

Fly with Me: Algorithms and Methods for Influencing a Flock

Katie Genter

The University of Texas at Austin
katie@cs.utexas.edu

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Bird Strikes in Aviation



\$3 billion per year (PreciseFlight)

Common Bird Strike Reduction Methods



Common Bird Strike Reduction Methods



Common Bird Strike Reduction Methods

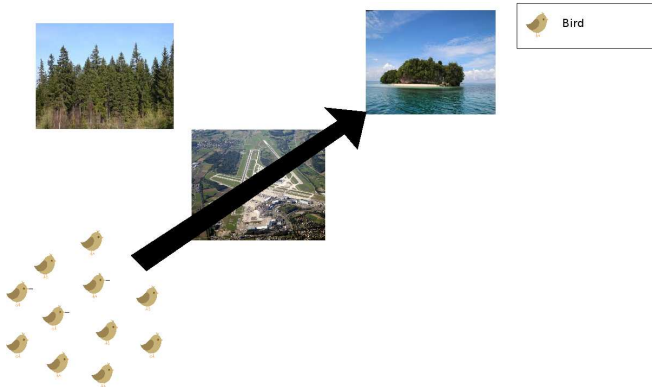


Clear Flight Solutions Robird

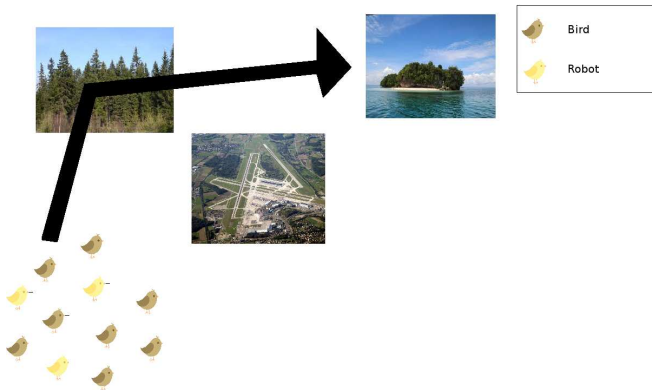
- ▶ Modeled after a peregrine falcon
- ▶ Will be first bird drone to operate on a daily basis for commercial operations at an airport
- ▶ 13 week trial at Edmonton begins July 1
- ▶ Remote controlled by a trained pilot



Motivating Example



Motivating Example



Thesis Question

Thesis Question:

How can influencing agents be utilized in various types of flocks in order to influence these flocks towards a particular behavior?

Background

- ▶ Artificial intelligence
 - ▶ Multi-agent systems
 - ▶ Teamwork
 - ▶ Ad hoc teamwork
 - ▶ Swarm behavior
 - ▶ Flocking



Multi-agent Teamwork

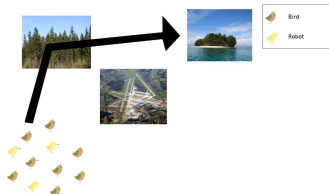
- ▶ All agents have the same **goal**
- ▶ Usually considers how to design **teams** of agents



Ad Hoc Teamwork

- ▶ Only in control of a single agent or subset of agents
- ▶ Shared goals
- ▶ No pre-coordination
- ▶ Limited or no communication

Scientific question: How to design an agent to **productively join** a pre-existing team while requiring **no pre-coordination**?



Flocking

- ▶ Emergent swarm behavior
- ▶ Simple local behaviors & interactions
- ▶ Group behavior appears well organized and stable



Thesis Question

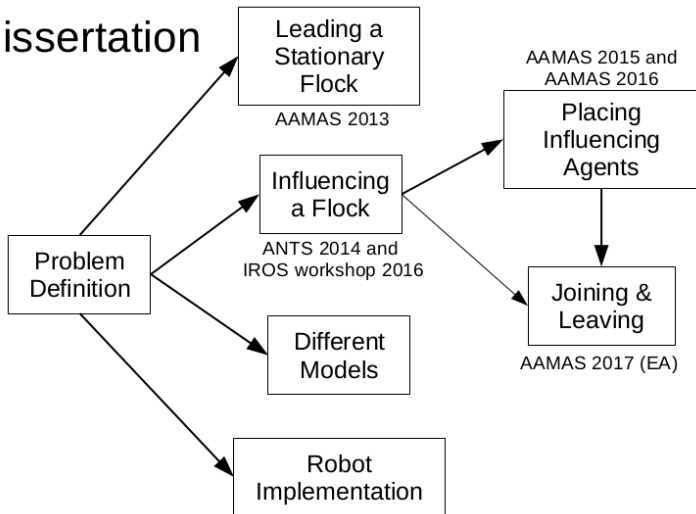
Thesis Question:

How can influencing agents be utilized in various types of flocks in order to influence these flocks towards a particular behavior?

