

# Surveillance strategies for autonomous mobile robots



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# Intelligence, surveillance, and reconnaissance (ISR) with autonomous UAVs



- ISR defines a class of real world applications where robots can be effectively employed
- UAVs are an emerging technology in this field:
  - Large number of real world deployments
  - High level autonomy is an open challenge
- Our reference application: cooperative surveillance by a team autonomous UAVs
- Key feature: robots have faulty sensors to do detect attacks

# Background and motivations

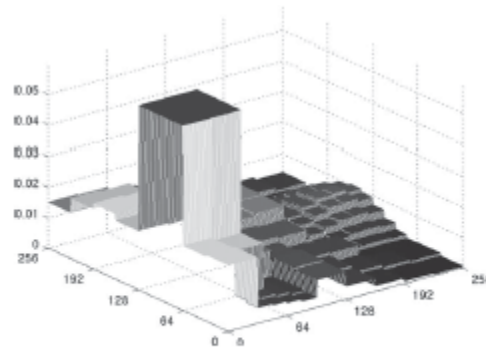
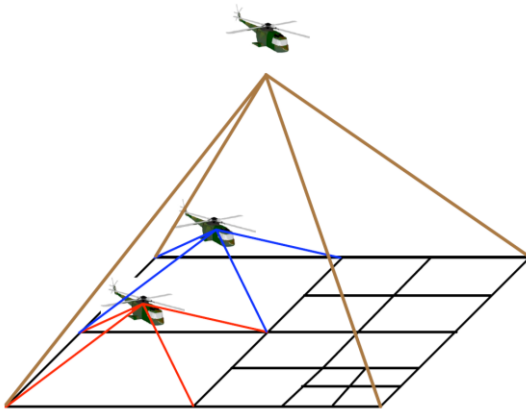
- One central problem in autonomous surveillance is the definition of a **surveillance strategy**: how to schedule environmental inspections in space and time
- It's an online problem: where to inspect next in general depends on what has been seen so far
- It has theoretical roots in search theory [Koopman, 1956], [Stone, 2007]
- It needs for practical and scalable methods that can cope with features of real world robotic settings

Decision theory

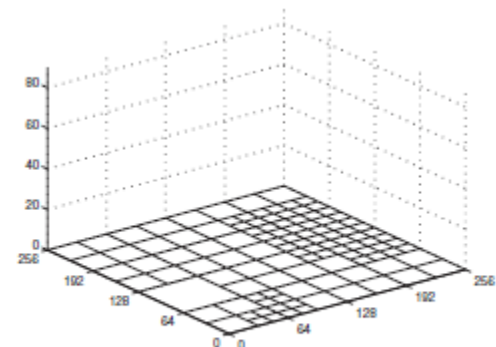
Game theory

# Background and motivations

- Variable resolution environment representations [Carpin et al 2011, 2012]:
  - Robots use a multi-scale, dynamically maintained, quadtree representation of the environment
  - More efficiency with small losses of performance



(b) PQ prior with 150 nodes



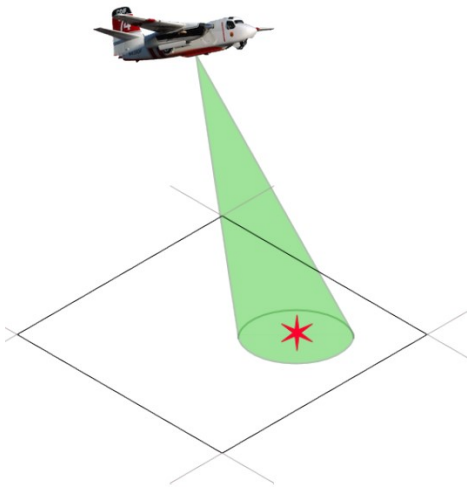
(c) Quadtree with 150 nodes

# Background and motivations

- Two-level robot coordination [Basilico et al., 2014]:
- UAVs are assigned roles:
  - Sentinels undertake broad-area surveillance
  - Searchers perform triggered local inspections

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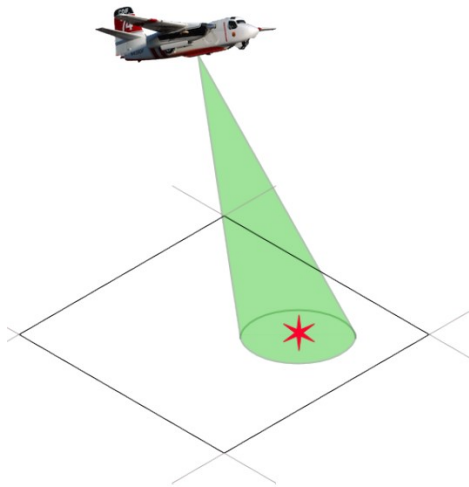
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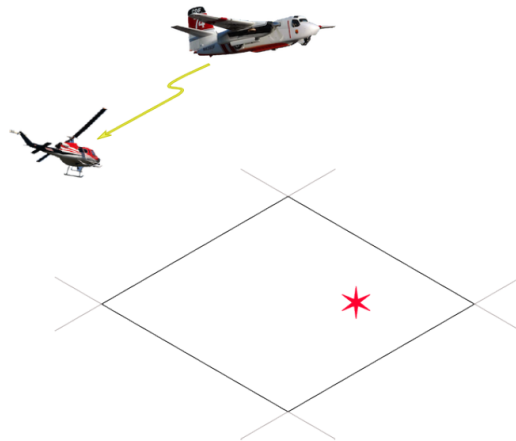
Broad area surveillance:  
detection

