PRIORITIZED FORAGING USING ROBOT SWARMS

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Abstract

This study aim to improve on the problem of prioritized foraging by using robotic swarms. The attempt will introducing clustering during the foraging process to improve the over all performance of foraging. The paper focuses on swarm intelligence and swarm robotics through using social insect colonies as the work force. Previous work has shown that ant cemetery clustering and honey bee foraging algorithms are successful in collaboration for this study. Honey bee foraging is used for the ability to adapt to dynamic environment and the ant cemetery clustering not requiring communication from its peers or pheromone deposits. These algorithms are highly successful at path finding, grouping and retrieval of prioritized object, this directly correlates to the optimization goal.

1   Introduction

The gold mining industry and grown complex where deep shaft mining is required to access richer gold deposits. At these depths practicality of humans work becomes impossible, highly expensive & dangerous. To retrieve these gold deposits using Swarm Robotics will be more efficient, safer & at lower maintenance costs. These is swarm robotics model is made possible by using Swarm Intelligence.

Swarm Intelligence (SI) is a broad field of modeling algorithms after animal & insect behavior in nature [6]. This paper will focus on the algorithms simulated after ant cemetery clustering and of honey bee foraging. The one this these colony based eco-systems has as a priority is foraging. This is defined as search and retrieve procees from the environment to a collection point [11].

Swarm robotics uses a large amount of robots with simplistic abilities working together in a collaborative function [4] as it is designed by SI. It can be seen as an entity with mechanisms where each is a specific behavior where together these entities collaborate in a swarm. These algorithms are build around the idea of self-organized social systems where the dynamic mechanics drive the entire system towards a goal [6].

The research methodology to be followed and the process to complete the study. A literature review of background research relevant to the topic helping to get a better understanding and show the knowledge obtained during preparation for the paper. The project plan showing the sub-section of the project and project projection from start to submission. Also to show the study timely and no other research has been done on this exact topic before.

I will follow up on a publication by Jade Abbott [1] about Prioritized foraging of multiple foraging methods. In her study the methods where tested against each other to determine their efficiency in different situations. From the study the honey bee algorithm has performed best under most situations, and indicates that it would perform best in dynamic environments. This study will introduce the property of clustering before and during foraging to improve the foraging process due to prioritized objects being in a closer vicinity of each other.

2 Problem Statement

Prioritized Foraging using Robot Swarms is required by the mining industry when looking at deep mining where humans can't retrieve the gold ore after blasting has occurred. This is due to the reduced amount of breathable air and over whelming heat, increasing operation cost and unacceptable working conditions. All of this is irrelevant when looking at using robots to retrieve the gold and working prolonged hours.

The empirical study will simulate the behavior which will be programmed on a swarm of robots to cluster and forage the gold and rubble out of the mine. The foraging aspect will come in with retrieving the objects. Clustering to the grouping of objects to simplify the foraging process and prioritization to the gold and rubble objects respectively. This work research foraging and clustering but limits itself to only Ant Cemetery Clustering and Honey Bee Foraging, to keep with the SI model applicable to robots and by results of the standing study [1].

I am aiming to determine how clustering of the prioritized items will improve the foraging process, division of labor will be introduced where robots switch between clustering and foraging as needed to improve performance.

This empirical study will prove that deep mining can be undertaken with the use of robots when physical simulations have been tested. No other study has been done on concurrent clustering of the prioritized object while foraging takes place to improve feasibility and performance

3 Methodology

The proposed paper will consist of Prototyping and Simulations to be discussed in section 3.1 and 3.2.

3.1 Prototypes

A prototype will be developed combining the algorithms being used to be able to run the simulations. It will consist of the following components of development:

- A controller which controls all the simulations to be run with all the variations of settings.

- Honey bee foraging mechanism module, where each robot entity explores the search space for the prioritized object to carry them out the collection point as a bee would honey.

- Ant cemetery clustering mechanism module, where each robot entity when discovering a prioritized object it has to cluster it as ant would his dead.

- Division of labor module controlling the decision process where each robot entity must decide between the two mechanic modules.

