

DYNAMIC FORMATION CONTROL WITH HETEROGENOUS MOBILE ROBOTS

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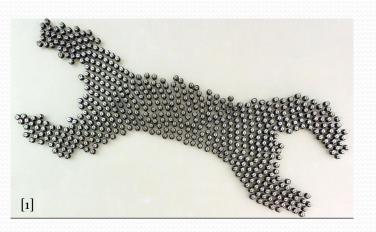
Outline

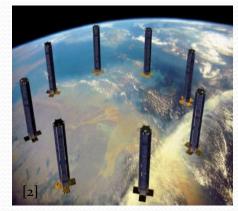
- Introduction
- Motivation
- System Overview
- Local Positioning System Design
- Formation Control System Design
- Results
- Conclusion and Future Works
- References



Introduction

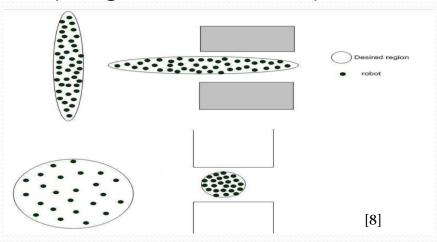
This thesis work focuses on dynamic adaptation to achieve changes in formation of swarms consisting of heterogenous mobile robots

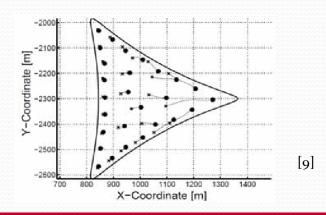






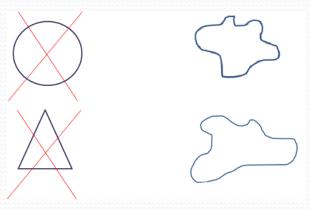
Formation control solutions are generally implemented with simple geometrical shapes which don't change with time.





Our aim

 Designing a formation control system for complex and dynamically changing shapes



The research about the formation control, mainly focuses on swarms with homogenous agents.

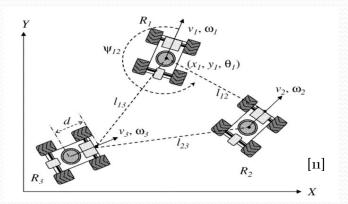


Our aim

 Designing a formation control system with heterogeneous agents

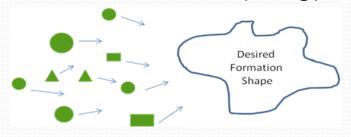
Centralized topologies create single point of failure type systems. We aim to implement a decentralized solution to increase the robustness of the system.



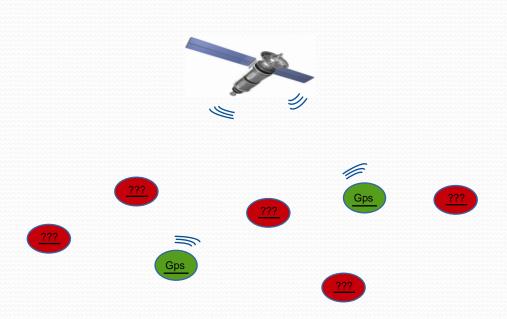


Our aim

 Designing a formation control system with a decentralized topology



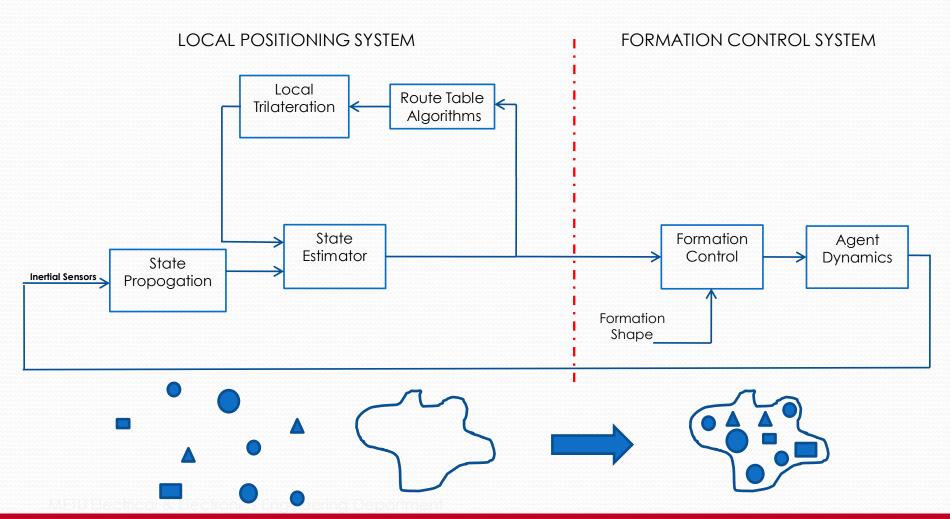
Most of the related works assume that the position data is always available (i.e. can be measured) for each agent in the workspace.



