

Preliminary Modeling for Anticipating Impacts with Stakeholders for Participatory AI Projects



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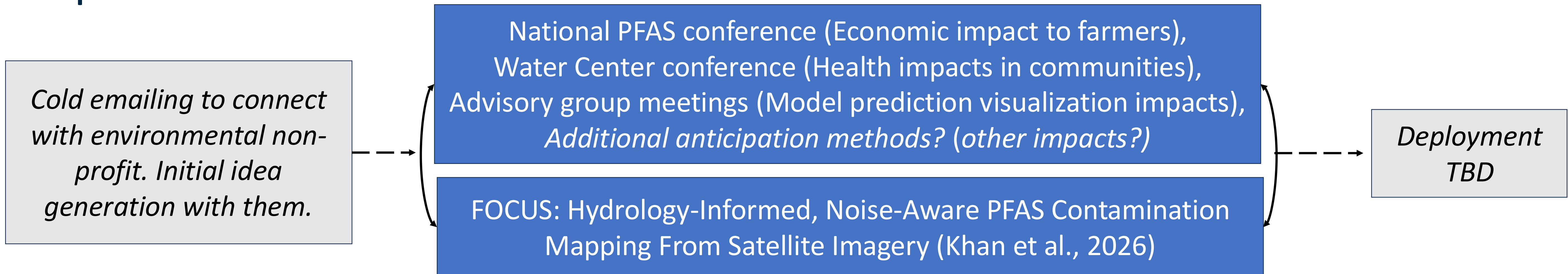
realize

Motivation



Key Question: How should we balance **resource constraints** and **discovering unknown impacts** in this loop?

Example: PFAS Prediction in Water Sources

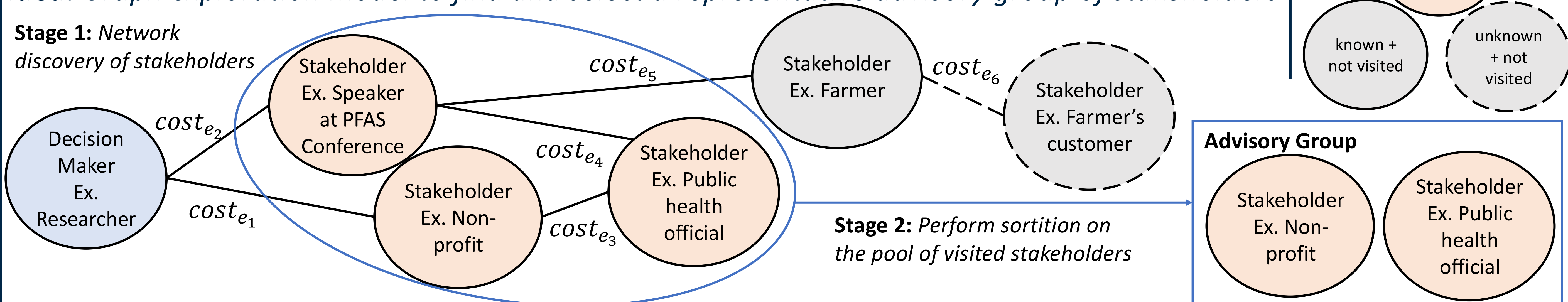


Khan, Jowaria, et al. "FOCUS: A Noise-Aware Geospatial Learning Framework for PFAS Contamination Mapping." 4th ICLR Workshop on Machine Learning for Remote Sensing (Best Paper), 2026.

