

Trusted AI Challenge

Phase II: Architectures and Approach

SERC Trusted Artificial Intelligence Systems Engineering Challenge,
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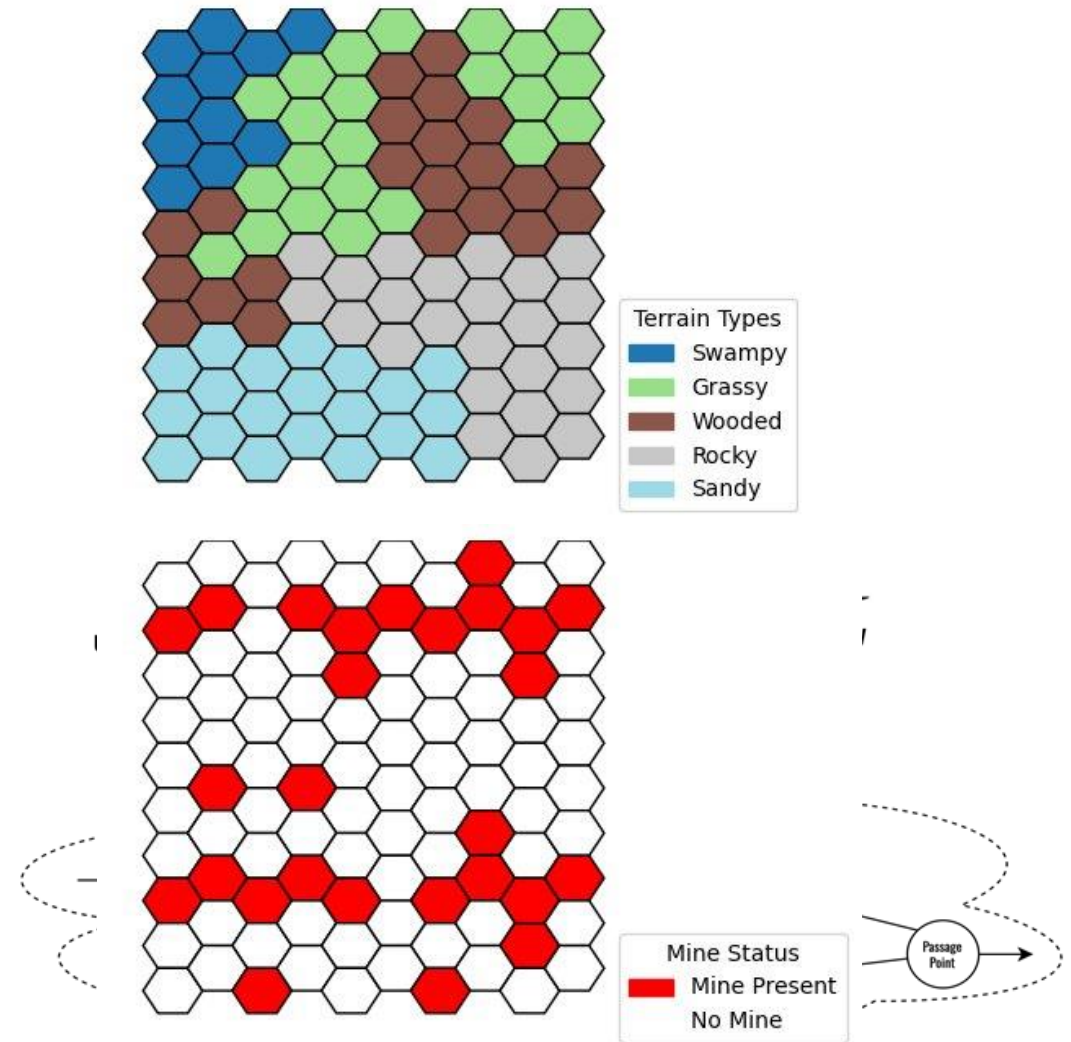
Design Engineering Laboratory @ Purdue
<http://engineering.purdue.edu/DELP>



Problem Statement

Objectives:

- Ensure safe passage from start node to terminus node along a defined network
- Develop the notion of trust between Artificial intelligence (AI) models and human operators
- Develop architectures to use human operators in conjunction with AI
- Identify the roles humans and AI should play in different architectures
- Compare different architectures across measures of performance



Characterizing Architectures

- We describe the architecture as the decision-making architecture for the task of navigating the network
- Each architecture is characterized by the flow of information and control instructions between the different stakeholders – who makes what decision? In which sequence?
- Communication between agents are of two types: information or control input