Our aim

Designing a local positioning system to provide position information to the second type agents with the help of position beacons.

System Overview



Local positioning system is composed of two main parts.



<u>Local Trilateration</u>

This process calculates the positions of the second type agents with the help of position beacons which are their direct neighbors.



Route Table Determination

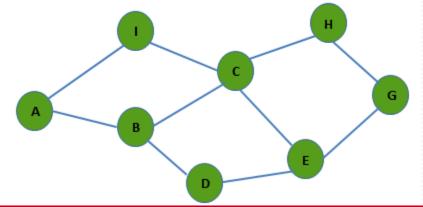
This process determines the route tables for agents in the swarm.

1)Route Table Determination

• Destination-Sequenced Distance Vector Routing Protocol (DSDV) algorithms are used to create the route tables.

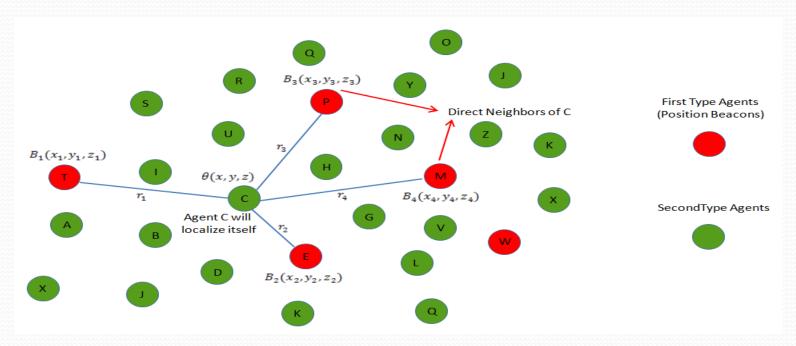
Route table for agent R

Moute table for agent b				
Destination	Next Hop	Cost		
Α	Α	1		
1	Α	2		
c	С	1		
н	С	2		
G	С	3		



2)Local trilateration

- Calculates the position of an agent with the help of position beacons which are direct neighbors.
- •The solution can be reduced to a problem of $\vec{Ax} = \vec{b}$



Three different approaches are used to design the formation control system in this thesis work.

Formation Control Strategies

Potential Field
Based
Approach
1) Artificial Forces Method

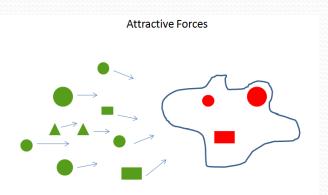
 Directly calculates control laws based upon potential fields Shape Partitioning Based Approaches

- 2) Bubble Packing Method
- 3) Randomized Fractals Method
 - Partitions the desired formation shape into goal states.
 - Control laws are implemented to reach these goal states.

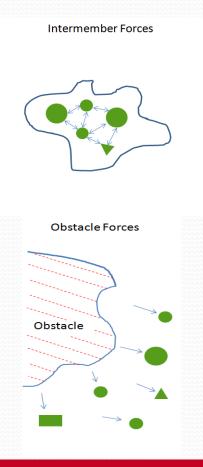


Artificial Forces Method

Directly defines the control law for individuals with different potential field components.



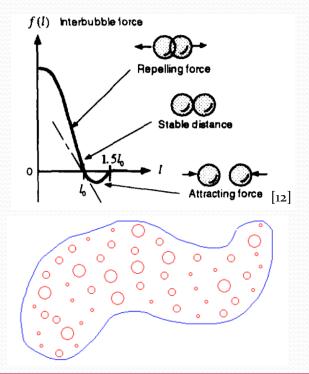
$$u_i = u_{att_i} + u_{rep_i} + u_{obs_i} + u_{int_i}$$



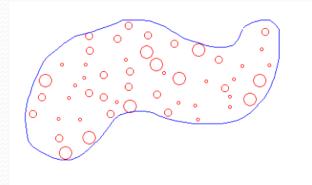
Bubble Packing and Randomized Fractals Methods

These two methods partition the desired formation shape into goal states with different approaches. The procedure of the assignment of the agents to these goal states are identical.