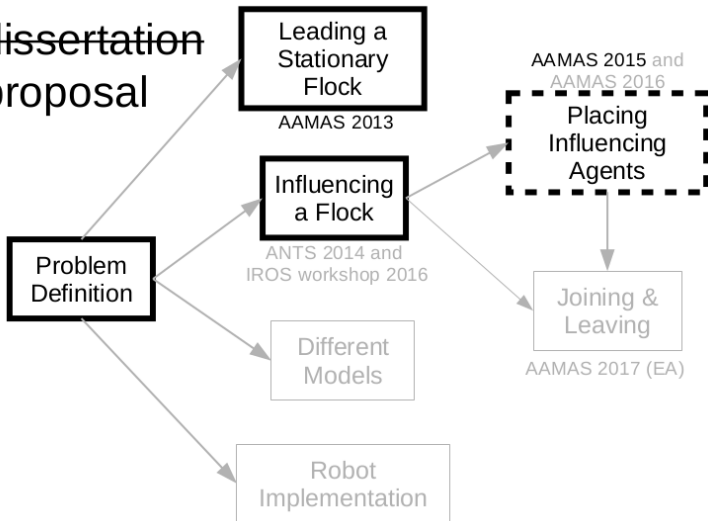
Relation to Existing Approaches

- ▶ Simulating flock behavior Reynolds 1987 and Vicsek et al. 1995
 - ▶ My work: **Influence** the flock using influencing agents
- ▶ Influence the flock to converge eventually Jadbabaie et al. 2003, Su et al. 2009, and Celikkanat and Sahin 2010
 - ▶ My work: Influence the flock to converge **quickly**
- ▶ Influencing agent moves much quicker than the flock Han et al. 2006
 - ▶ My work: Influencing agents move **no quicker than** the flock
- ▶ Influence a flock via heterogenous herders Lien et al. 2004, Lien et al. 2005, Pierson and Schwager 2015
 - ▶ My work: Influence the flock **from within**

My dissertation

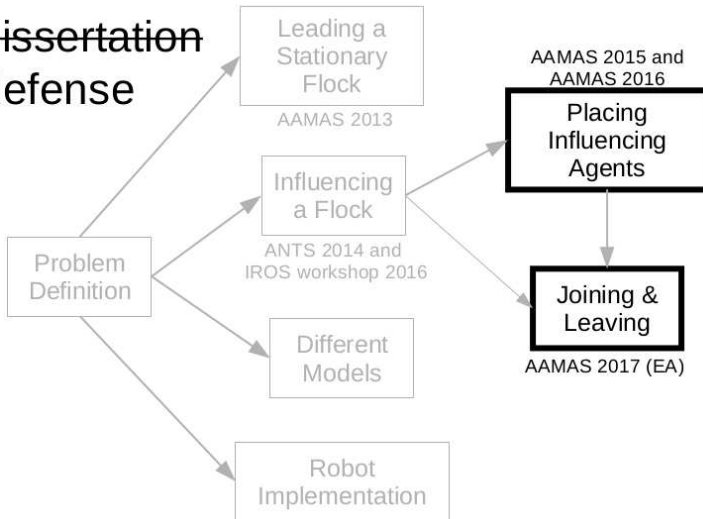


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My proposal

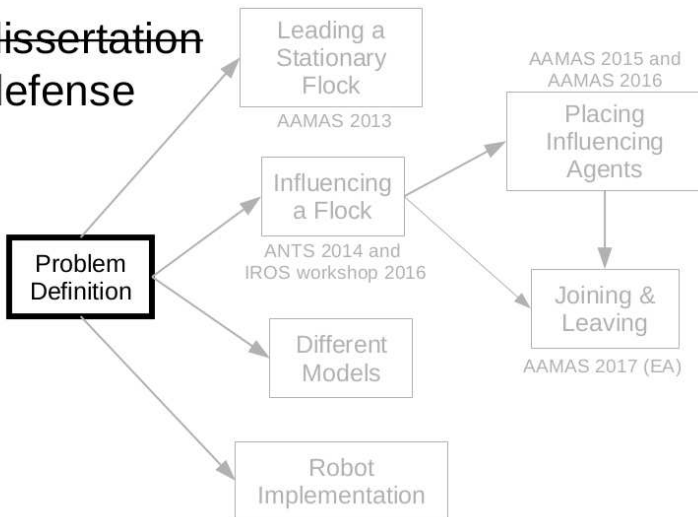


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My defense



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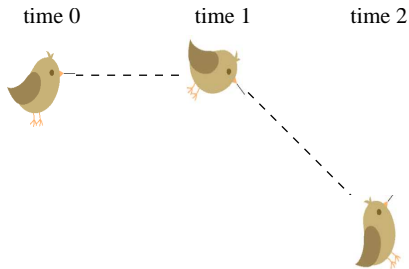
Problem Definition