Images Captured



Updated Route



AI Confidence



Actual Terrain Info

Overall method

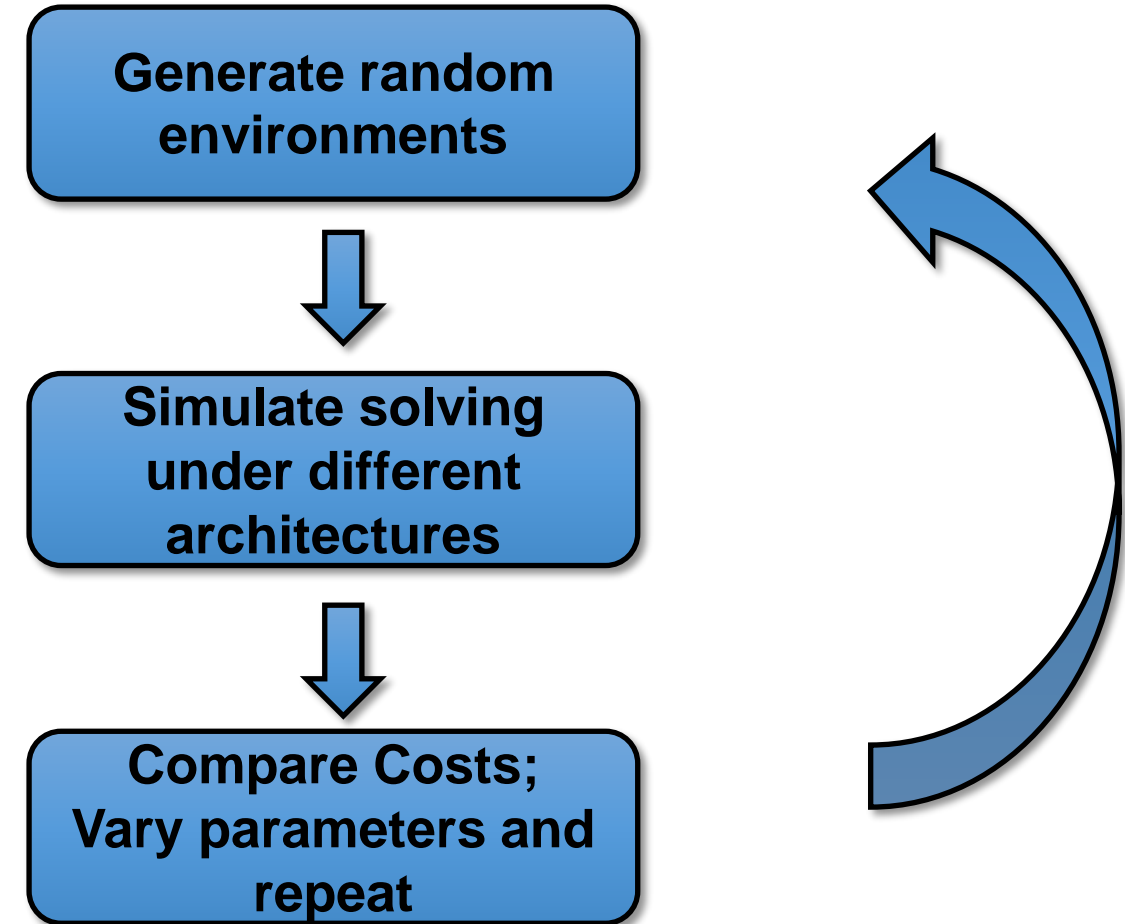
Architectures

A1: Centralized, Human

A2: Centralized, AI

A3: Centralized, Human-AI Teaming

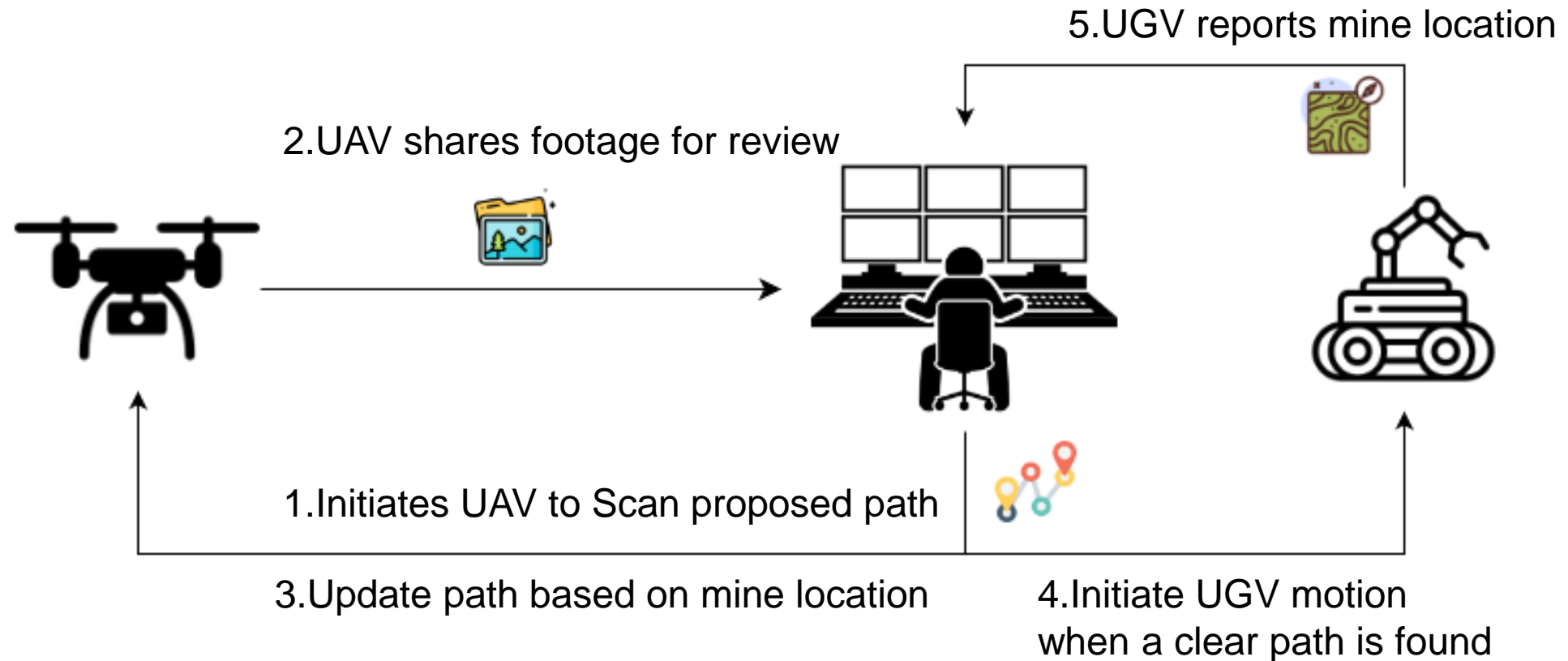
Architectures characterize the decision-making involved in navigating the terrain



How do we simulate and compare architectures?