Modeling Attempt and Pitfalls

Idea: Graph exploration model to find and select a representative advisory group of stakeholders

Stage 1: Network discovery of stakeholders



Pitfall: Estimating beliefs about impacts based on network position and defining an access mechanism to observe them **requires significant assumptions**

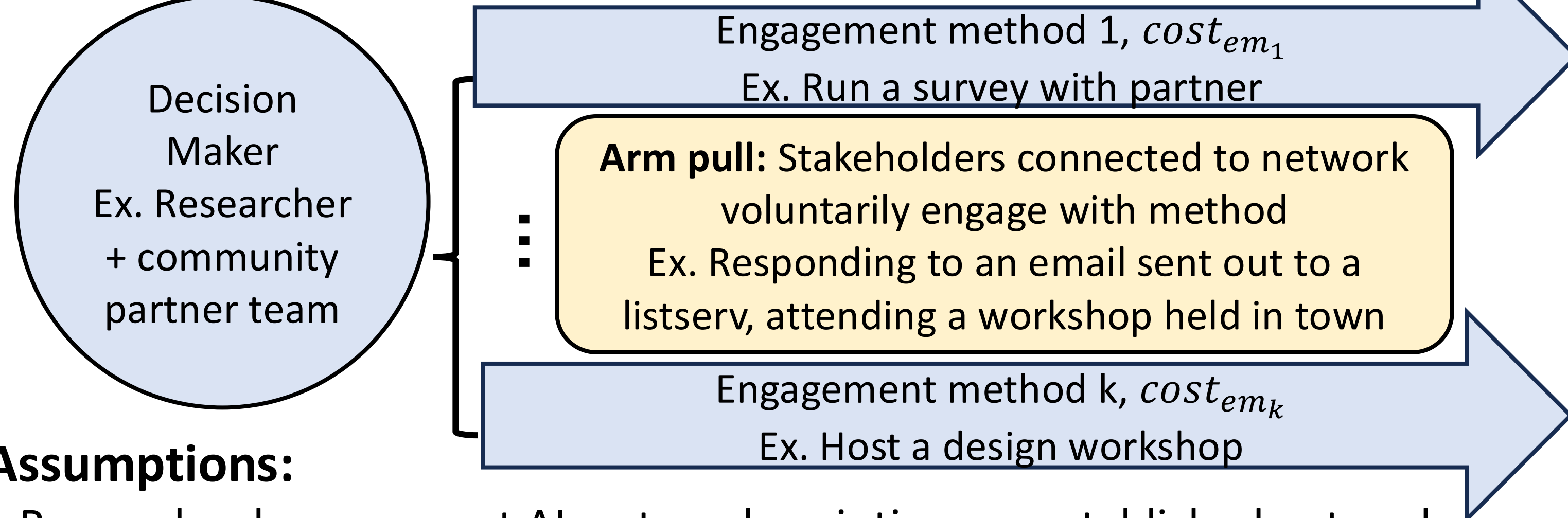
Pitfall: Approaching individual stakeholders to ask for their impact beliefs in exchange for a chance to be in an advisory group **imposes unnecessary burdens**

Pitfall: Defining and selecting a representative group **requires initial knowledge of what features can be found in the network**

