

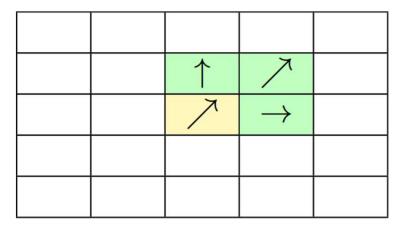
# Active Walkers

Jacob Toller; Mathematical Modelling II

### What are Active Walkers

- Dumb robots/creatures/thingymajigies with a very simple ruleset/decision-making process
- Exist in a 2D grid, with discrete timesteps ('turns')
- All walkers move at the same time during a turn.
- Can move into one of three forward positions:

		7	
	$\rightarrow$	$\rightarrow$	
		X	



(similarly for the other 6 facing directions.)

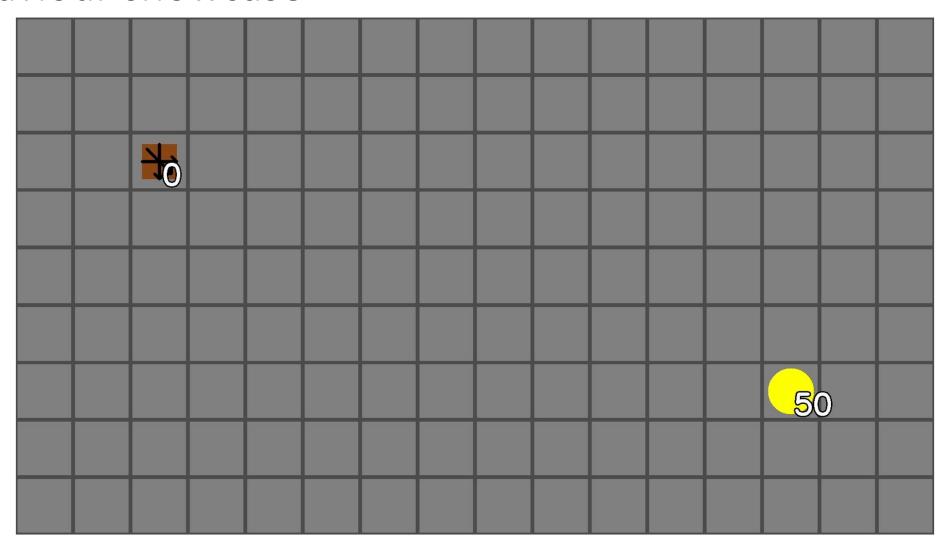
### Walker Objective

- To move resources from resource piles in the environment; to their home base
- Has two possible job roles that help them do this:

	Is sensitive to:	Deposits:
Scout	Marker B	Marker A
Carrier	Marker A	Marker B

(Markers are deposited by walkers after each turn, once they have made their move)

### Behaviour showcase



### How are the markers used

- No other way to navigate/self-orient
- Each grid cell has a concentration value, per marker type.
- Markers deposited to increase the concentration in the local area.

Most likely choice

 Walker chooses randomly from PDF; weighted so that higher concentrations ⇒ more likely. This is massively simplified; I have a more detailed explanation if requested.

	1.00	0.40	_
)	0.29	0.22	(
3	7	0.16	(
<u> </u>	<u>೧ 1                                   </u>	0 1 K	$\Gamma$

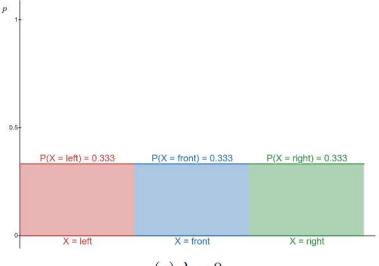
# Probability Density Function for Decision Making

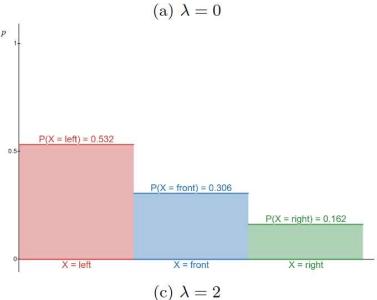
- Each concentration value raised to some power  $\lambda$ ; as a relative weight.  $C(\vec{x})^{\lambda}$
- Weights normalised to 1 to form valid PDF.
- Larger values of  $\lambda \Rightarrow$  Stronger bias towards large concentrations ('less random')