- ▶ Define the problem of adding influencing agents to a flock as new scenario for studying ad hoc teamwork
- ▶ Introduce assumptions and parameters

Problem Definition

All agents have:

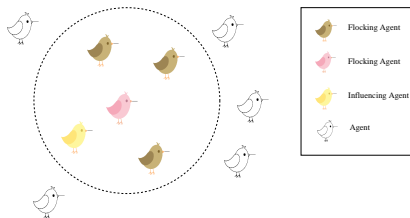
- ▶ Constant velocity
- ▶ 2D Position
- ▶ Global orientation



Problem Definition - Neighborhood Model

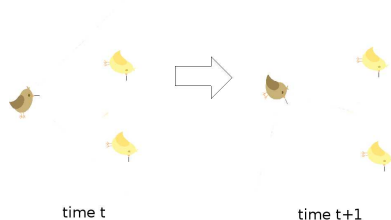
Each agent reacts only to agents within a certain **neighborhood** around itself.

- Characterized by a **visibility radius** for most of this dissertation



Problem Definition - Influence Model

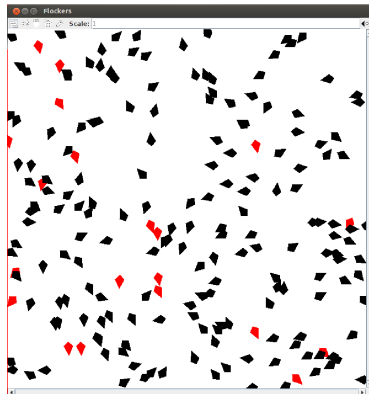
An agent's orientation at the next time step is set to be the **average global orientation** of all agents currently within its neighborhood.



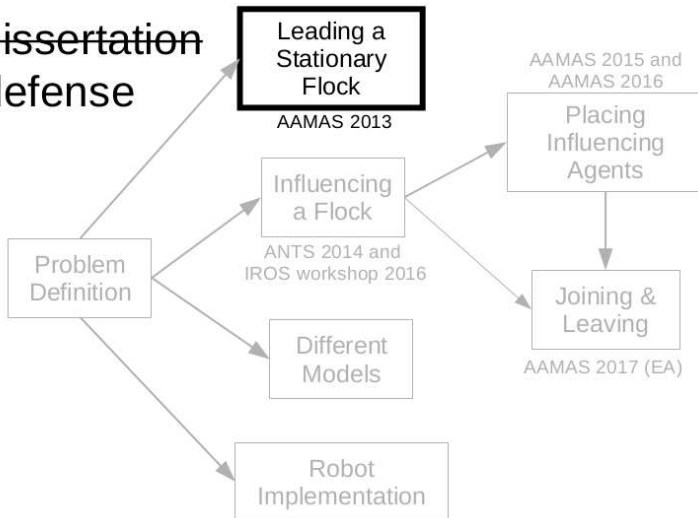
MASON Simulator

- ▶ Multi-agent simulation library core
- ▶ Flockers domain

S. Luke, C. Cioffi-Revilla, L. Panait, K. Sullivan, and G. Balan. MASON: A multi-agent simulation environment. In *Simulation: Transactions of the Society for Modeling and Simulation International*, 2005.



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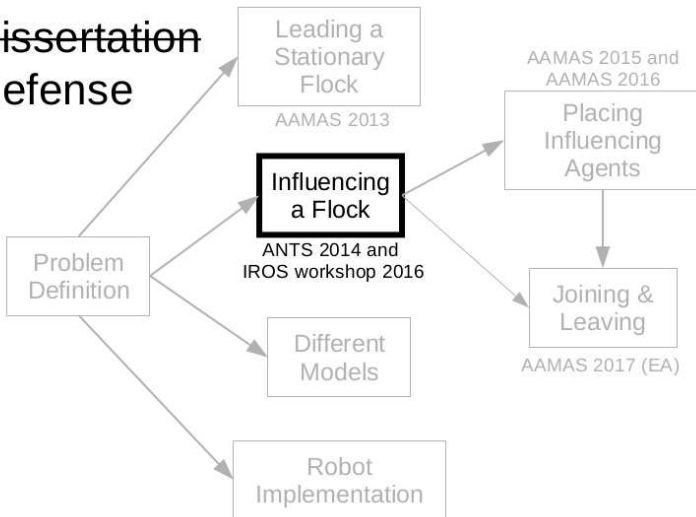
Leading a Stationary Flock

- ▶ Set **bounds** on the extent of influence the influencing agents can have when both the influencing agents and flocking agents are stationary
- ▶ Contributed an algorithm for **orienting a stationary flock** to a desired orientation using stationary or non-stationary influencing agents
- ▶ Introduced behaviors for influencing agents that are outside of any flocking agent's neighborhood

K. Genter, N. Agmon, and P. Stone. Ad hoc teamwork for leading a flock. In AAMAS'13, May 2013.

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Influencing a Flock

- ▶ Contributed algorithms that can be used to **influence a flock** to orient towards a desired orientation
- ▶ Experimentally considered how to **maneuver** the flock through turns quickly but with minimal agents being separated

K. Genter and P. Stone. Influencing a flock via ad hoc teamwork. In Ninth International Conference on Swarm Intelligence (ANTS'14), September 2014.