Proposed Model and Open Questions

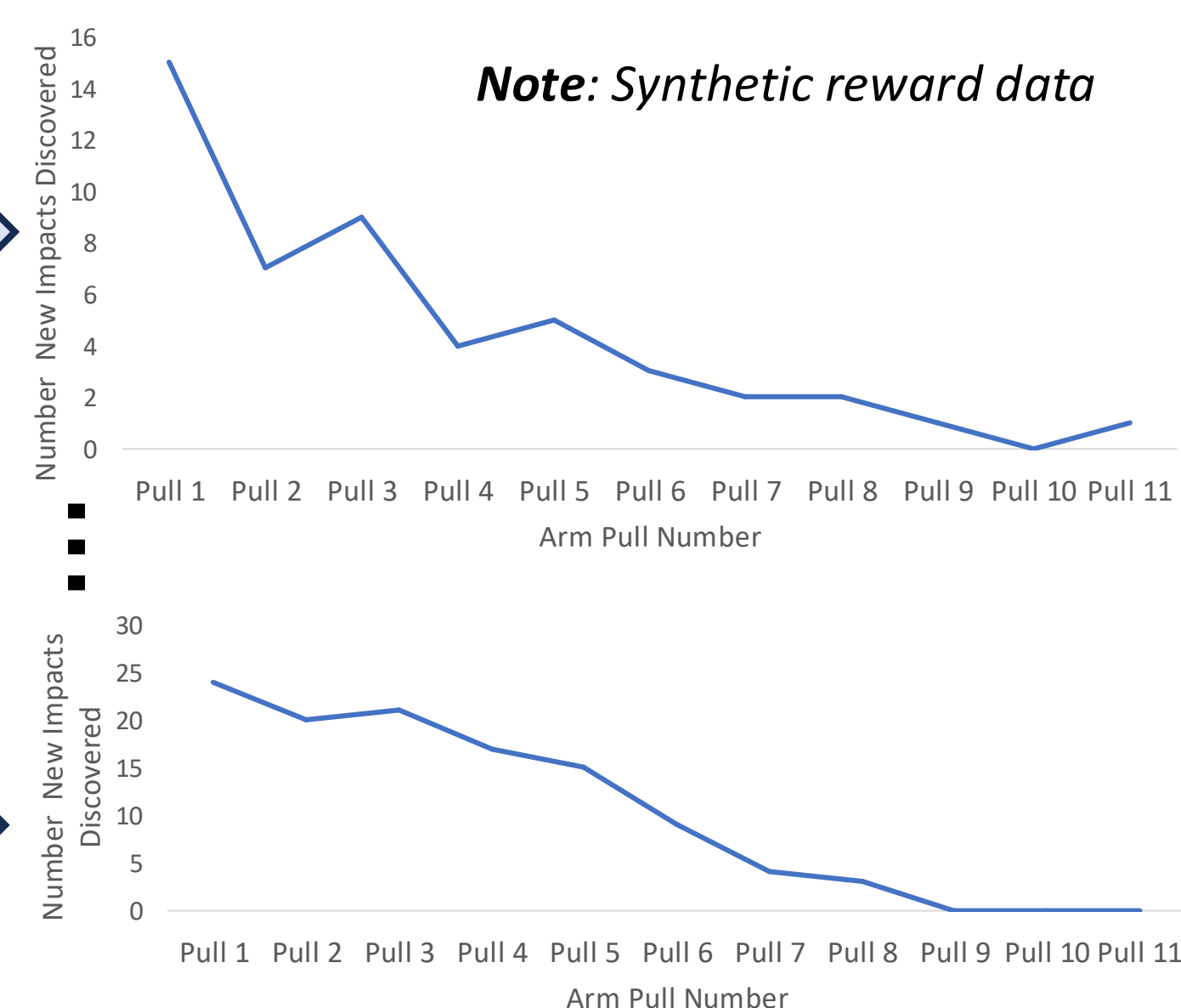
Idea: Bandit model for choosing stakeholder engagement methods to discover potential impacts with stakeholders connected to your established network of community partners

Learning Model: Stochastic non-stationary bandit with a knapsack



Assumptions:

- Researcher has a current AI system description, an established network of partner organizations, and an established set of engagement methods
- Number (and content) of potential impacts is *initially unknown* and there are *finitely many*
- For each engagement method, em , there is a (latent) fixed probability of independently discovering each impact. The (latent) joint distribution over discovering any set of impacts is denoted Θ_{em} .



Rewards: Each pull of an engagement method draws a sample from Θ_{em} . Given assumptions, number of new impacts discovered *decreases in expectation* with additional samples.

Data Requirements:

- Assume natural language feedback from each engagement method can be coded (e.g., researchers and community partners come to a consensus) to categorize potential impacts
- Cost of each engagement method is known
- Fixed budget constraining number of arm pulls is known

Open Questions:

- **Reward functions related to saturation**
 - Framework is analogous to *new species discovery in ecology*, can we adapt the statistical estimates from this literature in a *decision-focused learning approach* to estimate the reward functions while learning which arm to pull?
- **Limited arm pulls and potentially delayed feedback**
 - Can we *work with community partners* to define correlations between arms or otherwise provide the model with prior information or additional signals about rewards?