

# Void Reduction in Self-Healing Swarms

(Using proximity detection)

Neil Eliot<sup>1</sup>, David Kendall<sup>1</sup>, Alun Moon<sup>1</sup>, Michael Brockway<sup>1</sup>,  
Martyn Amos<sup>1</sup>

<sup>1</sup>Department of Computer and Information Sciences  
University of Northumbria

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## Video

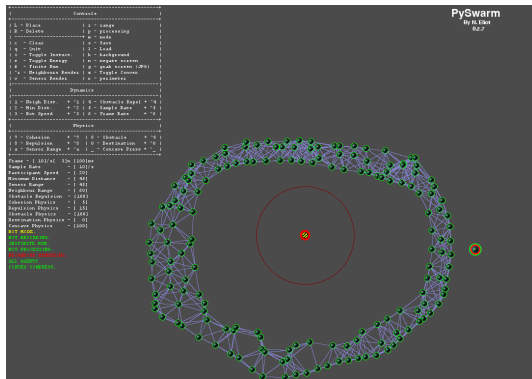


Figure: Simulator

<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:
  - 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 overhead.
  - $n_1 \rightarrow n_3$
  - $n_1 \rightarrow n_2$
  - $n_3 \rightarrow n_2$  (decision!)
- Message propagation takes time which limits swarm size.

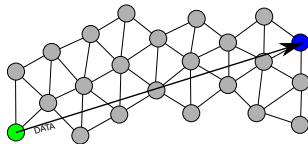
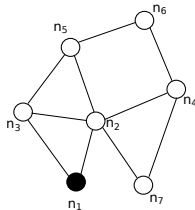
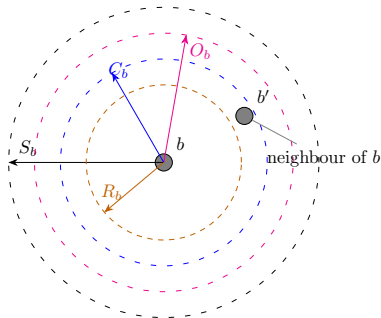


Figure: Swarm Communications

# Proximity Sensing

- Proximity detection only, no other form of communication required.
- Arbitrary sized swarms possible.
- Agent attributes include various ranges (as shown in figure).



**Figure:** Ranges -  $R_b$  - repulsion,  $C_b$  - cohesion,  $S_b$  - sensing, and  $O_b$  - obstacles avoidance.

# Agent Movement

- Agent movement is computed as the weighted sum of 4 vectors, as shown in equation 1

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

- $v_c(b)$  - cohesion to ensure agents remain part of the swarm.
- $v_r(b)$  - repulsion to ensure agents do not collide.
- $v_d(b)$  - destination vector for goal based swarms.
- $v_o(b)$  - obstacle avoidance vector.
- $k_c, k_r, k_d, k_o$  are weightings to allow modifications to the vector effects.

# The Swarm

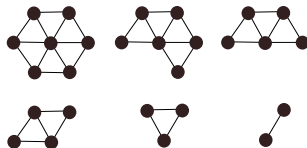


Figure: Stable Structures

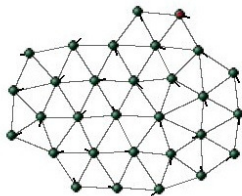


Figure: The Swarm - (Simulator)



# 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.

# What is a Perimeter?

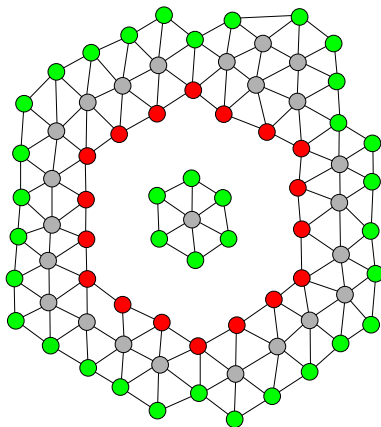


Figure: Internal (red) and external (green) perimeters

# Perimeter Detection (Concave)

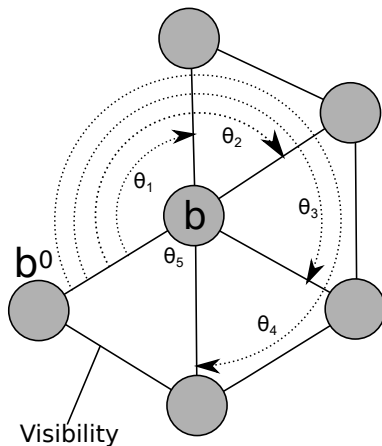


Figure: Concave gap (*Void Reduction*)