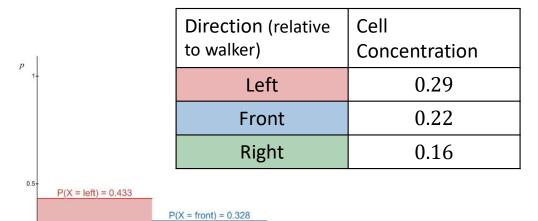
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$\overline{)}$	0.29	0.22	(
3	7	0.16	(
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Direction (relative to walker)	Cell Concentration $C(\vec{x})$	Relative Weight, $C(\vec{x})^{\lambda}$ ; $\lambda = 2$	Normalised Probability
Left	0.29	0.0841	0.532
Front	0.22	0.0484	0.306
Right	0.16	0.0256	0.162

### Effect of the $\lambda$ parameter

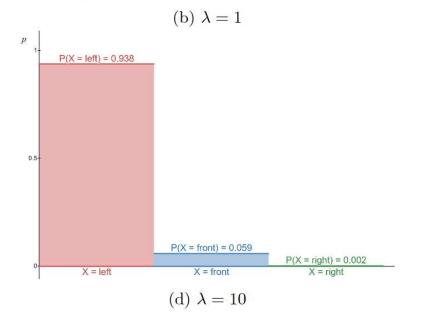






P(X = right) = 0.239

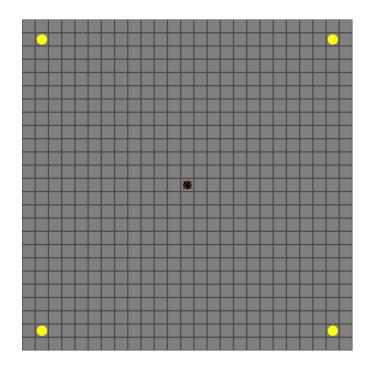
X = right



X = left

### $\lambda$ 's effect on walker efficiency

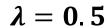
- Do the walkers perform better with more randomness, or less randomness? (maybe there's a sweet spot in the middle?)
- We can measure performance by how long it takes a group of walkers to move all the resources in the environment to the nest
- Same scenario for all trials, averaged performance over 10 trials.

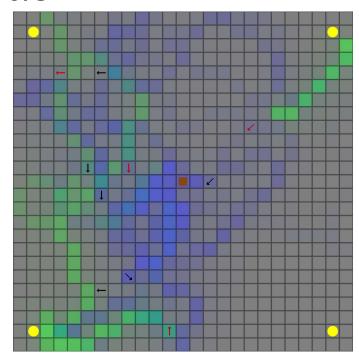


#### **Test Scenario:**

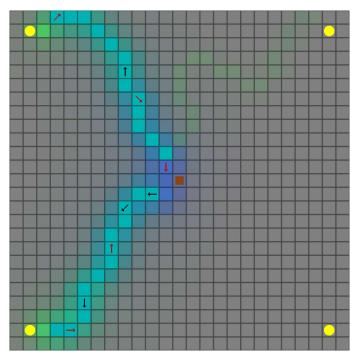
- 25x25 grid
- 10 walkers start at base in middle
- Resource pile in each corner, each with 20 units of resource
- Scenario ends once all 80 units have been moved to the base.

## Visual comparison of high/low $\lambda$

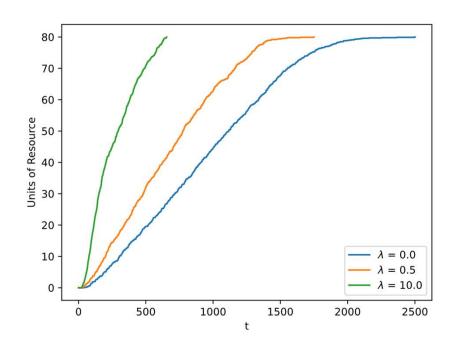


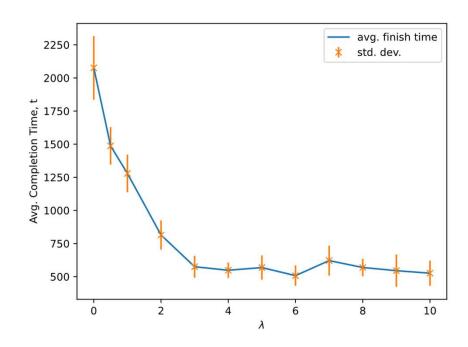


$$\lambda = 10$$



### Results





- Larger values of  $\lambda$  ('more deterministic') mean more efficiency in completing the objective.
- (Slightly) more consistent for larger  $\lambda$ .
- Diminishing returns at  $\lambda = 3$  onwards.

### Why is any of this useful?

# Modelling biological processes/phenomena

- Can be used to model ants and the way they discover and transport food
- Accurately models real phenomena (Ant mills/Death Spirals)

## Applications in low-performing robotics

- Replaces the use of AI/complex algorithms; with emergent behaviour from simple rules.
  - Similar applications; such as SAR
  - Power/cost efficient.
- Possible markers:
  - Sprayed chemical with a 'robotic nose'
  - LED's with small battery; sensed with a light intensity sensor.

