

Symbiotic Design for Cyber Physical Systems

Sandeep Neema, I2O

DESTION, CPS Week

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Innovation in Design – Creativity



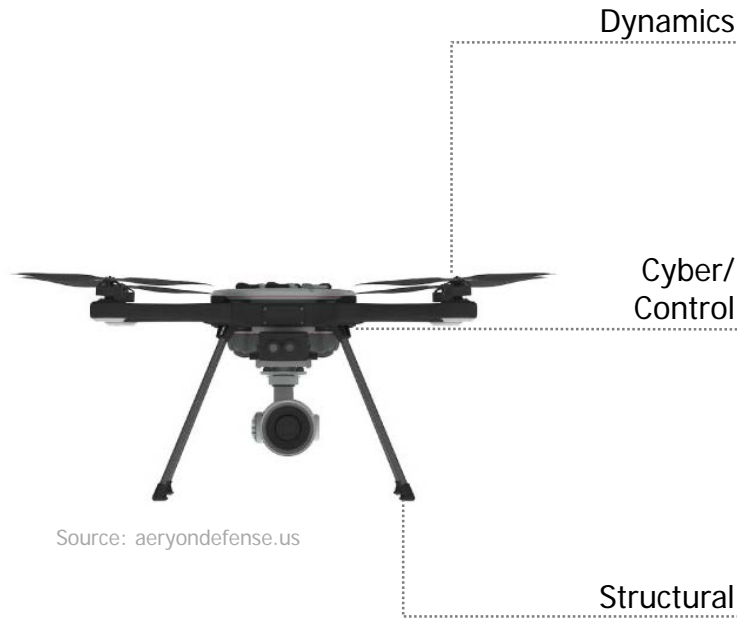
Source: Netflix, AlphaGo Movie

Can AI creatively solve the kinds of problems that innovation requires?



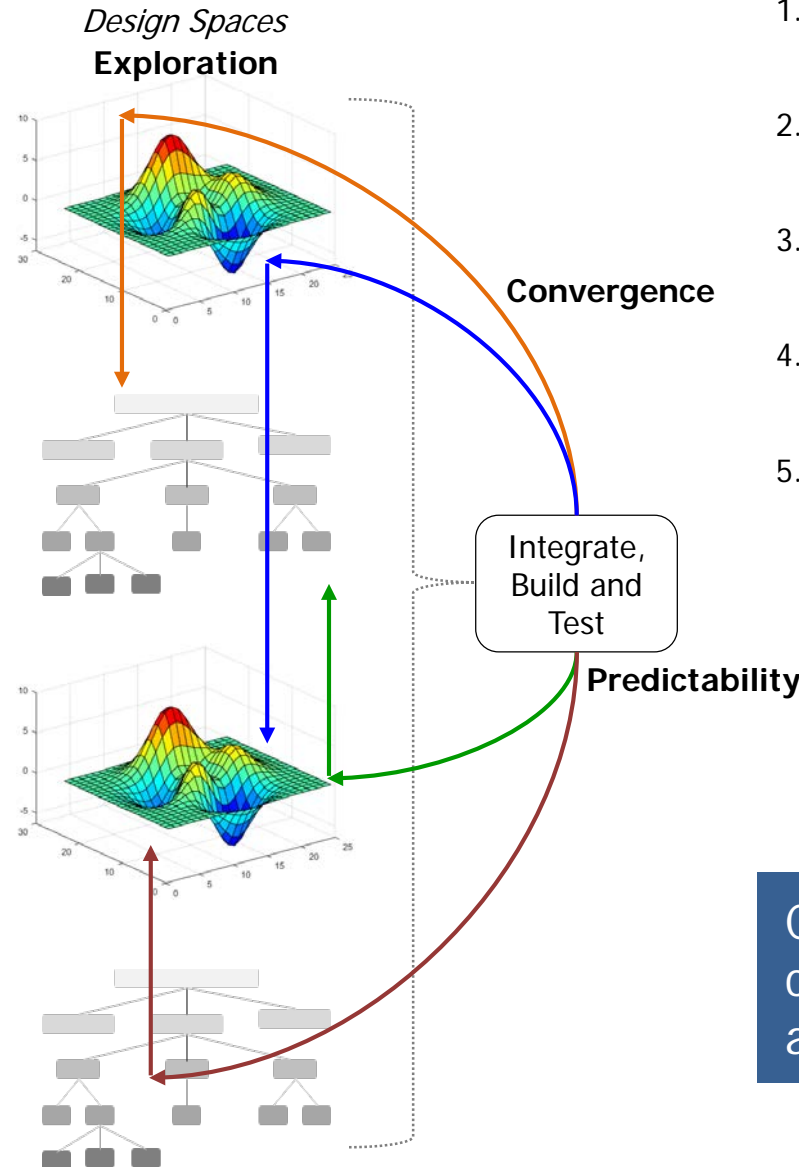
How is design done today?

