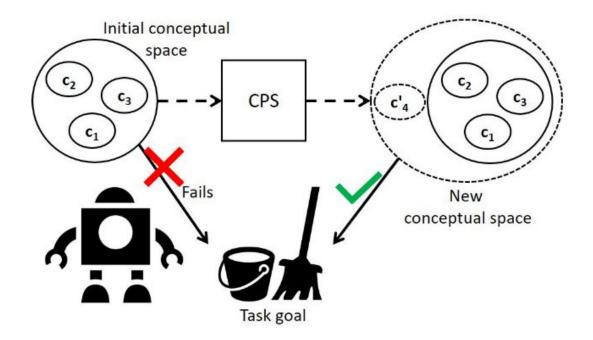




Creative Problem Solving (Novelty Resolution) in Intelligent Agents

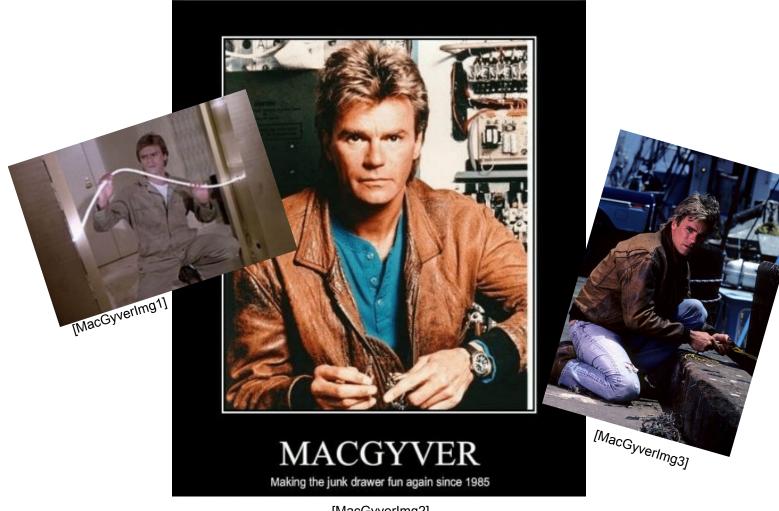
Evana Gizzi Conway Artificial Intelligence PhD Candidate, Tufts University Student Engineering Trainee, NASA Goddard Space Flight Center



Creative problem solving (CPS) is defined as the process by which the agent discovers new concepts that were not in the initial conceptual space of the agent, allowing it to accomplish a previously impossible goal. [Gizzi et al. 2020]

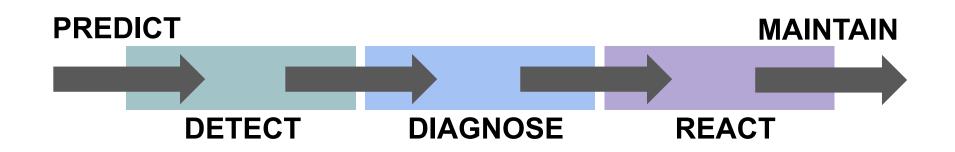
Puzzle Box Problems





[MacGyverImg2]

Progression of CPS [Anomaly Handling]



Research Questions

- 1. What is CPS?
- What kinds of problems require CPS, and how can they be detected/distinguished from GPS problems?
- 3. How can we diagnose the cause of such anomalies which necessitate CPS for problem solving?
- 4. How can CPS be implemented in intelligent agents, specifically in the context of anomalies?
- 5. How can an agent get better at CPS over time?

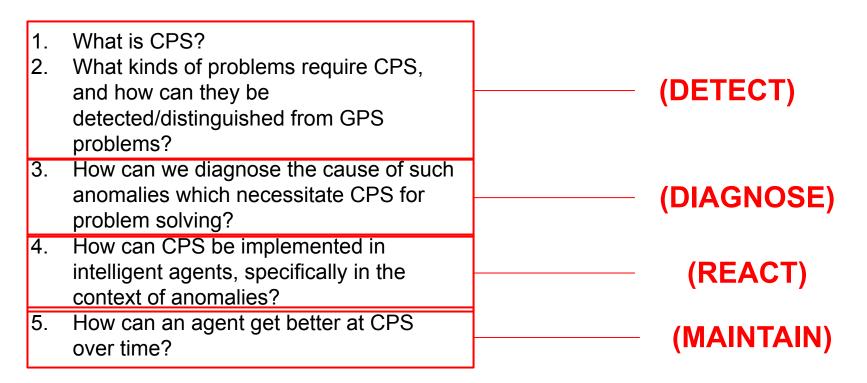
Research Questions

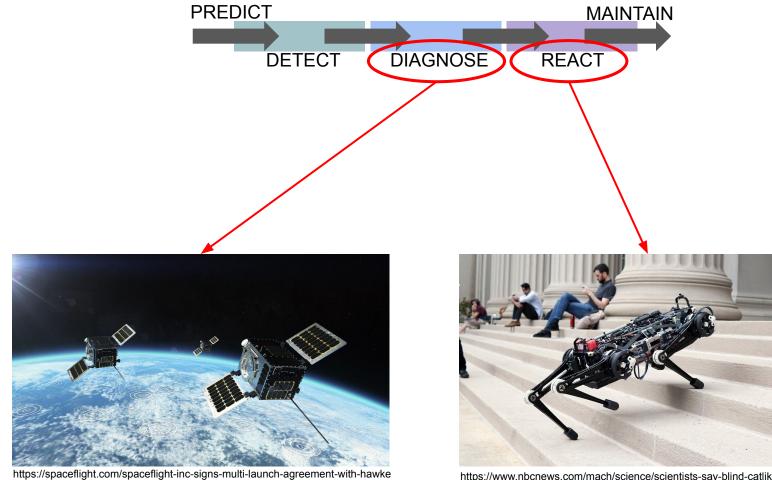
- 1. What is CPS?
- 2. What kinds of problems require CPS, and how can they be detected/distinguished from GPS problems?
- 3. How can we diagnose the cause of such anomalies which necessitate CPS for problem solving?
- 4. How can CPS be implemented in intelligent agents, specifically in the context of anomalies?
- 5. How can an agent get better at CPS over time?

