Heterogenous Swarming and Emerging Collective Motion of RVR and BOLT Robots:

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**Abstract**

**Swarming intelligence > swarms > boid > heterogeneous vs homogeneous > formations and collective motion > so what**

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# Introduction

As the field of robotics moved forward with advances in both the efficiency and capability of embedded systems utilised in robots so did the scope extend for robots to tackle an ever-growing set of larger and more complex challenges. An approach to these new challenges can be made through the implementation of swarm robotics, which is a branch of multi-agent robotics systems that is characterized by its emphasis towards the emulation of natural biological swarms, such as packs of wolves hunting prey; multi-agent swarms utilise a multitude of smaller and simpler agents that act together towards a collective intent.

Heterogeneous swarming presents an opportunity to broaden the scope and applications of robotic swarms, enabling robots with differing degrees of computational power, sensors and mobility to work together. While many classical tasks of robotics swarms, which are loosely based on patterns found in nature, such as path finding, source localization and area exploration and coverage. Furthermore nature utilizes collective motion and formations optimize efficiency of swarms with the implementation of formation control becoming an increasingly emerging research topic within the field.

While current research within swarm robotics that is aimed towards formation control is modelled, simulated and implemented with the use of homogeneous swarms, formations within heterogeneous swarms is mostly uncovered. With the advance of swarm robotics and the increasing complexity required of robot swarms to tackle the challenges presented, physical and behavioural heterogeneity within swarms presents a clear path to extend the capabilities of swarm robotics [1]. To both fully optimize as well as to further understand the opportunities presented by heterogeneous swarms it is important to research their full capabilities and limitations.

Formalise – aim of the paper is to create heterogeneous formations using homogeneous swarm code + generic formation code

# Background Info

## Boid Swarming Algorithm

The **boids** model was initially conceptualised by Craig Reynolds in 1986, which simulated the flocking motions of birds,

## Collective Motion using Deep Reinforcement Learning

asd

# Literature Review

## The Boid Model and Swarm Formations

Swarm robotics is defined and characterized by its attempt to emulate the swarm intelligences seen in the natural world through the lens of robotics; robots interact on a local level with simple rules, through this it is possible to design collective behaviors which are both scalable and robust [1]. One of the approaches towards this is the Boids Swarming model, which was originally conceptualized by Craig Reynolds in 1986, it sought to model the flocking of birds through a set of rules: separation – boids must not move too close to each other to avoid collisions, alignment – boids must move in the general direction of the rest of the flock, and cohesion – boids must move towards the center of the flock. By implementing these three rules Reynolds created a simulation that modelled the flocking of birds [2]. While Reynolds’ work in 1987 was centered from the view of computer graphics, the Boids model proved useful in the understandings of swarm intelligence and was first realized in the realm of robotics by Turgut et al. who developed a system for robots to determine the heading, position and velocity of their peers thus enabling them to form “flocking” swarms utilizing the Boid rules of separation, cohesion and alignment [3].

Collective behaviour within swarm intelligence is crucial to the real-world application and control of swarms, a key form of this is in creating defined formations. In recent history this challenge has been approached in many ways, however this paper aims to provide an extension of the existing formation control methods through the application to the real world in readily accessible and low complexity robotics, which present both easy access research [4], [5], practical [6] and educational opportunities [7]. Emergent behaviors of formations in swarms proves desirable as it seeks to provide swarms with optimal positioning for various tasks while maintaining the robustness and scalability of the swarm model. In the past swarming formations have been achieved through a variety of different methods, however many of these methods relied fully or partially on central controllers [8], [9], [10], this reduces the robustness and scalability of the swarm, additionally there have been many decentralized based approaches, those that utilised different forms machine learning through the use of a graphical/potential field type techniques seen in [11], [12], [13]. In addition to this there exists some prevailing research into formation control through emergent behaviours, which are implemented in a variety of methods seen in [14], [15]. These methods exist as a large wealth of knowledge in the development and implementation of swarming formation control systems.

However the current research into formation control through emergent behaviours is pervaded by homogeneous swarms [14], [15], with the new research presented in this thesis, we utilise deep learning to create completely emergent behaviour within heterogeneous swarms, this type of formation has been previously seen within homogeneous swarms and other types of multi-agent scenarios. For example, Hüttenrauch et al. controls the entire swarm through the use of deep reinforcement learning [16], while in a conference paper in 2021, Bezcioglu et al. demonstrates flocking through a global state space matrix utilizing deep reinforcement learning [17], these papers show that there is significant interest and investment in the conceptualization and realization of swarming formation control through a variety of different approaches, however there is not much literature to show research on a wide variety of formation control techniques being applied to heterogeneous swarms.