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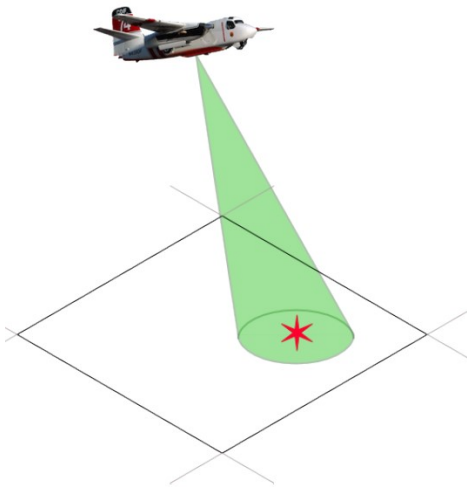
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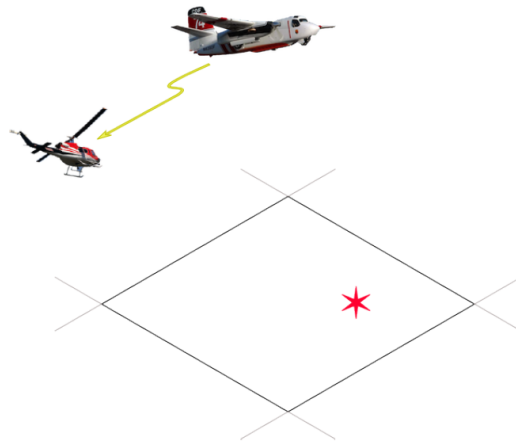
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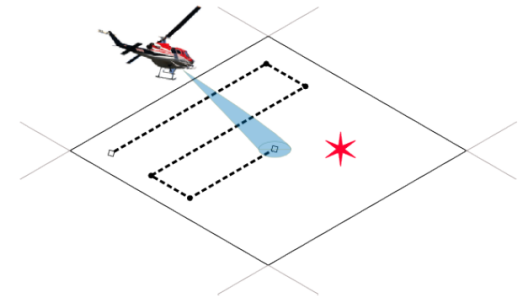
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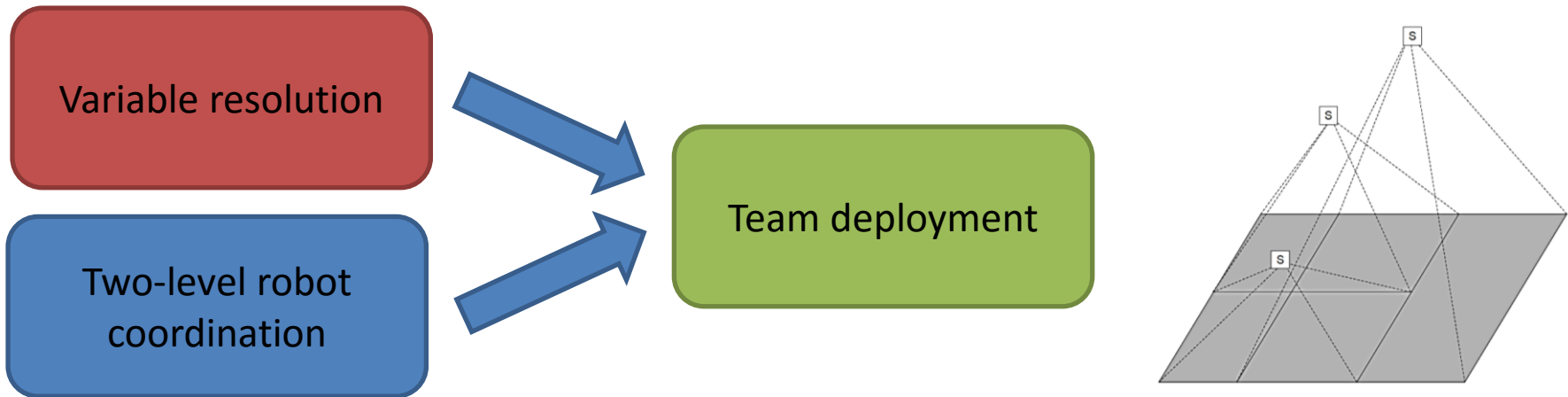
Broad area surveillance:  
dispatch



Local inspection



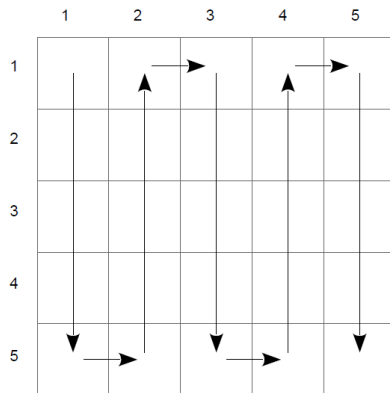
# Background and motivations



- Assign each sentinel a sub-area of the environment with the objective of:
  - Better distribute the common effort over the whole environment
  - Exploit synergies given by overlaps in the most sensitive regions
- Problem 1: given a candidate deployment, how to measure its goodness?
- Problem 2: how to efficiently search for an optimal deployment?

# Evaluating deployments

- Environment discretized in cells, each cell  $c$  has a loss  $l(c)$  (the higher the loss the higher the damage per time unit of having an attack there)
- Attacks follow a probabilistic behavior
  - Spatial: proportional to the loss
  - Temporal: exponential arrival times
- Once dispatched, a searcher will perform a “lawn mower” pattern over its assigned sub-area



If sentinel is placed at position  $s$ , the expected total loss from a cell  $c$  is

$$v_s(c) = \mathbb{E} \left[ l(c) \int_0^T a(c, t) dt \right]$$

Loss of cell  $c$

Attack function: equal to 1 if cell  $c$  is attacked at time  $t$  (unknown!)