3.2 Simulations

The simulations will test this prototype on a 2D grid that will simulate the environment where the robots might find themselves. The set of configurations to be tested are drawn from possible situations described in [1]:

- Environment layout of how the prioritized items are scattered around the search space.

- Grid size variation the search space and increasing complexity and run time.

- Coverage percentage of prioritized items over the open search spaces.

- Ratio of the amount of gold objects versus that of the rubble.

- Priority of foraging gold over rubble if in a close proximity.

- Number of robot entities in the search space.

Each simulation has a long run time due to iteration count with simplistic probabilities. This implies high concurrency is more important for running multiple simulation and not high performance processors are required. The simulations will be run on a Amazon compute module to improve testing times.

4 Literature Survey

The study will be based around a paper by Abbott [1], various foraging methods where tested against one another: naive foraging, honey ant & desert ant. This study will build on previous work by taking the proven best foraging algorithm and add a clustering process. It can be seen as a dynamic system where clustering and foraging are inter changed by each robot as the need occurs. The require research fields to complete this study is discussed in detail in the following sub-sections.

4.1 Swarm Intelligence / Robotics

Swarm Intelligence (SI) as defined by Gerando Beni [2] it is 'A group of non-intelligent robots ("machines") capable of universal material computation'. He further stated that there SI can be divided in to sets, where ACO and PSO algorithms are more suited to swarm automata due to its pattern analysis behavior, and termites behavior of structure building is suited to intelligent robot behavior. The foraging and clustering behavior used is more suited for swarm robots due to it does not follow any pattern analysis but more of an opportunistic behavior by its mechanical ability.

Swarm-based robotics is defined by Bonabeau *et al* [7] that it may be able to perform tasks without needing a explicit representation of the environment. These self organized social insects use a communication method call stigmergy [7], it can be in a form of direct and in-direct communication. Where direct is in the form of physical information exchange & in-direct by means of environment changes.

Stigmergy will be used as direct and in-direct by this study in the following ways:

- Direct: The honey bee broadcasts a foraging site by dancing when it arrives at the hive, this dance indicates the direction the other bees should follow to the site.

- In-direct: Ant cemetery clustering where each ant exams the environment to determine if an object should be picked up or dropped while clustering.

These communication methods will discussed in the expanded section of clustering and foraging.

4.2 Clustering

Ant colony clustering will be used due to the application where it needs to be develop for a large amount of cooperative robots. This possible by the social ability of ant and how they work together to achieve a goal. A basic ant colony clustering model was inspired by Chrtien [3] about the *Lasius niger ants* and develop by Deneubourg [9] as the first algorithm. This study will only focus on a improved ant cemetery algorithm by Lumer & Faieta [8].

These clustering algorithms use the ant cemetery model where ant pick up died ant from a low populated area & dropping it when its in a higher populated area or when the work load becomes to high. The algorithm aim to maximized inter-cluster distance (cluster separation) and minimize intra-cluster distance (cluster density). While performing clustering the ants will be moving around randomly in the search space monitoring a neighborhood of N^2 where N is the visual distance.

The "local" population density $\lambda(y_a)$ where i is the ants position at time step t and γ the scale of dissimilarity is given as [8]:

$$\lambda(y_a) = max \left\{ 0, \frac{1}{n_N^2} \sum_{y_b \in N_{n_N x n_N}(i)} \left(1 - \frac{d(y_a, y_b)}{\gamma} \right) \right\}$$

In neighborhood N and n_N the sites surrounding i. The constant γ determines is two object of priorities have to be placed together.

Ants use probability to decide it a object needs to be picked up if it isn't burdaned and dropped if the amount of work exceeds the limits [8]:

$$P_p(y_a) = \left(\frac{\gamma_1}{\gamma_1 + \lambda_{y_a}} \right)^2$$

$$P_d(y_a) = \begin{cases} 2\lambda(y_a) & \text{if} \lambda(y_a) < \gamma_2 \\ 1 & \text{if} \lambda(y_a) \geq \gamma_2 \end{cases}$$

These measures allows progression without any physical communication between the robots it can rely only on detection of the environment. The clustering process will not only be used as a process to improve the performance of foraging, but also as a testing condition where the search space is pre-clustered. This will especially test the ability of the division of labor to see if the ant realize it is already clustered and switch to a bee state.