Example: military-grade small unmanned aircraft system (R80D SkyRaider)



Source: aeryondefense.us

Manufacturing



1. Goal: build a UAV with mission-specific performance objectives (e.g., range, endurance, speed, payload)
2. Decompose: the design problem into sub-problems (i.e., domains) and recruit a team to address each sub-problem
3. Design: Each team produces design artifacts from domain-specific design spaces
4. Integrate: them into an overall system-level design and build a prototype
5. Integration uncovers numerous issues triggering change
 - a) Weight of system plus payload exceeds the lift
 - b) Power requirements for propulsion and payload exceeds power available in the battery
 - c) Acoustic coupling of motors induces high dynamic stress on the frame material
 - d) Light weighting requires use of carbon composites which can not be manufactured in-house

CPS design today – constrained by challenges in predictability, convergence, and exploration – is *construct by correction*



Today vs. Future of CPS Design

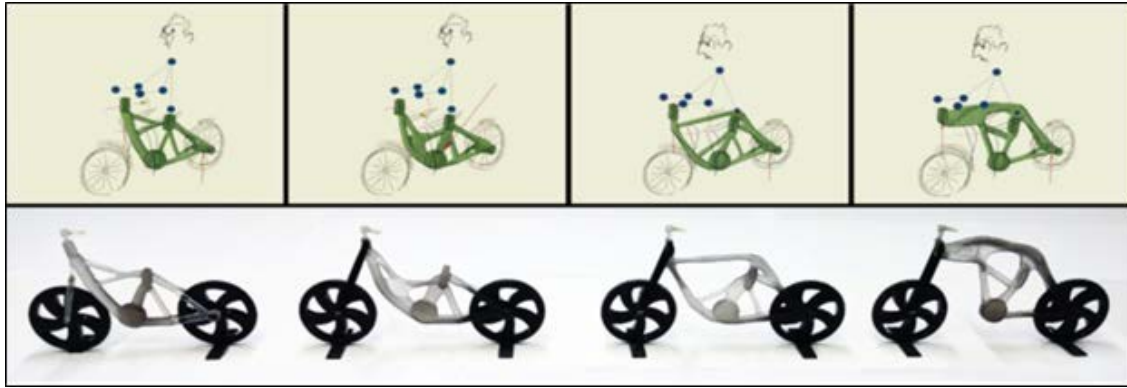
Challenge	Limiting Factors	State of the Practice	State of the Art		Symbiotic Design
		Model-based Design Tools	Commercial: Autodesk Dreamcatcher for CAD	DARPA: META Tools	
Predictability	Model construction (cost and fidelity)	Manual	Manual	Manual	Automated
Convergence	Model composition (intra-domain and cross-domain)	Manual	N/A	Partially automated	Automated
Exploration	Co-design (domains jointly explored)	Single	Single	Multiple	Multiple
	Designs explored	10's (manual)	1,000's (automated)	100's (partially automated)	1,000,000's (automated)
Cognitive Load	Interaction complexity	High	Medium	Medium	Low



State of the Art in CPS Design

Commercial (Autodesk Dreamcatcher for CAD)