(DIAGNOSE)

(REACT)

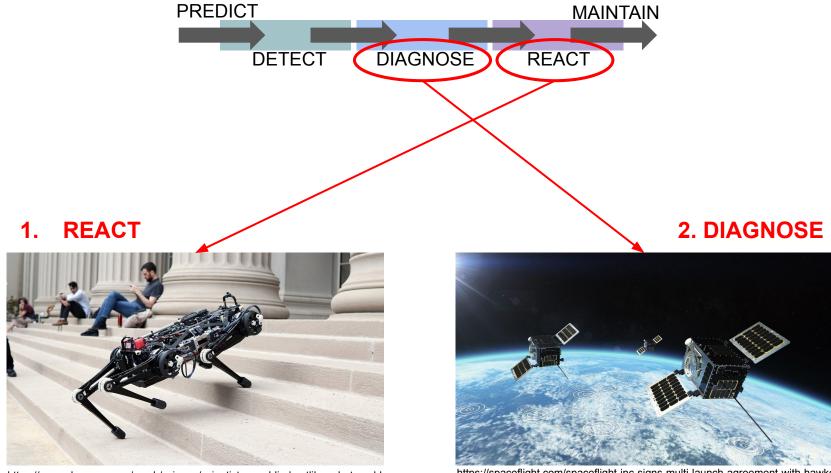
Research Questions





. ye-360/

https://www.nbcnews.com/mach/science/scientists-say-blind-catlike-robot-could-transform-search-rescue-ncna892116



https://www.nbcnews.com/mach/science/scientists-say-blind-catlike-robot-could-transform-search-rescue-ncna892116

https://spaceflight.com/spaceflight-inc-signs-multi-launch-agreement-with-hawke ye-360/

REACT: Action Discovery in Robotics

Gizzi, E., Castro, M.G., Lin, W., and Sinapov, J. (2021). A Framework for Creative Problem Solving through Action Discovery. In 2021 Robotics: Science and Systems (RSS 2021) Workshop on Declarative and Neuro-Symbolic Representations in Robot Learning and Control (DNR-ROB).

Robotics Example



[ANYmalBot]

begun(evanas_talk) in_process(evanas_talk) complete(evanas_talk)

begun(evanas_talk)
in_process(evanas_talk)
complete(evanas_talk)

ACTIONS:

```
call_elevator(:gripper, :elevator,
:button)
    preconditions: is_visible(:button)
    effects: called(:elevator)
        open(:elevator)
```

arguments

ACTIONS:

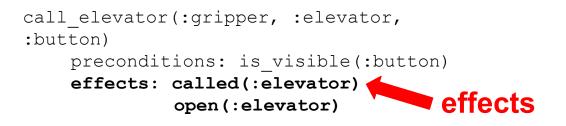
call_elevator(:gripper, :elevator, :button)
 preconditions: is_visible(:button)
 effects: called(:elevator)
 open(:elevator)

ACTIONS:

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:button)
    preconditions: is_visible(:button)
    effects: called(:elevator)
        open(:elevator)
```



ACTIONS:



ACTIONS:

```
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:button)
    preconditions: is_visible(:button)
    effects: called(:elevator)
        open(:elevator)
```

ORIGINAL SCENARIO

ACTIONS:

```
call_elevator(:gripper, :elevator,
:button)
    preconditions: is_visible(:button)
    effects: called(:elevator)
        open(:elevator)
```

ORIGINAL SCENARIO

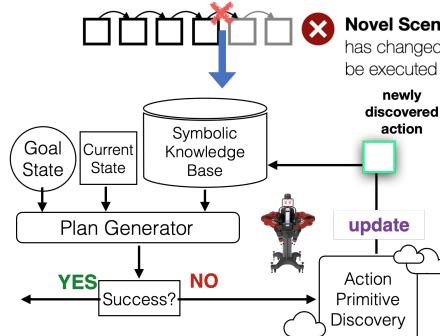
NOVEL SCENARIO

Original Scenario: Agent is able to plan for accomplishing a goal state

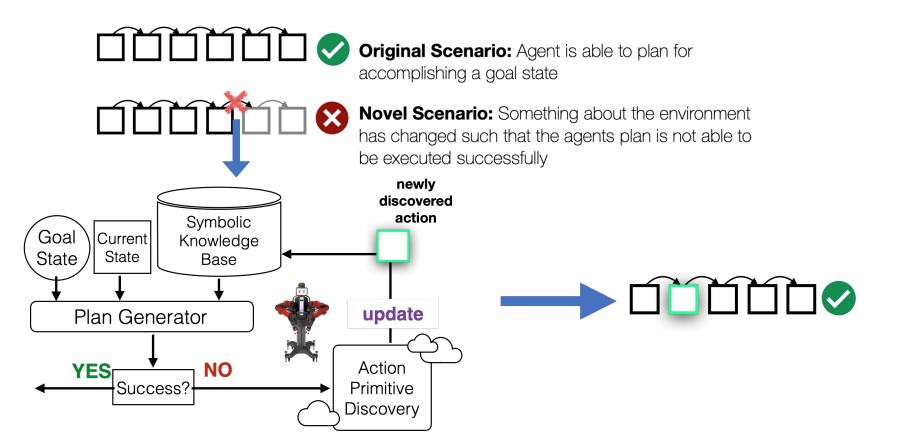


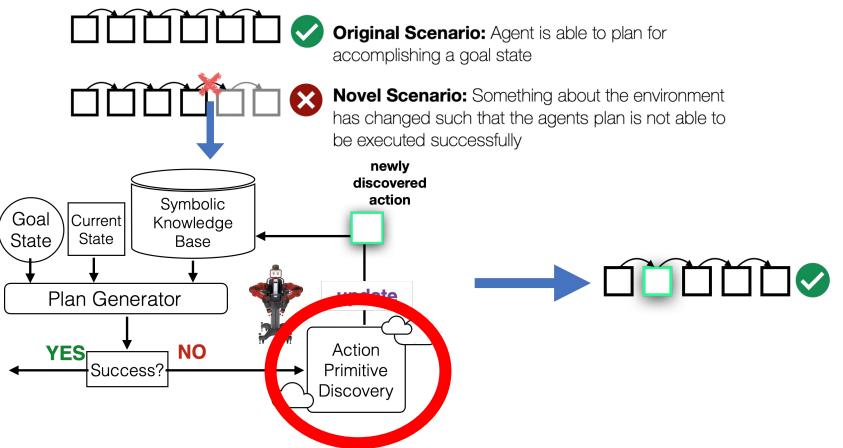
Novel Scenario: Something about the environment has changed such that the agents plan is not able to be executed successfully

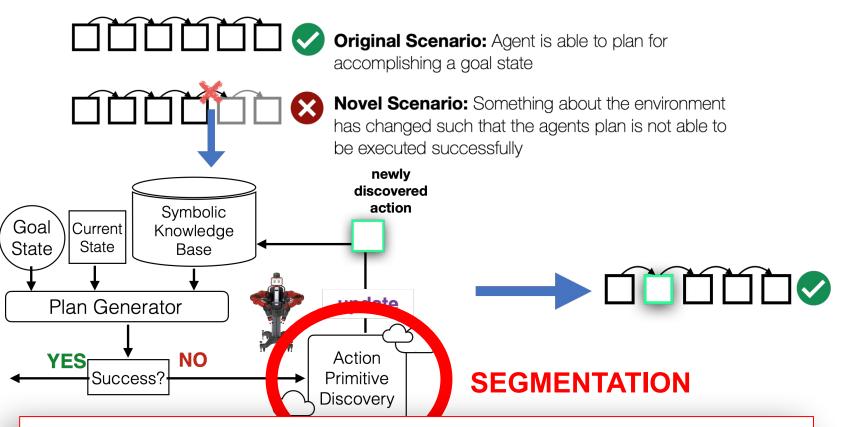
Original Scenario: Agent is able to plan for accomplishing a goal state



Novel Scenario: Something about the environment has changed such that the agents plan is not able to be executed successfully

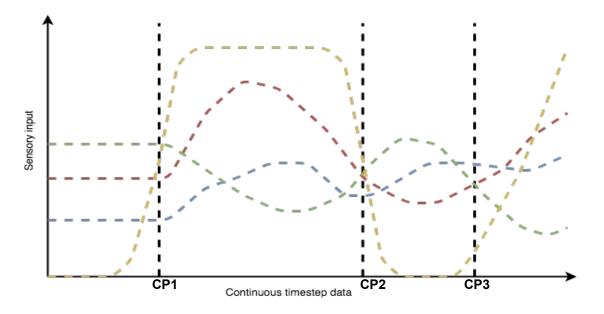






Gizzi, E., Castro, M.G., and Sinapov, J. (2019). Creative Problem Solving by Robots using Action Primitive Discovery. In 2019 Joint IEEE 9th International Conference on Development and Learning and Epigenetic Robotics (ICDL-EpiRob) (pp. 228-233). IEEE.

Bayesian Change Point Detection



- — = number of pixels of object_1 visible to agent
- = = = object_1's distance from left gripper
- — = object_1's distance from right gripper
- - = object_1's distance from robot center point

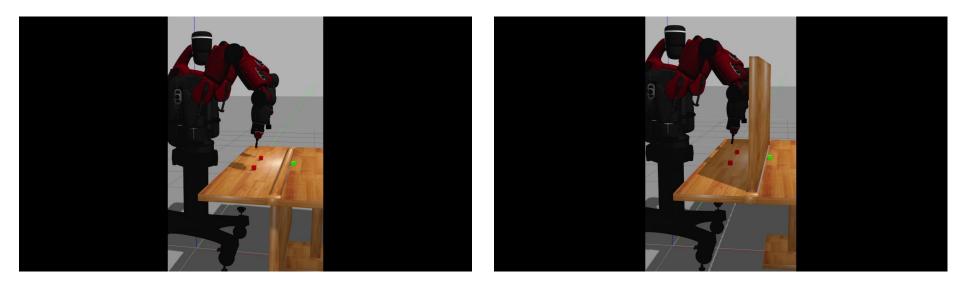


[ANYmalBot] edited

Illustrative Example



Illustrative Example



ORIGINAL SCENARIO

NOVEL SCENARIO

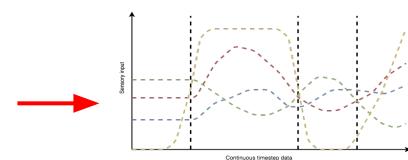
Action-Object Combination List (candidate actions to vary)

press_button(r_gripper, l_button)
press_button(r_gripper, r_button)
obtain_object(r_gripper, wall)
obtain_object(r_gripper, table)
obtain_object(r_gripper, block)
press_button(l_gripper, r_button)
press_button(l_gripper, r_button)
obtain_object(l_gripper, wall)
obtain_object(l_gripper, table)
obtain_object(l_gripper, block)

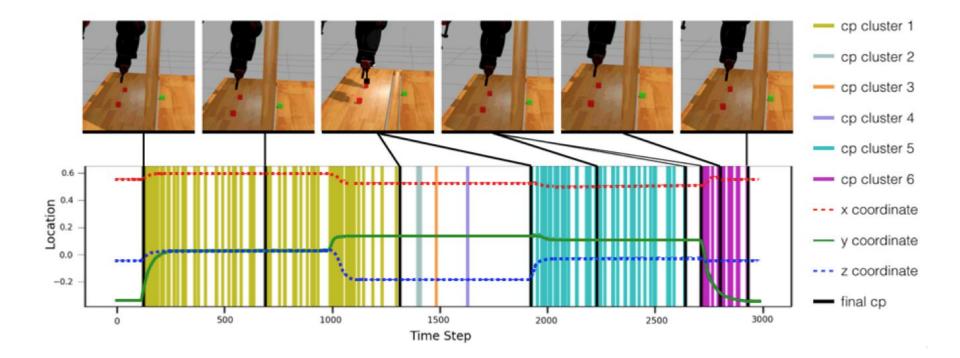
Action-Object Combination List (candidate actions to vary)

press_button(r_gripper, l_button)
press_button(r_gripper, r_button)
obtain_object(r_gripper, wall)
obtain_object(r_gripper, table)
obtain_object(r_gripper, block)

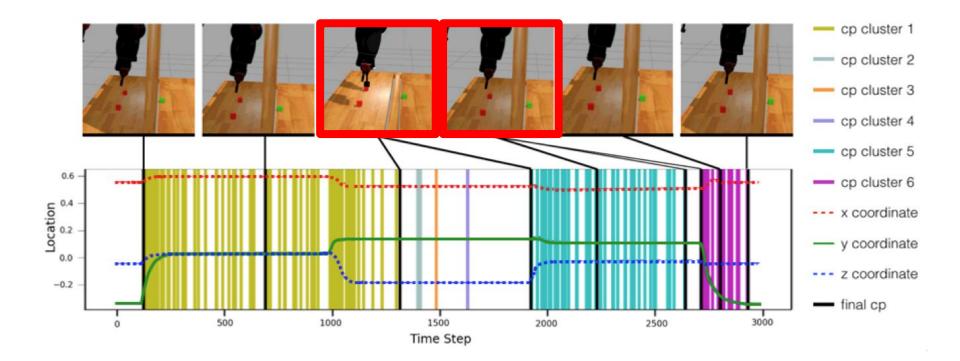
press_button(l_gripper, r_button)
press_button(l_gripper, r_button)
obtain_object(l_gripper, table)
obtain_object(l_gripper, block)



Results



Results



a_f =