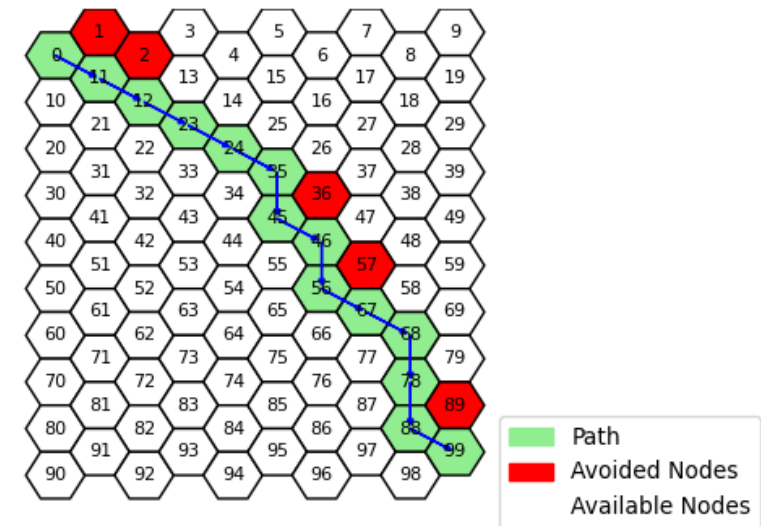
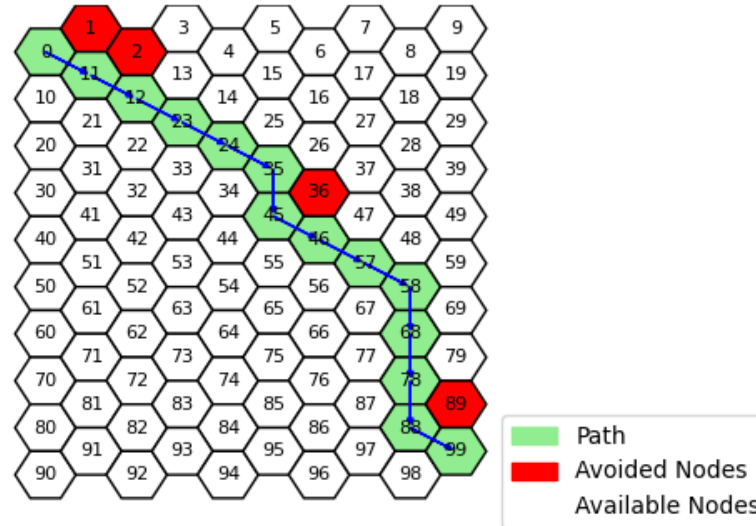
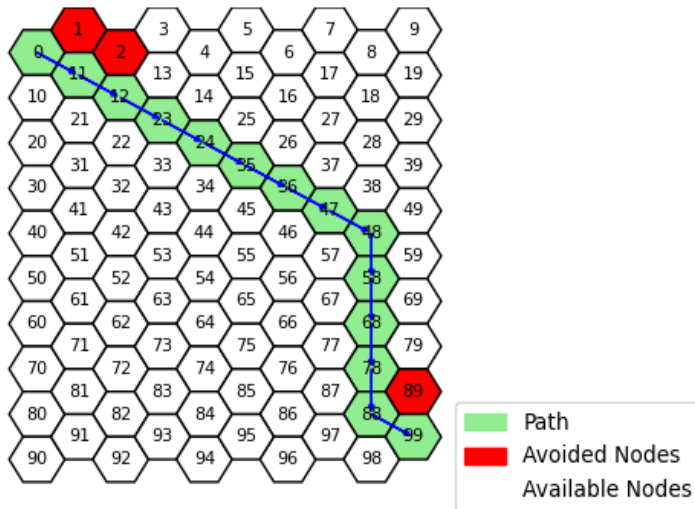
- Environment nodes are initialized with random terrain type, mine placement, start and end locations.
- Formalizing it as Markov Decision Process (MDP):
 - **States**
 - Final position (100x1)
 - Terrain classes (100x5)
 - UAV position (100x1)
 - UGV position (100x1)
 - UGV Trace (100x1)
 - Human responses (100x1)
 - AI responses (100x1)
 - **Actions**
 - 0-5: Move UAV
 - 6-11: Move UGV
 - 12: AI query
 - 13: Human query
 - **Reward**
 - UAV movement: -1
 - AI query: -5
 - Human query: -30
 - UGV movement: -20
 - UGV clearance: -60

Architecture A1: Centralized, Human

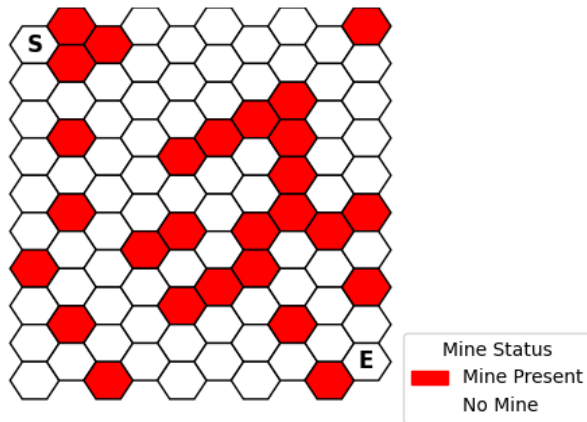
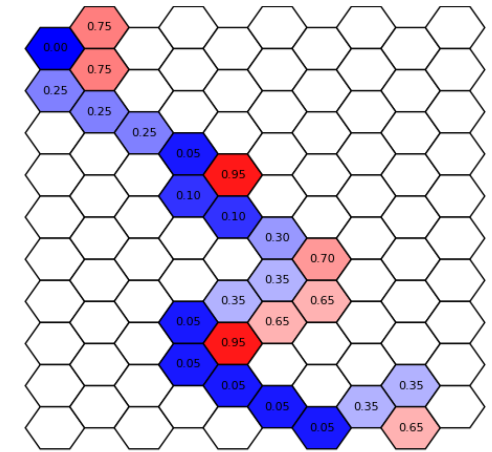
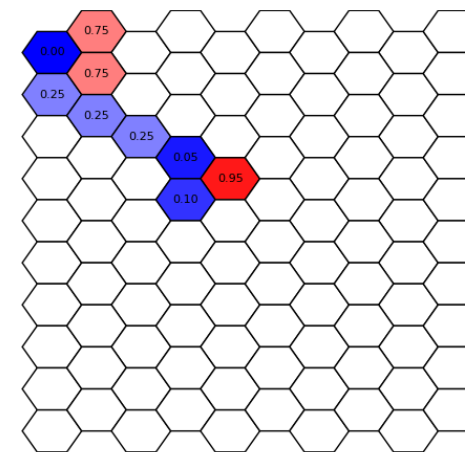
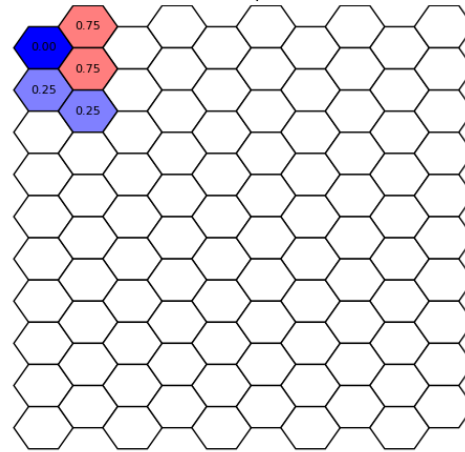
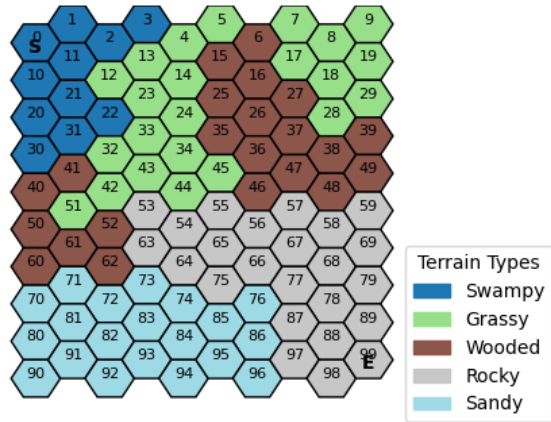