Manually specify baseline geometry and loading requirements

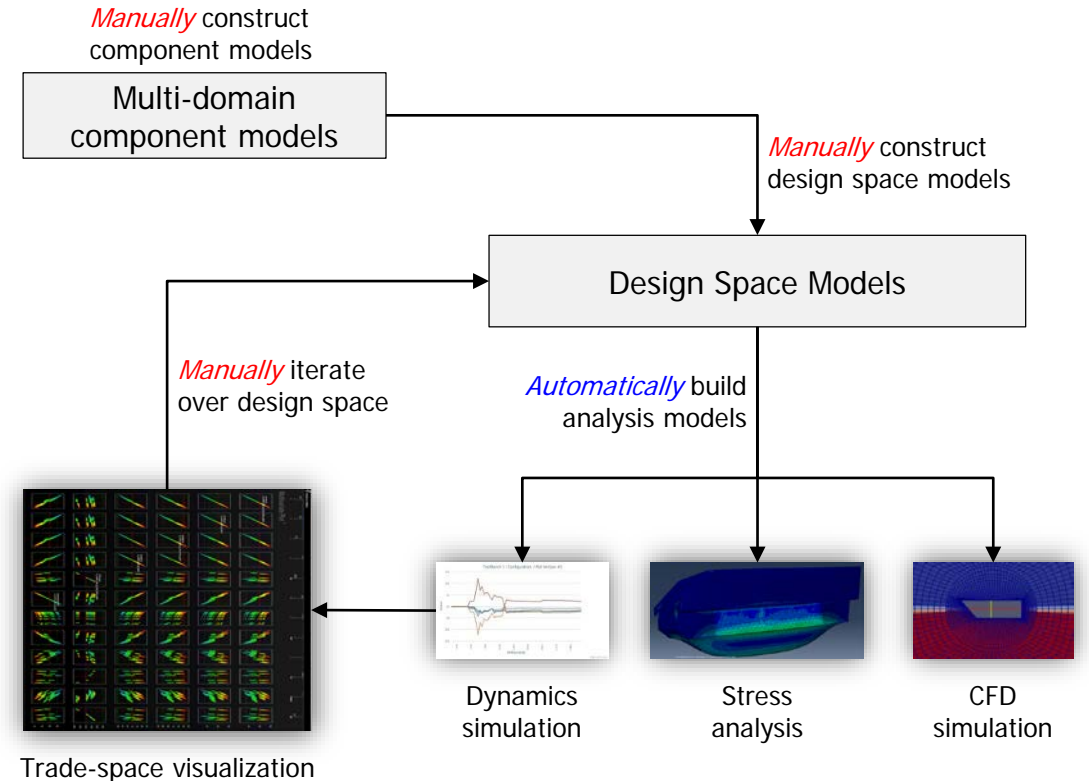


Source: Autodesk Dreamcatcher Project

Automatically evolve many variations of the baseline geometry satisfying loading requirements

Parametric topology optimization
Limited to single domain (geometry)

DARPA (META Tools)



Source: Vanderbilt University

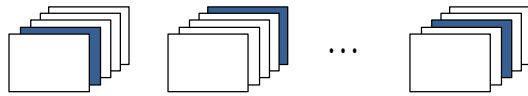
Rule-based model transformations
Manual design space construction and exploration



Challenges and Insights

Design Space Construction

Predictability Challenge: Automate construction of design spaces



New Insights:

- Novel machine learning approaches for mining large code corpora in DARPA MUSE program

Design Composition

Convergence Challenge: Automate composition of partial designs

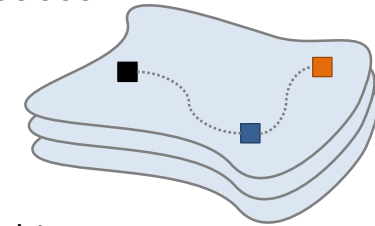


New Insights:

- Deductive model composition toolchains prototyped in DARPA META program
- Early results in combining deductive and inductive synthesis

Design Space Exploration

Exploration Challenge: Automate exploration of high-dimensional design spaces



New Insights:

- Recent results in scalable design-space exploration using learning-based approaches
- AlphaGo inspired approaches for exploration