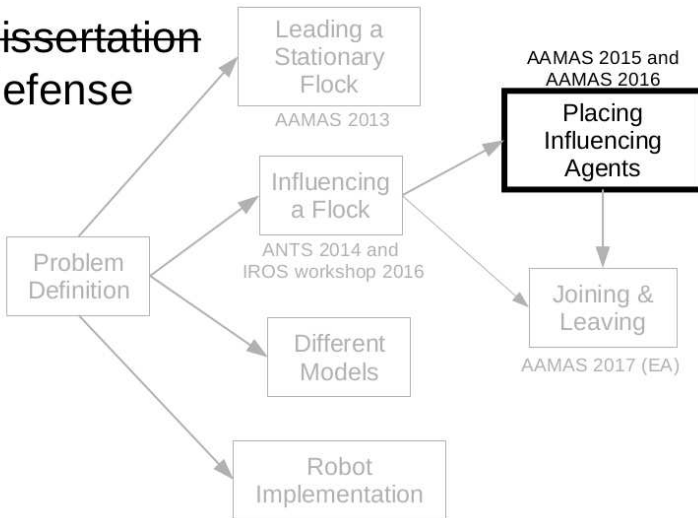
K. Genter and P. Stone. Ad hoc teamwork behaviors for influencing a flock. Acta Polytechnica, 2016.

1-Step Lookahead Behavior

- ▶ For a set number of discrete orientations the influencing agent could adopt, how will each **neighbor** be influenced?
 - ▶ Must estimate how other neighbors of each neighbor will influence the neighbor
- ▶ Choose orientation that results in the least average **difference** between the goal orientation and each neighbor's new orientation after one time step

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Placing Influencing Agents into a Flock

- ▶ The position of influencing agents within the flock affects their influence
- ▶ Contribute various approaches for placing influencing agents directly into a flock
- ▶ Evaluate on various flock sizes and compositions

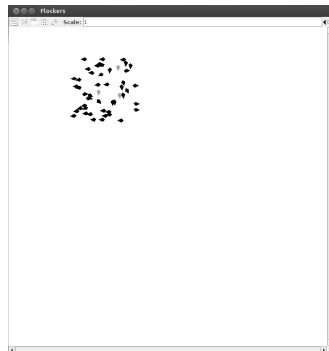
K. Genter, S. Zhang, and P. Stone. Determining placements of robots in a flock. In AAMAS'15, May 2015.

K. Genter and P. Stone. Adding influencing agents to a flock. In AAMAS'16, May 2016.

Placing Influencing Agents into a Flock

Assumptions:

- ▶ Flock begins in set area
- ▶ Non-toroidal
- ▶ Flock may or may not be cohesive
- ▶ The flock will interact with each other according to the assumed model



Determining Desired Positions

- ▶ Constant-time placement methods
- ▶ Graph placement method
- ▶ Hybrid placement methods
- ▶ Two-Step placement method
- ▶ Clustering placement methods

Determining Desired Positions - Constant-time Methods

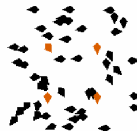
- ▶ Constant-time placement methods
 - ▶ Random Placement
 - ▶ Grid Placement
 - ▶ Border Approach
- ▶ Graph placement method
- ▶ Hybrid placement methods
- ▶ Two-Step placement method
- ▶ Clustering placement methods

Determining Desired Positions - Constant Time Methods

- ▶ Random Placement
 - ▶ Randomly place influencing agents within the dimensions of the flock
- ▶ Grid Placement
- ▶ Border Approach

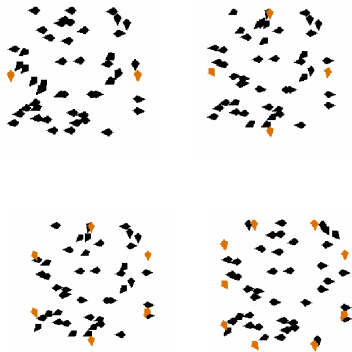
Grid Placement Method

Place **influencing agents** at predefined **gridded positions** throughout flock