# Any Questions?

If you know what the term 'nerd-sniping' means; I'd advise against asking questions.

I have 8 more slides prepared...

### Environment

- 2D grid;  $\vec{x} = (x, y) \in [0, N] \times [0, M] : N, M \in \mathbb{Z}$ 
  - N wide by M tall
  - Represents a physical space, e.g. 1 cell = 1m x 1m
  - Multiple walkers can occupy a space at a time

- Timesteps;  $t \in \mathbb{Z}^+ = \{0,1,2,...\}$ 
  - One turn represents however long it takes a walker/robot/ant to cross the space

### Boundary conditions

0	0	0	
0.07	$\rightarrow$	0.12	
0.13	0.16	0.16	

- Edges of the grid act as a solid wall
- Cells 'outside of the grid' considered to have 0 probability
  - Chooses from the remaining options

If the walker is facing into the wall with no valid options; turn around without moving

0	0	0
0.07	<b>↑</b>	0.12
0.13	0.16	0.16



0	0	0
0.07	$\rightarrow$	0.12
0.13	0.16	0.16

### Concentration Sub-Model

#### For a given type of marker:

• Cell value calculated as sum of individual strengths of each marker.

$$C(\vec{x}) = \sum_{i=1}^{n} c_i(\vec{x})$$
 Strength decay over time 
$$c_i(\vec{x}) = \frac{\alpha_i(t)}{(\|\vec{x} - \overrightarrow{p_i}\| + 1)^{\beta}}$$
 Distance falloff; e.g.  $\beta = 2$  follows inverse square law

### Concentration Sub-Model

### **Effect of a single marker**

### **Effect of multiple markers**

0.17	0.25	0.17	0.10	0.06	0.04
0.25	1.00	0.25	0.11	0.06	0.04
0.17	0.25	0.17	0.10	0.06	0.04
0.10	0.11	0.10	0.07	0.05	0.03
0.06	0.06	0.06	0.05	0.04	0.03
0.04	0.04	0.04	0.03	0.03	0.02

0.19	0.27	0.20	0.12	0.09	0.06
0.27	1.03	0.28	0.15	0.10	0.08
0.20	0.29	0.22	0.15	0.12	0.10
0.13	0.16	0.16	0.16	0.16	0.13
0.10	0.12	0.15	0.22	0.29	0.20
0.08	0.10	0.15	0.28	1.03	0.27

(With 
$$\alpha_i(t) = 1$$
;  $\beta = 2$ .)

### Real World Phenomena: Ant mills

- Occur in real life when ants follow a pheromone trail that runs in a circle; constantly reinforced by the ants it's trapped.
- Is replicated by our model

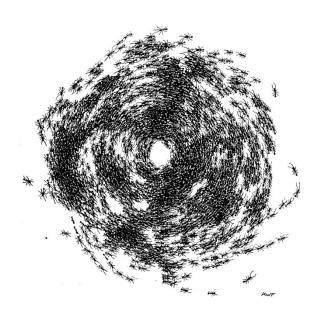
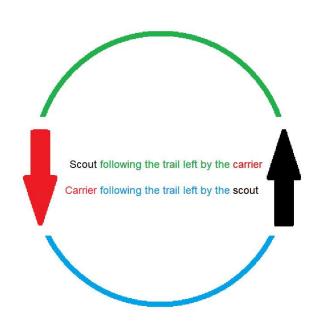
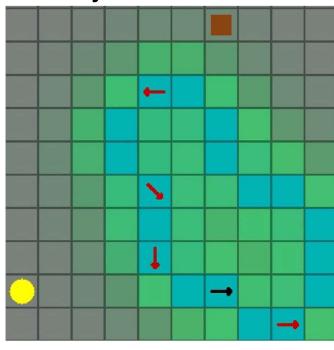


Diagram from: Schneirla, Theodore Christian et al. (1944). "A unique case of circular milling in ants, considered in relation to trail following and the general problem of orientation"

### Real World Phenomena: Ant mills

• In our model; we need at least one walker of each job





### Preventing Ant Mills from forming

- Make new markers weaker, relative to the amount of steps the walker has done without touching an objective (resource or home base).
- Walkers stuck in a mill cant refresh their step counter; leaving incredibly weak markers
- Walkers will eventually break free, and explore randomly until reaching an objective to lay a new marker trail at full strength.

$$c_i(\vec{x}) = \frac{\alpha_i(t)e^{-\gamma s_i}}{(\|\vec{x} - \overrightarrow{p_i}\| + 1)^{\beta}}$$

•  $\gamma$  is some small parameter,  $s_i$  as the step count of the walker at the point when the marker is created.

### Ant Mill demonstration & prevention