Path Planning Scheme

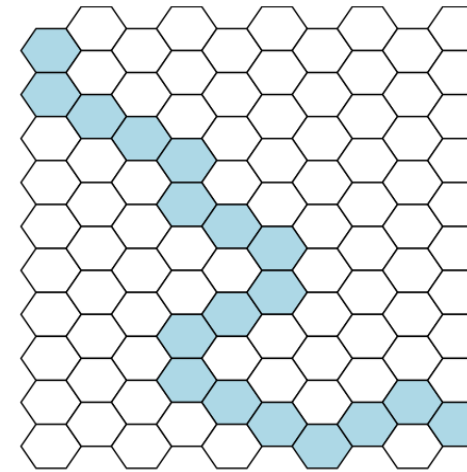
A* search algorithm is used to determine the path based on start node (current node), end node (final node) and a list of nodes to avoid (mines)



Architecture A1 – Centralized, Human

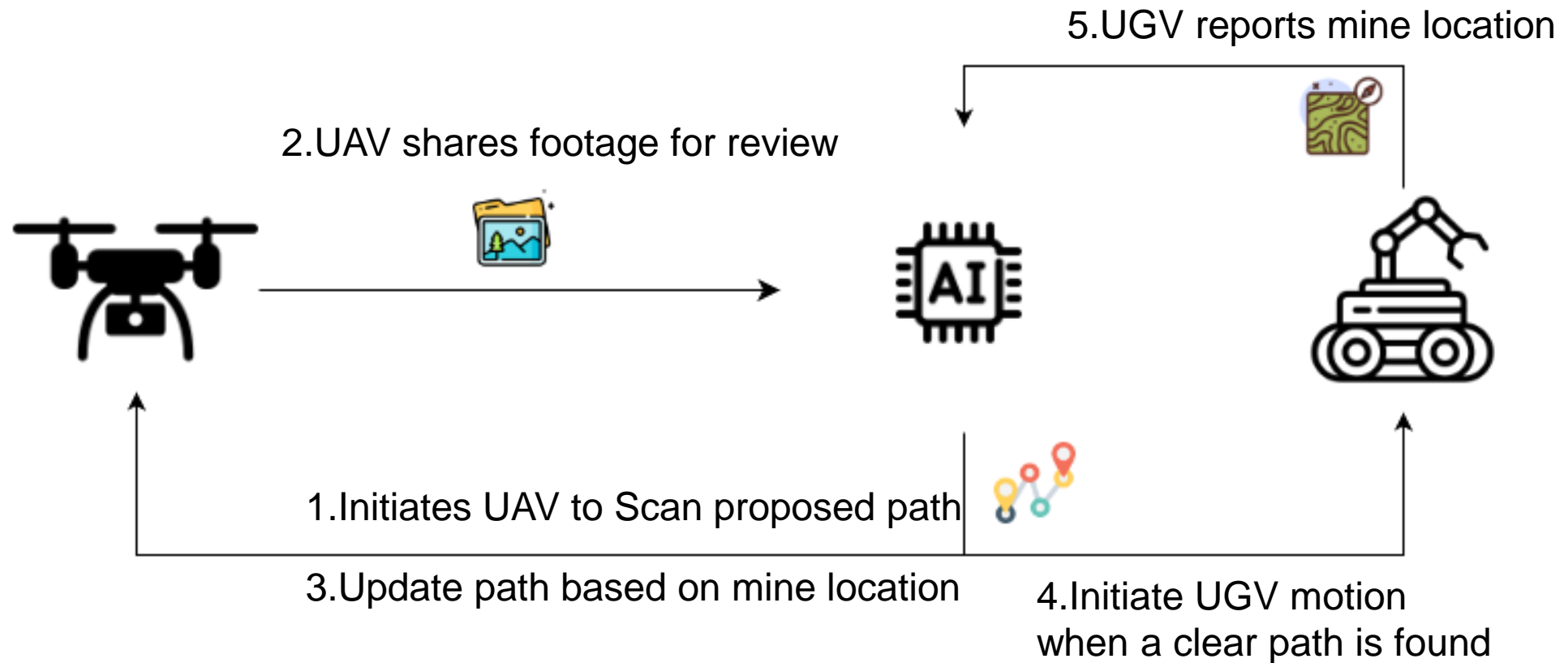


UAV Estimates