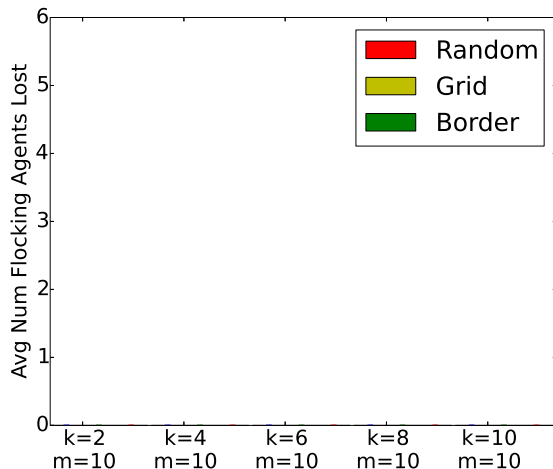


Border Approach Method

Place **influencing agents** around the space covered by the flocking agents

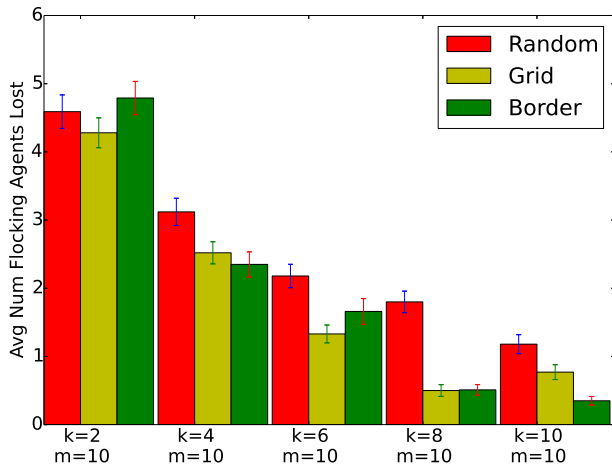


Experimental Results - Constant-time Placement



100 trials, error bars = standard error of the mean.

Experimental Results - Constant-time Placement



100 trials, error bars = standard error of the mean.

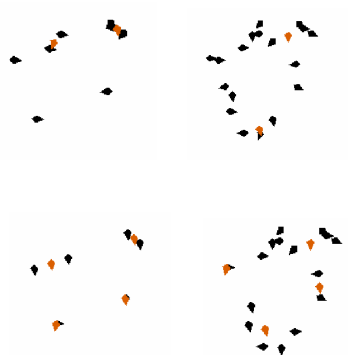
Determining Desired Positions - Graph Method

- ▶ Constant-time placement methods
- ▶ Graph placement method
- ▶ Hybrid placement methods
- ▶ Two-Step placement method
- ▶ Clustering placement methods

Graph Placement Method

Consider sets of positions in which the **influencing agents** could be placed.

Evaluate how well these sets connect the flocking agents with the influencing agents.



Graph Placement Method - Steps (1)

- ▶ Consider potential influencing agents positions
 - ▶ Mid-points between flocking agents
 - ▶ Only for flocking agents within 2 neighborhood radii
 - ▶ Near flocking agents



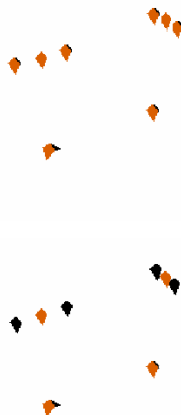
Graph Placement Method - Steps (2)

- ▶ Evaluate each set of influencing agent positions:
 - ▶ minimize flocking agents not *connected* to an influencing agent
 - ▶ maximize *connections* between flocking agents and influencing agents



Graph Placement Method - Steps (2)

- ▶ Evaluate each set of influencing agent positions
 - ▶ minimize flocking agents not *connected* to an influencing agent
 - ▶ maximize *connections* between flocking agents and influencing agents



Experimental Results - Graph Placement

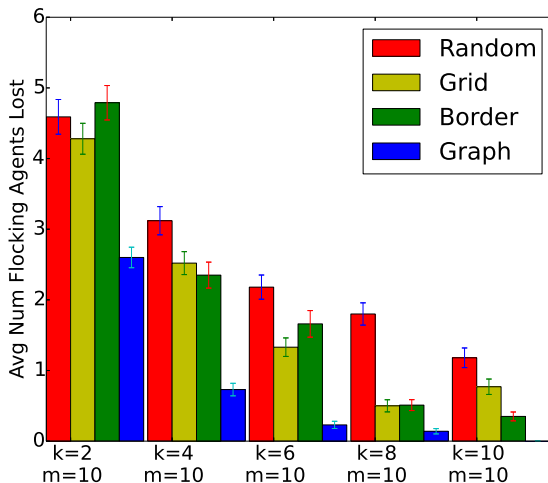


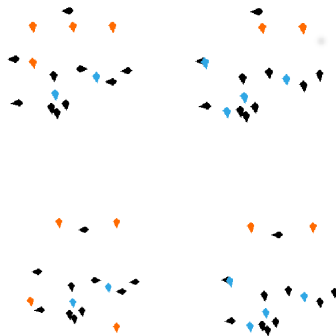
Figure : 100 trials, error bars = standard error of the mean

Determining Desired Positions - Hybrid Methods

- ▶ Constant-time placement methods
- ▶ Graph placement method
- ▶ Hybrid placement methods
- ▶ Two-Step placement method
- ▶ Clustering placement methods

Hybrid Placement Method

Use the **Graph placement method** to choose the first k_g influencing agent placements. Then select the remaining $k - k_g$ placements using an **constant time placement method**.



