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Going Beyond Mono-Mission Earth Observation: Using the Multi-Agent Paradigm to Federate Multiple Missions Blue Sky Ideas

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AAMAS 2024

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Federated Observations
Federated Communications
Wrap-up

Earth Observation Ecosystems Evolution

- More advanced instruments and satellites
- Growing needs for higher responsiveness [EUSPA, 2022]
- E.g. order surveillance, maritime monitoring, and disaster response
- From mono- to multi-missions
- Ground Segment as a Service (GSaaS)







Case Study: Reactive Large Area Acquisition







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Optimize acquisition time of large areasOptimize download time of acquisitions



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Case Study: Reactive Large Area Acquisition

Optimize acquisition time of large areas
Optimize download time of acquisitions
⇒ Multiple missions and ground stations





















Client K















DOMINO-E Approach

Add a Federation Layer to Coordinate Multiple Systems







DOMINO-E Approach (cont.)

Add a Federation Layer to Coordinate Multiple Systems







Pederated Observations

"How to divide a large area and to assign these subdivisions to different missions in order to minimize time of acquisition and maximize the quality of the images, even though we do not know the future workload of missions?"







Federated Observations

System Composition

- *K* end-users, referred to as *clients*, who request images over large areas on the Earth's surface
- *N* observation agents, referred to as *missions* with own *planners*
- Objective = complete the observations requested by the K clients using the N observation missions available
- → Dispatch observation tasks to missions than merge results
- Seamless access to numerous resources without having to care about details







Federated Observations

Multi-agent Decisions and Interactions

- Multi-agent negotiation [KRAUS, 2001]
 - Send queries for sub-areas to missions (that could accept or reject)
 - Each sub-area is allocated to a specific mission
 - The response time of a mission is high during the negotiation part
- \Rightarrow Surrogate model to assess the impact of including obs into mission schedules [Tull et al., 2022]

• Combinatorial auctions [CRAMTON et al., 2006]

- Limit the number of interactions between the agents
- · First phase: the missions to bid on possible observation sub-areas
- Second phase: observation tasks are dispatched given the bids received
- 1 bidding system for each mission is time-consuming
- ⇒ The federation skips the bidding phase and directly allocates observation tasks based on its current knowledge of the capacity and load of each mission (a surrogate model)





Federated Observations

Challenges

Highly combinatorial problems

- Hundreds or thousands of requests, tens or hundreds of satellites, numerous ways to partition the area of each request into a set of sub-areas, etc.
- Study both coarse-grain dispatching strategies and fine-grain dispatching strategies
- Multiple objectives
- Compatibility between the requirements associated with each request and the capabilities of the missions







Federated Observations (cont.)

Challenges

– Need for a model of the other agents -

- The federation must handle a model of the current capabilities of each mission
- Can be imprecise, especially for *external missions*
- Even for the *legacy missions*, the federation layer does not have a full control
- \Rightarrow Learn a high-level model of the capabilities
 - Exploiting a model of the density of the high-priority requests

How to assess the load of each mission? How to predict a mission response?







Federated Observations (cont.)

Challenges

Uncertainty management

- Various sources of uncertainty
- e.g. Clouds lead to failed images (approx. 50%) [Hadj-Salah et al., 2019]
- How to exploit short-term meteorological forecast
- How to exploit historical weather data
- ⇒ Learn a model of the *long-term reward* provided by a dispatching decision
- ⇒ Re-dispatch online to automatically update the coverage strategy based on the actual execution status







"How to assign bundles of communication windows to satellites in order to meet data flow requirements, to minimize jamming, and to minimize costs induced by booking services?"







System Composition

- N satellites (held by mission agents), seeking for communication windows for data transfer
- M independent communication site agents, referred to as *sites*, implementing GSaaS interfaces
- · Several contacts (communication opportunities) for each satellite-station pair
- Objective: allocating contacts to satellites of federated missions (optimizing costs and jamming)
- Pre-defined to respected when interacting with each GSaaS provider, based on an agreed SLA







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Multi-agent Decisions

Multi-Agent Resource Allocation [CHEVALEYRE et al., 2006] -

- Allocating communication windows (contacts) to satellites for each request
- Each allocation generates non additive costs
- Each allocation generates non additive jamming

Multi-Agent Planning [SHOHAM and LEYTON-BROWN, 2008]

- Planning satellites' communication activities
- Under visibility constraints
- Joint decisions may lead to jamming

Jamming model is unknown (depends on external assets)

Loads of some stations are unknown (GSaaS)





Challenges

Highly combinatorial problems

- Large number of satellites and contacts
- Long planning horizon (> week)
- Multi-objective
- Externalities: individual utility strongly depends on other agents' bundles
- ⇒ Dedicated solution methods have to be devised







Federated Communications (cont.)

Challenges

Need for a model of the other agents -

- The federation must handle a load model of each GSaaS
- The federation must handle a jamming model for each pair of satellites
- ⇒ Learning the probability of a contact request being rejected
- \Rightarrow Learning the jamming model
 - Handle strategic behaviors
 - Multi-Agent Reinforcement Learning (MARL) problem [ALBRECHT et al., 2024]

How to assess the load of each GSaaS? How to predict a GSaaS response? How to assess jamming?









- Key terms for NewSpace: multi-asset, multi-user, multi-system...
- Asset sharing means cost-efficiency, but requires automated coordination and privacy/sovereignity preservation







Wrap-up(cont.)

- How to coordinate such composite systems?
 - Efficiency
 - Fairness
 - Explainability







Wrap-up(cont.)

- How to coordinate such composite systems?
 - Efficiency
 - Fairness
 - Explainability
- Multi-agent Systems
 - Resource allocation and combinatorial auctions
 - Distributed optimization
 - Federated and multi-agent learning
 - ...





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Acknowledgements

Part of this work has been performed within the DOMINO-E project which received funding from the European Union's Horizon Europe Programme for Research and Innovation under Grant Agreement n°101082230.









Thank you for your attention! Any question?

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