## Applications of Heterogeneous Swarms

Throughout nature both homogeneous and heterogeneous swarms develop to solve distinct and unique problems [18], [19], these are emulated within swarm robotics, with many papers demonstrating swarms of robots solving natural problems with key benefits being those of robustness, flexibility and scalability [6], [20], [21], [22], [23], [24]. These key features of swarming homogeneous and heterogeneous systems make them well suited to solving novel challenges and scenarios that traditional single and multi-agent systems do not perform as well in. Swarm robotics can be applied to a multitude of different applications, the applications that are typically well suited to this are ones that cover regions, involve high risk, scale up or down in time required, or in applications that required redundancy [24]. For example, swarm robotics have been increasingly researched as a useful asset in mine clearing, this is due to both the high risk and large time and size scales, as minefields are both vast but also having by having a large swarm accidental mine detonations do not result in significant losses to a project’s capability [25], heterogeneity could be applied to this through utilizing more complex drones to perform mission planning while swarms are used to collect data on a smaller scale.

Both physical and behavioural heterogeneity within swarms opens up new dimensions in terms of capabilities and possibilities for swarm intelligence [26]. While hallmark papers such as Swarmanoid [27] conceptualize and implement models for some of the possible applications of heterogeneous swarms, formation control is a key element to the application of heterogeneous swarms with formations providing added stability, efficiency and task – based optimization to swarms. Recent research has shown success in formations of heterogeneous swarms with differing levels of diversity within the swarm populations [28], [29], [30]. However, many of these heterogeneous swarm formations implement centralized, semi-centralized or computationally heavy decentralized formations in comparison to the implementations presented in this research project.

To introduce swarming formation this project will utilise prior research done using homogeneous swarms. Boid rules were modified using deep reinforcement learning AI to generate emergent collective behaviours [31], within the paper by Abpeikar et al. this is shown to be a viable implementation within simulation with the use of a homogeneous swarm and thus proves promising for adaptation to a heterogeneous swarm in similar conditions.

## Benefits of Heterogeneous Swarms

Swarm intelligence and robotics is heavily steeped in the thought of biomimicry, a practice that seeks to learn and copy nature to solve a variety of problems, with many problems having existed in nature and with the evolution of swarm behaviour within less singularly cognitive species, these provide an example to how lower complexity robots can complete challenges requiring higher level thinking through collective behaviour [32]. While studies are largely dominated with the use of homogeneous swarms [1], the idea of heterogeneous swarms, swarms made of different types of robots, has become increasingly researched in recent times as the challenges posed to robotic swarms have evolved [33]. An approach to the widely opened horizons of heterogeneous swarming can be seen within the experiment by Dorigo et al. [27], where three different classes of robots, “eye”, “hand” and “foot” robots, in which different elements of the swarm provide significantly different and more capabilities to the heterogeneous swarm over that of a homogenous swarm, other novel approaches to develop useful heterogeneous swarms include ones utilizing a “shepherding” method in which powerful, less mobile robots collect and slave groups of less powerful, more mobile robots [34], synergizing the benefits robustness and scalability of swarm robotics with the complexity and power of classical single agent systems. Furthermore, it has been demonstrated by Prorok et al. that heterogeneous swarms can use decentralized control to divide and conquer across multiple ‘species’ specific tasks, which shows how larger heterogeneous swarms can provide significantly more modularity and the ability to solve multiple problem types simultaneously [28].

Formations within swarms can provide both crucial placements to optimize efficiency as well as organization within swarms, this is particularly crucial to heterogeneous swarming. These could be seen useful across **talk more about the applications of heterogeneous swarms here and how these benefits can apply [was struggling to find actual sources on heterogeneous swarms being used in real life]**

# Planning

## Project Plan

The project’s upcoming deliverables are the interim report, viva voce and project seminar, due on 17 May 2024 and 28 May 2024 respectively. The remainder of the project’s timeline is dictated in Appendix [ insert gantt chart] which shows a Gantt chart of different tasks required as well as their beginning and expected completion dates. This timeline was designed with the efficient and thorough completion of the project deliverables by the 25 Oct 2024, which entails the submission of python code package, documentation as well as a summary report, by aiming to finish this by the 1 Oct 2024 this allows the possibility for either polishing of the final submission or to work on potential extensions before the due date. To maximise the efficiency towards completing the project the tasks are placed in both a sequential and logical order as well as in parallel to ensure that during particular weeks it is possible to complete tasks that required lab access, such as testing, debugging and datalogging that requires Vicon as well as completing tasks outside of lab hours such as simulations, code development or optimization and analysis of results as well as literature review.