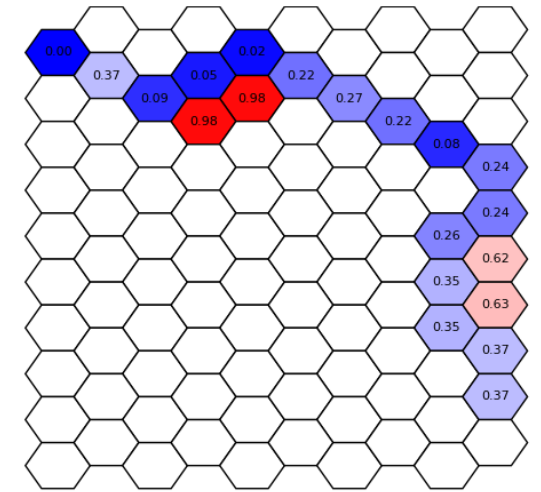
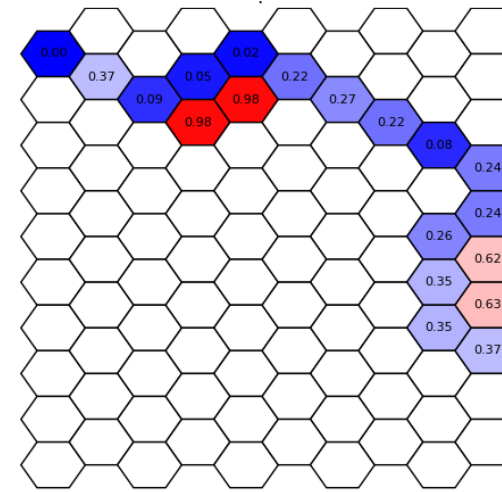
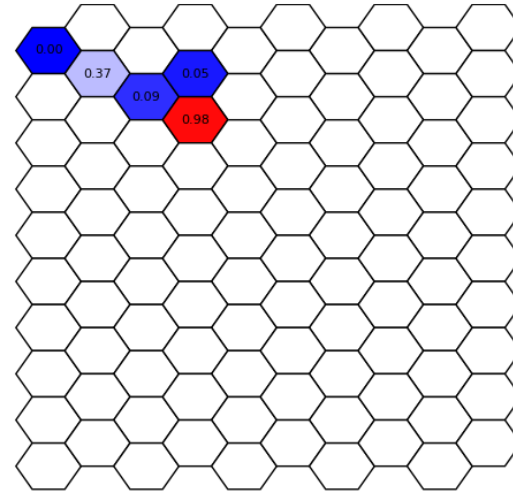
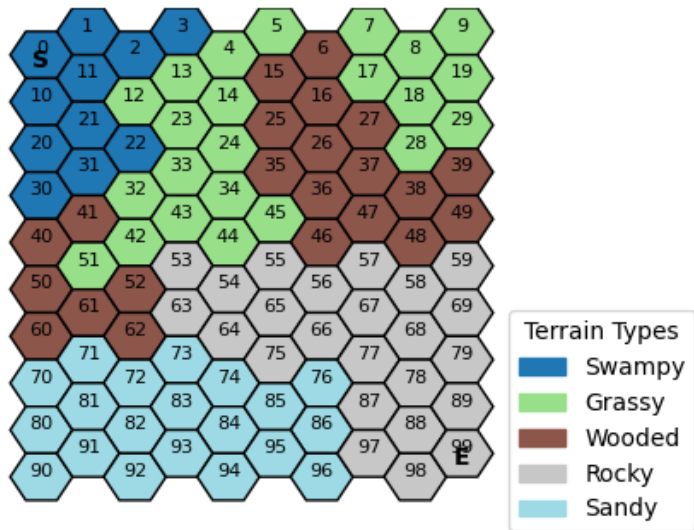


UGV Path

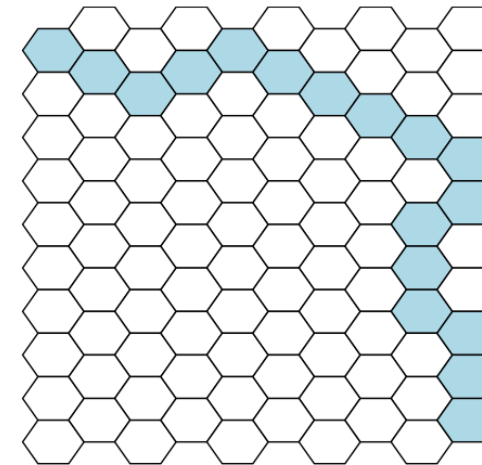
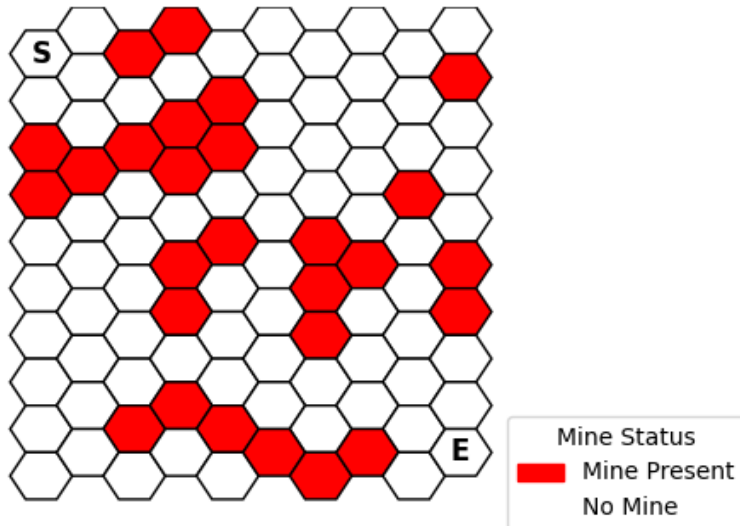
Architecture A2: Centralized, AI