Experimental Results - Hybrid (Grid/Graph) Placement Method

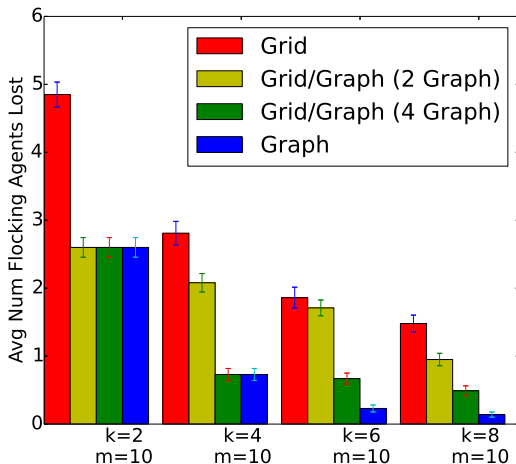


Figure : 100 trials, error bars = standard error of the mean

Determining Desired Positions - Two-Step Method

- ▶ Constant-time placement methods
- ▶ Graph placement method
- ▶ Hybrid placement methods
- ▶ Two-Step placement method
- ▶ Clustering placement methods

Two-Step Placement Method

- ▶ Constant-time methods for defining a set S of possible positions
 - ▶ Grid Set
 - ▶ Border Set
- ▶ Methods for selecting k influencing agent positions from S
 - ▶ Random
 - ▶ OneNeighbor
 - ▶ MaxNeighbors
 - ▶ MinUninfluenced



Determining Desired Positions - Clustering Methods

- ▶ Constant-time placement methods
- ▶ Graph placement method
- ▶ Hybrid placement methods
- ▶ Two-Step placement method
- ▶ Clustering placement methods

Clustering Placement Methods

Well-recognized clustering methods to identify k clusters of flocking agent positions



Figure : Farthest First



Figure : K-Means

Clustering Placement: Farthest First

- ▶ Randomly choose a flocking agent and place the first influencing agent
- ▶ Then, on each subsequent placement, place an influencing agent at the flocking agent **farthest** from all previously placed influencing agents



Clustering Placement: K-Means

- ▶ Choose k flocking agents as cluster centers
- ▶ Until convergence:
 - ▶ Assign all flocking agents to nearest cluster center
 - ▶ Calculate centroid for each cluster, centroid becomes a new cluster center



Experimental Results - Clustering Placement

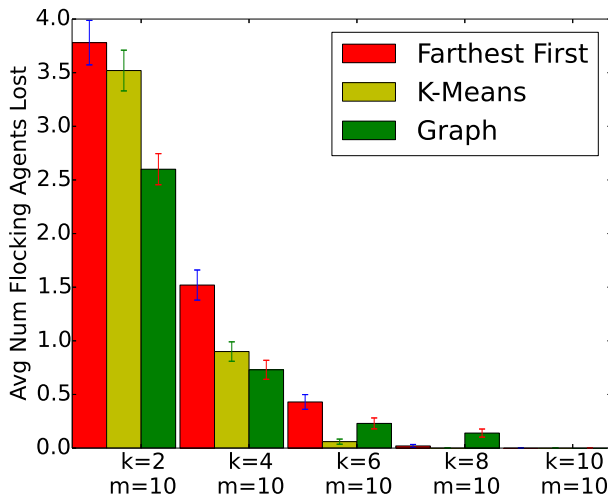


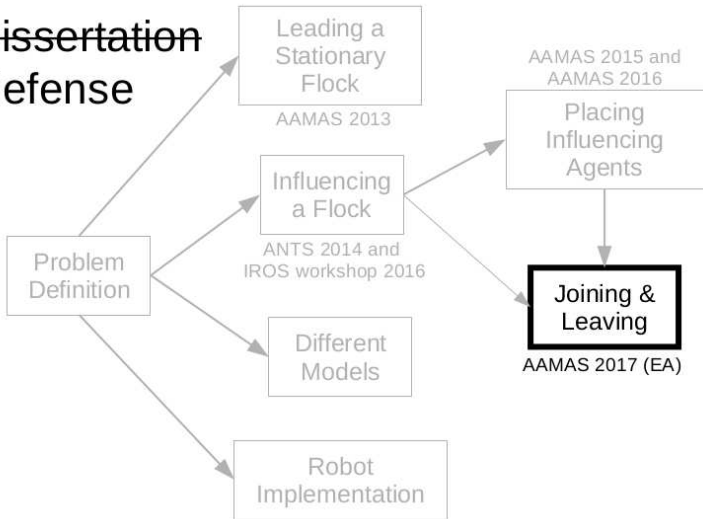
Figure : 100 trials, error bars = standard error of the mean.

