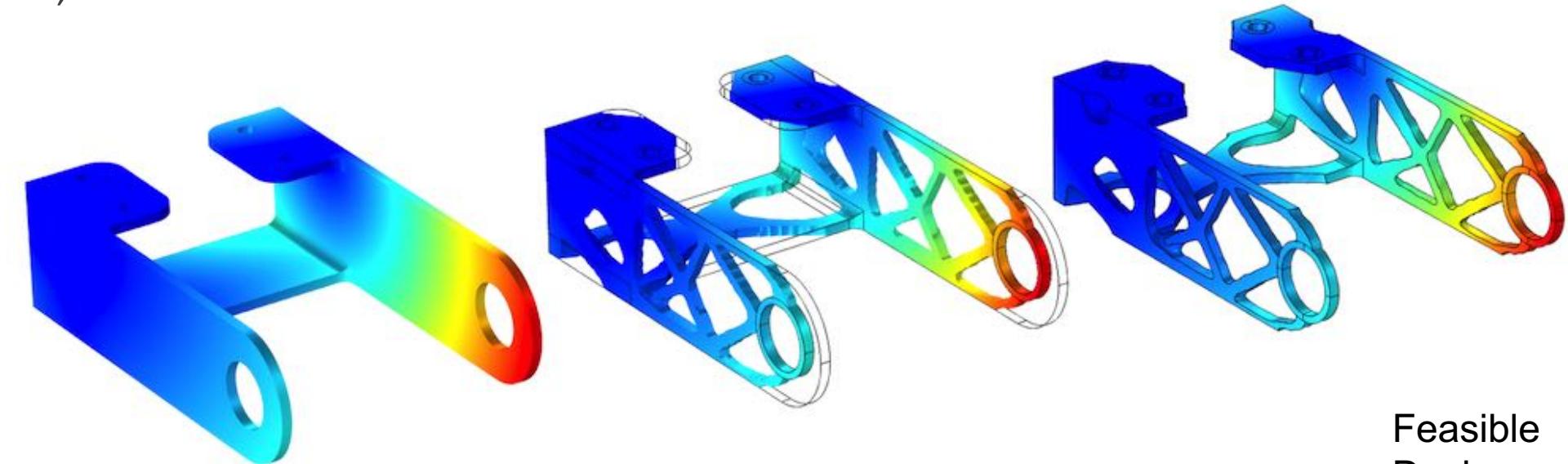


Software 2.0, Andrew Karpathy
<https://medium.com/@karpathy/software-2-0-a64152b37c35>

Augmented Intelligence in Material Science



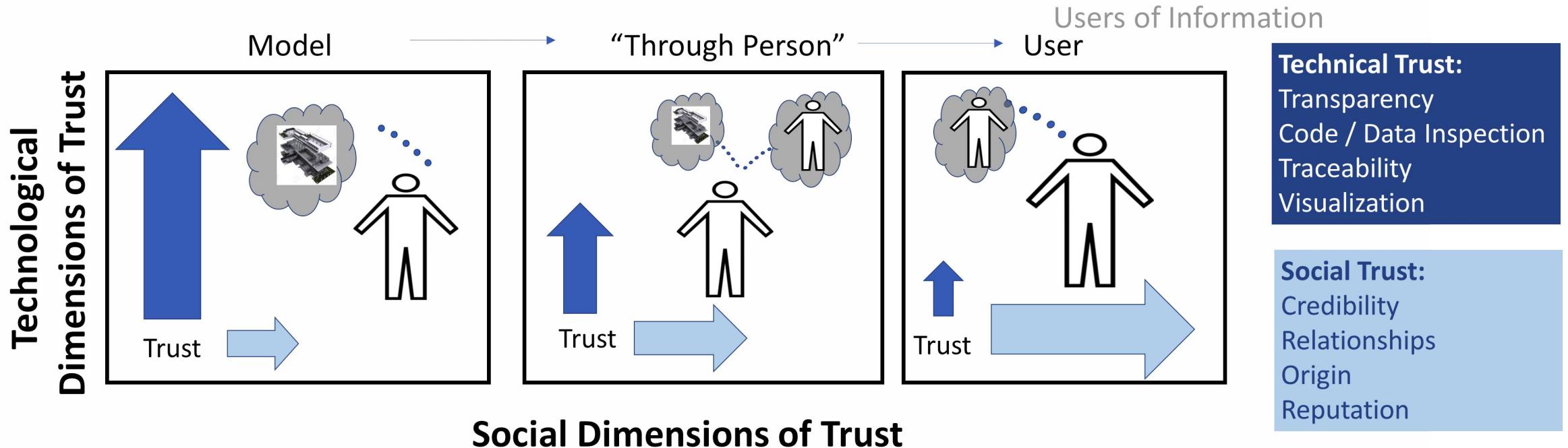
A bracket geometry is topology optimized, leaving only 50% of the material, which contributes the most to the stiffness.