Bubble Packing



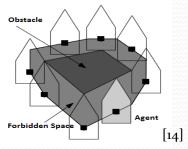
Randomized Factals



Bubble Packing and Randomized Fractals Methods

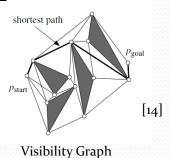
Decision of Goal States

1)Calculation of Free Configuration Space

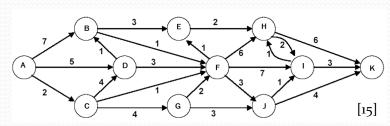


Forbidden Space

2)Visibility Graphs

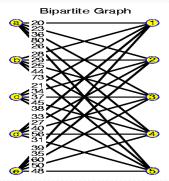


3)Dijkstra's Algorithm



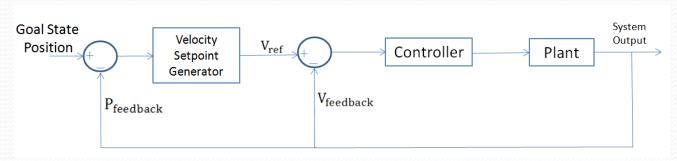
Calculation of Minimum Shortest Path From A to K

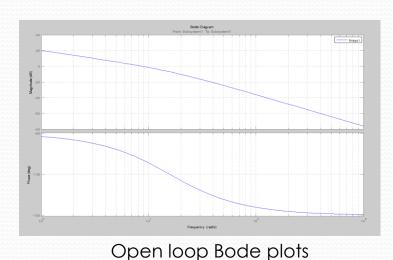
4) Hungarian Algorithm (Munkres Assignment Algorithm)



Bubble Packing and Randomized Fractals Methods

Navigation to Goal States





Step Response
From: Subsystem/1 To: Subsystem/1

1.2

1.2

0.8

0.8

0.4

0.2

0.0

0.1

0.2

0.3

0.4

0.5

0.6

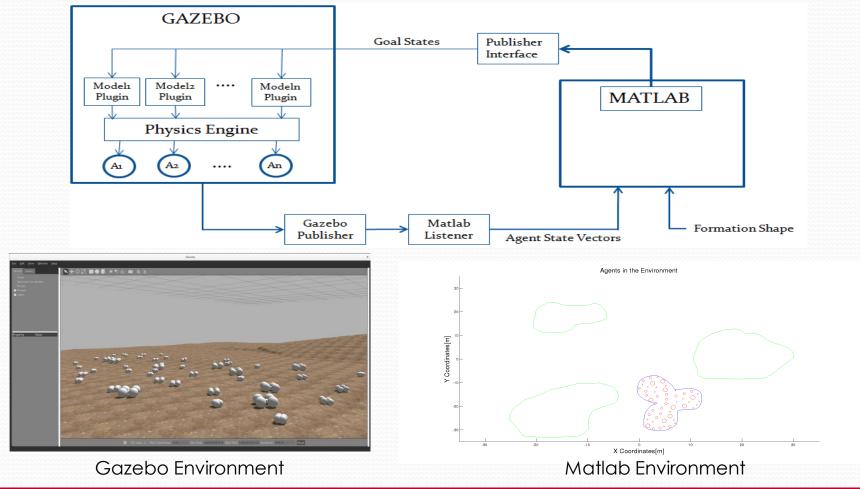
7

Time (seconds)

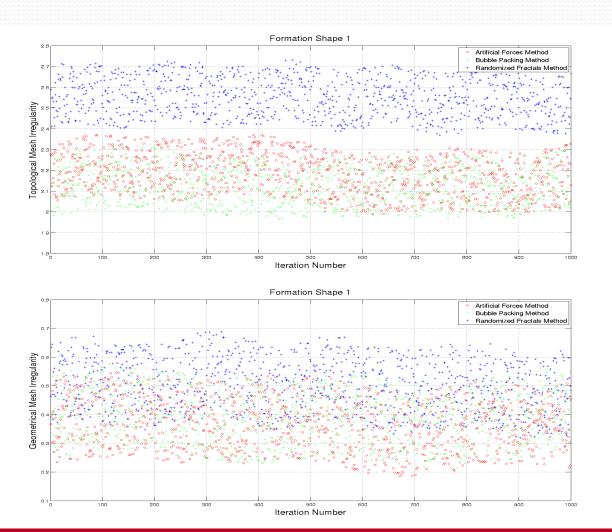
Step response

Results

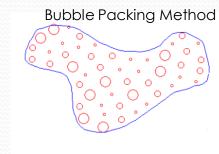
Proposed solutions are implemented in a simulation environment.

