

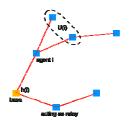
Outline

- □ Shortest Communication-Path Searching
 - Bellmann-Ford algorithm
 - Algorithm for dynamic case
 - Modifications to our algorithm
- Synchronized Task-Switching
 - Combining tasks
 - An algorithm for synchronized task-switching
 - Time complexity
- Summary

Bellman-Ford Algorithm Algorithm Restrictions in Dynamic Case

Communication

- □ Setup:
 - Network of agents, transmitting data to the base.
 - Communication costs, which increase with increasing distance between agents, should be kept low

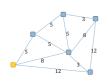


Routing protocol needed, to find the shortest path to the base

Bellman-Ford Shortest Path [1]

Setup: Set of edges which are connected over vertices.
 Goal: Find shortest path from each agent to base.

Notation:



 c_i = communication cost to base of agent i

 N_i = set of neighbors of agent i

 $|v_{ij}| = \text{com. cost from agent i to agent j with } j \in N_i$

 d_i = downstream neighbor of agent i

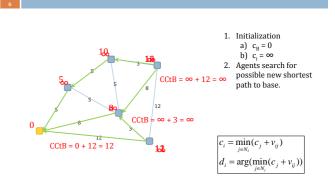
Update rule for every agent:

$$c_i = \min(c_i + v_{ii})$$

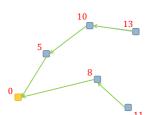
 $d_i = \arg(\min_{i \in \mathcal{N}} (c_j + v_{ij}))$

[1] Richard Bellman - On a Routing Problem - 1958

Bellman-Ford - Example



Bellman-Ford - Example



- 1. Initialization
- a) $c_B = 0$ b) $c_i = \infty$ Agents search for possible new shortest path to base.
- 3. Shortest path is found after maximum of N -1 iterations

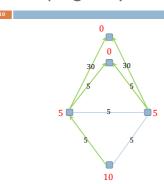
Bellman-Ford Algorithm

Algorithm Restrictions in Dynamic Case

Dynamic shortest path search

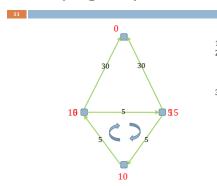
- Changes to static case:
 - Topology of network changes
 - Weightings of vertices vary over time
 - Certain problems occur in dynamic case • looping
 - communication loops occur, connection to base gets lost
 - "longest path" search because of old information in the network, agents choose wrong path to base

Looping in dynamic case

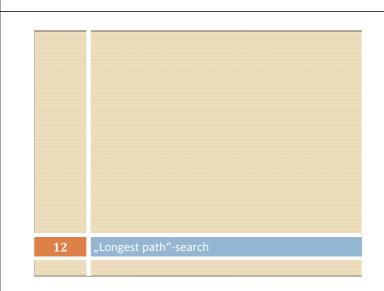


- 1) Agents in steady state
- 2) Downstream Neighbor moves away → communication cost
- increases
 3) Agents search for new shortest path

Looping in dynamic case



- Agents in steady state
 Downstream Neighbor moves away → commun<mark>icatio</mark>n cost increases
- 3) Agents search for new shortest path





Idea of Dynamic Shortest Path Search

- Problem: changing weights and time delayed information propagation leads to loops and wrong pathes
- But: No problem in static case, because here, the communication cost only decreases while converging to shortest path
- Idea: Fix the communication costs and topology between agents and use static computation to find shortest path
- Advantages:
 - Finds real shortest path for given setup (no "longest path", loops)
- Disadvantages:
 - Needs time to converge, during this time, not optimal path

Realization of Dynamic Shortest Path Search

Procedure:

- Fix communication costs to all neigbor agents
- Start new static shortest path search
- When shortest path search is finished, set new found downstream neigbhor as new d.n.
- Go to first step.

Two Problems:

- How do agents know when to finish the shortest path search and start with a new one with updated communication costs?
- How assure that new found shortest path downstream neighbor is still in communication range?

Realization of Dynamic Shortest Path Search

First Problem:

Idea: Base is central processing unit and therefore can be used as a quasi synchronisation module to start the new search.

- New search should start, after shortest path to base was found.
- → Wait worst case time for shortest path search.
- New search signal will be propagated over whole communication range from each agent.
 - → New search signal is faster than shortest path search.

Base sends new search signal to all agents in communication range after worst case computation time for shortest path

Realization of Dynamic Shortest Path Search

- □ **Worst case time** for new shortest path to base:
 - Worst case topology is connected chain
 - Worst case update is, if agents farest away from base update first.