Architecture A2: Centralized, AI



UAV Estimates

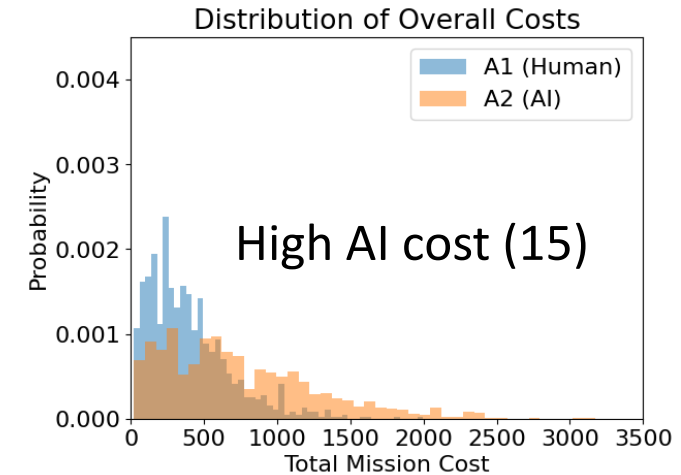
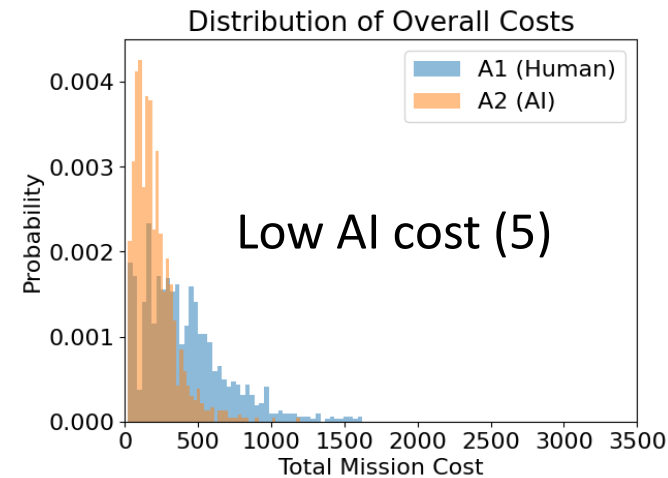
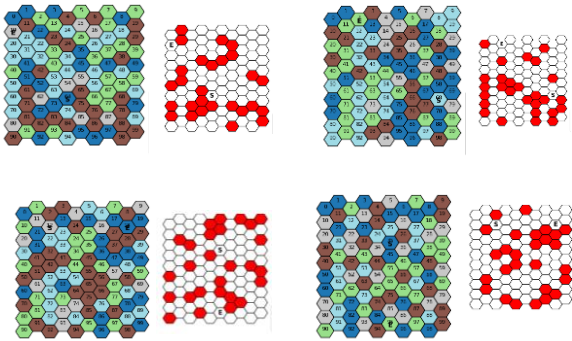


UGV Path

Comparison of A1 and A2

A1 is better when AI cost is high and A2 is better when AI cost is low.

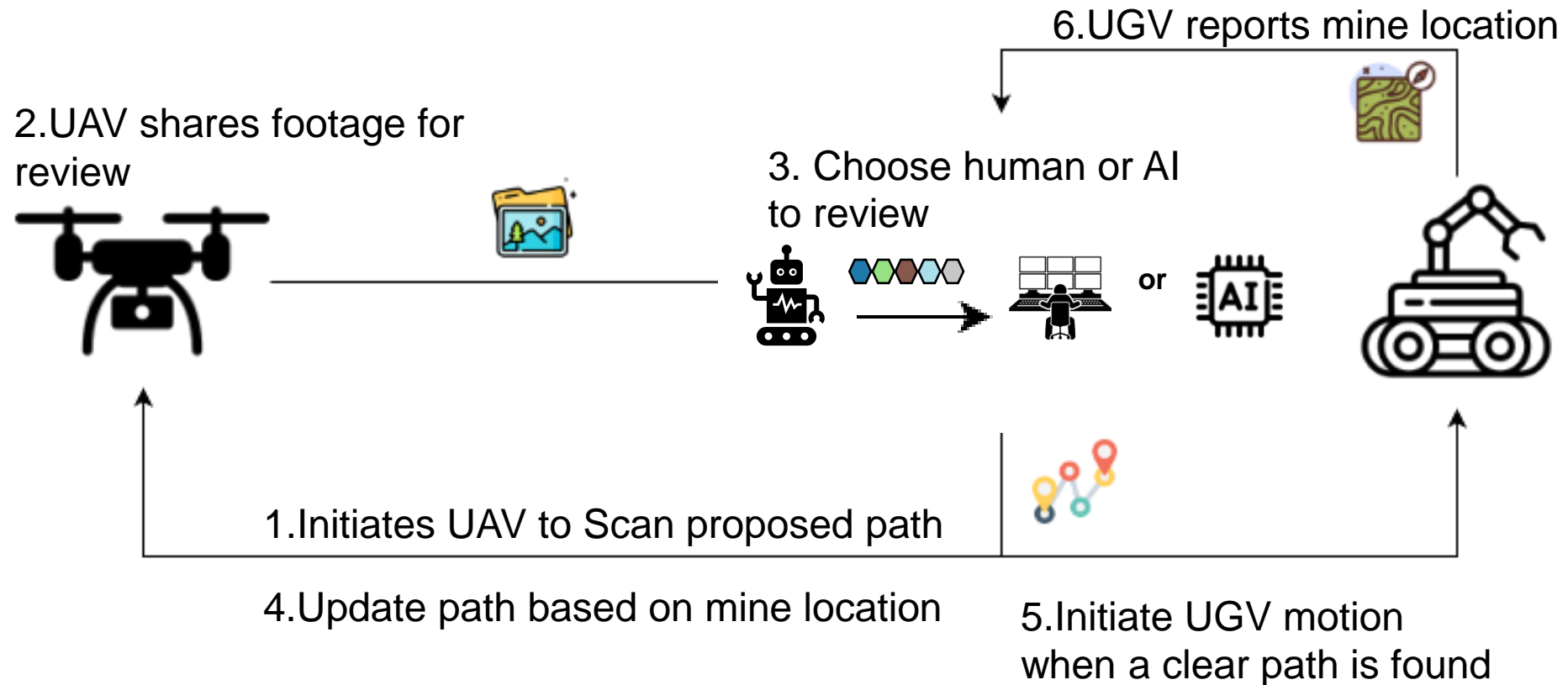
Test A1 and A2 on 10,000 randomly initialized missions



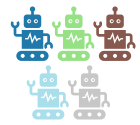
Can we learn when to trust AI or human based on the **terrain information** and **cost**?