```
a<sub>s</sub> =
```

```
action_attempt_1_trial4_seg12(:gripper, :button, :object)
    preconditions:
    effects: is_visible(:object)
        pressed(:button)
```

a_f =

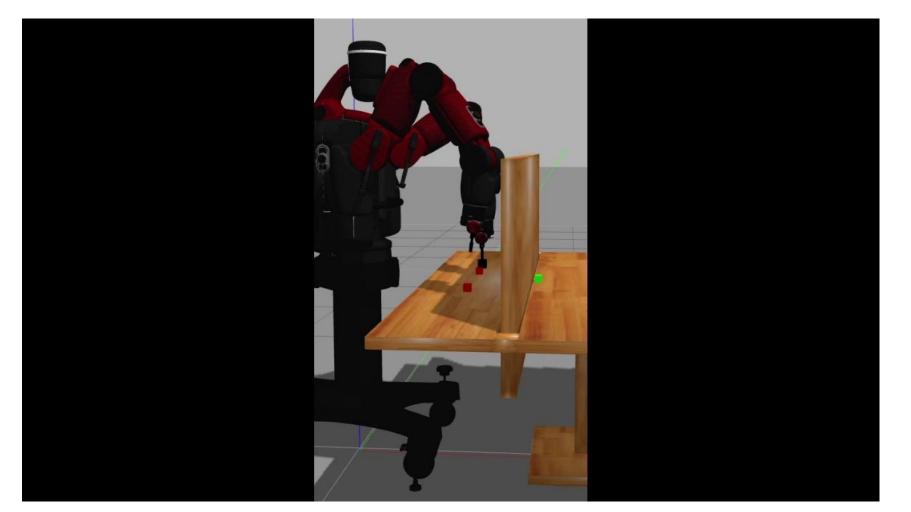
a_s =

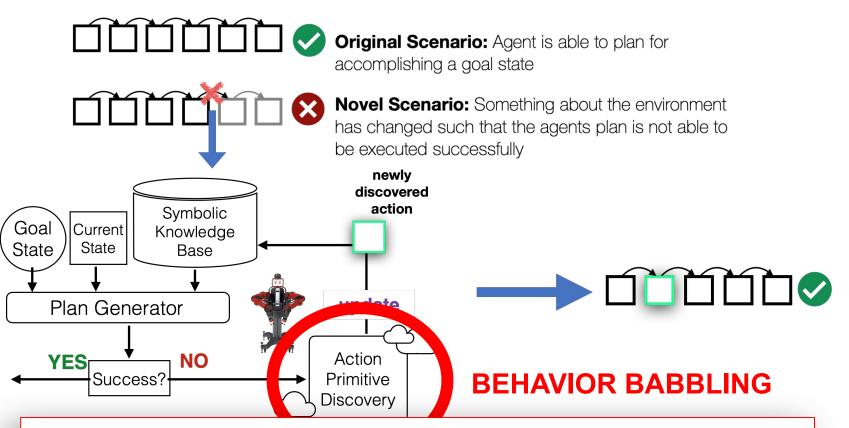
```
action_attempt_1_trial4_seg12(:gripper, :button, :object)
    preconditions:
    effects: is_visible(:object)
        pressed(:button)
```

a_f =

a_s =