Placing Influencing Agents into a Flock - Summary

- ▶ The position of influencing agents within the flock affects their influence
- ▶ Contributed various approaches for placing influencing agents directly into a flock
- ▶ Evaluated on various flock sizes and compositions

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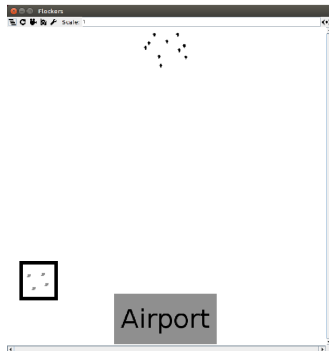
Joining and Leaving a Flock

- ▶ Consider how influencing agents should **join** a flock in motion with the goal of altering its orientation
 - ▶ Why don't placement methods work?
- ▶ Consider how influencing agents can **leave** a flock

K. Genter and P. Stone. Agent behaviors for joining and leaving a flock (Extended Abstract). In AAMAS'17, May 2017.

Joining and Leaving a Flock

- ▶ Leave a station positioned ahead of the flock
- ▶ Assumptions:
 - ▶ Influencing agents can not travel **faster** than the flocking agents
 - ▶ Flock is cohesive and flocking in a particular direction



Approaches for Joining and Leaving a Flock

Two different scenarios:

- ▶ **Hover**: the influencing agents are able to hover with a particular orientation at a set position
- ▶ **Intersect**: the influencing agents maintain the same velocity as the flocking agents

Approaches for Joining a Flock

Two different scenarios:

- ▶ Hover: the influencing agents are able to hover with a particular orientation at a set position
- ▶ **Intersect**: the influencing agents maintain the same velocity as the flocking agents

Intersect Approach

- ▶ Target formation
- ▶ Calculate the desired time to join the flock and begin influencing
 - ▶ All influencing agents join together
 - ▶ Constant velocity

Intersect: Target Formations



(a) Push to Goal Line



(b) Forward Line



(c) Push to Goal Funnel



(d) Forward Funnel



(e) L Corral

Approaches for Leaving a Flock

Three different scenarios:

- ▶ Hover: all of the influencing agents hover in place facing the desired orientation of the flock
- ▶ **Nearest Edge**: leave the flock at the nearest edge
- ▶ **Influence while Leaving**: trade-off between influencing and leaving

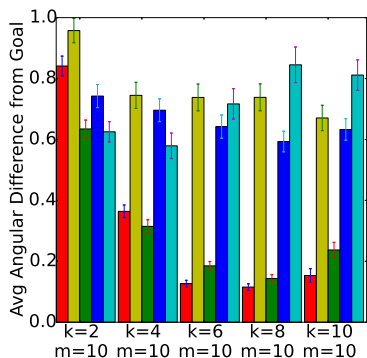
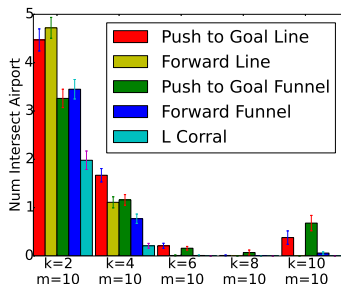
Nearest Edge Approach for Leaving

- ▶ Influencing agents know approximate position in flock
- ▶ When time to leave, orient towards the nearest edge
 - ▶ **Nearest 3-edge**: don't leave towards the flock's desired orientation
 - ▶ **Nearest 2-edge**: don't leave towards the airport or the flock's desired orientation

Influence while Leaving Approach for Leaving

- ▶ Alternate between influencing and leaving
- ▶ If no neighbors, leave
- ▶ Else if the neighbors aren't aligned to the goal or the influencing agent has left the flock for too long, influence
- ▶ Else, leave