Δt

maximum time within communicator updates → all communicators update at least once within timespan Δt

 $\Delta t_{\text{total}} = (N-1) * \Delta t$

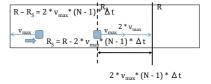
is worst case time to find shortest path

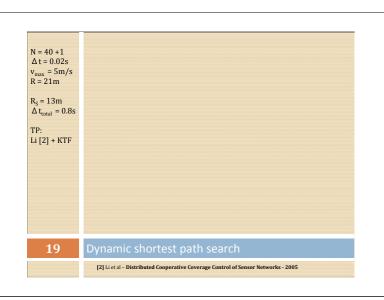
Realization of Dynamic Shortest Path Search

Second Problem:

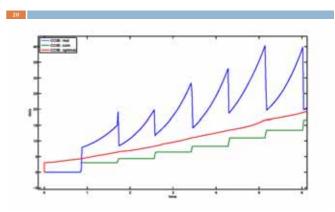
Idea: Agents should not move out of communication range while shortest path search

- assume maximum speed of agent v_{max}
 worst case if agents move in opposite direction with maximum speed, during whole worst case computation time









Dynamic Shortest Path Search - Limitations

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Worst case waiting time:

- \blacksquare Increases with growing number of agents N and computation time Δt .
 - Time between new searches becomes to big, and therefore the error increases.
- Limited neighbor range:

$$R_S = R - 2 * v_{max} * (N - 1) * \Delta t$$

7 steps

- \blacksquare Decreases with growing number of Agents N, computation time Δt and $v_{\rm max}$
 - Maybe R_S becomes to small for a proper shortest path search.

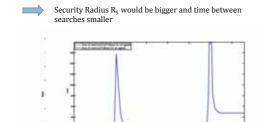
Dynamic Shortest Path Search

Regular Search High frequency search

High search frequency - Motivation

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□ Idea: Increase the frequency with which a new search starts.



5 steps

N = 40 + 1 $\Delta t = 0.02s$ $v_{max} = 5m/s$ R = 21m

 $R_{S} = 20m$ $\Delta t_{total} = 0.1s$

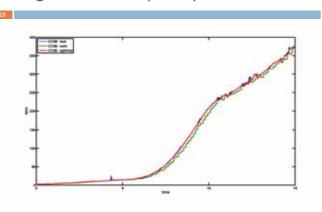
TP: Li [2] + KTF

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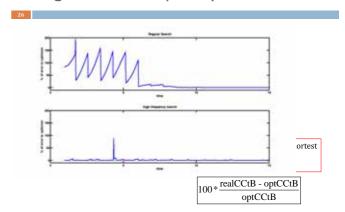
igher new search frequency

 $\hbox{\cite[2]$Li$ et al-Distributed Cooperative Coverage Control of Sensor Networks-2005}$

High search frequency



High search frequency



Discussion

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Regular search

Positive:

- Finds shortest path using only local information (no looping etc.)
- Negative:
 - Strong dependence on number of agents etc.
 - Error while waiting for worst case convergence time

High frequency search

- Positive:
 - Reduces distance to optimum
- Negative:
 - Shortest path is not guaranteed to be found in computation time
 - Rough knowledge of topology needed

Synchronized Task-Switching

Combining Tasks An Algorithm for Synchronized Task-Switching Time Complexity Simulation Results

Combining Tasks



- Coverage control:
 - Maximizing the probability of detecting events.
 - Most important areas of the mission space are well covered.
- □ Exploration of the mission space:
 - $\hfill \blacksquare$ Use of deployment algorithms.
 - $\hfill \square$ Maximize the area covered by all agents.

Combining Tasks



- □ Combine both tasks:
 - First explore the mission space.
 - Then cover the most important areas.
- Enables the agents to cover areas unreachable if only using coverage control.
- □ Switch task when the exploration task is finished.

Combining Tasks

- How do the agents know that the exploration task is finished?
 - □ For each agent, only local information is available.
 - But, information about all agents (=global information) is necessary.
- □ Use of consensus-like algorithms
 - Enables each agent to determine the state of the
 - Task-switch is performed, when all agents agree that the exploration task is finished.

Notations

□ Bi-directional communication between agent i and its neighbors Ni

$$N_i(k) = \{ j \in \{1, ..., n\} | || s_i - s_j || < R \}$$

□ Communication topology is undirected graph G:

A ... Adjacency matrix of G

$$a_{ii} = a_{ii} = 1$$
 if $j \in N_i(k)$

D ... Degree matrix of G

$$d_{ii} = \sum a_{ij}$$

State variables for consensus:

z ... Task state of agents

 $z_i = 1$ if agent i has finished first task

x ... Consensus state

An Algorithm

- Assumptions:
 - There exists a time k_0 such that $A(k)=A(k_0)$ for all $k \ge k_0$
 - There exists a time $k_1 \ge k_0$ s.t. z(k) = 1 for all $k \ge k_1$.
 - □ If $A(k+1) \neq A(k)$, that is there exists i s.t. $N_i(k+1) \neq N_i(k)$, then $z_i(k+1)=0$ even if agent i has finished first task.
- More Notations:
 - $Z \dots diag(z(k))$
 - I ... $n \times n$ identity matrix
 - $\underline{1}$... n × 1 vector with all elements equal to 1
 - $d_i ... d_i = |N_i|$ is the cardinality of the set N_i

