# Void Reduction in Self-Healing Swarms (Using proximity detection)

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## Video

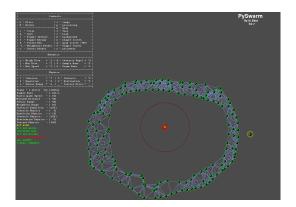


Figure: Simulator

https://www.youtube.com/watch?v=iyMSpj10elk

#### Introduction

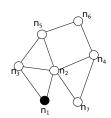
- 1 How did we do that?
- 2 Set the Ground Rules!
  - Swarm Rules
- 3 Communications and/or Sensing
  - Communications
  - Sensing
- 4 Void Reduction
  - Model
  - Local Effect
  - Global Effect
  - Simulated Results

## Swarm Rules

- Swarms consist of many agents (mobile robots or drones) that interact according to a simple set of rules.
- We consider swarms of agents that:
  - are capable of detecting their neighbours (proximity detection).
  - do not require any another form of communication.
- Swarms can be made fault tolerant (resilient to agent loss).

# Why no communications?

- Communication propagation protocol demands computing overhead.
  - $n_1 \rightarrow n_3$
  - $\bullet$   $n_1 \rightarrow n_2$
  - $n_3 \rightarrow n_2$  (decision!)
- Message propagation takes time which limits swarm size.



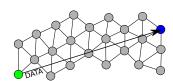


Figure: Swarm Communications

# **Proximity Sensing**

- No communication is necessary apart from proximity detection.
- Arbitrary sized swarms are possible.
- Agent attributes include various ranges as shown in figure.

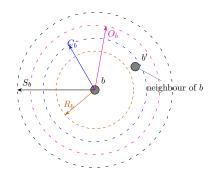


Figure: Ranges:  $S_b$  = sensing,  $O_b$  = obstacles avoidance,  $R_b$  = repulsion,  $C_b$  = cohesion.

## Agent Movement

 Movement of agent b is computed as the weighted sum of 4 vectors, as shown in equation 1

$$v(b) = k_c v_c(b) + k_r v_r(b) + k_d v_d(b) + k_o v_o(b)$$
 (1)

- $v_c(b)$ : cohesion term ensures agents remain part of the swarm.
- $v_r(b)$ : repulsion term ensures agents do not collide.
- $v_d(b)$ : destination vector for goal based swarms.
- $v_o(b)$ : obstacle avoidance vector.
- Numerical weights  $k_c$ ,  $k_r$ ,  $k_d$ ,  $k_o$  to allow tuning of relative effects.

#### The Swarm



Figure: Stable Structures

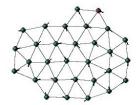


Figure: The Swarm - (Simulator)

# Perimeter Detection

#### NOTE

Perimeter detection is used as part of the *void reduction* process. This will be discussed later.

- Perimeter detection allows for directional coordination with reduced resource usage.
  - 'Internal' agents don't need to use their GPS.
- Reduces computational overhead in agents.

Model

# What is a Perimeter?

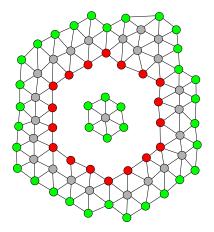


Figure: Internal (red) and external (green) perimeters

Model

# What is a Void?

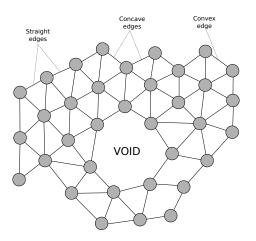


Figure: Concave components of a perimeter (voids)

# Perimeter Detection (Concave)

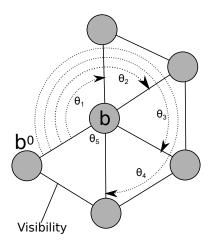


Figure: Concave gap (Void Reduction)

# Perimeter Detection (Convex)

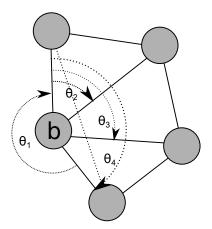


Figure: Convex gap

#### Void Reduction Movement

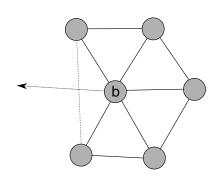


Figure: Concave detection

As part of the perimeter detection a pair  $G_b = \{n_1, n_2\}$  of agents is generated. These are the first two agents identified as creating a 'gap' in agent b's neighbours. Equation 2 calculates the centroid of the identified 'gap'.