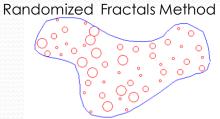


Results — Mesh Quality



Artificial Forces Method







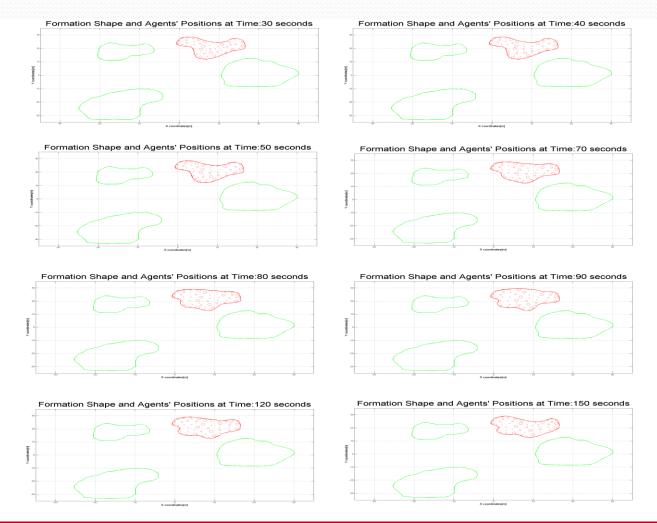
Results – Comparison of 3 Different Methods

•Comparison of different solution methods are illustrated in Table -1

Table -1

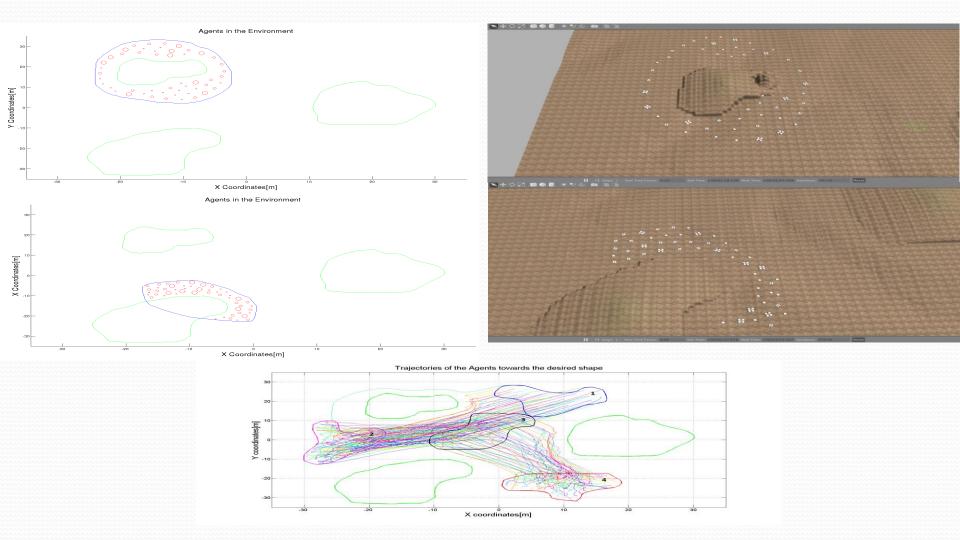
Method/ Metric	Total Displacement	Settling Time	Mesh Quality
Artificial Forces			
Bubble Packing			
Randomized Fractals			

Results - Dynamically Changing Formation Shapes





Results – Various Formation Shape Trials



Results - Sensitivity

• Mesh irregularities are expected to be increasing at boundary layers due to discontinuties[17]. It is possible to see this condition with a formation shape dynamically changing its perimeter by keeping its coverage area constant.

Shape	Area	Perimeter	Topological Mesh Irregularity
1	314[m^2]	62,8[m]	2,46
2	314[m^2]	82,8[m]	2,96
3	314[m^2]	116,82[m]	3,45
4	314[m^2]	128,36[m]	3,95

^{*}In 2D, circle has the minimum perimeter to cover a fixed area[18], thus it has the lowest irregularity.

