

AI4CE - Automated Space Mission Design Concepts Generation with Reinforcement Learning

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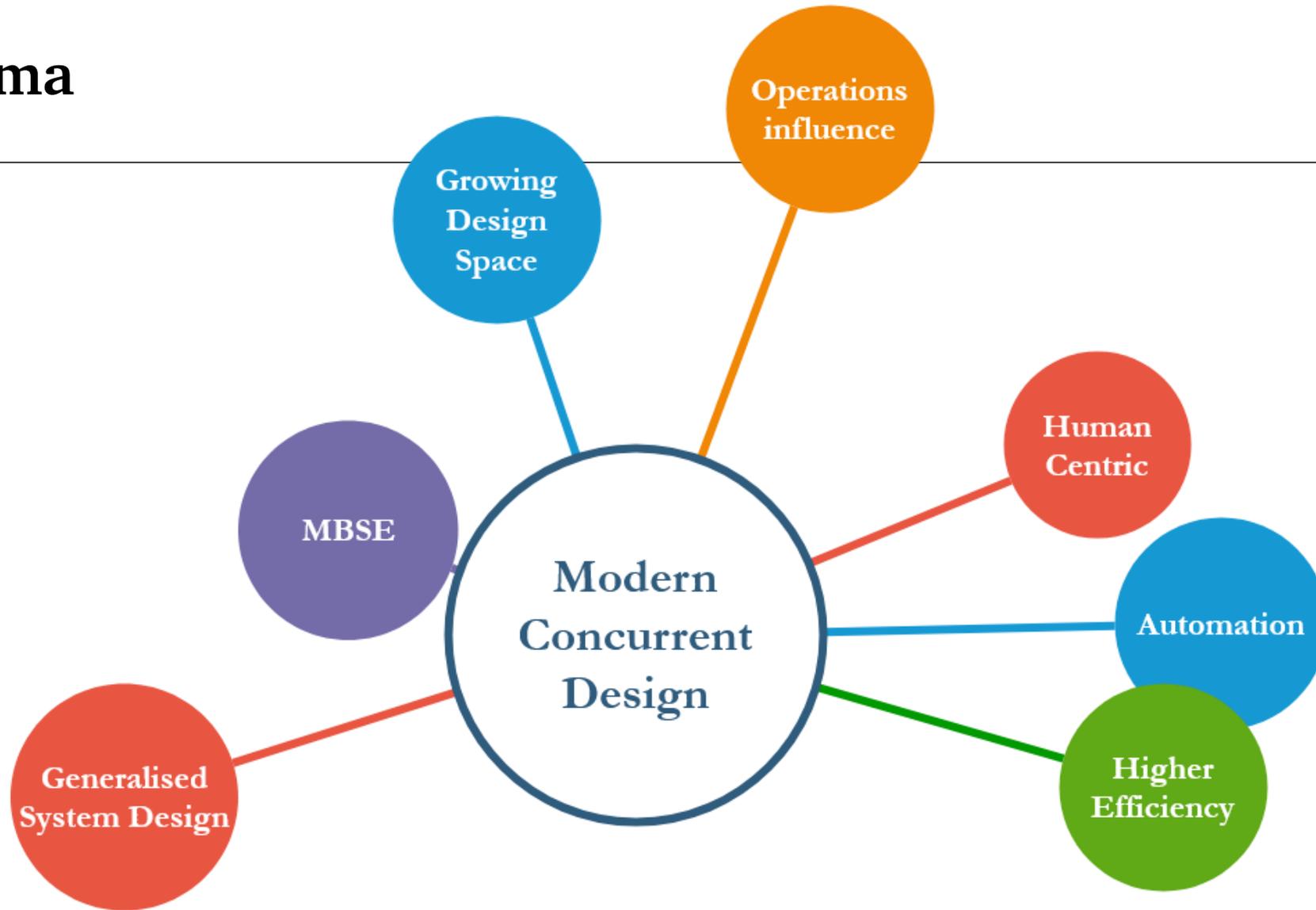
parametry.ai

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Parametry.ai
Concurrent Engineering Lab @ TU Darmstadt
ESALab@TU Darmstadt



[1]

The Dilemma



What would be the perfect solution?