```
action_attempt_1_trial4_seg12(:gripper, :button, :object)
    preconditions:
    effects: is_visible(:object)
        pressed(:button)
```





Gizzi, E., Hassan, A., Lin, W., Rhea, K., and Sinapov, J. (2021). Toward Creative Problem Solving Agents: Action Discovery through Behavior Babbling. In 2021 IEEE International Conference on Development and Learning (ICDL) (pp. 1-7). IEEE.







ORIGINAL SCENARIO mass: 5.0 kg push(arm, object)



ORIGINAL SCENARIO mass: 5.0 kg push(arm, object)

NOVEL SCENARIO

mass:20.0 kg
push(arm, object)



ORIGINAL SCENARIO

mass:5.0 kg
push(arm, object)

- **rate:** 2.0 m/s
- MM: 1.0 meter
- orientation: 180°

NOVEL SCENARIO

mass:20.0 kg
push(arm, object)



ORIGINAL SCENARIO

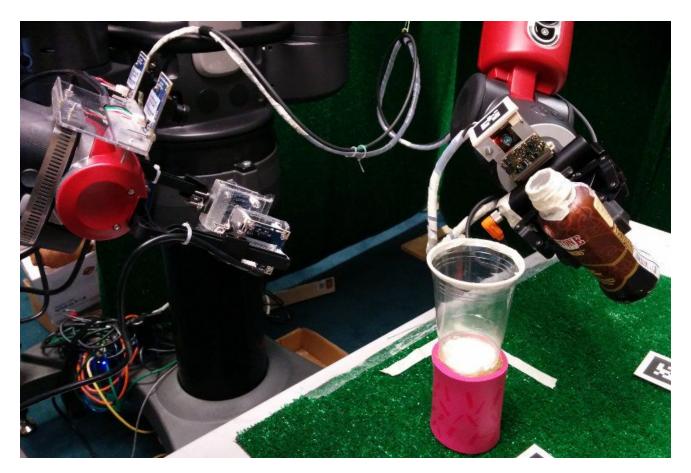
mass:5.0 kg
push(arm, object)

- **rate:** 2.0 m/s
- MM: 1.0 meter
- orientation: 180°

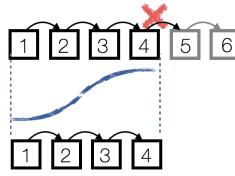
NOVEL SCENARIO

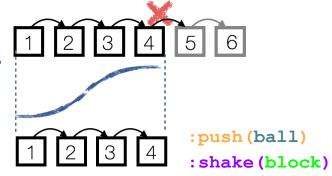
mass:20.0 kg push -> strike

- **rate:** 10.0 m/s
- **MM**: 1.0 meter
- orientation: 180°

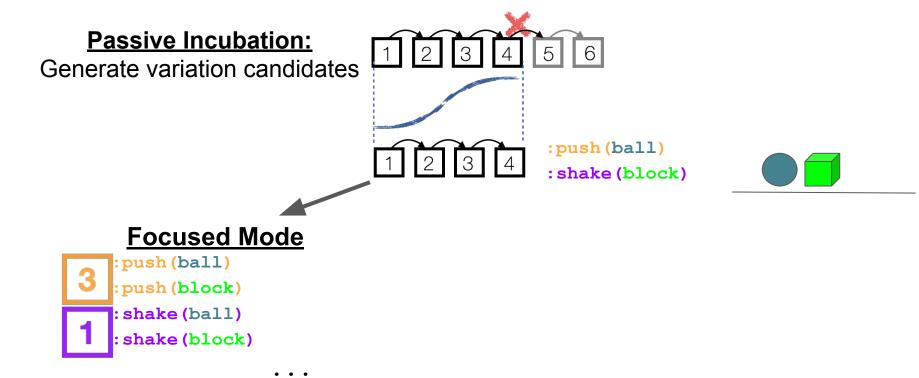


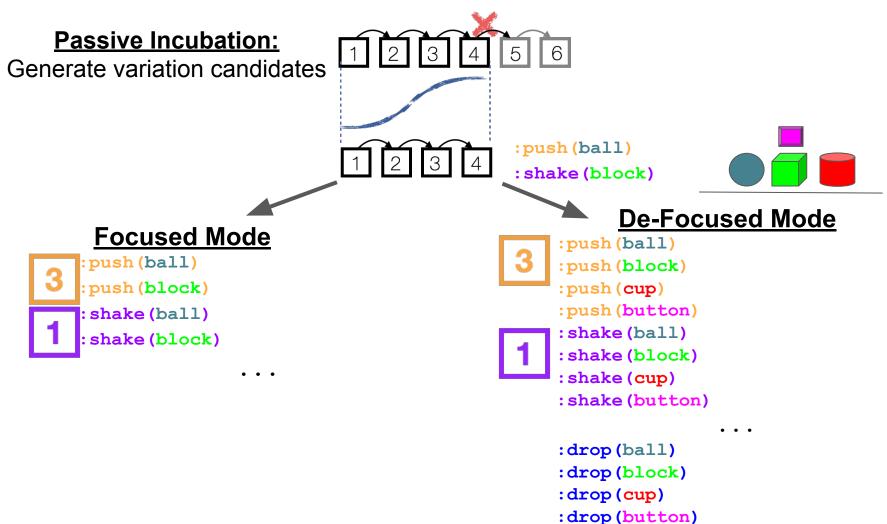
https://www.researchgate.net/publication/308994545_Stereo_Vision_of_Liquid_and_Particle_Flow_fo r_Robot_Pouring/figures?lo=1&utm_source=google&utm_medium=organic







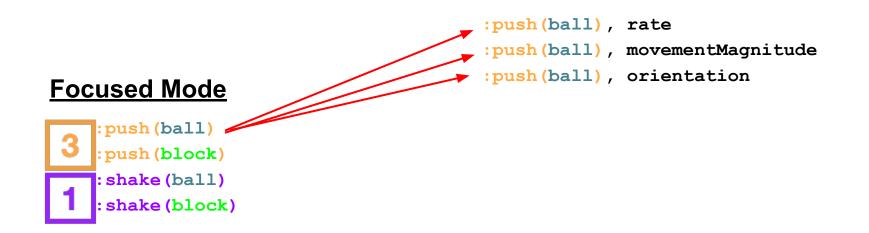


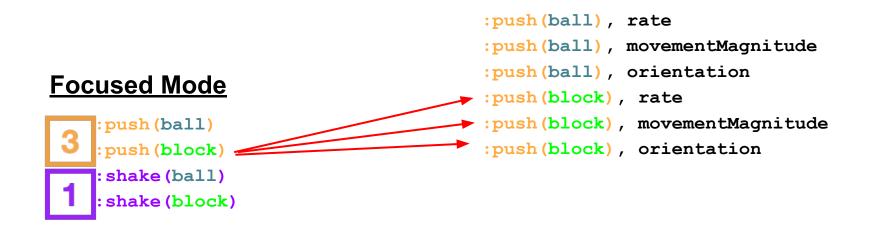


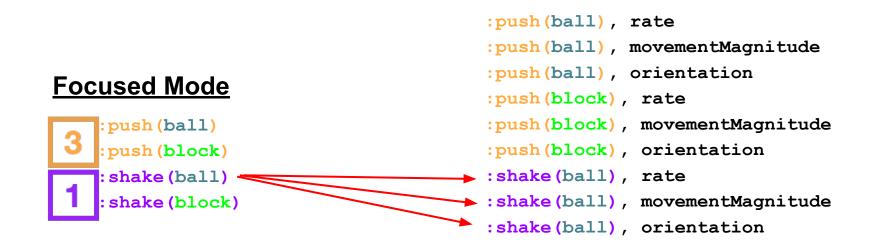
Generate variation candidates

Focused Mode

3 :push(ball) :push(block) :shake(ball) :shake(block)







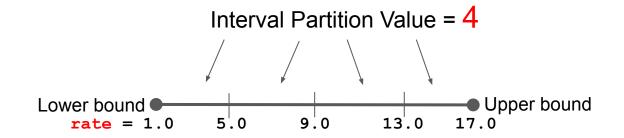
Generate variation candidates

Focused Mode

3 :push(block) :shake(ball) :shake(block)

- :push(ball), rate
- :push(ball), movementMagnitude
- :push(ball), orientation
- :push(block), rate
- :push(block), movementMagnitude
- :push(block), orientation
- :shake(ball), rate
- :shake(ball), movementMagnitude
- :shake(ball), orientation
- :shake(block), rate
- shake(block), movementMagnitude
- **: shake (block)**, orientation

Active Incubation: Evaluate variation candidates





- :push(ball), rate=9.0
- :push(ball), rate=13.0
- :push(ball), rate=17.0

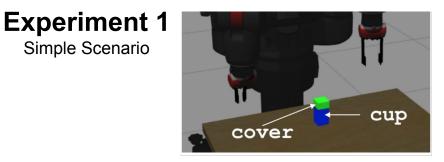
Active Incubation: Evaluate variation candidates

Inheritance Condition: Actions which accomplish the same results as the original action are added. (Focused Mode) Novelty Condition: Actions which accomplish results which are novel to the original candidate action are added. (De-Focused Mode)



Experiment 2 Focused Case





GOAL Uncover the cup **NOVEL SCENARIO** Cover is too heavy NOVEL ACTION : push -> : strike

(increased **<u>rate</u>** parameter)