Symbiotic Exchange



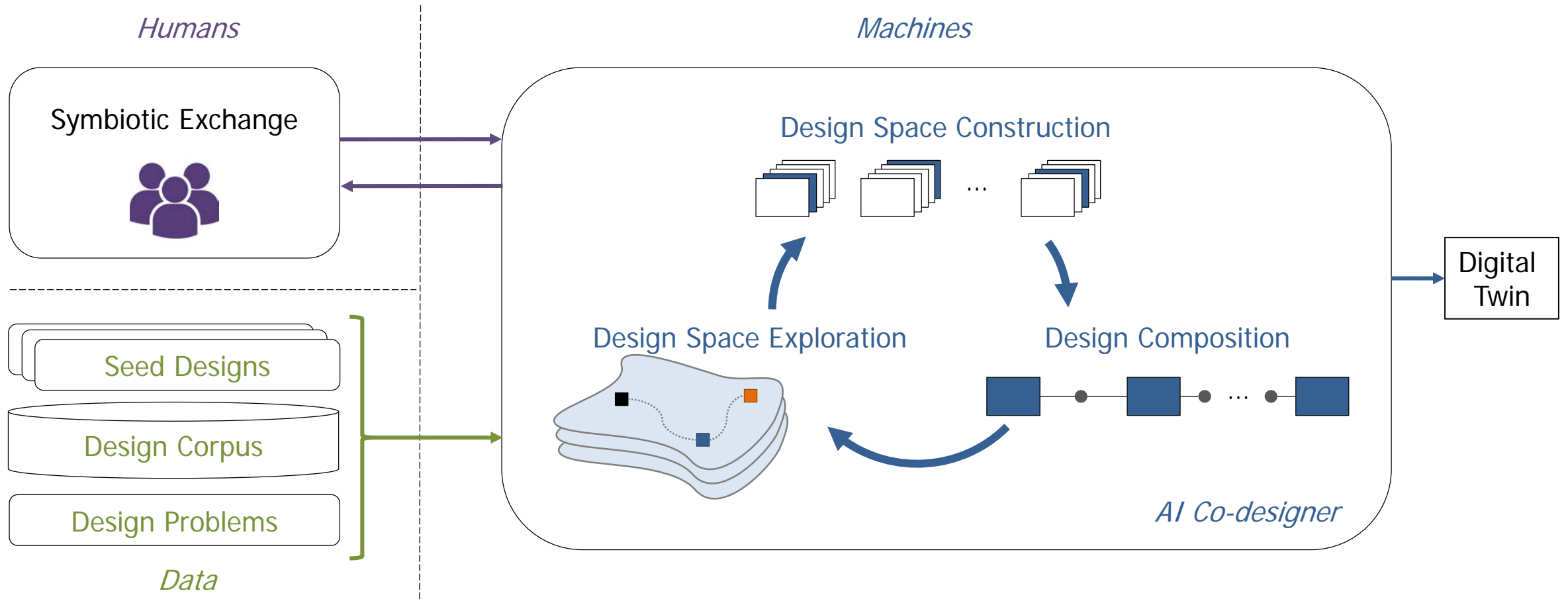
Cognitive Load Challenge: Natural and anticipatory interfaces to enable effective human-machine partnership

New Insights:

- Visualization and interpretation of high-dimensional data
- Cueing reinforcement learning by reward modeling
- Sketch-based interfaces



Symbiotic Design: Achieving correct-by-construction at scale



Develop symbiotic AI-based technologies for correct-by-construction design of military-relevant CPS in order to reduce the time from inception to deployment from years to months, and increase innovation