$$D_{pos}(b) = \frac{1}{2}(n_1 + n_2) \qquad (2)$$

 $(n_1, n_2 \text{ label agents and also denote their position vectors.})$ 

# Agent movement

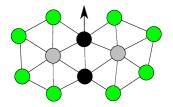


Figure: Initial positions

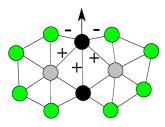


Figure: Agent movement

Global Effect

# Agent movement

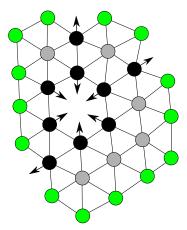


Figure: Initial positions

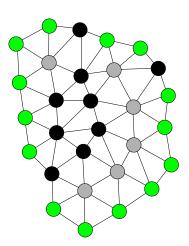


Figure: Agent movement

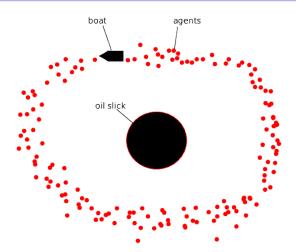


Figure: Oil Slick Encapsulation

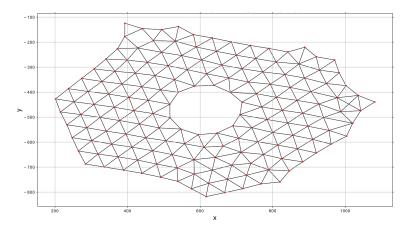


Figure: Oil Slick Encapsulation

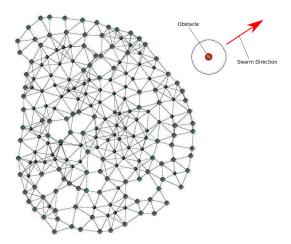


Figure: Mobile Swarm

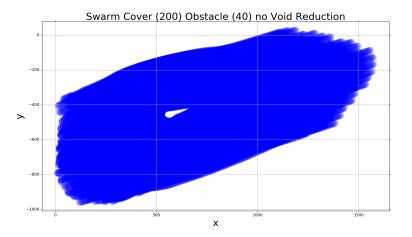


Figure: No void reduction - path plot

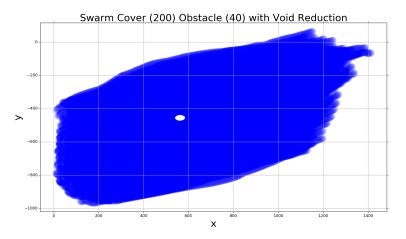


Figure: Void reduction - path plot

# Summary

- Void reduction has a local effect creating a global emergent behaviour that improves the shape and structure of a swarm.
- Void reduction removes anomalies on internal and external perimeters.
- Void reduction can be applied to both static swarms and directional swarms.

### Thank You

THANK YOU! QUESTIONS?

## Agent Movement

Here are the equations defining the movement vectors for cohesion and repulsion.

b, b' denote agents and also (inside the  $\sum$ s), their position vectors.  $\mathcal{R}_b$  is the set of agents within agent b's repulsion range;  $\mathcal{C}_b$  is the set of agents within its cohesion range.

$$v_r(b) = \frac{1}{|\mathcal{R}_b|} \left( \sum_{b' \in \mathcal{R}_b} \left( 1 - \frac{|b'|}{R_b} \right) b' \right) \tag{3}$$

$$v_c(b) = \frac{-1}{|\mathcal{C}_b|} \left( \sum_{b' \in \mathcal{C}_b} b' \right) \tag{4}$$

## Agent Movement

$$v_d(b) = d (5)$$

d is a constant vector which points to a destination.

$$v_o(b) = O_b \left( \sum_{o \in \mathcal{O}_b} \hat{o} \right)^{\wedge} \tag{6}$$

 $O_b$  is b's detection detection range;  $O_b$  is the set of obstacles within this range.

o denotes an obstacle and (within the  $\sum$ ) its position vector. The caret  $\hat{}$  denotes normalisation of a vector to unit length.