Results – Hardware Implementation

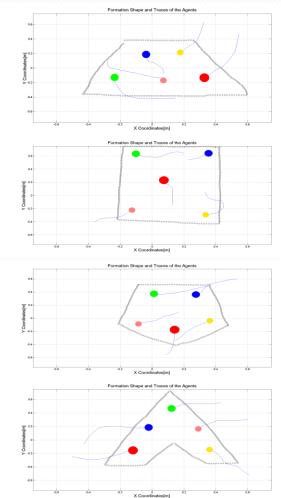
•Hardware applications which implements the methods discussed in this thesis work are also developed

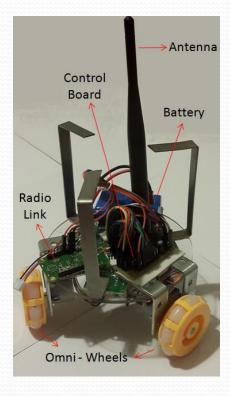












Conclusion

- In this project, we have proposed a complete solution, including localization process and formation control.
- •We aim to implement a formation control system with heterogenous agents and complex geometrical shapes which are changing dynamically.
- We have proposed different solutions to the formation control problem and discuss about their performance.
- •We make hardware applications to demonstrate that the proposed methods can be implemented in real time applications

Future Works

- Hardware implementation will be done with more agents
- We use heterogenous agents just to cover different formation shapes. We will try to achieve collaborative tasks, by using different functionalities of the agents.
- •Obstacle avoidance is implemented with potential fields. To avoid unwanted equilibrium states, obstacle avoidance feature will be implemented with a more appropriate way (e.g. Tangent bug algorithm).

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Thank you for your attention.

Bubble Packing and Randomized Fractals Methods

Velocity Controller

The dynamical system of agents is augmented with an artificial error state, to design an State feedback with LOR controller;

$$\begin{bmatrix} \dot{v} \\ \dot{e} \end{bmatrix} = \begin{bmatrix} -b/m & 0 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} v \\ e \end{bmatrix} + \begin{bmatrix} 1/m \\ 0 \end{bmatrix} F_{nst}$$

$$y = \begin{bmatrix} 1 & 0 \end{bmatrix} \begin{bmatrix} v \\ e \end{bmatrix}$$

Q and R matrices used in solving Riccati equations,

$$Q = \begin{bmatrix} q_1 & 0 \\ 0 & q_2 \end{bmatrix}$$
; $R = \rho r_1$

$$Q = \begin{bmatrix} q_1 & 0 \\ 0 & q_2 \end{bmatrix}; R = \rho r_1 \qquad q_1 = \frac{1}{t_{s_1}(x_{1max})^2}; q_2 = \frac{1}{t_2(x_{2max})^2} \text{ and } r_1 = \frac{1}{(u_{1max})^2}$$

where,

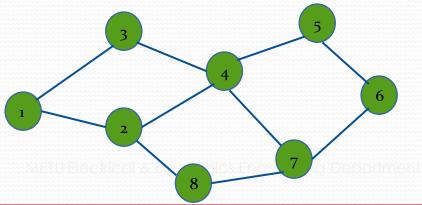
 t_{s_i} : desired settling time for x_i p: tradeoff regulation vs control effort

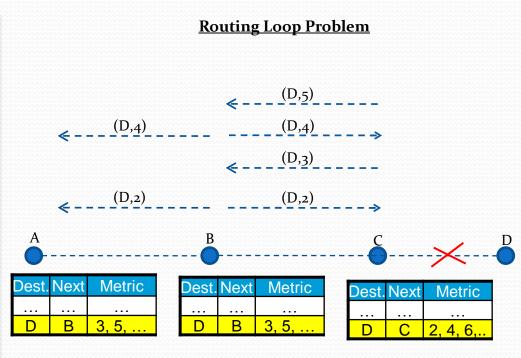
2) Route Table Determination

- DSDV is a table driven routing scheme based on Bellman Ford algorithm
- Used to create wireless mesh networks and ad-hoc mobile networks
- Solves routing loop problem in route table algorithms