J.A.R.V.I.S

[2]

Functionality

- General, automated design creation
- User-friendly and interactive
- Integration of the OPS experience

Integration

- Interactive CE integration
- MBSE-native

} higher efficiency
+
MBSE integration

→ *A system that helps CE study experts interactively and seamlessly, to design the best possible concept design.*

Q

What about using AI ?

State of the Art: AI for CE support

	 Top-Down	 Bottom-Up
Creating Design knowledge based on	<ul style="list-style-type: none">• Historical mission• Extracting knowledge	<ul style="list-style-type: none">• System requirements• Text book calculations
Technology	<ul style="list-style-type: none">• NLP• Expert Systems• Knowledge Base	<ul style="list-style-type: none">• Combinatorics• Reinforcement Learning RL<ul style="list-style-type: none">• DeepRL DRL

Proposition: AI4CE

- PhD Research Project
- Implementing AI-based bottom-up system creation
 - Deep Reinforcement Learning
 - Generalised system creation → Abstract building blocks
 - MBSE/CE integration
- Achieved in 3 modules:
 - DRL Concept Creator
 - MBSE/CE Integration
 - OPS Experience Integration

DCC

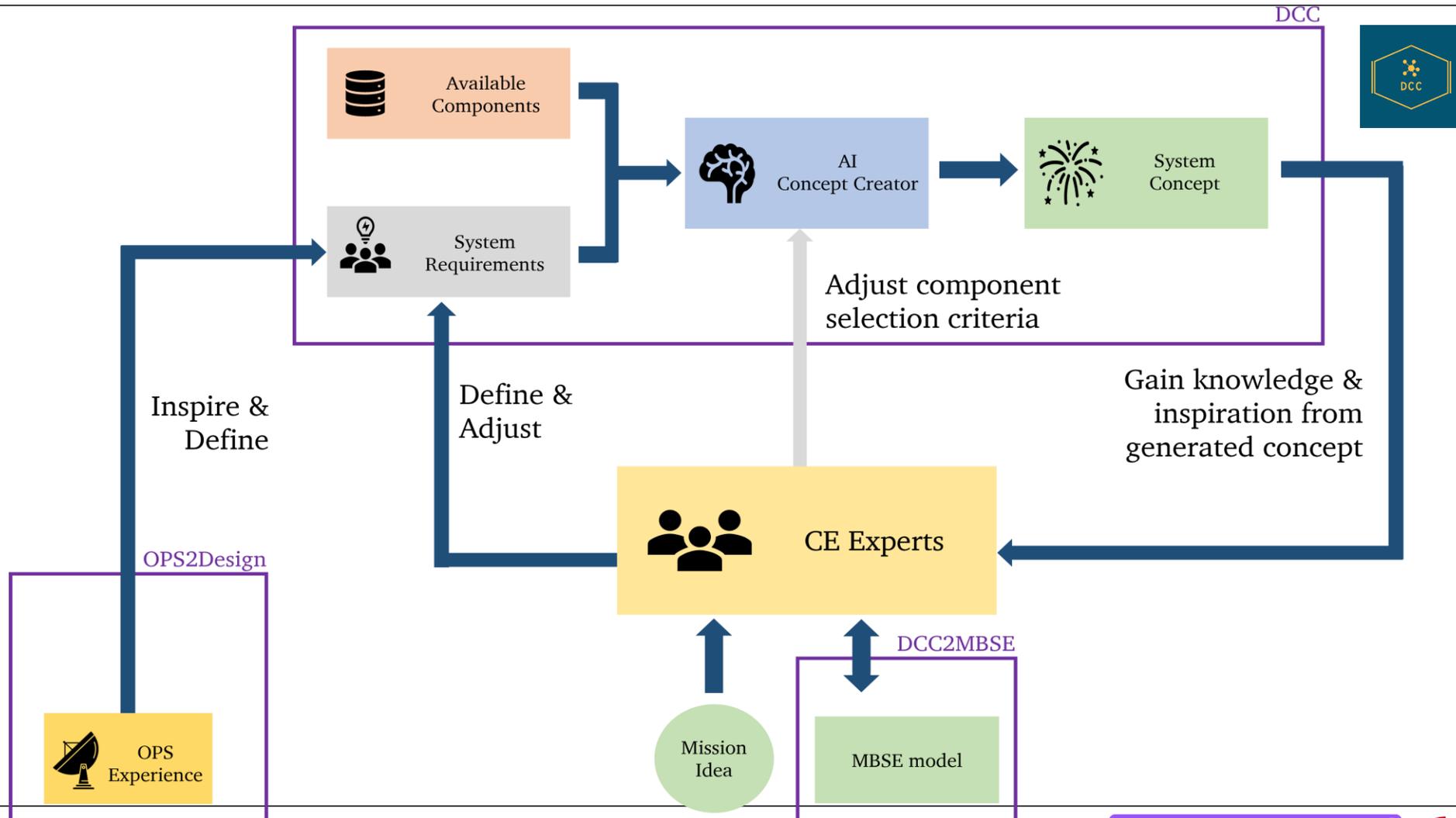
MBSE2DCC

OPS2Design



[3]

AI4CE Overview



[WIP] Research Questions

1. Can **DRL** be used for concept creation? If yes, how?
2. What is the **potential** when applying bottom-up AI methods?
3. How to **model the knowledge base** of component interactions?
4. How to **integrate** automated concept creation in CE workflow?