Sample Results



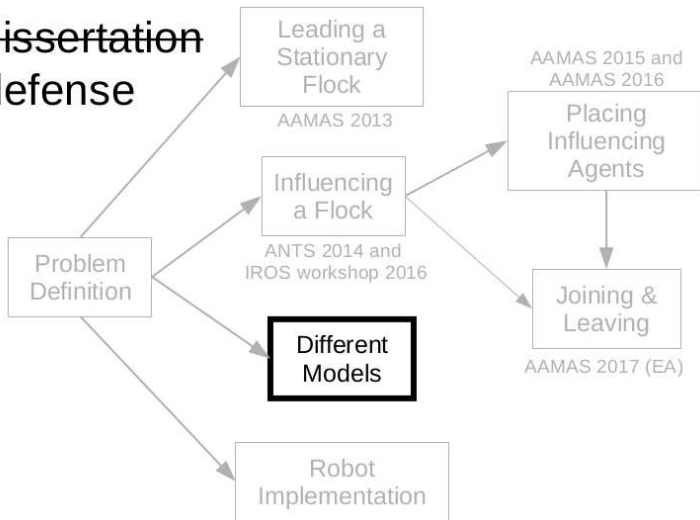
Intersect for joining and *influence while leaving* for leaving. 100 trials, error bars = standard error of the mean.

Joining and Leaving a Flock - Summary

- ▶ Considered how influencing agents should **join** a flock in motion with the goal of altering its orientation
- ▶ Considered how influencing agents can **leave** a flock

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Evaluation on Different Flocking Models

- ▶ How well do the methods generalize to other flocking models?
 - ▶ Alternate neighborhood models
 - ▶ Full Reynolds flocking algorithm
- ▶ Do the influencing agents need to know the **true** model?

Different Neighborhood Models

- ▶ Considered alternate neighborhood models:
 - ▶ Visibility Sector
 - ▶ N-Nearest Neighbors
 - ▶ Weighted Influence
- ▶ The 1-Step lookahead algorithm generalized to different neighborhood models
- ▶ It is not critical that the influencing agents know the true neighborhood model

Different Influence Models

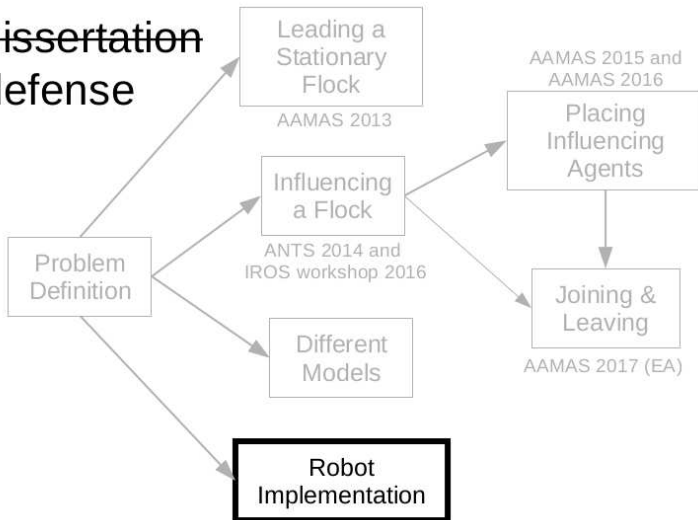
- ▶ Considered three aspects of Reynolds flocking algorithm:
 - ▶ Alignment: towards average heading of neighbors
 - ▶ Separation: away from neighbors to avoid collisions
 - ▶ Cohesion: towards the average position of neighbors
- ▶ Best for influencing agents to believe the flocking agents are behaving according to the **Alignment** influence model

Evaluation on Different Flocking Models - Summary

- ▶ The methods generalize well to other (similar) flocking models
 - ▶ Alternate neighborhood models
 - ▶ Full Reynolds flocking algorithm
- ▶ Knowing the **true** model utilized by the flocking agents is not critical

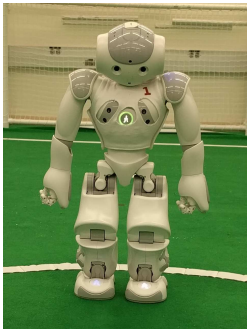
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Robot Implementation

How can influencing agents influence a flock of bipedal robots to avoid a particular area?



Robot Implementation - Summary

- ▶ The 1-Step lookahead algorithm can be utilized to influence a flock of bipedal robots with small alterations
 - ▶ Only consider influencing agent orientations that can be reached in one time step

Thesis Question

Thesis Question:

How can influencing agents be utilized in various types of flocks in order to influence these flocks towards a particular behavior?

Future Work

- ▶ Additional coordinated influencing agent behaviors
- ▶ Placement based on agent heading
- ▶ How to enter a flock
- ▶ Learning flocking models and how to influence
- ▶ Extending robot implementation to robot birds
- ▶ Extensions to other domains: endangered species, validating theories of biologists

Conclusion

Thesis Question:

How can influencing agents be utilized in various types of flocks in order to influence these flocks towards a particular behavior?

- ▶ How should the influencing agents behave?
- ▶ Where should the influencing agents be placed?
- ▶ How should the influencing agents join and leave the flock?
- ▶ How well do the methods generalize?
- ▶ How well do the methods transfer from simulation to a robot platform?