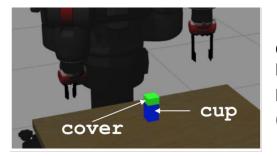
Simple Scenario



De-Focused Case

Experiment 1

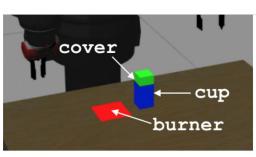
Simple Scenario



GOAL Uncover the cup **NOVEL SCENARIO** Cover is too heavy **NOVEL ACTION : push -> : strike** (increased **rate** parameter)

Experiment 2

Focused Case



GOAL Cook food
NOVEL SCENARIO Cover gets knocked off table (light
weight)
NOVEL ACTION :push -> :nudge

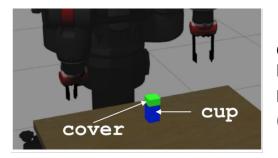
(decreased **<u>rate</u>** or **<u>movementMagnitude</u>** parameter)



De-Focused Case

Experiment 1

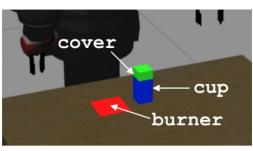
Simple Scenario



GOAL Uncover the cup **NOVEL SCENARIO** Cover is too heavy **NOVEL ACTION : push -> : strike** (increased **rate** parameter)

Experiment 2

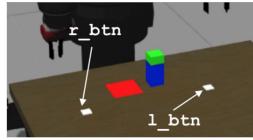
Focused Case



GOAL Cook food
NOVEL SCENARIO Cover gets knocked off table (light
weight)
NOVEL ACTION :push -> :nudge
(decreased rate or movementMagnitude parameter)

Experiment 3

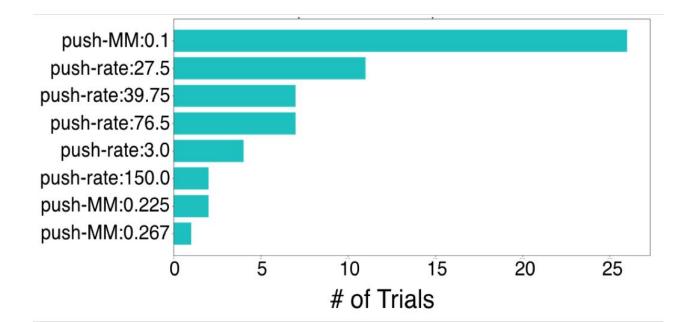
De-Focused Case



GOAL Cook food NOVEL SCENARIO Burner must first be turned on NOVEL ACTION :push -> :press (variation of <u>orientation</u> parameter NOVEL ACTION :shake -> :press (variation of <u>orientation</u> parameter)

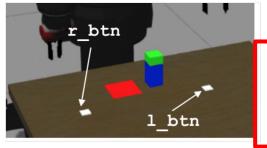




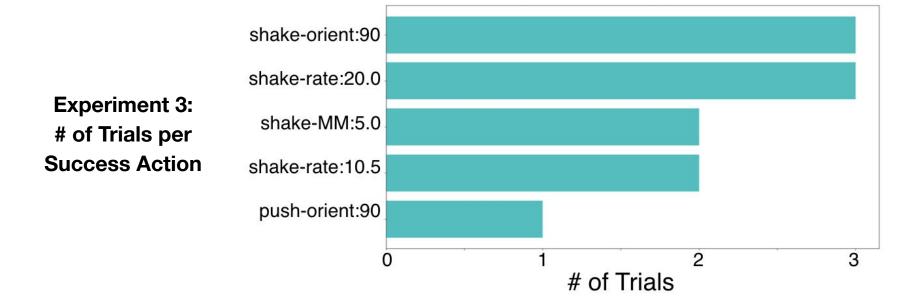


Experiment 3

Defocused Case



GOAL Cook food NOVEL SCENABIO Burner must first be turned on NOVEL ACTION : push -> : press (variation of <u>orientation</u> parameter NOVEL ACTION : shake -> : press (variation of <u>orientation</u> parameter)

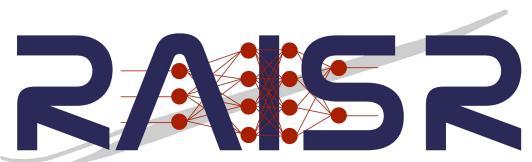


Summary on **REACT**

- Action discovery is a viable path toward novelty resolution
 - Not always necessary
- Important to examine the relationship between levels of knowledge abstraction
 - Hybrid approaches
- Limitations
 - These methods rely on many interaction trials/counterfactual reasoning (not always possible without a simulator)
 - Not capable of learning entirely new controllers from scratch
- Future Work
 - Life long creative problem solving (LLCPS)

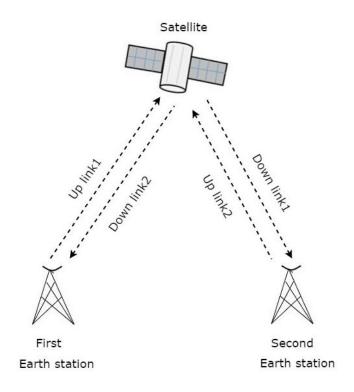
DIAGNOSE: Fault Diagnosis in Spacecraft

Gizzi, E., Owens, H., Pellegrino, N., Rasskin, G., Trombley, C., Marshall, J., and Sinapov, J. (2022). Autonomous System-Level Fault Diagnosis in Satellites using Spacecraft Telemetry. Under review

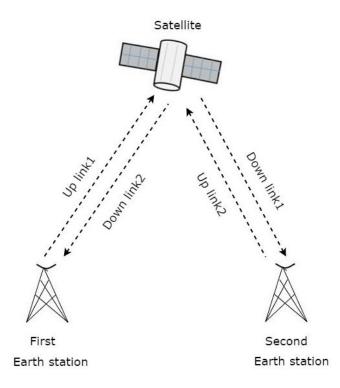


Research in Artificial Intelligence for Spacecraft Resilience

Background and Motivation



Background and Motivation

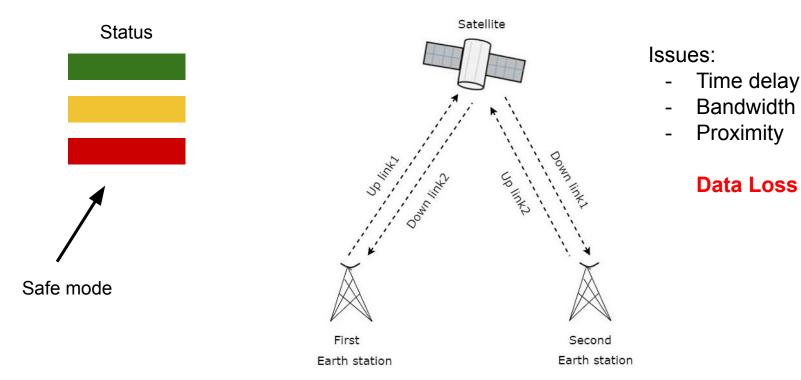


Issues:

- Time delay
- Bandwidth
- Proximity

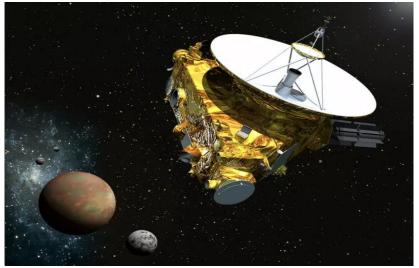
