

Master's Degree in Computer Science and Engineering

# Unity for Collektive: Reducing Reality Gap in the Simulation of Collective Adaptive Systems

Thesis in:  
SOFTWARE PROCESS ENGINEERING

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# **Abstract**

Max 2000 characters, strict.



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# Chapter 1

## Introduction

Modern computing is moving away from the era of powerful and isolated machines toward one composed by massively interconnected ensembles of devices. We can observe this transition everywhere, from global Internet of Thing (IoT) sensor networks to smart city infrastructures. In such scenarios, the focus shifts from ‘how to compute’ to ‘how to coordinate’.

As the number of devices in these systems grows into the thousands or millions, traditional centralized management becomes a bottleneck. The latency, bandwidth constraints, and single-point-of-failure risks of a ‘command-and-control’ architecture make it unsuitable for the dynamic, often unpredictable environments these systems inhabit. Instead, we must look toward decentralized coordination, where collective intelligence arises from local interactions rather than global oversight.

This thesis explores the intersection of high-level collective programming and high-fidelity simulation. Specifically, it addresses the engineering gap between abstract coordination models, such as Aggregate Computing (AC), and the practical requirements of developing, testing, and deploying these models within realistic 3D environments. By leveraging the power of modern game engines and automated development workflows, this work aims to provide a robust infrastructure for the next generation of collective system design.

## 1.1 Motivation: Swarm Behaviour

The natural world provides the strongest precedence for the goal of resilient decentralized coordination. From the coordinated flashing of fireflies to the intricate architectural achievements of termite mounds and the smooth collective motion of starling murmurings, biological systems exhibit an efficiency that is frequently difficult for classical engineering to match. These phenomena, which are collectively referred to as Swarm Intelligence (SI), arise from the interaction of many simple agents that follow localized rules rather than from a global supervisor.

In a natural swarm, intelligence is inherently distributed and emergent. Individual agents (be they ants, bees or birds) possess only a partial perception of their surroundings. The collective however can solve high-order problems such as finding the shortest path to a food source or executing rapid evasive maneuvers against predators. From an engineering perspective, these systems offer three indispensable properties:

- the absence of a central controller; the loss of individual units does not compromise the mission.
- The logic governing ten agents often remains functional for ten thousand, as interactions remain local regardless of total population size.
- Swarms autonomously adapt to dynamic environments, re-configuring their behaviour in response to external stimuli.

As we attempt to port these characteristics into the digital and physical domains (specifically through paradigms like AC) we face a significant translation gap. While the mathematical models for collective logic are maturing, the infrastructure to test them in realistic, high-fidelity environments remains fragmented. To truly harness the potential of swarm behaviour in human-made systems, we must develop tools that can simulate the complex interplay between decentralized algorithms and the physical world.

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1.2. PROBLEM STATEMENT: ENGINEERING CHALLANGES IN  
SIMULATION

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## 1.2 Problem Statement: Engineering Challanges in Simulation

### 1.3 Contributions

#### 1.3.1 A Standardized Automation Workflow for Unity Pack- ages

#### 1.3.2 A 3D Aggregate Computing Simulator

### 1.3. CONTRIBUTIONS

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# **Chapter 2**

## **Background and State of the Art**

### **2.1 Distributed Systems and Organizational Complexity**

### **2.2 Self-Organizing Frameworks**

#### **2.2.1 Aggregate Computing**

### **2.3 Simulation Landscape**

#### **2.3.1 Paradigms**

#### **2.3.2 The Reality Gap**

#### **2.3.3 Reéalism vs. Scalability**

### **2.4 Game Engines as Simulators**



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# **Chapter 3**

## **Unity-Package-Template: Automated Unity Development Infrastructure**

### **3.1 Requirements**

### **3.2 Features**



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# **Chapter 4**

## **Collektive×Unity: Designing a 3D Simulator for Collective Systems**

### **4.1 Goal**

### **4.2 Requirements**

### **4.3 Architecture**



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# **Chapter 5**

## **Implementation of Collektive×Unity**

### **5.1 Design**

### **5.2 Implementation Details**



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# Chapter 6

## Case Study: Environment-aware Gradient Ascent



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# Chapter 7

## Results

### 7.1 Comparison with Socket-based Communication



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# Chapter 8

## Conclusions and Future Work



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