The project is effectively approached in two phases:

1. **Phase 1: BOLT and RVR Swarm Boid Integration:** Review and understand existing RVR and BOLT swarming codes, modify and develop code to connect Bolt as a boid agent to the code. Validate swarm formation through Vicon tracking and debug console.
2. **Phase 2: Heterogeneous Formations:** Research, review and develop formation control for RVR and BOLT swarm to generate multiple different heterogeneous formations through the implementation of .

The practical element of this research project will entail the material requirements of Sphero Bolt and RVR robots to form a heterogeneous swarm. The Bolt robots run standalone off Bluetooth, however the RVR robots require the use of Raspberry Pi 3B+ single board computers to control them. Additionally, the research project also requires the use of the Vicon system for both datalogging and code function, these materials are pre-existing and/or on loan from UNSW Canberra thus the project will require no significant budget.

## Potential Difficulties

As the research project entails a practical component the nature of the project may evolve as certain different limitations come into play. These could appear within any stage of the project, simulations, design, testing and validation all have the potential reveal possible limitations. Additionally, there may be further project difficulties as the practical nature of the project adds to a variable of uncertainty. Hence potential project difficulties that may arise are:

Delays in the completion of tasks resulting in overall right-shifting of the timeline, this can result from issues such as additional testing and debugging being required as well as faulty, damaged or missing hardware requiring repair, replacement or purchase.

Initial difficulty in understanding, modifying and implementing existing swarming code as well as time required to further understand code to debug.

Bluetooth and Vicon integration may require more powerful hardware as well as extending troubleshooting time.

Compatibility issues may arise between RVR mounted Raspberry Pi 3B+s and Bolts.

By having a good situational awareness of the possible difficulties that may be encountered within this research project, these difficulties can be mitigated through a variety of precautionary measures to avoid issues during the progress of the research project:

1. Tracking tasks allows for both pre-planning and insights into current progress for task completion, this will result in less delays and keep progression through milestones timely.
2. By consulting with subject matter experts on the code base (thesis supervisors) as well as using up-to-date libraries the impact of this potential project difficulty can be mitigated
3. Testing data rates and reliability and consistency of connection early on, to discover what issues may arise as the project progresses.
4. Testing connections before beginning of running main code, implementation of alternative and contingency methods for Bluetooth.

Through the implementation of these strategies, as well as maintaining a good level of situational awareness, the potential difficulties mentioned above should be addressed and mitigated to a degree of acceptability, this is to ensure both the efficiency and thoroughness of the project.

# Methodology

## Research Design

The main objective of this research project is to achieve heterogeneous swarm formations using Sphero BOLT and RVR robots. In order to achieve this research and literature review has been conducted to analyse and understand different kinds of swarming and formation control in both homogeneous and heterogeneous swarms to gain background knowledge and a point of refence in the design and implementation of heterogeneous swarm formations.

## Background Theory and Analysis

Within the Boid swarming model, individual agents move with reference to the rest of the swarm, this is achieved through three fundamental forces: separation, cohesion and alignment [2].

* **Separation:** move away from nearby agents to prevent collision

A circular object with arrows pointing to the center

Description automatically generated with medium confidence

Figure 1A: Separation Diagram [2]

* **Alignment:** move towards the average heading of nearby agent

A clock with arrows pointing at the time

Description automatically generated

Figure 1B: Alignment Diagram [2]

* **Cohesion:** move towards the average position of swarm (center of swarm)

A circular object with arrows and triangles

Description automatically generated

Figure 1C: Alginment Diagram [2]

# Current Progress

## Sphero BOLT and RVR capabilities

The Sphero BOLT and RVR are both low cost consumer grade products that have differing capabilities in regards to both their sensors and effectors [35], [36], [37].

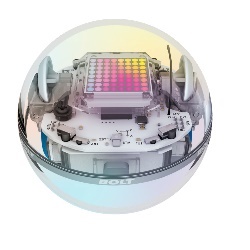
 

Figure 2A (left) and Figure 2B (right): Sphero BOLT (left) and RVR (right)

The BOLT (figure 2A) is a robot that operates within a spherical housing it communicates with the RVR via bluetooth. It has 2 drive motors with encoders and a sensor suite consisting of a light sensor to detect ambient light between 0 – 100 000 lux, four infrared emitter and receiver pairs for communication with other Sphero products, an inertial measurement unit which consists of an accelerometer, gyro and magnetometer which is used to read the robots acceleration and heading to then calculate velocity and position through dead reckoning [35].

