

Search and Task Allocation

Swarm Intelligence

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1 Formalism

We define a search A as an area bound by two points. Over the search area, n_T tasks T are spawned at random positions. The tasks have a task capacity T_c indicating how many agents R are required to solve a task. The task is solved immediately if T_c agents are within the task radius T_r . The agents R move randomly around the search area at a speed R_v . There are n_R agents. When an agent is inside the task radius T_r of a task, the agents will wait for other agents to complete the task. The agents can also call for aid in solving a task. The communication distance R_d determines how far an agent can send a call for aid to another agent.

The position of the agents are updated in a discrete Langevinian manner,

$$r_{R,t+1} = r_{R,t} + N(a_R + b_{R,t}), \quad (1)$$

where $r_{R,t}$ is the position of agent R at time t , a_r is a drift term, $b_{R,t}$ is a stochastic velocity term and N is a normalising operator, ensuring equal velocity of all agents. The drift term is assigned at the start of each simulation and the stochastic velocity term is updated for each time step. The trend is added in order to make sure that the agents will venture away from their origin point, as the expected value of a sum of equally distributed stochastic variables with mean zero will be zero. That is, even though there is a likelihood for agents to eventually cover the entire area, one would expect them to meander about the point they were initially placed.

In addition to walking randomly across the area, the agents have the possibility of calling other agents in the vicinity, if they are within a communication distance R_d .

2 Implementation

We implement a tool for visualising the problem described above in PyGame [1]. The full implementation is available at github.com/gregwinther/mas, and consists of three classes. Two of these, Task and Agent, both inherit from PyGame’s Sprite class. The last class, Simulation, takes care of the game mechanics such as updating states and drawing the area with tasks and agents. Initialising and running a particular game (simulation) can be done with very few lines. For example;

```
simulation = Simulation(1000)
simulation.cycles = 1000
simulation.communicate = True
simulation.write = True
simulation.n_T = 3
simulation.Tr = 50
simulation.n_R = 10
simulation.Tc = 3
simulation.start()
```

A screenshot of this example is shown in Figure 1.

The most important methods of the Task and Agent classes is the `update()` method, which is called for each iteration of the game. For a Task, this method is quite simple. It simply checks if there are enough agents within the task radius T_r to complete the task, if so the task sprite is killed.

The `update()` method for agents is more subtle. Firstly, it has to update the positions of the agents according to Equation 1. This is done by generating a random movement step, or finite difference velocity, which is added to the trend and then normalised to have length equal to R_v .

Since it is possible for an agent to wander outside the area, we take care of the boundaries in the following manner. When a moving agent hits one of the boundaries of the area, the trend unit vector which is perpendicular to the boundary line is reversed. This has the effect of “soft bounce” off the boundary.

More than the mere movement of the agents, it is possible for them to be in different states. They may be *searching*, *tasked* or *called*. The default state, *searching*, is what we just described above. Whenever an agent is within the task radius T_r of a task T its state will change to *tasked*. If the task capacity T_c required more agents than the one(s) present, the agent will stay put until enough agents arrive at the task to help. With the mechanic enabled, an agent may also be *called* if it is within the communication dis-

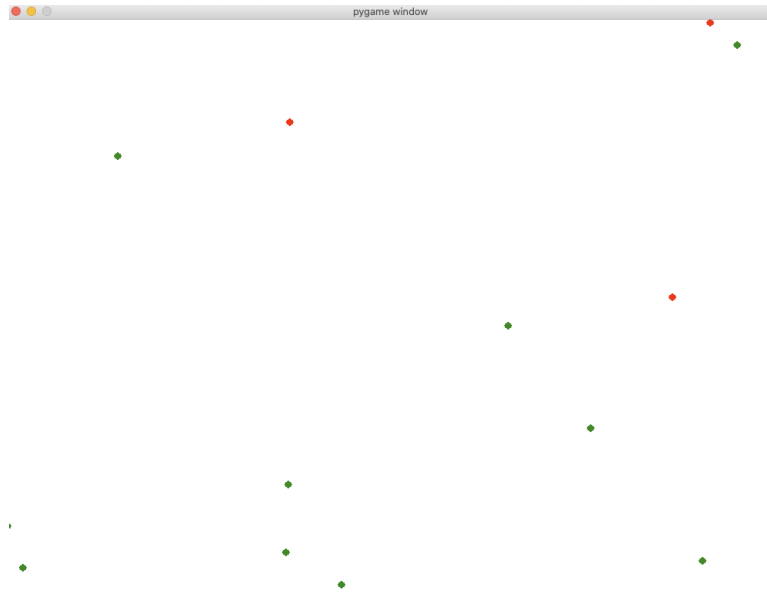


Figure 1: Sample screenshot of a simulation with $n_T = 3$ and $n_R = 10$. The green dots are agents R , and the red dots are tasks T .

tance R_d of another agent. The agent will then move at speed towards the position of the signalling agent. All of the transitions between agent states are handled within the main game loop of the Simulation class.

3 Results

4 Discussion

References

1. Shinnars, P. *et al.* *PyGame* <http://pygame.org/>. 2020.