Mission time

# Evaluating deployments

- We can provide an upper bound to the expected loss by combining the following expected quantities:
  - Time between attack arrival and scan in that cell
  - Num. scans to be performed before dispatching a searcher
  - Num. of searchers dispatches to be performed before the successful one
  - Time to detect the attack by the successful searcher

$$v_s(c) \leq W_c(\mathbb{E}[\zeta] + \mathbb{E}[t_{tr}^1] + (\mathbb{E}[n_s] - 1)\mathbb{E}[t_{tr}] + \mathbb{E}[s])$$

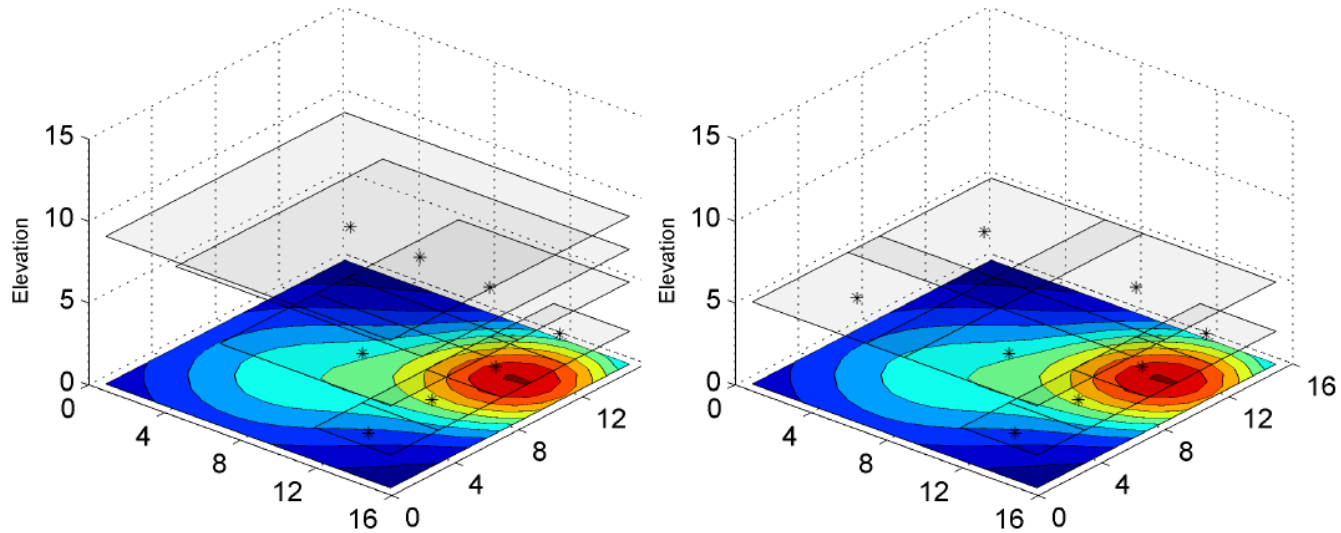
- Each cell  $c$  can be covered by multiple sentinels: we take the tightest UB (min)
- The whole performance in the environment  $P$  is given by the largest UB (max)
- We aim at minimizing  $P$

# Searching for optimal deployments

- We devised an iterative algorithm searching in the possible space of feasible deployments for  $M$  sentinels
- Tradeoff between exhaustive search (  $\rho = 1$  ) and greedy construction (  $\rho = M$  )
- Contrarily to what intuition would suggest submodularity cannot be exploited to provide quality guarantees to the greedy method

# Results

Example with 8 sentinels (red regions have higher losses)



(a)

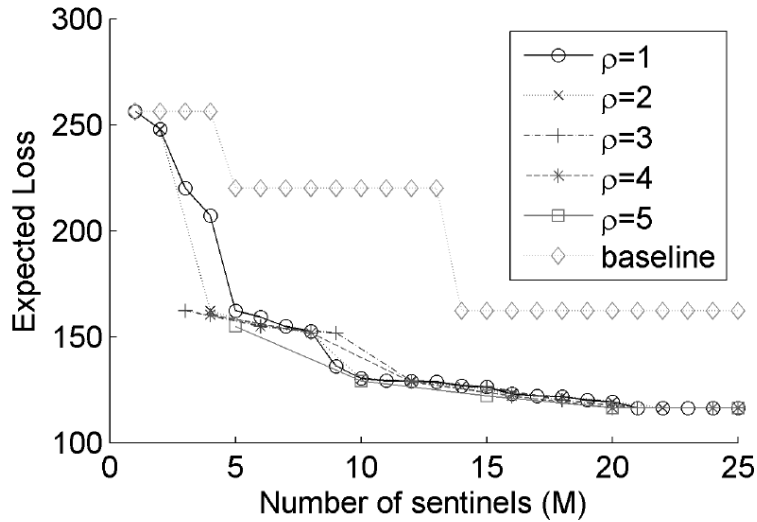
Greedy deployment

(b)

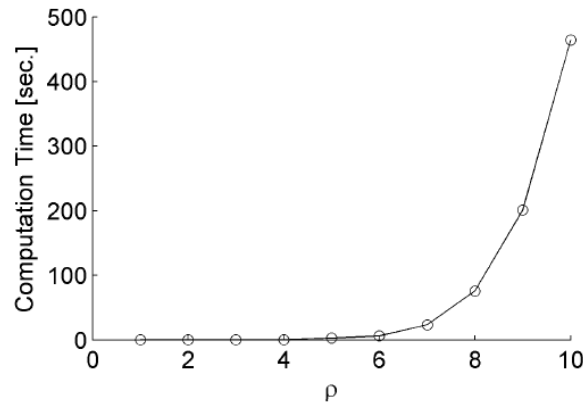
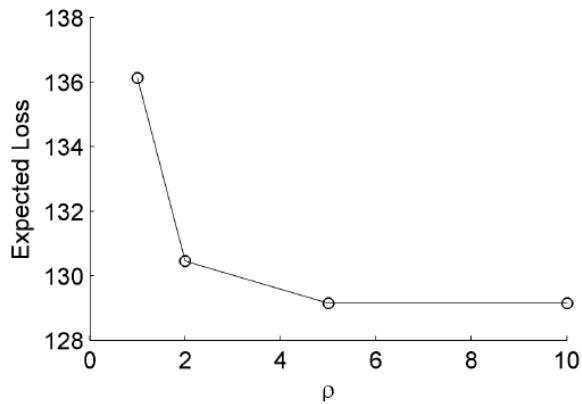
Optimal deployment