The other agent utilised in the heterogeneous swarm is the Sphero RVR with a Raspberry Pi 3B+, the Raspberry Pi communicates with RVR utilizing a USB A for 5V power, as well as the UART GPIO pins on the Raspberry Pi for communication. The Sphero RVR utilizes 2 drive motors with encoders that are controlled through a drive controller utilizing a skid steer drive motion. The RVR also has a sensor suite with a similarly constructed IMU to the BOLT while also incorporating a colour detection sensor and ambient light sensor [37].

To effectively operate both RVR and BOLT in conjunction with each other and to avoid the inherent errors that develop throughout the experiment run time using the IMU system, the Vicon system is used, the UNSW Canberra RAS Laboratory Vicon system consists of 12 cameras that use vision processing to create a GPS – like system within the laboratory environment [38].

## Sphero BOLT and RVR Communications

To establish communications between the Sphero BOLT and RVR requires Bluetooth 4.0+, this was initially tested with the Raspberry Pi’s onboard Bluetooth device, however the Bluetooth service was not able to be started within the Raspbian OS environment (known issue). A current workaround this was to utilise a Bluetooth USB dongle connected to the Raspberry Pi. To implement this the Raspberry Pi utilizes the bleak (Bluetooth low energy platform agnostic client) library to connect to the Sphero BOLTs, with Bluetooth connected the BOLTs can send and receive data through to the RVRs.

## Swarming with Vicon

The implementation of the basic Boid swarm incorporating RVR and BOLT utilizes each RVR hosting a group of BOLTs. In this situation the term ‘hosting’ refers to the Raspberry Pi handling all of the BOLTs location data, movement commands as well as communications with other RVRs and BOLTs. To ascertain the BOLT and RVR’s true position within the local positioning system Vicon is used to determine where each robot is currently operating. See figure below add diagram or photo to demonstrate this

This data is then sent through a local network connection to the Vicon server which passes this through the ‘vicon\_bridge.py’. This code initializes the Vicon data stream on the server as well as transmitting this data through local network connection to the RVR. This has been modified so that the data is transmitted using the following scheme: more flowcharts?

|  |  |  |  |
| --- | --- | --- | --- |
| Index | 0 | 1 | 2 |
| Data | X Coordinate | Y Coordinate | Robot Identifer String |

The Raspberry Pi’s run the script ‘rvr\_swarm\_controller.py’, this script initializes the RVR and BOLTs as **Agent** objects, each Agent object is then treated as a separate Boid swarm member, with the control loop taking its location:

* **if RVR location received:** update RVR Agent object and calculate Boid vector, broadcast location to unique known Boid ID over local network.
* **if BOLT location received:** update BOLT Agent object and calculate Boid vector, broadcast location to unique known Boid ID over local network.

This code takes place recursively and allows for each Raspberry Pi to calculate the correct Boid vector for each individual, with each Agent method having specific BOLT and RVR code for either robot, this swarm system is depicted in the figure.

This code was run and the results shown are shown below: [insert the results here and perform some analysis] comment -> results will show some swarming however will have required a maximum speed rule implementation

## Collective Motion in Simulation

Add to this section once simulation code is obtained, modify dimensions of each robot in Copellia, ie throw in the simulation results, talk about efficacy and draw back references to paper as well as impacts of differing size and speeds of robots

# Future Work

## Implement Emergent Collective Motion

The simulation has shown promise in the implementation of the current method, with functioning heterogeneous swarming working, the reinforcement learning model can be applied directly to the ‘Boid\_rules.py’ file and can hence achieve the desired collective motion that is the capitulation of this research project.

## Extension: Asymmetric Behaviour

Heterogeneous swarms are comprised of multiple different types of robots, the benefit of this is that swarming can draw upon different types robots and their respective strengths to create a cohesive swarm that exploits the strengths of each agent. In this case, a key strength of the BOLT is its simplicity and disposability in a high risk scenario. To model the conceptual idea the Boid swarm will implement a rule within formations to place BOLTs in lead positions as ‘scouts’ in front of the RVR, due to the RVR having a slower top speed but being more capable in realms of communications and sensing due to the Raspberry Pi.

## Extension: Heterogeneous Plug and Play

Comment: random idea, but modifying the code to be able to use any kind of robot, so that the only code that the Raspberry Pi would need to host would be the boid swarming control, with other robots simply needing to know where and when to move to a new location, with all of that data being handled on the Raspberry Pi and distributed around.

# Conclusion

Objective

Summarise the research proposal

Summarise work done and future work

Value of research, including responding to the gap in literature

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