5. How to **formalise** design requirements?
6. How can **OPS experience** be used for the design process?

Master

PhD

Status Quo

- 2020 TU Delft: early steps and RL
- 2021 TU Darmstadt: progression
- First DRL implementation
 - Prototype
 - Simplified CubeSat
 - Validation
 - Open Source
- 2022 Preparation for PhD research



The Future

2022

Preparation for PhD research

2023

Research @ Parametry/TU Darmstadt

- **Analyse** and define model use cases
- **Model** abstract component interactions
- Designing and **implementing** the AI
- Defining **validation** criteria and use cases
- **Testing** design efficiency, usability and feasibility in CE sessions

2024/25

IAC 2022 – Vision

AI4CE: BOTTUM-UP AI SUPPORT FOR CONCEPTUAL DESIGN

SECESA 2022 – MasterThesis

BUTTOM-UP AI-SUPPORT TO GENERATE CONCEPTUAL DESIGNS FOR CONCURRENT ENGINEERING STUDIES WITH DEEP REINFORCEMENT LEARNING

MBSE 2022 – MBSE Integration

Automated Space Mission Design Concepts Generation with Reinforcement Learning

SpaceOPS 2023 – OPS Integration

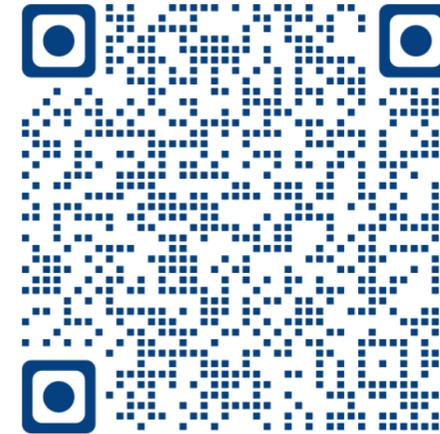
AI4CE - Closing the design-operation-loop: design, operate, learn, repeat

Where you can help me!

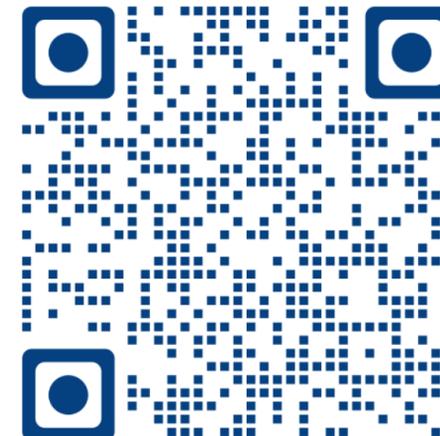
- Modelling
- CE integration
- All things MBSE
 - SysML vs X
 - Import expert MBSE models
- Creation Validation

Contacts

- AI Support for Conceptual Design
 - Concurrent Engineering, MBSE
 - AI, Deep Reinforcement Learning
 - Like Jarvis from Iron Man
- Jan-Peter Ceglarek, Parametry.ai
 - jan-peter@parametry.ai