The optimal deployment better exploits overlap of different sentinels, also sentinels occupy lower positions (less energy required)

# Results



Loss reduction as the number of sentinels increases: the choice of deployment can be critical



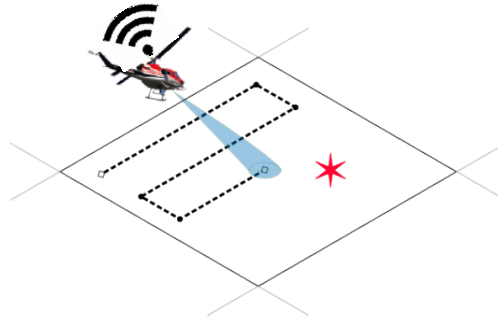
Quality vs computation time tradeoff of our method

# Remarks

- Team deployment is a critical issues besides the definition of good surveillance strategy
- Seeking effective and informed effort distribution can provide advantages during the on-line execution of surveillance missions

# Surveillance in environments with restricted communication

- Searchers must promptly report what they find



- In communication-restricted environments some works (e.g., [Mosteo et al., 2009, DARS]) try to maintain a multihop network



# Surveillance in environments with restricted communication

Idea: exploit an existing communication infrastructure ([Tortonesi et al., 2012, IEEE Commun Mag] , [Ochoa and Santos, 2015, Inform Fusion]) providing partial coverage to the environment to make reports to a Mission Control Center (MCC)

Given some locations to monitor, two objectives:

- 1) Maximize the frequency of visit to the locations
- 2) Receive periodic and fresh reports at the MCC



# Situation Aware Patrolling (SAP)

- Given a discretized environment  $G=(V,E)$  (weighted, metric)
  - $V_m$ : locations to be monitored
  - $V_c$ : locations with available communication

and:

- K UAVs
- a time budget T
- a starting depot  $d \in V_c$
- Determine K cyclic walks of length  $\leq T$  on G minimizing the **average communication latency** of *m-type* vertices.

NP-hard and inapproximable

Inspection (by any robot)

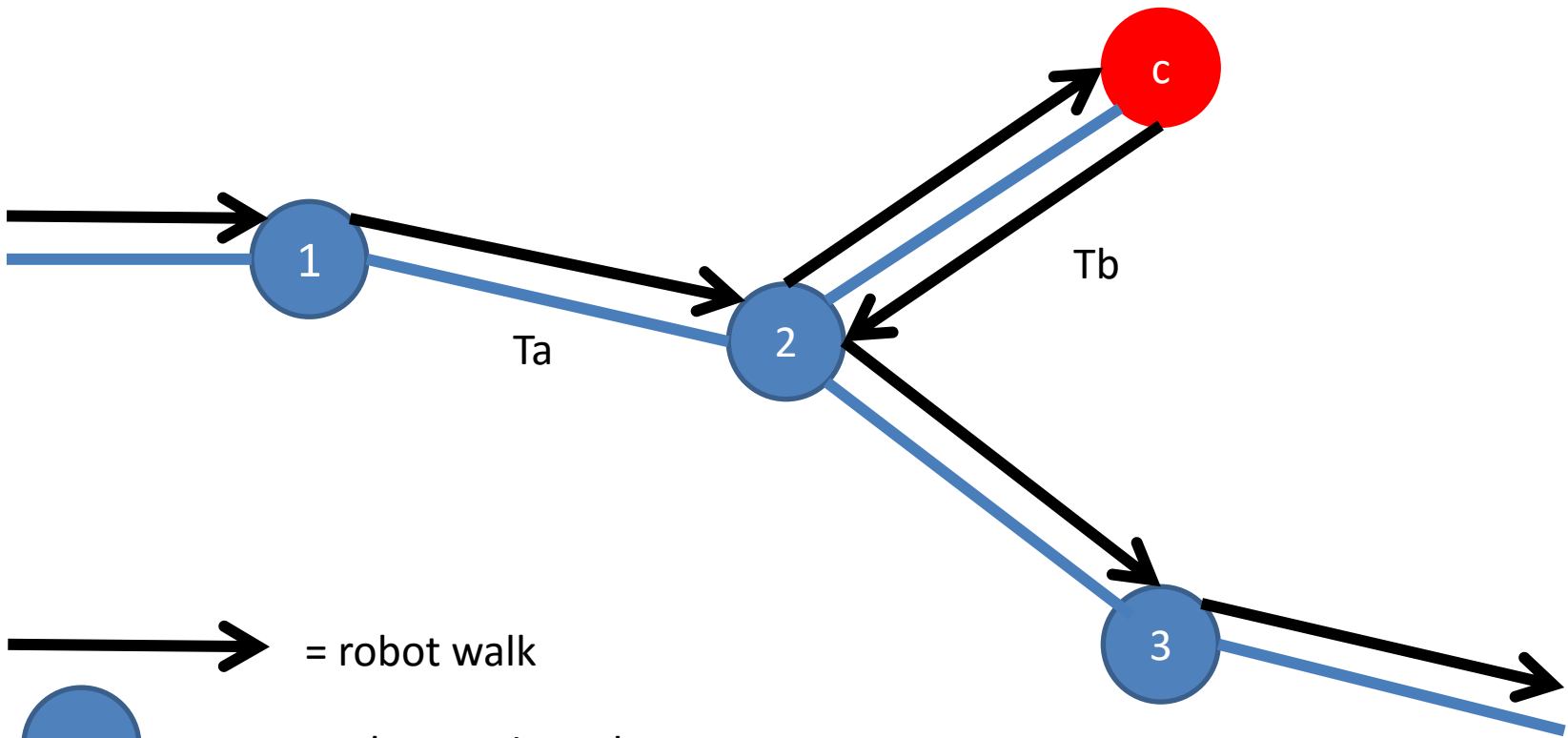
Passage by a c-type vertex (by any robot holding the information)