Data Loss

Background and Motivation



Safe Mode Examples

"The investigation into the anomaly that caused New Horizons to enter "safe mode" on July 4 has concluded that no hardware or software fault occurred on the spacecraft."



"NASA's New Horizons Plans July 7 Return to Normal Science Operations." NASA Blogs. July 5, 2015.

Cofield, Calla, "NASA's New Horizons Spacecraft Wakes Up for Pluto Encounter in 2015." Space.com. December 7, 2014.



Grecius, Tony, "Mars 2020 Perseverance Healthy and on Its Way." NASA Blogs. July 30, 2020.

"We set the limits for the temperature differential conservatively tight for triggering a safe mode. The philosophy is that it is far better to trigger a safe mode event when not required, than miss one that is."

Goal of Project RAISR:

Intelligently diagnose cause of fault, given the symptom.

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Intelligently diagnose cause of fault, given the symptom.

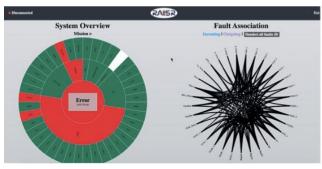
Evana Gizzi [GSFC-587] Dr. James Marshall [GSFC-587] Nicholas Pellegrino [GSFC-587-OSSI] Jeff St. Jean [GSFC-587-OSSI] Christopher Chapman [GSFC-587-OSSI] Hayley Owens [GSFC-587-OSSI] Gabriel Rasskin [GSFC-587-OSSI] Christopher Trombley [GSFC-587] Carrie Kuzio [WFF-589] Ahmed Ghalib [WFF-811]

Artificial Intelligence (AI) Constructs Examined

- Classification and Predictive
 - Long Short-term Memory (LSTM) Neural Networks
 - Autoencoders, Variational Autoencoders, Shapley Values
 - Convolutional Neural Networks (CNN)
 - Kalman Filters
 - Telemetry vector Clustering
- Graph-based
 - Association Rule Mining
 - Function classification (curve smoothing, function fitting, feature extraction)
 - Causality
 - Walkdown Algorithm
- Reinforcement Learning
 - Partially Observable Markov Decision Processes (POMDP)
 - Proximal Policy Optimization (PPO), forest of POMDPs

Algorithms and Tools Developed

- Importance Sampling
- 2 Diagnostic Algorithms
- Diagnosis GUI
- python/cFS adapter (Marshall)



Artificial Intelligence (AI) Constructs Examined

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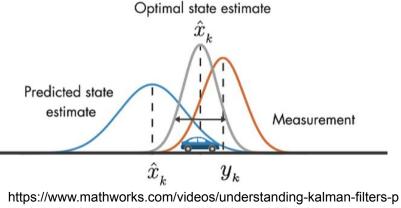
Algorithms and Tools Developed

- Importance Sampling
- 2 Diagnostic Algorithms
- Diagnosis GUI
- python/cFS adapter (Marshall)

Preliminaries: Kalman Filters

- Dynamic linear estimation models able to provide a prediction from previously measured data
- Account for noise in sensor readings

How "broken" is a mnemonic, relative to itself?



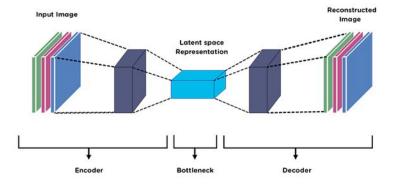
art-3-optimal-state-estimator--1490710645421.html

Preliminaries: Autoencoders

- Unsupervised neural network representation learners