4.3 Foraging

Honey bee foraging will be used for collecting the prioritized objects from the environment as they are clustered. The honey bee specific algorithm is used due to the results from [1] where it has shown good results in a most cases and in dynamic environments. This process was derived from a study done by Seeley [10]. There are three roles bees take during foraging:

- Scouts: Exploring the environment to find new sources of the prioritized objects. Once a source is found it return to the "hive" and broadcasts the message of the source. This message contains the direction vector to be used by the remainder bees to follow towards the source and its distance. The process is call recruiting of unemployed foragers.

- Unemployed foragers: Waits in the "hive" for a dance to be detected. Once the message is received it decides towards which source it will move, becoming an employed forager.

- Employed foragers: Using the source information it attempts to find the prioritized objects. If it was able to find the objects it picks up an object and returns to the hive where unemployed foragers are waiting to offload it.

During the scouting process the bee is constantly aware of the hives location and moves randomly through the search space until objects are found. If a source is found it moves directly as possible back to the hive to preform the dance for unemployed foragers.

Scouts use density to determine the viability of a forage site as μ_t, this is detected within the visible distance of the robot [10].

$$\mu_t = \frac{1}{n} \sum_{i=1}^{n} k_{i_t}$$

Where the distance value received from the sensor of item t, k_{i_t} is calculated as:

$$k_{i_t} = \begin{cases} k_i & \text{if item } i \text{ is type } t \\ 0 & \text{otherwise} \end{cases}$$

The following cases when non-prioritized item are foraged as from [1]:

- If no prioritized items was located within time f_{max}.

- Non-prioritized items are foraged until the site can not be located where it switches to prioritized items.

- It will not communicate the location of non-prioritized objects.

4.4 Division of labor

A colony has proven to be efficient at dividing labor between its working within the ecological system, this is done by giving a subset of tasks to each

robot.

Multiple task selection [5] uses a threshold vector θ_k and θ_{kj} the stimulus threshold, where k denotes the number of worker on the task j. After a specified time a robots will have a chance to select a new task, it may happen that it is the same task, the probability of the function selected is [5]:

$$P_{\theta_{khj}}(s_j) = \frac{s_j^\omega}{s_j^\omega + \theta_{kj}^\omega}$$
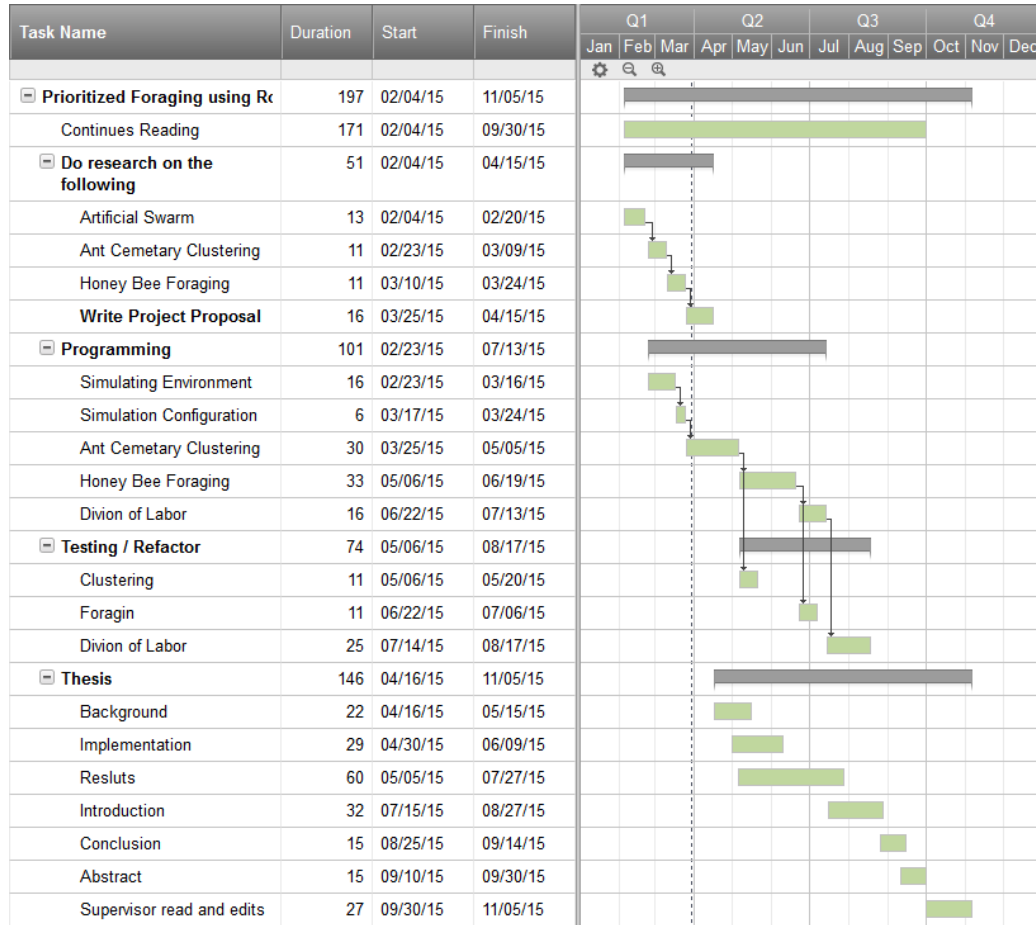
The division of labor process has not been finalized upon of which algorithm will be used by each robot to switch between clustering and foraging. This decision will only be made if times allows for it after the other process have been thoroughly been tested and deemed completed.