LinkedIn



Slides
Paper

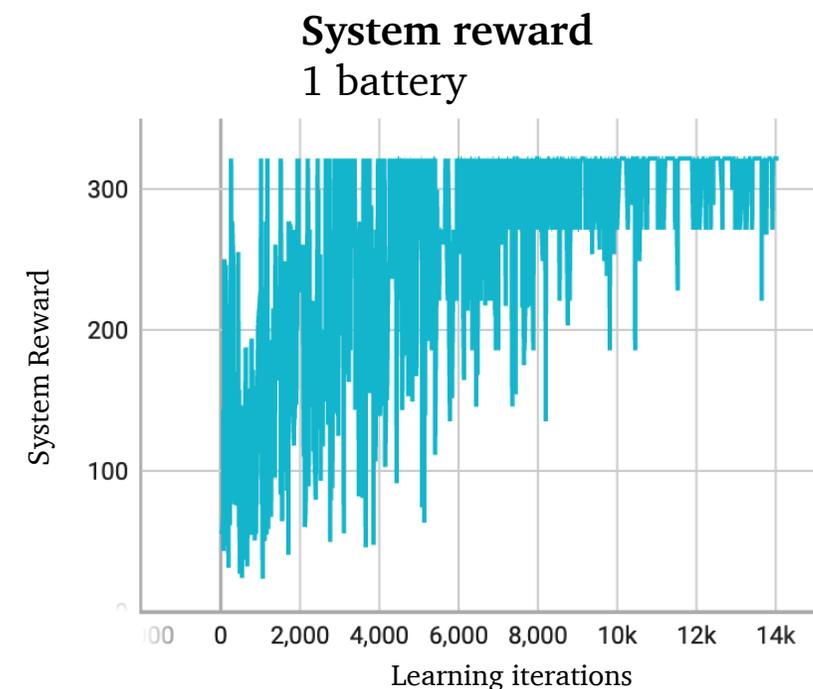
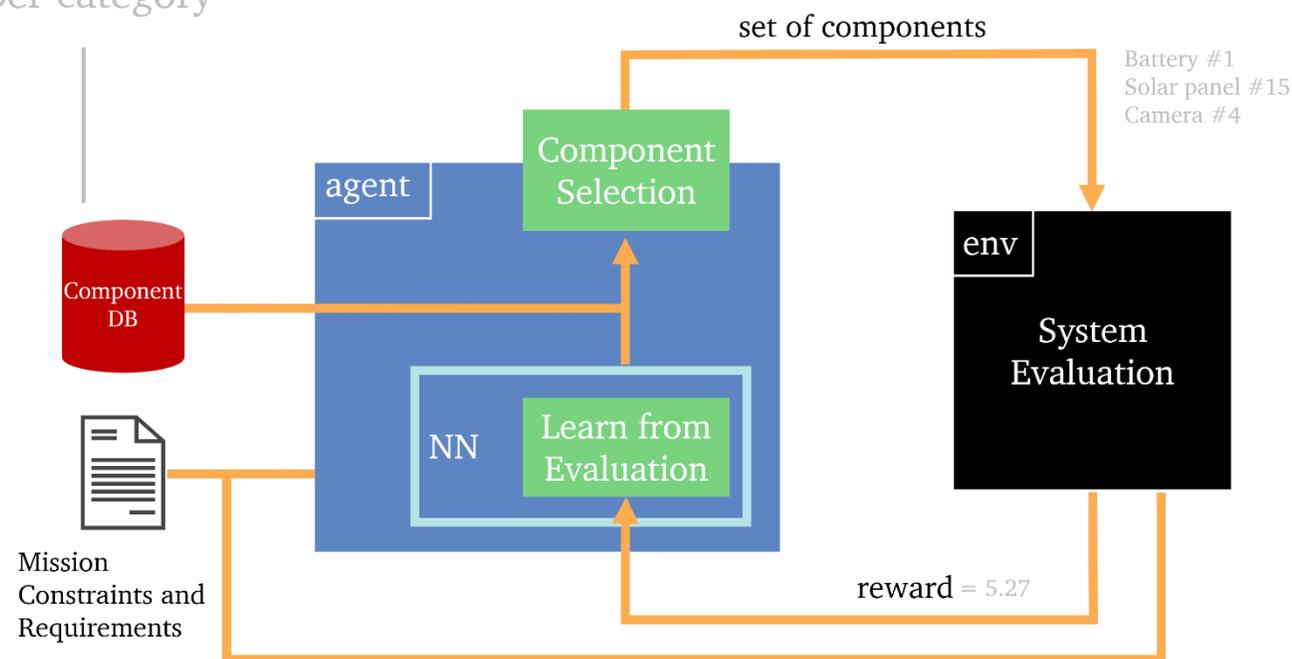
Resources