- Can encode information into a latent space representational form, and decode latent space back into original high dimensional form
- Use Shapley Values and reconstruction error

How "broken" is a mnemonic, relative to the telemetry landscape?

(contextual)

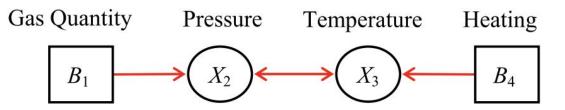


https://medium.com/@birla.deepak26/autoencoders-76bb49ae6a8f

Preliminaries: Causality

- Measures the level of directional "instantaneous relatedness" among mnemonics
- Utilizes conditional probabilities and derivatives (do-calculus) to capture event changes
- Similar to traditional correlation calculation, with more restrictions to induce causal relationships

Which mnemonics are related to which other mnemonics, in an "instantaneous" sense?



Zhang, Q., Dong, C., Cui, Y., & Yang, Z. (2013). Dynamic uncertain causality graph for knowledge representation and probabilistic reasoning: statistics base, matrix, and application. IEEE Transactions on Neural Networks and Learning Systems, 25(4), 645-663.

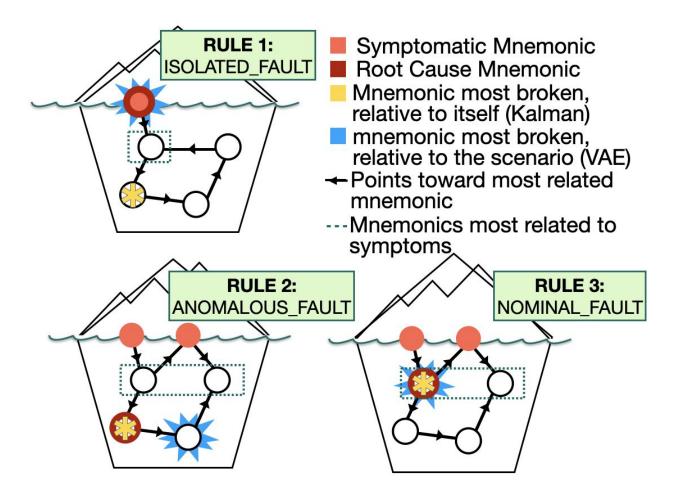


Figure 1: Classes of faults diagnosed by RAISR using spacecraft telemetry

Outcomes

- Software Platform
 - Performance Tested on 4 datasets of varying complexity levels
 - Baseline Tested against state of art AI approaches (0.7 vs 0.55 in worst case performance of RAISR)
- Hardware-in-the-loop
 - Benchmark Tests on Raspberry Pi ARM architecture (COVID contingency plan)
 - Ablation Study on constituent parts
- Paper in review process

Distributed Spacecraft Mission (DSM)



https://spaceflight.com/spaceflight-inc-signs-multi-launch-agreement-with-hawkeye-360/

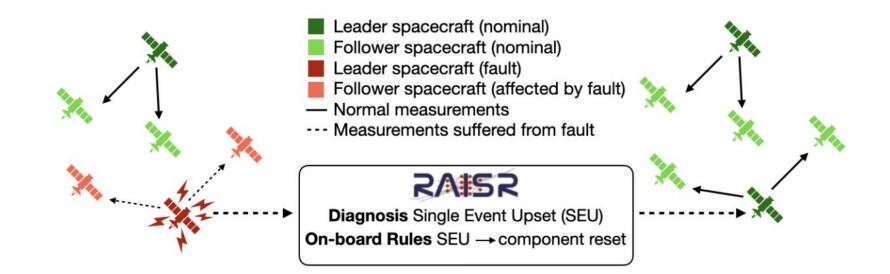
Distributed Spacecraft Mission (DSM)



https://www.space.com/europea-plumes-stealth-particles-mystery.html

- Intelligent Science Payload
- Onboard Processing
- Fault Detection and Mitigation
- autoNGC
- Communication Systems
- DSM Emulation
 - Environment
- System Engineering & MBSE

RAISR applied to DSM





Summary & Future Work



• Fault diagnosis is a necessary step toward

Resilient space flight (true autonomy)

Enabling next gen space flight, like constellation missions

• Future Work (DSM)

Algorithmic Thrust/TRL Boosting thrust Broadening RAISR to "system resilience"

References - Images

[MacGyverImg1] https://sharetv.com/shows/macgyver/episodes/284074 [MacGyverImg2] https://me.me/i/macgyver-making-the-junk-drawer-fun-again-since-1985icanhascheezburger-com-a5626df3836149ed923388c614003da3 [MacGyverImg3] http://www.fanpop.com/clubs/macgyver/images/41776422/title/1985-macgyver-photo [ANYmalBot] https://youtu.be/gM1z60aeunU

Thank you! Questions?