<https://www.comsol.com/blogs/performing-topology-optimization-with-the-density-method/>

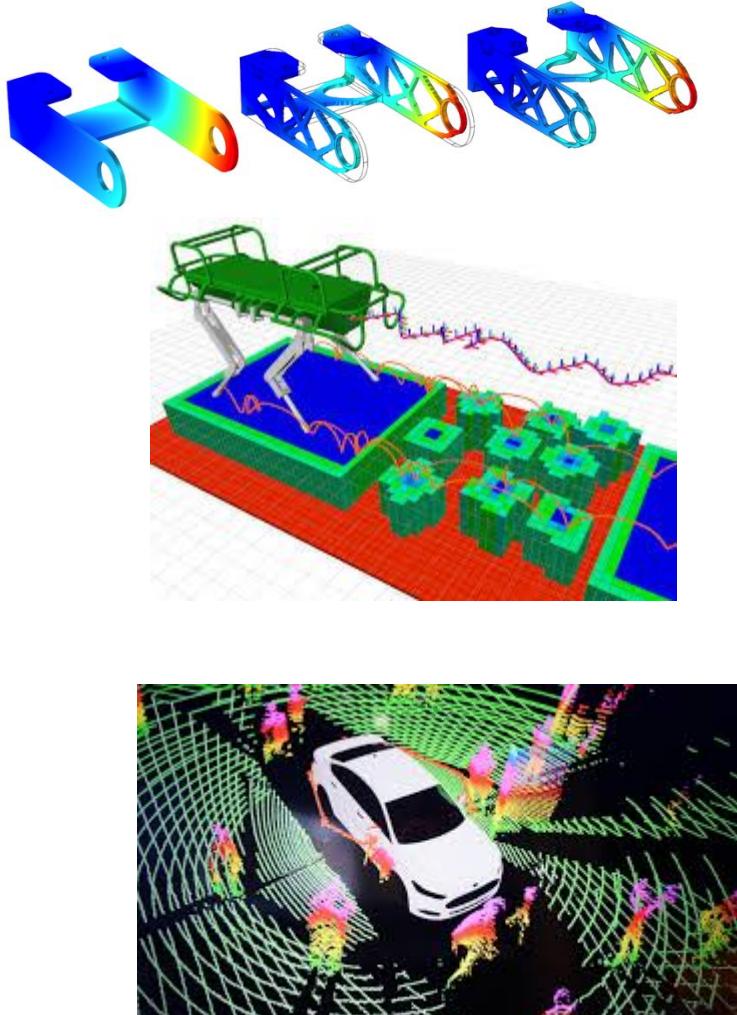
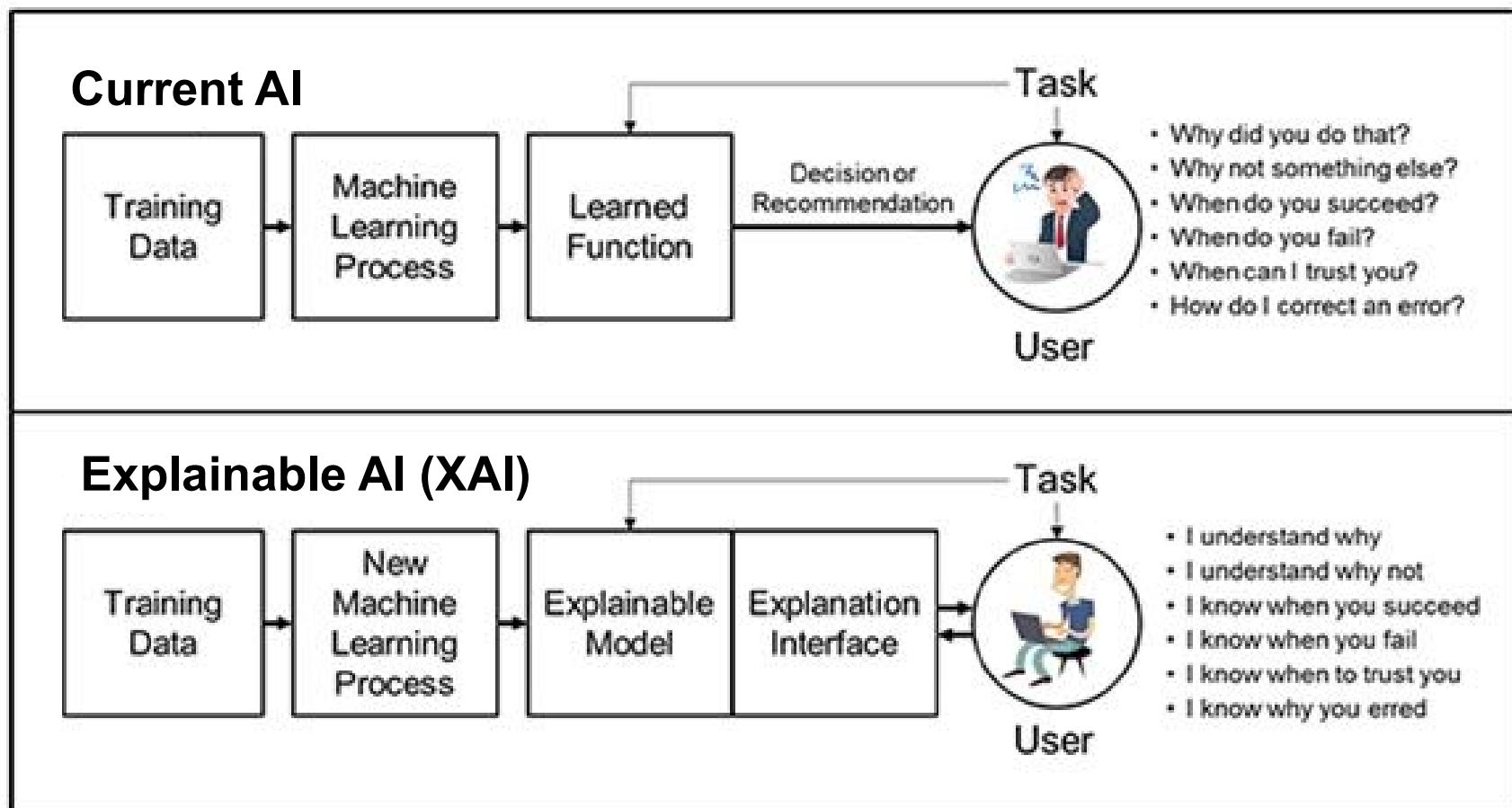


Through Person Trust Model



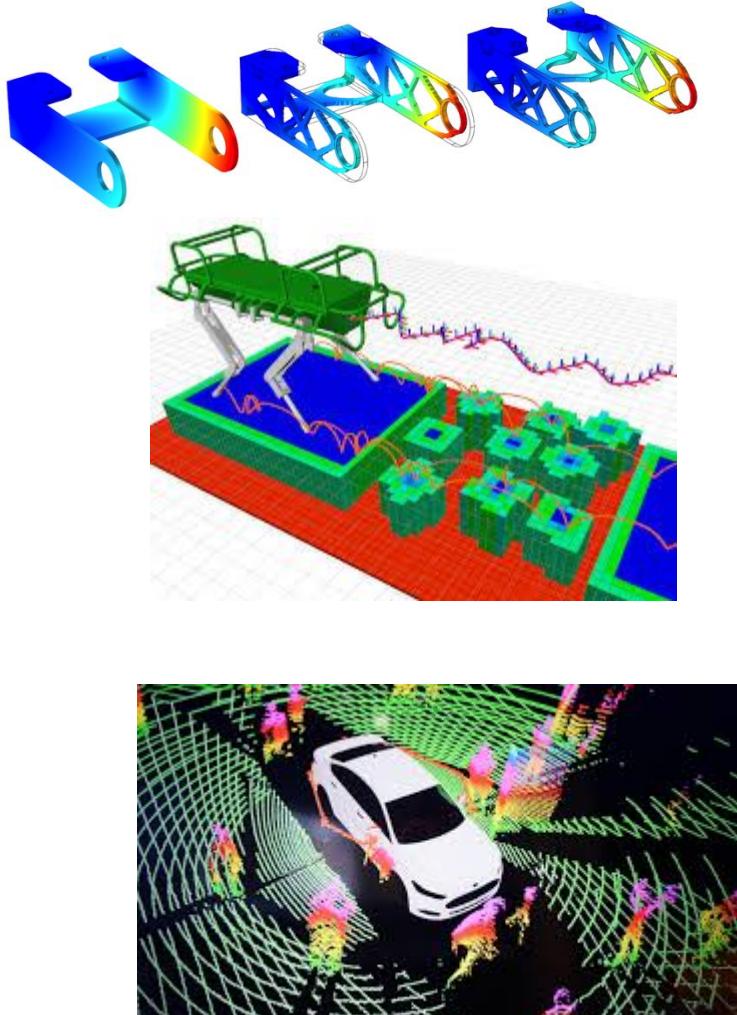
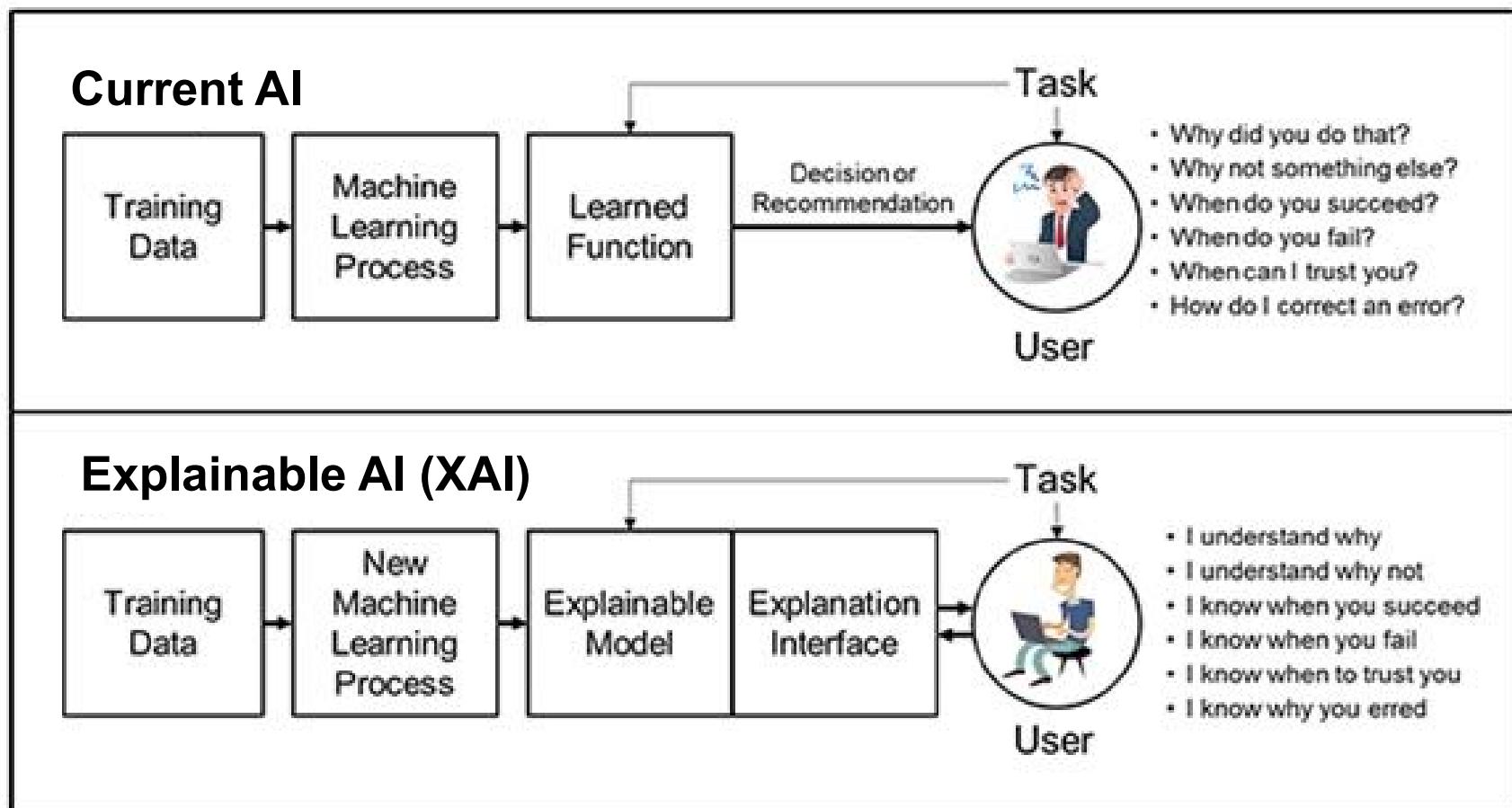
Model-Centric Decision Making: Insights from an Expert Interview Study,
Donna H. Rhodes, E. Shane German, Massachusetts Institute Of Technology

Transparent and Explainable Models



<https://www.darpa.mil/program/explainable-artificial-intelligence>

Transparent and Explainable Models



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