Communication latency

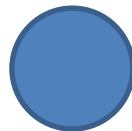
Time



# Simple example



 = robot walk

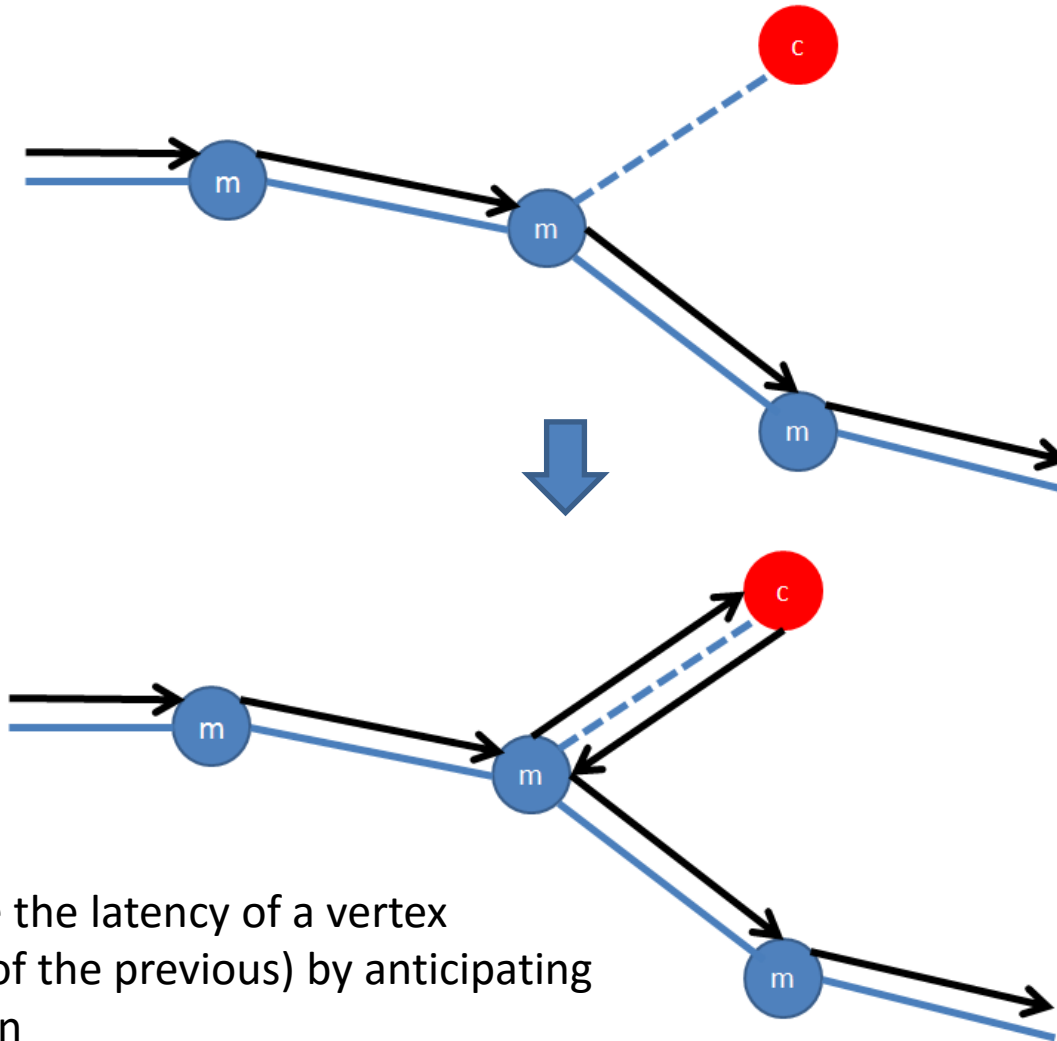
 = vertex to be monitored

 = communication vertex

# Resolution methods

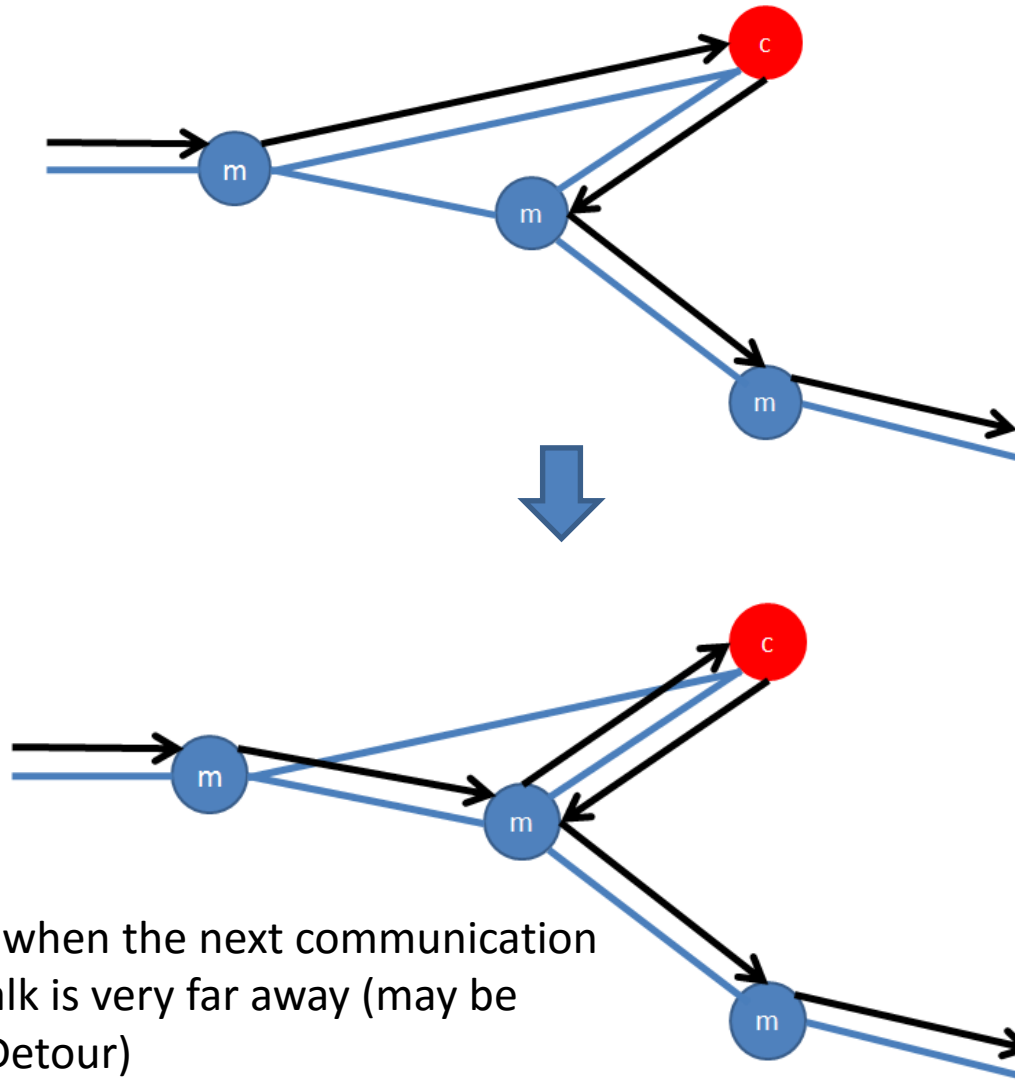
- Heuristic algorithm:
  - Start from a feasible solution of  $K$  walks (e.g., use Frederickson's