- [1] DLR Report, Böning 2021, “The Current State of Research and Technology of Digitalization in the Space Industry”
https://www.dlr.de/sc/en/desktopdefault.aspx/tabid-5135/8645_read-8374/
- [2] Jarvis <https://i.pinimg.com/originals/ec/9d/bc/ec9dbccee1ca0cc5c93af15032bb1d5c.jpg>
- [3] <https://gitlab.com/jan-peter/drl-concept-creator>
- [4] EQuiSat https://dl.airtable.com/.attachments/bf8aadaf84b824b4ab8ebf997f3f7cd5/9f5d80eb/EQUISat_2.jpg

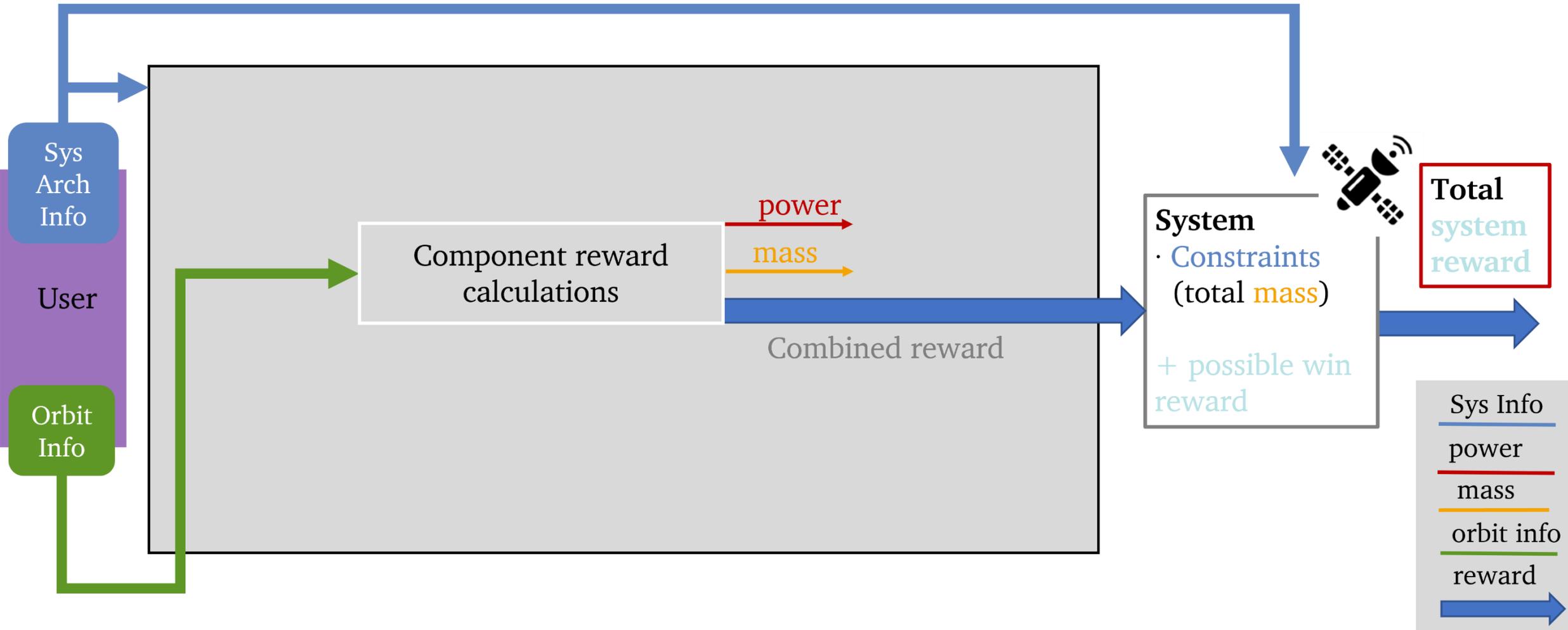
DRL Concept Creator



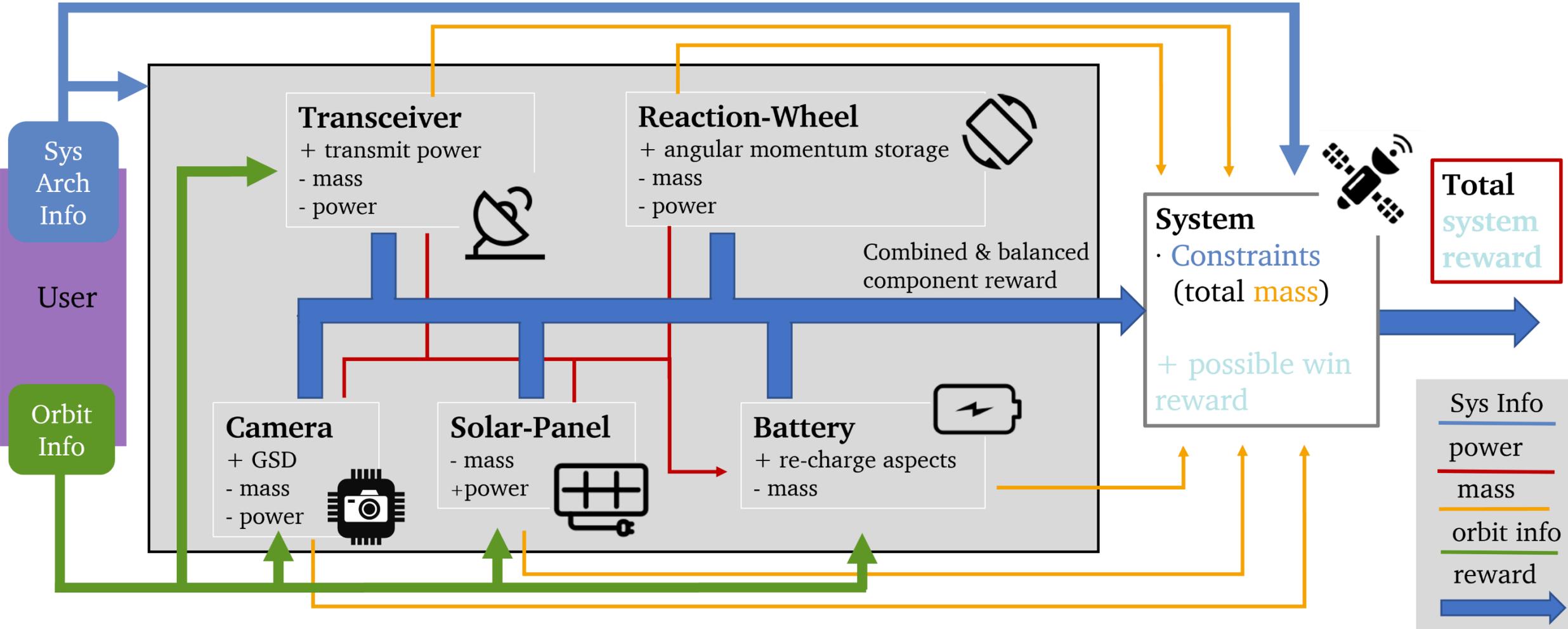
SatSearch WebShop
10s .. 100s components
per category



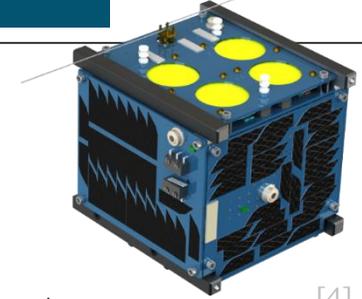
DCC Prototype: Reward Building



DCC Prototype: Reward Building



DCC Prototype: Validation



[4]

- Multiple global system configurations
- Multiple orbits
- Against
 - real-world missions
 - Brute-force combinatorics

1U=best case

→ Promising Results

EQUIsat

transceiver, solar panels

Batteries + payload + structure

AI tool

1U solar panel

mass = 800 g

$P_{\text{balance}} = -2.3 \text{ W}$

EQUIsat

1U solar panel

mass = 1350 g

$P_{\text{balance}} = -5 \text{ W}$

$\text{mass}_{\text{system}}$	- 32 %
P_{balance}	- 54 %

→ Fitting within limitations

- Missing components
- Different database