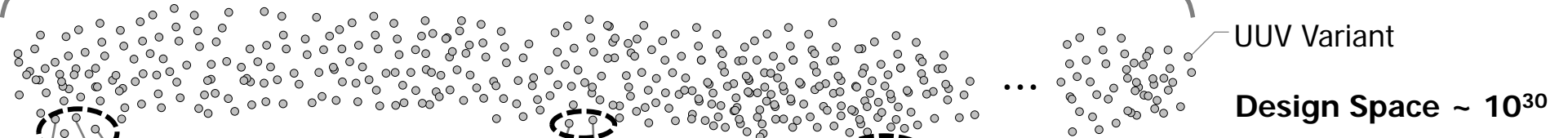
An Example of Symbiotic Design flow

↓ Humans provide
Seed Design



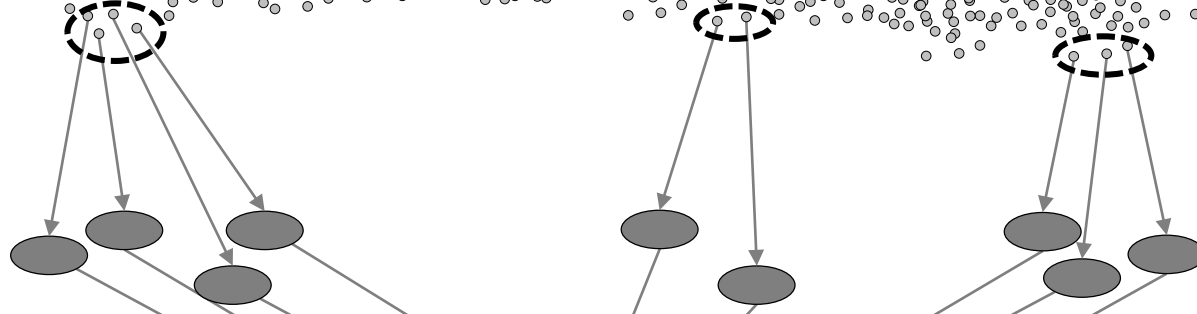
↓ Computer mines design corpora
Humans curate design options

Design Space



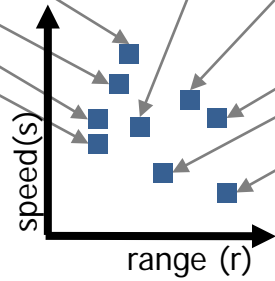
↓ Computer composes domain models
Humans provide rules and examples

Models



↓ Computer simulates and analyzes models

Evaluation Results



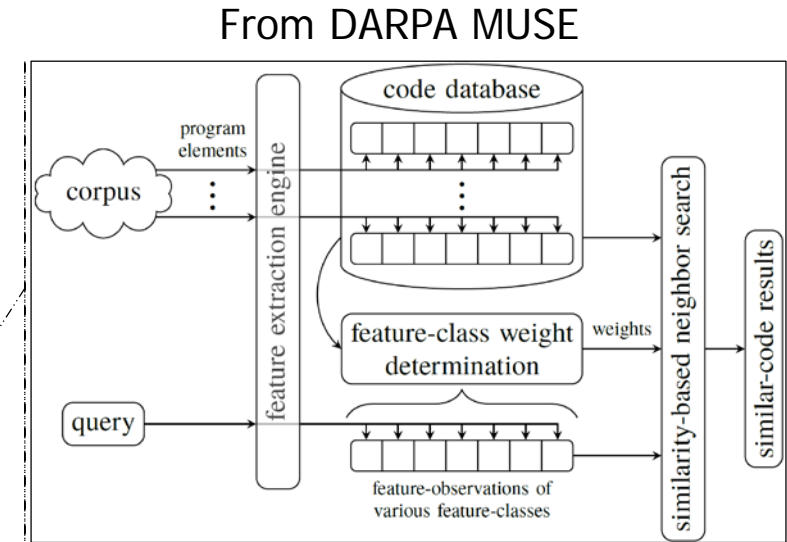
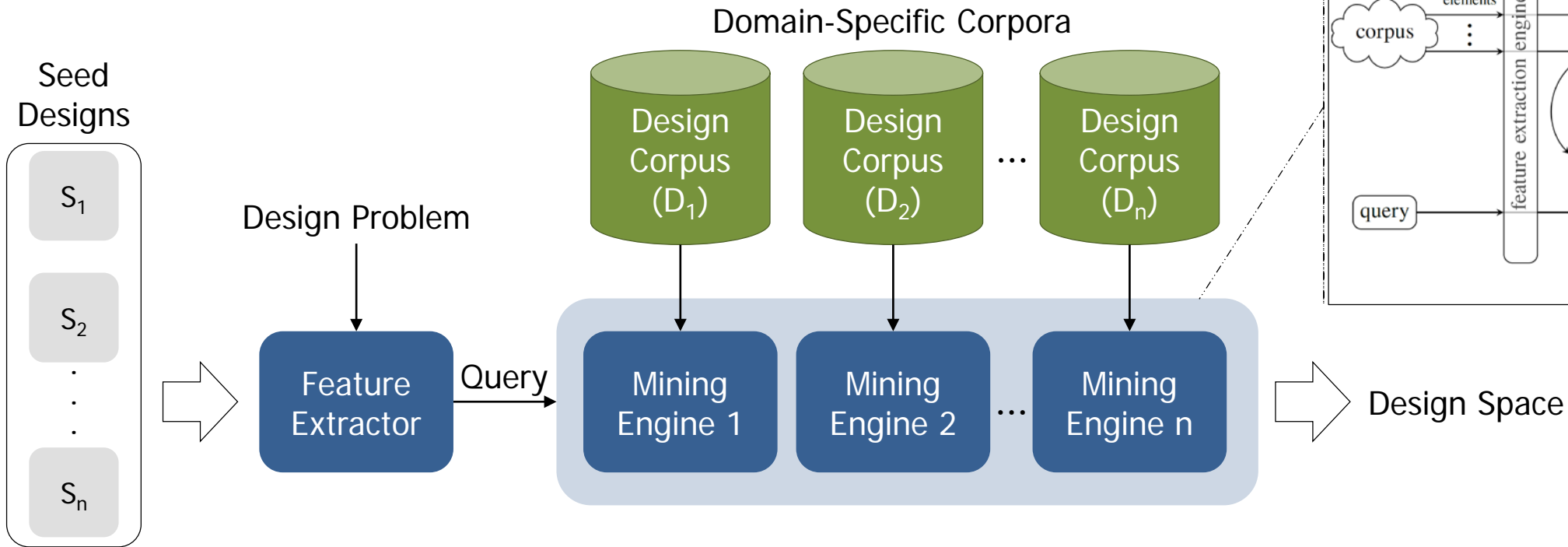
$$(s, r) = f(h, n, c, m, b)$$