Approach A3: Actions for querying are made by 5 multi-armed bandits (MAB) with two arms corresponding to AI and human

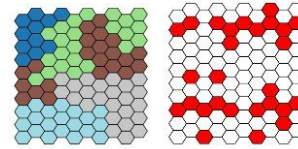
Architecture A3: Centralized, Human-AI Teaming



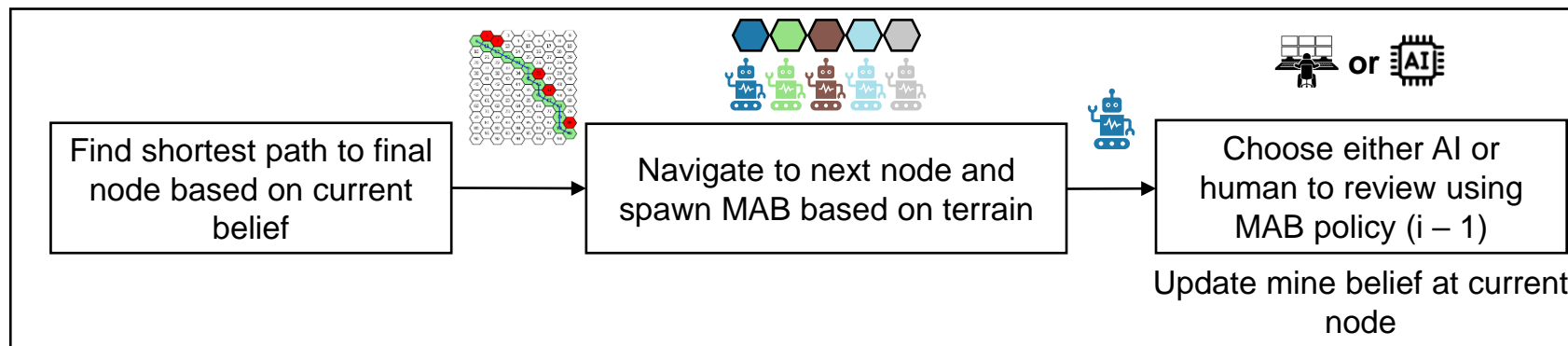
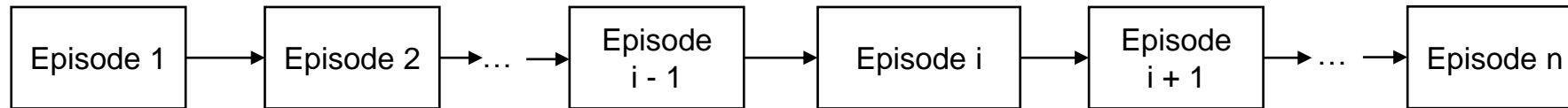
A3: Using AI4SE for SE4AI



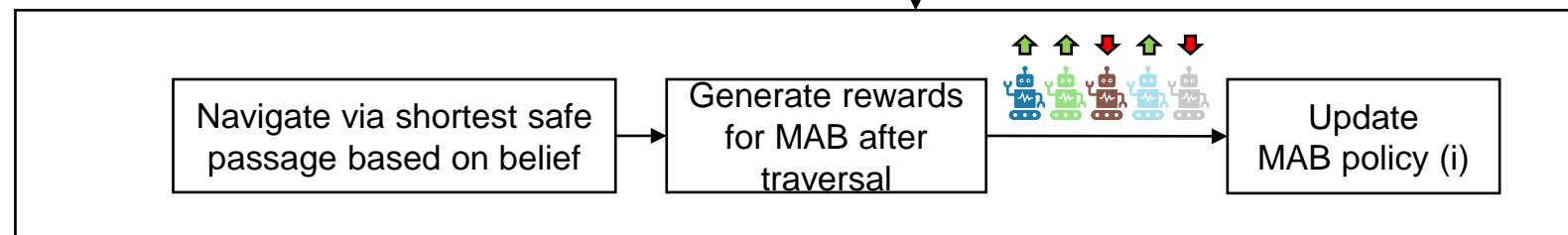
Initialize multi armed bandits (MABs) with a random policy



Generate simulation environment with randomized terrain and mine locations for each episode