# Perimeter Detection (Convex)

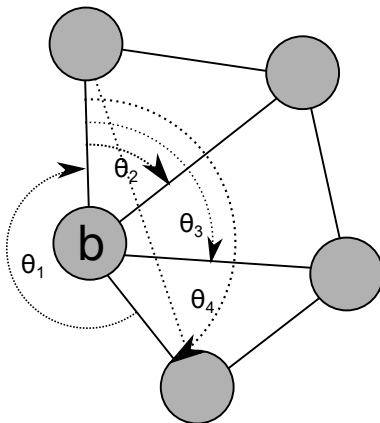


Figure: Convex gap

# Void Reduction Movement

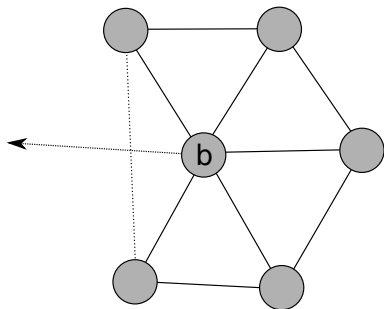


Figure: Concave detection

As part of the perimeter detection a pair ( $G_b$ ) of agents is generated. This is the first two agents identified as creating a 'gap' in agent  $b$ 's neighbours. Equation 2 calculates the centroid of the identified 'gap'.

$$D_{pos}(b) = \frac{1}{2} \sum_{n \in G_b} n \quad (2)$$

# Agent movement

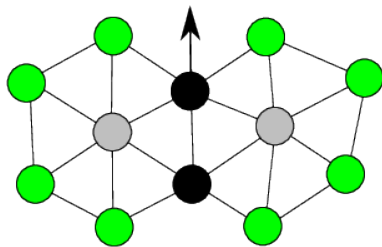


Figure: Initial positions

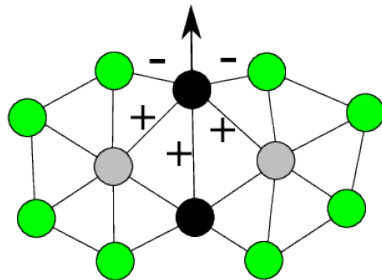


Figure: Agent movement

# Agent movement

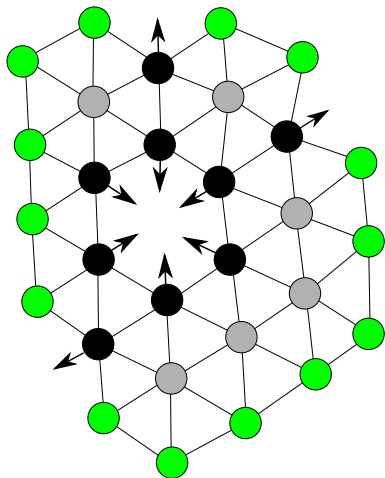


Figure: Initial positions

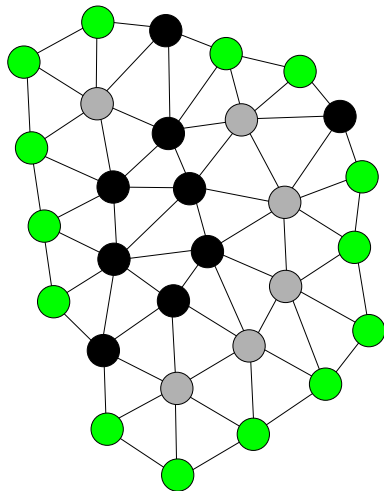


Figure: Agent movement

# Scenario

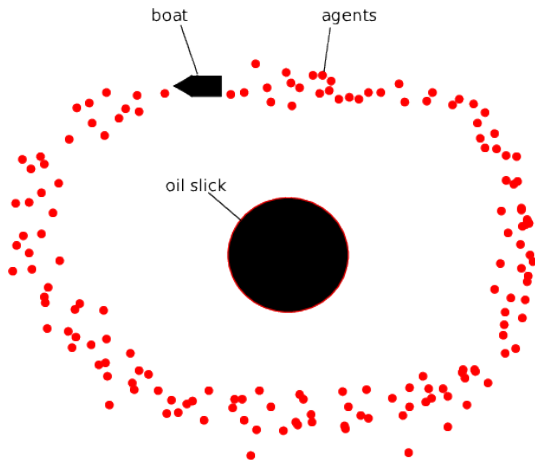


Figure: Oil Slick Encapsulation



# 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

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

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

# Agent Movement

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

$$\begin{aligned} v_o(b) &= O_b \hat{q}_o \\ \text{where } q_o &= \sum_{o \in \mathcal{O}_b} \hat{o} \end{aligned} \quad (6)$$

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