heuristic for the  $k$ -TSP) consuming as few budget as possible
  - Iteratively apply local modifications until some budget is available

# Modifications - Detour



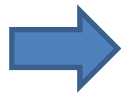
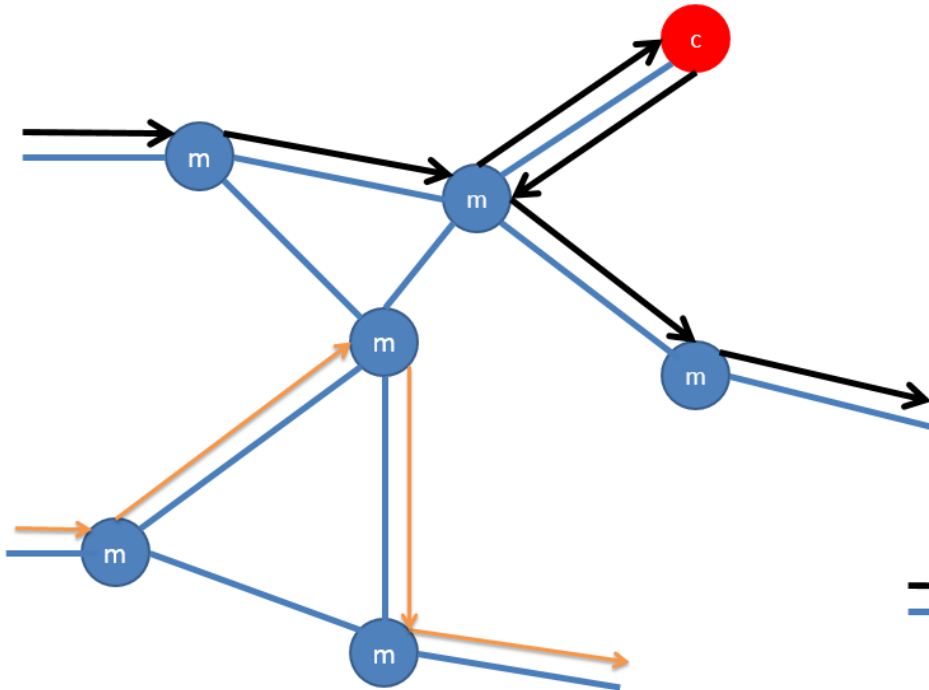
Idea: decrease the latency of a vertex  
(and possibly of the previous) by anticipating  
communication