Route	table	for	age	ent	2

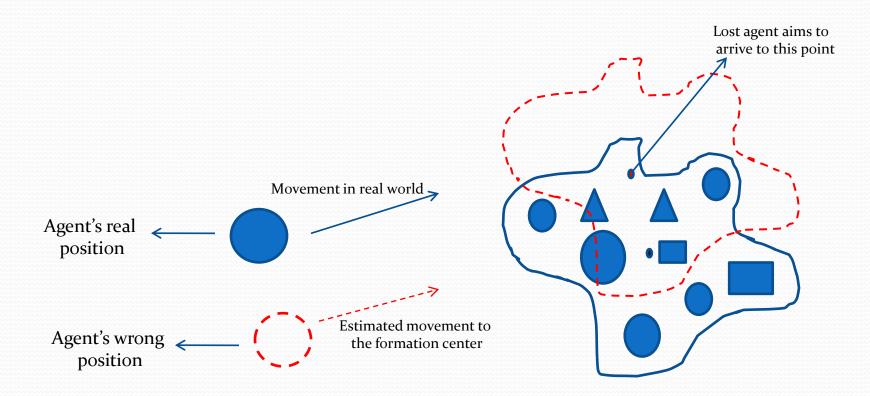
Destination	Next Hop	Metric	Dest. Seq. No
1	1	1	123
3	3	2	516
4	4	1	212
5	4	2	168
6	8	3	372
7	8	2	432





1)Local trilateration

Return to Home Mode



1)Local trilateration

The solution of the position(P(x,y,z)) with the help of positions of neighbors can be reduced to a problem of;

$$\overrightarrow{Ax} = \overrightarrow{b}$$

We have an A matrix with a dimension of $[n-1] \times 2$ (where n is the number of neighbors). There are three options for the solution of the problem related with the condition of A matrix,

- 1) $\hat{x} = A^{-1}.b$, unique solution (if there are 3 neighbors and A is full column rank matrix)
- 2) $\hat{x} = (A^T A)^{-1} A^T b$, minimum norm solution (if there are more neighbors and A is full column rank matrix
- 3) Find the minimum error/norm solution with nonlinear least squares method, if rank(A) = 1

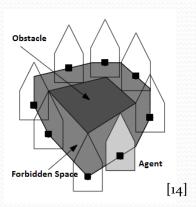
Bubble Packing and Randomized Fractals Methods

Decision of Goal States

1)Calculation of Free Configuration Space

$$C(R_i) = C_{free}(R_i, S) + C_{forb}(R_i, S)$$

Forbidden Space : $S_1 \oplus S_2 := \{ p + q : p \subset S_1, q \subset S_2 \}$

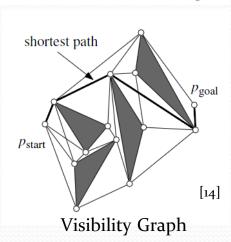


Forbidden Space

2)Visibility Graphs

The shortest path between a start and goal among a set S of augmented polygonal obstacles consists of arcs of the visibility graph [14]

$$\gamma_{vis}(S^*)$$
 where $S^* := S \cup \{p_{start}, p_{goal}\}$

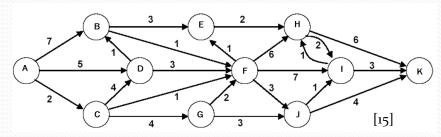


• Bubble Packing and Randomized Fractals Methods

Decision of Goal States

3)Dijkstra's Algorithm

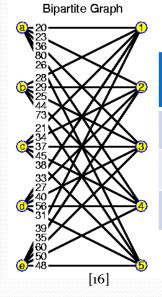
Dijkstra's algorithm is a tree search algorithm for finding the shortest paths between nodes in a graph



Calculation of Minimum Shortest Path From A to K

4) Hungarian Algorithm (Munkres Assignment Algorithm)

The shortest path between a start and goal among a set S of augmented polygonal obstacles consists of arcs of the visibility graph

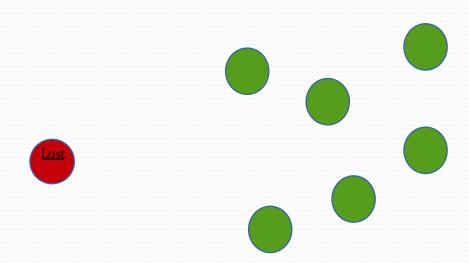


	Clean Bathroom	Sweep Floors	Wash Windows		
Jim	\$3	\$2	\$7		
Steve	\$2	\$5	\$3		
Alan	\$4	\$3	\$2		

2)Local trilateration

Lost agent handling rules;

- •An agent is called 'lost' when it doesn't have minimum 3 neighbors
- •If an agent is lost it cannot enter the localization process, and it enters 'Lost' mode in which it is directed to the center of formation shape.



Results

Local Positioning System (LPS)