Exploration

Computer learns f , searches for "best" design
Humans guide search

$s = \text{speed}; r = \text{range}; h = \text{hull form};$
 $n = \text{navigation sensors}; c = \text{control algorithm};$
 $m = \text{mission payloads}; b = \text{battery}$

Goal: Develop technologies to automatically and incrementally construct design space for a given specification and seed designs



[Kashyap et al, 2017]

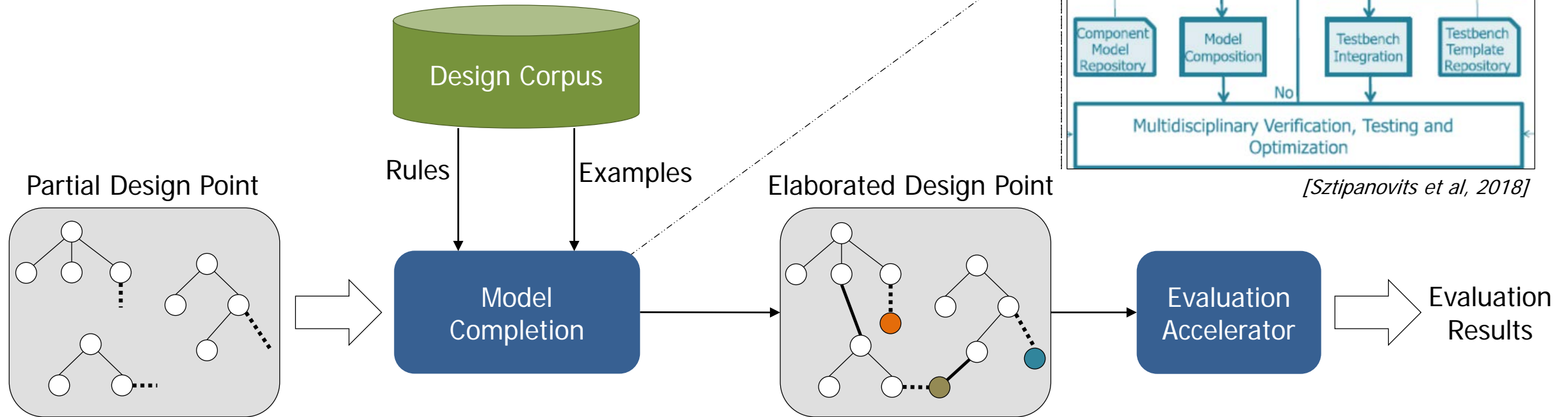
Challenges

- Query generation from seed designs and design problem
- Mining heterogeneous model-based design artifacts
- Incremental construction of design space