# Modifications - Shift

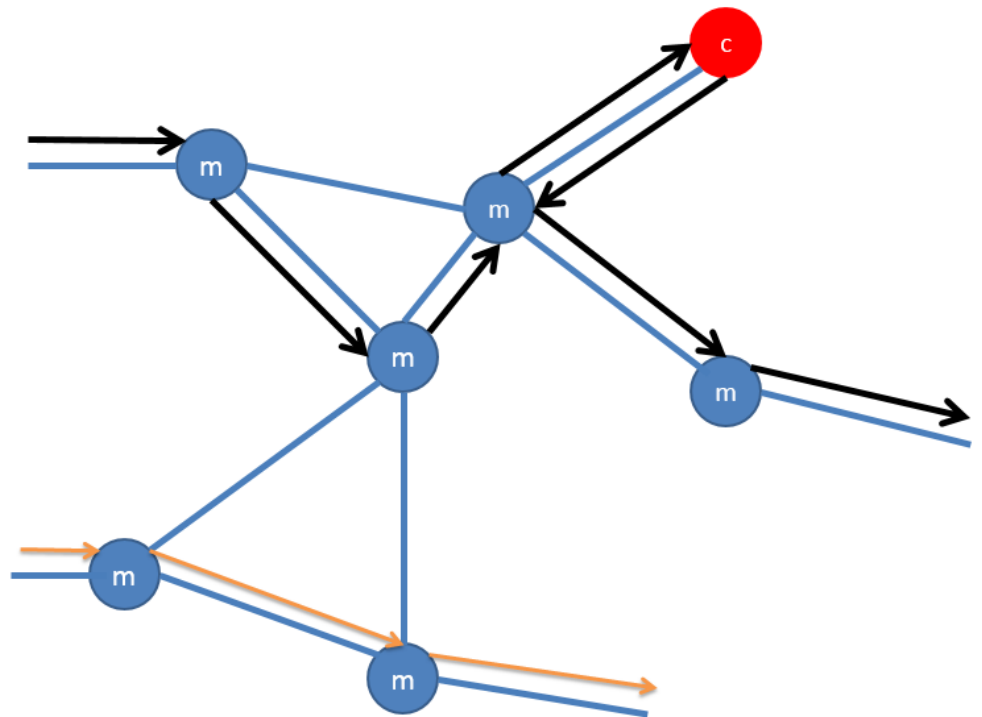


Idea: effective when the next communication node in the walk is very far away (may be better than a Detour)

# Modifications - Overlap



Idea: useful when robots start to run out of time budget



# Heuristic algorithm

- Start from a feasible solution
- Repeat:
  - Construct all the possible modifications
  - Take the modification with the highest (improvement in latency)/(cost) ratio
  - Clean the solution (“Shortcut” modification removes useless portions of walks to regain some budget)
- Until some budget is available



# Some results

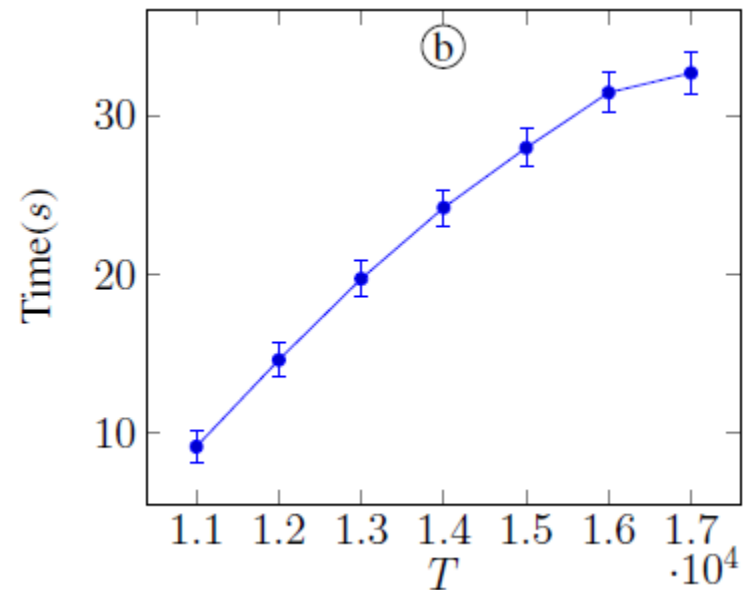
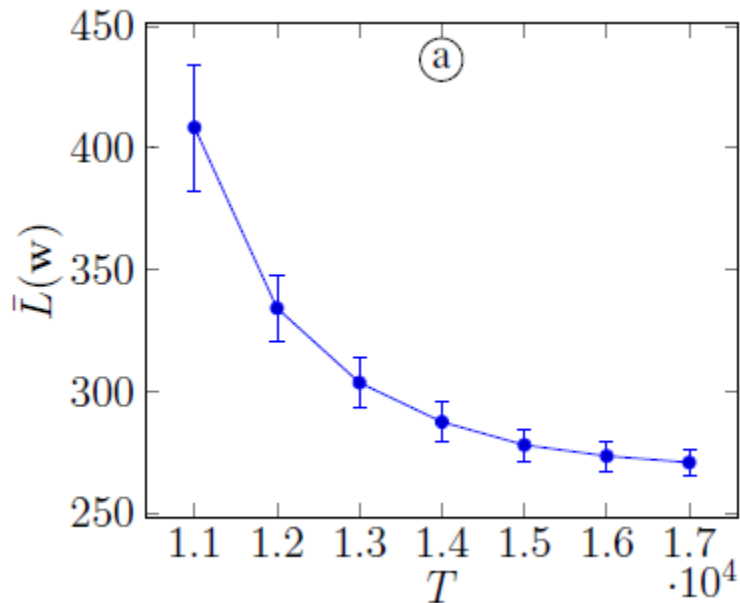
- Effect of local modifications in percentage (100 random graphs of 100 vertices on a 1000x1000 area, 4 robots)

$\epsilon$  = graph edge density

$\epsilon$	Detour	Shift	Overlap	Shortcut	Overall	
0.1	2.9 $\pm$ 2.9	1.6 $\pm$ 1.6	1.7 $\pm$ 1.6	1.8 $\pm$ 1.7	62 $\pm$ 7	T = 13000
0.3	2.5 $\pm$ 2.8	1.7 $\pm$ 1.3	1.7 $\pm$ 1.6	1.8 $\pm$ 2.0	68 $\pm$ 5	T = 8000
0.5	2.4 $\pm$ 2.7	1.6 $\pm$ 1.3	1.6 $\pm$ 1.6	1.8 $\pm$ 2.0	70 $\pm$ 5	T = 7000
0.7	2.4 $\pm$ 2.8	1.5 $\pm$ 1.2	1.5 $\pm$ 1.6	1.9 $\pm$ 1.8	71 $\pm$ 5	T = 5000