5   Project Planning

The project plan covers research, prototyping, testing, re-factoring, writing of the article & supervisor review. These section cover the following:

- Research :
  Read up on all the required topics starting at the that this research follows up on, then to the references of that paper and further.

- Prototyping / Programming :
  Implementation of the prototype to be able to test the improvement of clustering for the foraging and division of labor between the processes.

- Testing / Re-factoring :
  Running simulations of each configuration clustering and foraging. Also to implement changes required if it is necessary after a testing cycle is completed.

- Writing article :
  Writing the article to be submitted of approximate 15 pages.

- Supervisor review & edits :
  The supervisor will review the article and request chances if needed.

The project time-line is constructed as follows:

Task Name	Duration	Start	Finish	Q1 Jan Feb Mar	Q2 Apr May Jun	Q3 Jul Aug Sep	Q4 Oct Nov Dec
⊟ Prioritized Foraging using Ro	197	02/04/15	11/05/15				
Continues Reading	171	02/04/15	09/30/15				
⊟ Do research on the following	51	02/04/15	04/15/15				
Artificial Swarm	13	02/04/15	02/20/15				
Ant Cemetary Clustering	11	02/23/15	03/09/15				
Honey Bee Foraging	11	03/10/15	03/24/15				
Write Project Proposal	16	03/25/15	04/15/15				
⊟ Programming	101	02/23/15	07/13/15				
Simulating Environment	16	02/23/15	03/16/15				
Simulation Configuration	6	03/17/15	03/24/15				
Ant Cemetary Clustering	30	03/25/15	05/05/15				
Honey Bee Foraging	33	05/06/15	06/19/15				
Divion of Labor	16	06/22/15	07/13/15				
⊟ Testing / Refactor	74	05/06/15	08/17/15				
Clustering	11	05/06/15	05/20/15				
Foragin	11	06/22/15	07/06/15				
Divion of Labor	25	07/14/15	08/17/15				
⊟ Thesis	146	04/16/15	11/05/15				
Background	22	04/16/15	05/15/15				
Implementation	29	04/30/15	06/09/15				
Resluts	60	05/05/15	07/27/15				
Introduction	32	07/15/15	08/27/15				
Conclusion	15	08/25/15	09/14/15				
Abstract	15	09/10/15	09/30/15				
Supervisor read and edits	27	09/30/15	11/05/15				

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