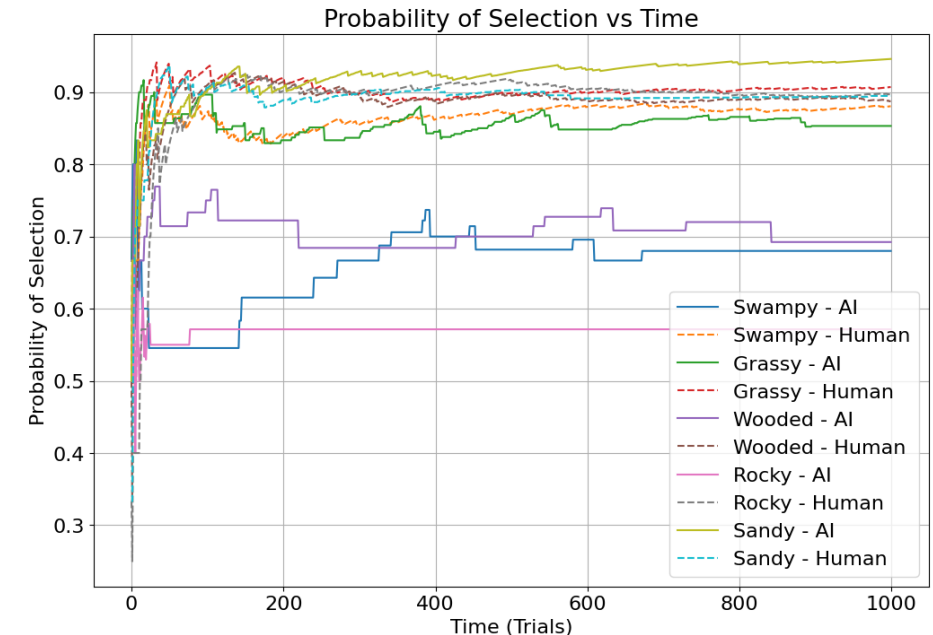
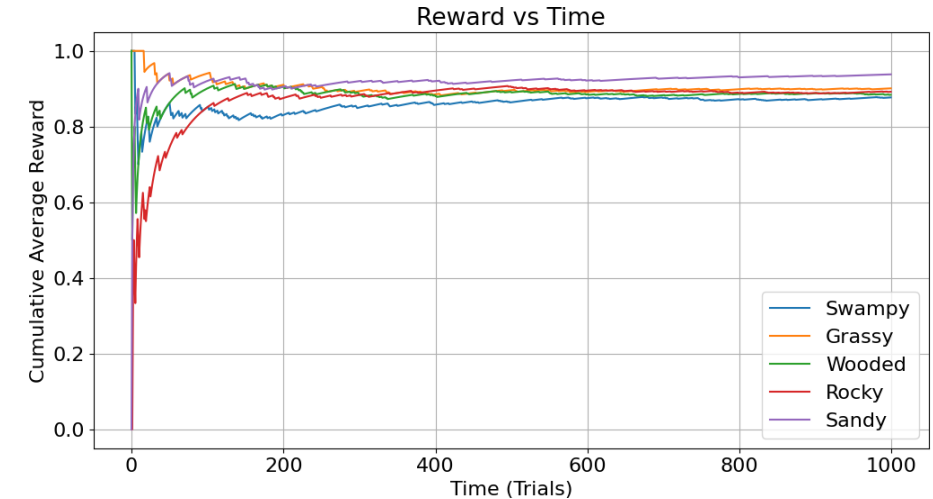
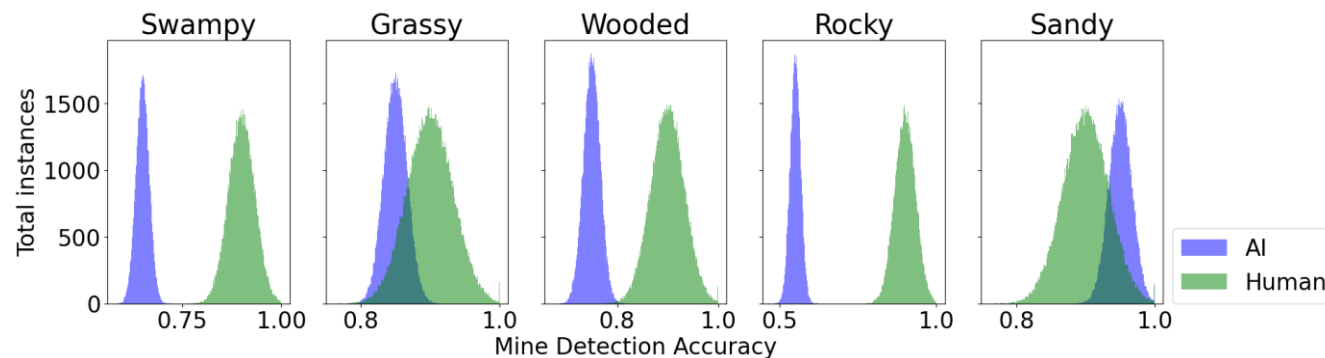


After UAV finds safe passage



A3: Centralized, Human-AI Teaming

- Actions for movement are predetermined by pathfinding algorithm (A^*)
- Actions for querying are made by multi-armed bandits (using UCB policy)
 - Total bandits: terrain types (5)
 - Arms corresponding to AI and human query (2)
 - $\text{Reward} = (1 - \lambda) \times \text{accuracy} + \lambda \times (-\text{normalised_cost})$
- Environment is sampled for 1000 episodes ($\lambda=0$)



A3: Heuristics for terrains

	Prioritize Accuracy ←		Prioritize Cost →	
	$\lambda = 0$	$\lambda = 0.33$	$\lambda = 0.67$	$\lambda = 1$
Rocky	Human	Human	Human	AI
Swampy	Human	Human	AI	AI
Wooded	Human	AI	AI	AI
Grassy	Human	AI	AI	AI
Sandy	AI	AI	AI	AI

$$\text{MAB reward} = (1 - \lambda) \times \text{accuracy} + \lambda \times (-\text{normalised_cost})$$

Results

- Successfully generated multiple simulations to test and validated three approaches
- **Centralized Human (A1):**
 - Effective for high AI costs
- **Centralized AI (A2):**
 - Superior when AI costs are low
- **Human-AI Teaming (A3):**
 - Leveraged reinforcement learning (RL) using multi armed bandits
 - Balanced accuracy and cost
 - Learnt heuristics for trust based on the terrain

Possible Extensions

- Account for changes in the operational environment during the task, limited bandwidth and errors/loss of information during communication
- Expand to decentralized architecture with a UAV AI
- Consider Lethality while making decisions
- Expand strategy to leverage multiple UAVs/UGVs and human movement along with UGVs
- Hardware implementation plan:
 - Set up a test network structure with obstacles simulating mines
 - Use UAVs to scan the network, with view obfuscated in parts
 - Benchmark the architectures in terms of expected time, computational power, bandwidth required
- Hardware
 - Turtle Bots
 - Drones
- Leverage existing facilities at Purdue for test and evaluation
 - Purdue UAS Research and Test Facility (PURT)



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