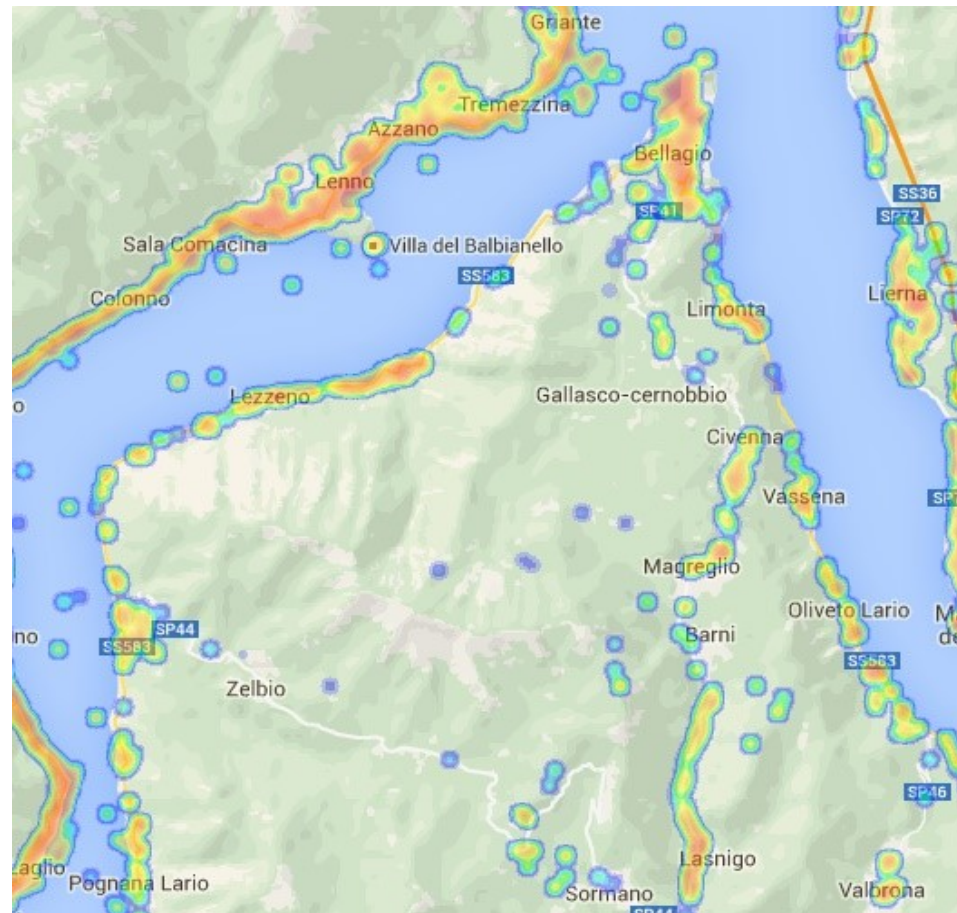
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- Increasing the time budget ( $\varepsilon = 0.1$ , 4 robots)



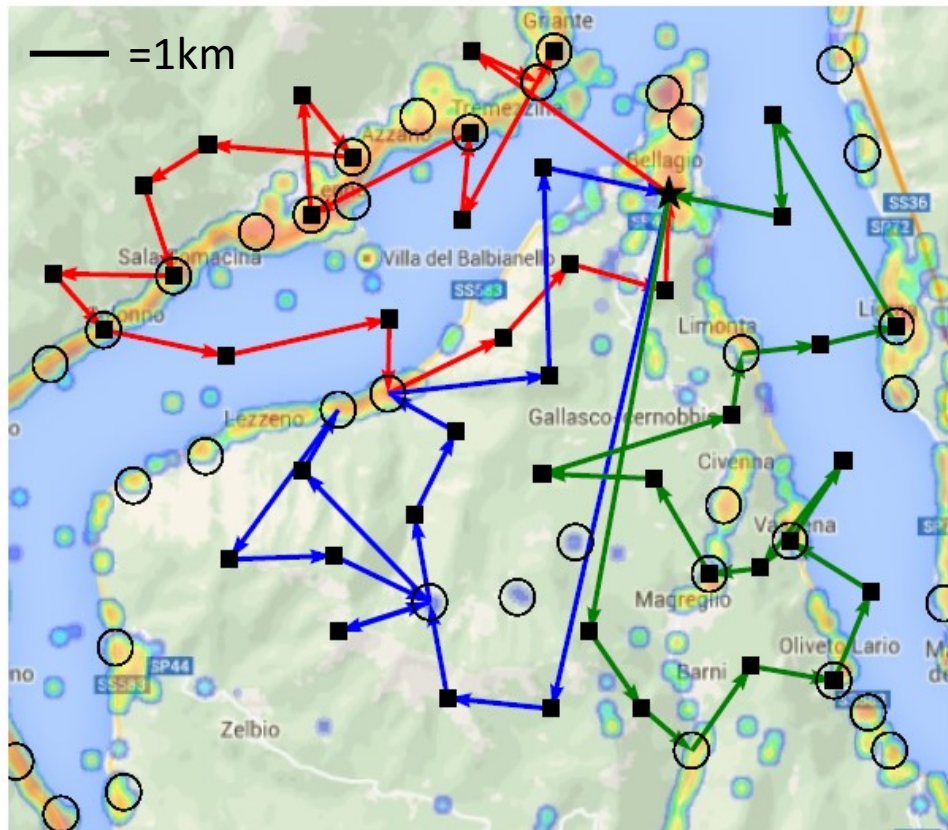
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Como lake (time budget = 1h20min, 3 UAVs, speed 30 km/h)



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Avg latency:  
from  $\approx 10$  min  
to  $\approx 3.5$  min

# Final discussion

- Agents, robots ...
- Experiments ...
- Projects ...