Design Composition

Goal: Develop technologies to automatically compose and evaluate a design point



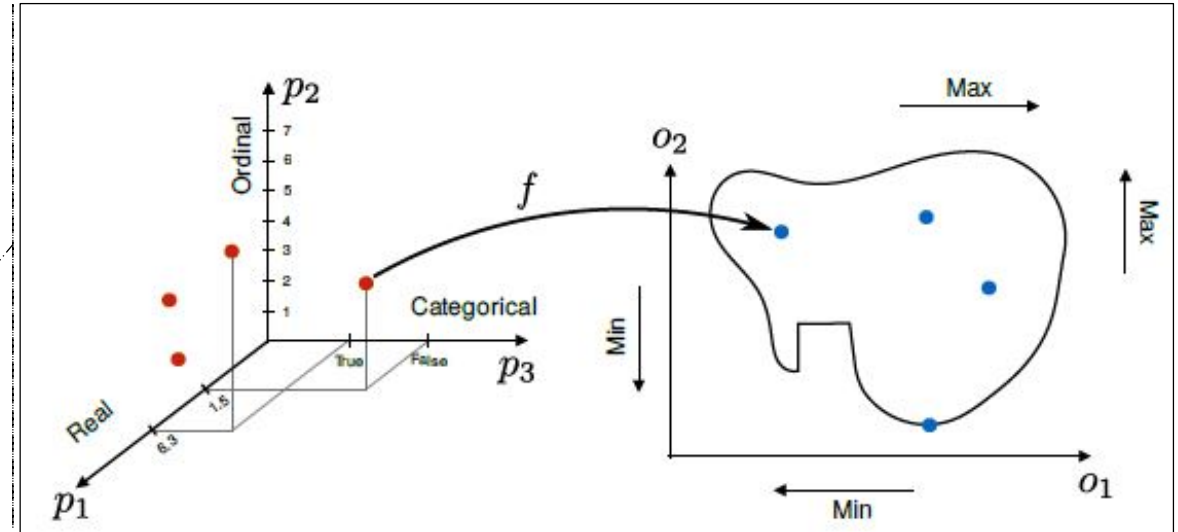
Challenges:

- Automated model completion in heterogeneous domains
- Automated cross domain reasoning and model learning
- Accelerated design analysis and simulation



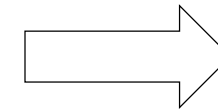
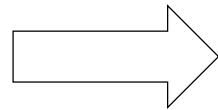
Design Space Exploration

Goal: Develop technologies to explore high-dimensional multi-domain combinatorial design spaces



[Nardi et al, 2018]

Evaluation Results
Design Space

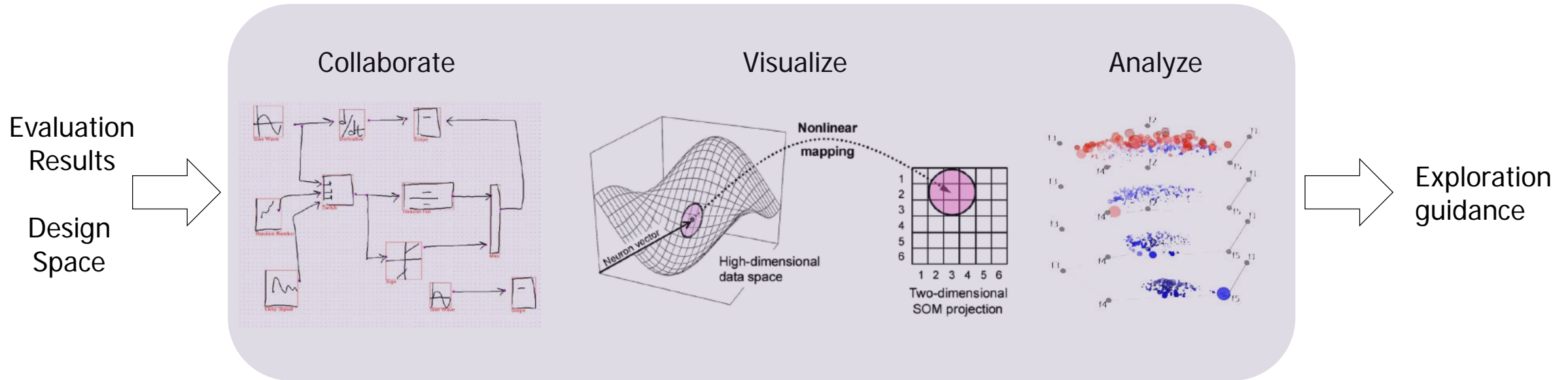


Next Design Point

Challenges:

- High-dimensionality of the space
- Heterogeneous domain models
- Objective functions defined over multiple abstraction layers

Goal: Develop technologies to enable effective partnership between human and machines to solve complex design problems

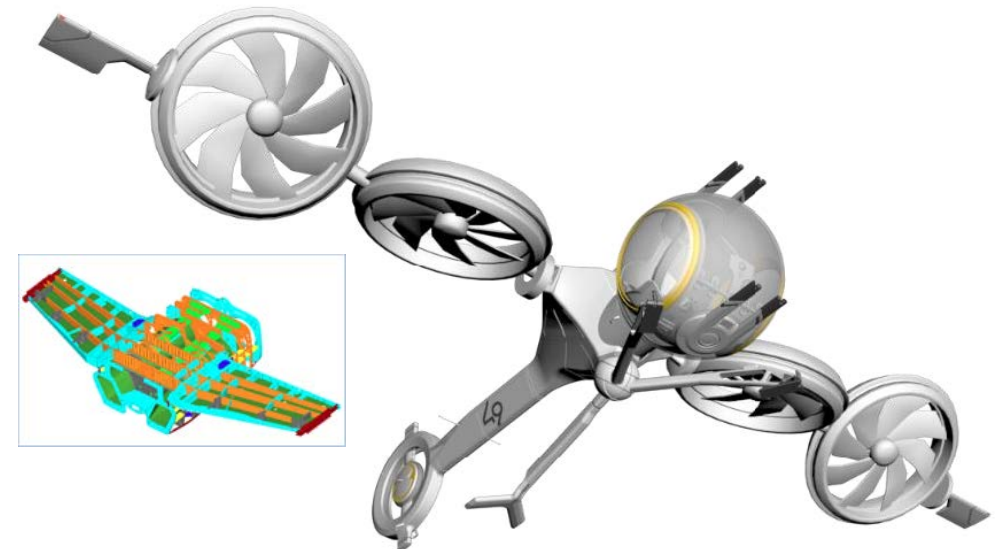


Challenges

- Visualization and understanding of high-dimensional design spaces
- Shaping and guiding exploration
- Interaction complexity of engineering design tools

- CPS Design is complex, highly manual, and iterative
- Novel AI-based approaches combining automated design mining, composition, and scalable exploration offer promise
- Symbiotic Design program geared to explore and demonstrate the potential of AI-based approaches for revolutionizing CPS design in DoD applications

<p>Reference Designs</p>	<table border="1"> <tr> <td>Packaging</td> <td>Bulkheads / Structures</td> </tr> <tr> <td>Nav Sensors</td> <td>SW Systems •Mission C2 •Nav •Autonomy •Sensors</td> </tr> <tr> <td>Comm Modems</td> <td>Energy/Power Mgmt Batteries Buoyancy Control Payload Pressure Hull</td> </tr> <tr> <td>Launch & Retrieval</td> <td>Propulsion Control Surfaces</td> </tr> </table>	Packaging	Bulkheads / Structures	Nav Sensors	SW Systems •Mission C2 •Nav •Autonomy •Sensors	Comm Modems	Energy/Power Mgmt Batteries Buoyancy Control Payload Pressure Hull	Launch & Retrieval	Propulsion Control Surfaces	<p>Novel, efficiently-derived Designs</p>
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Launch & Retrieval	Propulsion Control Surfaces									
<p>CPS Component Models, Libraries, Specs</p> <ul style="list-style-type: none"> • Pressure hull, structure & materials, thermal • Buoyancy, static & dynamic stability • Hydrodynamics, lift, drag, control • Propulsion, motors, power curve, propeller drag • Energy/power management, batteries, distribution • Navigation, INS sensors & control, comms, payload • C2 / operations / autonomy • Launch & retrieval • Cost, SWaP, manufacturability, reliability, safety 										





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