An Algorithm

- □ Algorithm:
 - □ If z_i =1, set each agent i's consensus variable x_i to the average value of the sum of its own task state z, and the consensus states of its neighbors.
 - Else, set $x_i=0$.
- Update rule:
 - Update rule: For each agent: $x_i(k+1) = z_i \cdot \frac{1}{d_i + 1} \cdot \left[\sum_{i \in N_i} x_j(k) + 1 \right]$
 - Whole network: $x(k+1) = Z \cdot (D+I)^{-1} \cdot \left[A \cdot x(k) + \underline{1}\right]$

An Algorithm

- □ State of the network and task-stwitch:
 - If at least one agent has not finished the first task, $x_i(k)$ <1 for all agents.
 - If all agents have finished the first task, z(k)=1 and for every agent i, $x_i(k) \rightarrow 1$ for $k \rightarrow \infty$.
 - Perform task-switch if x_i is sufficiently close to 1.
- □ Convergence of Algorithm:
 - For constant z(k), system is an asymptotically stable LTI-system with constant input.
 - $\hfill\Box$ There is always one unique equilibrium point x_{EP} and $x_{EP}=1$ for z=1.

Threshold for Task-Switch



- \square When is x_i sufficiently close to 1?
 - Task-switch if $x_i > \delta$, with $\delta < 1$
 - $\hfill \blacksquare$ If δ is too small, false task-switch might happen.
- □ How to determine δ?
 - Derive from static case where no topology changes happen.
 - Show that even under switching topology, x_i(k) is never larger than $\max_{x_i} x_i^{wc}$ in the worst case static topology.

Time Complexity

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- □ In [3] the term Time Complexity is introduced:
 - □ The **Time Complexity TC** is the time an algorithm needs to perform, depending on the *number of agents* **n**.
 - For the **task-switch**, a sensible notion is the time from when the last agent finishes the first task until the last agent starts with the second task.
- Upper bound:
 - An upper bound to the order of the time complexity is given by:

$$TC_{A1} \in O(\frac{\ln(1-\delta(n))}{\ln(\frac{n-1}{n})})$$

[3] Martinez, Bullo, Cortes, Frazzoli - "On synchronous robotic networks - Part II

Time Complexity

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- □ Proof:
 - At step k_1 , let i be the agent such that $x_{min}(k_1) := x_i(k_1) \le x_i(k_1)$ for all agents j.
 - Then in the next step for agent i:

$$x_i(k_1+1) = \frac{1}{d_i+1} \cdot \left(\sum_{j \in \mathcal{N}_i(k_1)} \sum_{\substack{i \in \mathcal{N}_i(k_1) \\ i \in \mathcal{N}_i(k_1)}} x_i(k_1) + \sum_{i=1}^{i} (k_1) \right) \geq \frac{1}{d_i+1} \cdot \left(d_i \cdot x_i(k_1) + 1 \right)$$

 \square The smallest possible value for $x_i(k_i+1)$ is achieved by maximizing the number of neighbors.

$$x_i(k_1+1) \ge \frac{1}{n} \cdot ((n-1) \cdot x_i(k_1) + 1)$$

Time Complexity

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□ Proof:

□ The value $x_{min}(k_1+1) := x_i(k_1+1)$ provides a lower bound on the consensus values of all agents j in step k_1+1 :

$$x_{j}(k_{1}+1) \ge x_{\min}(k_{1}+1) = \frac{1}{n} \cdot ((n-1) \cdot x_{\min}(k_{1}) + 1)$$

■ This can easily be seen: Suppose there exists $x_i(k_1+1) < x_{min}(k_1+1)$

$$\begin{split} x_{i}(k_{1}+1) &= \frac{1}{d_{i}+1} \cdot \left(\sum_{j \in \mathcal{N}_{i}(k_{1})} x_{j}(k_{1}) + 1 \right) \geq \frac{1}{d_{i}+1} \cdot \left(d_{i} \cdot x_{\min}(k_{1}) + 1 \right) \\ &\geq \frac{1}{N} \left((n-1) \cdot x_{\min}(k_{1}) + 1 \right) = x_{\min}(k_{1}+1) \end{split}$$

Time Complexity

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- □ Proof of Time Complexity:
 - This lower bound on the consensus value of all agents can be generally described by:

$$x_{\min}(k+1) = \frac{1}{n} \cdot ((n-1) \cdot x_{\min}(k) + 1)$$

 $\hfill\Box$ The solution to this difference equation for $k \geq k_1$ is:

$$x_{\min}(k) = 1 - \left(\frac{n-1}{n}\right)^{k-k_1} \cdot \left(1 - x_{\min}(k_T)\right)$$

 \blacksquare With the switching condition $x_i > \delta$ and k_T the step when all agents have switched to the second task it follows:

$$TC = k_T - k_1 \ge \frac{\ln\left(1 - \delta(n)\right)}{\ln\left(\frac{n-1}{n}\right)}$$

Time Complexity

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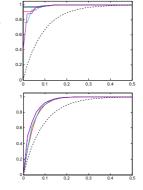
Simulations:

- Simulation of task-switch for different topologies and numbers of agents in task 1 before switch.
- Chain topology:
 - Waiting for only one agent:



Waiting for all agents:





Time Complexity

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- □ Simulations:
 - Random topology:
 - Waiting for one agent:



Waiting for all agents:



