



Eidgenössische Technische Hochschule Zürich
Swiss Federal Institute of Technology Zurich

Lecture with Computer Exercises:
Modelling and Simulating Social Systems with MATLAB

Project Report

**Modelling Desert Ant Behaviour
with a special focus on desert ant movement**

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Agreement for free-download

We hereby agree to make our source code for this project freely available for download from the web pages of the SOMS chair. Furthermore, we assure that all source code is written by ourselves and is not violating any copyright restrictions.

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Abstract

This paper is the final result of the course MODELING SOCIAL SYSTEMS WITH MATLAB which aimed to offer an insight into the MATLAB programming language and to use said language to model social systems with various different approaches. The timeframe of the course is one semester.

In this paper we will try to show how to replicate the behaviourr of desert ants in a MATLAB simulation. Furthermore we will discuss our results and compare them to experimental results obtained by biologists.

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1 Individual contributions

The whole project was done in a cooperative manner.

2 Introduction and Motivations

We think ants are exciting animals because despite their small body mass and therefore small brain they form very huge and complex social structures. Very large numbers of them work together efficiently like one body. This requires a high level of coordination. We have already seen some videos which show the great achievements of ant colonies in building and hunting. Now we found out about their navigation abilities and are curious to learn how ants are able to cover extreme distances. The human being would definitely get lost when trying to journey this far in the desert without GPS or any other form of modern help, so one of our main goals will be to find out how ants can master this difficult task.

Ants have been subject of modern research since 1848, the motivations were often interest in their instincts, society and of course the hope to learn from them. Studies in ant movement became even more compelling when scientists started to look for algorithms that solve such fundamental tasks like finding the shortest way in a graph (Graph Theory). The class of ant colony optimization algorithms was introduced 1992 and has since been a field of active study.

However, those algorithms are using the behaviour of forest ants of the western hemisphere, which is not similar to the behaviour of desert in terms of choosing a good path and finding food. Since we are studying desert ants we had to take a different approach. Desert ants rely much more heavily on the few landmarks they find in their environment and less on pheromone tracks other ants have laid out before them, like forest ants do. Also they make use of a path-integrator with which they are able to track their position in reference to where they started the journey, most likely the nest.

Results of interest are:

- How optimized is navigation by vectors
- What is the most energy-consuming task
- Out of which states is it possible for the ant to find the nest (e.g. dropping the ant somewhere else, outside of her regular path etc.)
- How well does the ant learn in the course of repeated journey towards the food and back

Of course we were as well motivated to improve our knowledge of MATLAB™

3 Description of the Model

We would like to create a model of desert ant behaviour. This will include their search for food, their returning to the nest and their orientation with global and local vectors. Also we will see how close our algorithms are to real ant movement. Therefore we want to simulate the experiments described in the papers. Our model should be able to deal with different numbers of landmarks, obstacles and starting points. We would like to give our ants the ability to learn and improve their efficiency when searching and finding food.

Because of the nature of our problem we choose to design our simulation around a time-discrete step-based model of an ant. We chose to let only one ant run at a time, because we don't think that a higher number of ants would make much of a difference considering the vast space in the deserts. Therefore we can leave out influences of near ants like separation and cohesion (compare Agent Based Modeling).

The simulation should be capable of finding a good path between nest and feeder and use a simple learning process to achieve that. We want to create a model, that can autonomously avoid obstacles and not get stuck in a corner. In order to meet this requirements we split our simulation in two parts:

Landscape

Our landscape should contain all the information about

- Position of the nest
- Position of the feeder
- Obstacles (stones, trees, cacti, oases, sand dunes and many more), from which some can be used as landmarks

We chose to limit our landscape: We implemented fixed boundaries, which hinder the ant from escaping out of our experiment area. This is important to limit the time the ant needs to find food and thus making our simulation very less time-consuming. A matrix stores information about taken and free points by the values true or false, where false stand for an obstacle. Nest, feeder, landmarks and local vectors are saved separately as vectors, to make them easy to reach.

Ant

Our ant should follow certain, simple rules to move according to the studies we received as part of the project description. Such are basic rules like avoiding obstacles or a little more specific rules like following the global vector when returning to the

nest and using the local vectors of the landmarks when finding the food again. During the simulation and after the ant has had success in finding food our local vectors should as well change according to the new found and better path.

3.1 Simplifications

There will be simplifications and assumptions, the most important ones are:

- We decided to create fixed boundaries on our Landscape.
- For our model we strictly separate navigation by global vector (feeder to nest) and by local vectors (nest to feeder). This is due to the fact that this behaviour can differ from ant to ant and there is no consistent result true for all desert ants.
- The model will have a detection-radius in which landmarks, nest and feeder are considered for moving and navigating.

4 Implementation

As described above our simulation consists of two main parts: The landscape and the ant. Both of these were implemented as separate classes. A third class the simulation-class should handle the rendering, initialising and iterations. We also used a main-file in which we declared variables that would have impact on the outcome of our simulation like the detection-radius of the ant or information on the map, that should be loaded.

4.1 Landscape

The landscape class only contains information about the map, the nest and the feeder as well as some spots which are landmarks, used by the ant as anchor points for local vectors.

We implemented different versions of loading landscapes into our simulation. Beside the possibility of creating the landscape-matrix in a separate m-file and the random-map generator we often used a simple but elegant method for generating maps out of arbitrary made generic Portable Network Graphics. This method finds specific color values and translates them into their meaning in the context of the landscape.

	Color in png-file	Color in Matlab
Obstacle	black	red
Nest	green	black circle
Feeder	blue	black cross
Landmark	turquoise	blue circle

Table 1: Color values and their meaning

4.2 Ant

The class `ant` mainly contains the current position of the ant, the local vectors on landmarks and the path integrated global vector which should always point to the nest (as long as the ant moves are coherent). We built our ant around the most important method: `move`. The `move` function is called out of two different methods the `find_food` and the `return_to_nest`. In the following all methods of the class `ant` are described:

4.2.1 Find food

This loop iterates the move-method until the ant reaches the food. Depending on how often the ant has already been on the track, it uses the aggregated local vectors to calculate a direction which the ant should follow to reach the food sooner. As soon as the feeder is in a certain distance (the detection radius) the ant runs straight towards it.

4.2.2 Calculate the direction from landmarks

$$\vec{v}_{direction} = \sum_{i=0}^n \vec{l}_i \quad \forall ||\vec{l}_i||_2 < r_{detection} \quad (1)$$

where \vec{l}_i are the local vectors, i ranging from the first to the last landmark and $r_{detection}$ is the view radius of the ant.

4.2.3 Return to nest

When returning to the nest, the model uses the same move method as when searching, but instead of calculating a general direction out of the occurring local vectors the ant uses the global vector, which always leads straight back to the nest. While returning to the nest it updates all local vectors while passing the related landmarks. In our implementation the local vectors always points to the last landmark the ant has passed or are adjusted toward this position. Thereby the ant develops a steady route that is a close to the optimal route. Of course there is no possibility to find out how real ants remember the exact direction and length of the local vectors and therefore this way of implementation must be tested for reliability later on.

4.2.4 Updating the local vectors on all landmarks

For the implementation we decided, that our model of the ant, would be able to remember only the last global vector where a landmark was spotted. This simplification seems to be adequate because of the limited brain complexity of real ants. For this reason the model, when spotting a new landmark always calculates a vector pointing to the latest landmark and thus developing a path that leads from the first landmark, the nest, to the latest, which should be quite close to the feeder. This implementation however will result in non-changing local vectors after the first run. So we included a *grow-factor*, which only allows a small adjusting every time the ant passes the landmark. As a result the learning curve of the ant became interesting, as described below in the experimental results.

$$a = 0.5 * \exp\left(-\frac{\|\vec{l}_i\|_2}{10}\right) \quad (2)$$

$$\vec{l}_i = a * \text{round}\left(\vec{l}_i + (\vec{g}_i - \vec{g}_{i-1})\right) \quad (3)$$

4.2.5 move

The move method is heart of the ant class: it accepts any general direction vector as input and sets the new position of the ant as a result. A general direction input can be calculated in the method find food or return to nest. It also handles all the checking for obstacles. Move is invoked in every time-step. Because the ant has only a choice of 8 possible next positions the method must calculate a new direction vector to one of the first order Moore neighbours:

To calculate the matching Moore neighbour from the general direction vector we use the following formula and call the result the main-direction.

$$\vec{m} = \text{round}\left(\vec{dir} * \frac{1}{\max|\vec{dir}_i|}\right) \quad (4)$$

In case the general direction is not exactly a multiple of a vector given by the Moore neighbourhood this calculation will result in a non-natural path (s. picture below). So we calculated a second-direction which is chosen as move direction with a certain probability depending on the angle between the general direction and the main direction. This allowed to walk directly towards a target. Some limit cases are handled separate.

$$\vec{s} = \vec{m} - \vec{dir} * \min(|\vec{dir}_i|) \quad (5)$$

$$p = \frac{\min|\vec{dir}_i|}{\max|\vec{dir}_i|} \quad (6)$$

If there is no general direction given to the method move, or if the general direction is zero a vector is generated based on the previous move direction. This vector then is turned around +/- 45 degree with a certain probability. This probability defines how twisted the ants path is. The following picture was taken with a low turning-probability (10 percent).

In the picture it is easily seen, that the ant can not move trough obstacles. This is also part of the method move. Therefore the method checks the desired position on the map. If the position is not available the move-vector is turned around 45 degree

clockwise or counter-clockwise, then the desired position is checked again until a possible step is found.

4.3 Simulation

The simulation class main purpose is to serve as a holder for the landscape and the ant. It also handles everything that has to do with output. The most important functions are the run-method, and the render-method, these two are described below.

4.3.1 Run

In this method there are basically two while-loops checking whether the ant is searching for food or trying to return to the nest. This is indicated by two Boolean values in the class ant. In each cases the corresponding ant-methods are invoked. Here is a simple example of using parts of the run-method (pseudo code):

```
1 while the ant has no food
2     search for food and move one step
3     if the simulation needs to be rendered
4         render the actual position of the ant on the map
5     end
6 end
```

5 Simulation Results and Discussion

6 Summary and Outlook

A Research Plan

B MATLAB Code

B.1 main.m

```
1 %% Mainfile
2 % for common configurations of the simulation (mostly testing
3 % purposes
4
5 % clear everything
6
7 clc;
```

```

8 clear all;
9 clf;
10 close all;
11
12 runduration = 100; % Duration of simulation
13
14 addpath('Maps');
15
16 %% Option1 saved Map
17 % all saved Maps can be found in the code-folder/Maps
18
19 %% two Obstacles - Experiment 1
20 % map1
21
22
23 %% map2
24 % noch erstellen.
25
26 %% Option2 random Map
27 %mapsize = 100;
28 %s = simulation(mapsize);
29 %s.l.generateLandscape(50, 50, 0.8);
30 %s.a.position = [5 5];
31 %s.l.nest = [5 5];
32 %s.l.feeder_radius = 50;
33
34 s = simulation(100);
35
36 s.l.load_image('test', 'png')
37 s.a.position = s.l.nest;
38
39 s.a.createGlobalVector(s.l);
40 s.a.createLocalVectors(s.l.landmarks);
41 s.init();
42 s.run(0);

```

B.2 simulation.m

```

1 %% Simulation Class
2 % Handles everything simulationwise e.g. run the simulation, define ...
   simulation wide parameters
3 %% Variables
4 % * l
5 %   Landscape
6 %   defines the Landscape of the simulation
7 % * a TODO decide if should/could be an array or not (simulate more than ...
   one ant in a given simulation)

```

```

8 % Ant
9 % defines the ant of the simulation
10
11
12 classdef simulation < handle
13     properties (SetAccess = private)
14         l;
15         a;
16         r_ant
17         r_ant_view
18     end
19     methods (Access = public)
20         %% Initialization
21         % Initializes a simulation with landscape size N
22         % Ant is at the moment placed in the center of the map
23         function S = simulation(N)
24             if(nargin == 0)
25                 S.l = landscape(1);
26                 S.a = ant(1);
27             else
28                 S.l = landscape(N);
29                 S.a = ant(N);
30             end
31         end
32         %% Run
33         % Runs simulation for specified amount of iterations
34         function init(S)
35             S.init_render();
36         end
37         function reset(S)
38             S.a.has_food = 0;
39             S.a.nest = 0;
40             S.a.obstacle_vector = zeros(100, 100, 2);
41         end
42         function run(S, render)
43             S.reset();
44             while S.a.has_food == 0
45                 S.a.findFood(S.l);
46                 if render
47                     S.render()
48                 end
49             end
50             while S.a.nest == 0
51                 S.a.returnToNest(S.l)
52                 if render
53                     S.render()
54                 end
55             end % while ant is not at nest.
56         end % run
57         function init_render(S)

```

```

58     figure(1)
59     imagesc(S.l.plant)
60     axis off, axis equal
61     colormap ([0 1 0; 1 0 0; 1 0 0])
62     hold on
63     plot(S.l.nest(1), S.l.nest(2), 'o', 'Color', 'k')
64     plot(S.l.feeder(1), S.l.feeder(2), 'x', 'Color', 'k');
65
66     plot(S.l.landmarks(:,1), S.l.landmarks(:,2), 'o', 'Color', 'b');
67
68     S.r_ant = plot(S.a.position(1), ...
69                   S.a.position(2), '.', 'Color', 'b');
70     S.r_ant_view = plot(S.a.position(1) + ...
71                       S.a.view.radius*cos(2*pi/8*(0:8)), ...
72                       S.a.position(2) + S.a.view.radius*sin(2*pi/8*(0:8)), ...
73                       'Color', 'k');
74
75     hold on
76 end
77 %% Render
78 % renders the simulation (plant & ant)
79 function render(S)
80     figure(1)
81
82     %plot(S.a.position(1)-S.a.move_direction(1), ...
83           S.a.position(2)-S.a.move_direction(2),...
84           '.', 'Color', 'w')
85
86     set(S.r_ant, 'XData', S.a.position(1));
87     set(S.r_ant, 'YData', S.a.position(2));
88     set(S.r_ant_view, 'XData', S.a.position(1) + ...
89           S.a.view.radius*cos(2*pi/20*(0:20)));
90     set(S.r_ant_view, 'YData', S.a.position(2) + ...
91           S.a.view.radius*sin(2*pi/20*(0:20)));
92
93     drawnow
94     % Global Vector plotten?
95     % pause(0.01)
96 end % render
97
98 function render_local_vectors(S)
99     S.init_render();
100     for i=1:length(S.l.landmarks)
101         line([S.l.landmarks(i,1) S.l.landmarks(i,1) + ...
102              S.a.local_vectors(i,1)], [S.l.landmarks(i,2) ...
103              S.l.landmarks(i,2) + S.a.local_vectors(i,2)]);
104     end
105 end
106 end
107 end
108 end

```

B.3 landscape.m

```
1 %% Landscape class
2 % A class for handling the landscape of a simulation
3 %% Properties
4 % * size:
5 %   int, size of quadratic landscape
6 % * plant(size, size):
7 %   int-array map of landscape
8 % * feeder(1,1):
9 %   int-array position of
10
11 classdef landscape < handle
12     properties (SetAccess = public)
13         size;
14         landmarks;
15         plant;
16         feeder;
17         feeder_radius
18         nest;
19     end
20     methods (Access = private)
21     end
22     methods (Access = public)
23         %% Initialize Landscape
24         % size = n
25         function L = landscape(N)
26             L.size = N;
27             L.feeder = round([1/3*N 2/3*N]);
28             L.nest = round([2/3*N 1/3*N]);
29         end % init
30
31         %% set Feeder Radius for better observability;
32         function setFeederRadius(L, r)
33             L.feeder_radius = r;
34         end
35
36         %% Stump for external generateLandscape function
37         function generateLandscape(L, obstaclecount, obstaclesize, ...
38             obstacleprobability)
39             L.plant = generateLandscape(L.size, obstaclecount, ...
40                 obstaclesize, obstacleprobability);
41         end
42
43         %% Function to set nest and feeder positions (not always required)
44         % Nest = nestposition, Feeder = feederposition
45         function setNestAndFeeder(Nest, Feeder)
46             L.nest = Nest;
```



```

45         L.feeder = Feeder;
46     end
47
48     %% Set Landmarks
49     function setLandmarks(Landmarks)
50         L.landmarks = Landmarks;
51     end
52
53     % Load a map with a specified plant and feeder/nest positions
54     function load_map(L, P)
55         L.plant = P; % Set plant
56         L.size = length(P);
57     end % load_map
58
59     function load_image(L, image, type)
60         img = imread(image, type);
61         L.size = length(img(:,:,1));
62         L.plant = ~img(:,:,1); % use hex #ffffff
63         [y, x] = find(img(:,:,2) == 153);
64         L.landmarks = [x, y];
65         [y, x] = find(img(:,:,2) == 238, 1, 'first'); % use hex #1100ee
66         L.nest = [x, y];
67         [y, x] = find(img(:,:,3) == 238, 1, 'first'); % use hex #11ee00
68         L.feeder = [x, y];
69         L.plant(1,:) = ones(1,L.size);
70         L.plant(L.size,:) = ones(1,L.size);
71         L.plant(:,1) = ones(1,L.size);
72         L.plant(:,L.size) = ones(1,L.size);
73     end
74
75     end % methods
76     methods (Static)
77     end % Static functions
78 end % classdef

```

B.4 ant.m

```

1  %% Ant class
2  % This class defines the behaviour/movement of an ant in a given landscape
3  %% Variables
4  % * position
5  %   1x2 int matrix
6  %   Position of ant in landscape
7  % * move_radius
8  %   nx2 int matrix
9  %   Defines "move radius" (neighbor fields for ant)
10 %   e.g. [-1 -1; -1 0; 0 -1; 0 1; 1 0; 1 1] ...

```

```

11 % * landmarks (TODO not implemented yet)
12 %   nxn int matrix
13 %   Defines local landmark-vectors for ant, should have the
14 %   size of the landscape
15 % * velocity
16 %   Is a 1x2 vector defining the x-y-velocity of our ant
17
18 classdef ant < handle
19     properties (SetAccess = public)
20         position
21         move_radius = [1 1; 1 0; 0 1; 1 -1; -1 1; -1 0; 0 -1; -1 -1];
22         move_direction
23         global_vector
24         has_food
25         nest
26         obstacle_vector
27         rotation
28         view_radius = 20;
29         local_vectors
30         updated_local_vectors
31         last_global_vector = [0 0]
32     end
33     methods (Access = private)
34         % creates the move_radius matrix
35         function create_moveradius(A, movewidth)
36             k = 1;
37             n = round(movewidth/2);
38             for i=-n:n
39                 for j=-n:n
40                     if i == 0 && j == 0
41                         break
42                     end
43                     A.move_radius(k,1) = i;
44                     A.move_radius(k,2) = j;
45                     k = k + 1;
46                 end
47             end
48         end
49         %% Function to update local vectors on seeable landmarks (only ...
50         when returning)
51         function update_lv(A, landmarks)
52             for i = 1:length(landmarks)
53                 if norm(landmarks(i,:) - A.position) < A.view_radius && ~...
54                     A.updated_local_vectors(i)
55                     A.local_vectors(i,:) = A.global_vector - ...
56                         A.last_global_vector;
57                     A.last_global_vector = A.global_vector;
58                     A.updated_local_vectors(i) = true;
59                 end
60             end
61         end
62     end
63 end

```

```

58     end
59     %% Function to calculate a second direction from given local vectors
60     function temp = calc_lv_direction(A, landmarks)
61         temp = [0 0];
62         for i=1:length(landmarks)
63             if norm(landmarks(i,:) - A.position) < A.view_radius
64                 temp = temp + A.local_vectors(i,:);
65             end
66         end
67         disp(temp);
68     end
69 end % private methods
70 methods (Access = public)
71     %% Initialization of ant
72     % x,y: starting positions
73     % movewidth: size for created generated move_radius matrix
74     function A = ant(x,y,movewidth)
75         if nargin == 1
76             A.position(1) = round(x/2);
77             A.position(2) = round(x/2);
78         elseif nargin > 1
79             A.position(1) = x;
80             A.position(2) = y;
81         end
82         A.rotation = -1;
83         A.move_direction = [0 1];
84         A.nest = 0; % True or False
85         A.has_food = 0;
86         A.obstacle_vector = zeros(100,100,2);
87     end
88
89     %% createGlobalVector from Landscape
90     function createGlobalVector(A, L)
91         A.global_vector = L.nest - A.position;
92     end
93     %% init local vectors
94     % only for coding & plotting convenience
95     % no ant predeterminately knows all landmarks on map
96     function createLocalVectors(A, landmarks)
97         A.local_vectors = zeros(length(landmarks), 2);
98         A.updated_local_vectors = zeros(length(landmarks), 1);
99     end
100     %% findFood
101     % Moves ant randomly in landscape to find the feeder
102     % Ant should learn landscapes and path integrate the global
103     % vector
104     % return true if found food
105     % return false if not
106     % calculate localvectors into move vector
107     function findFood(A, L)

```

```

1108         if A.position(1) == L.feeder(1) && A.position(2) == L.feeder(2)
1109             A.has_food = 1;
1110             A.last_global_vector = A.global_vector;
1111             disp('found food');
1112             return
1113         end
1114         dir = A.calc_lv_direction(L.landmarks)
1115         if dir(1) == 0 && dir(2) == 0
1116             dir = A.move_radius(randi(length(A.move_radius)),:);
1117             while dir * A.move_direction' ≤ 0
1118                 dir = A.move_radius(randi(length(A.move_radius)),:);
1119             end
1120         end
1121
1122         if norm(A.position - L.feeder) < A.view_radius
1123             dir = L.feeder - A.position;
1124         end
1125
1126         A.move_direction = dir;
1127         A.move(L, dir);
1128         A.has_food = 0;
1129     end
1130
1131     function init_returnToNest(A, landmarks)
1132         A.update_local_vectors = zeros(length(landmarks), 1);
1133     end
1134
1135     %% returnToNest
1136     % Ant returns to nest after she found food
1137     % Tries to go the most direct way with global_vector
1138     % which points straight to the nest
1139
1140     function returnToNest(A, L)
1141         % if the ant reached the nest no move is needed.
1142         if A.global_vector == 0
1143             A.nest = 1;
1144             disp('reached nest')
1145             return
1146         end
1147         A.update_lv(L.landmarks);
1148         A.move(L, A.global_vector);
1149
1150     end
1151
1152     %% move(A,L)
1153     % Moves ant in landmark, according to typical ant behaviour.
1154     % A: Ant
1155     % L: Landscape
1156     function move(A, L, move_vector)
1157         for i = 1:8

```

```

158         move_vector(1) = move_vector(1)...
159         + A.obstacle_vector(A.position(1) + ...
          A.move_radius(i,1), A.position(2) + ...
          A.move_radius(i,2), 1);
160         move_vector(2) = move_vector(2)...
161         + A.obstacle_vector(A.position(1) + ...
          A.move_radius(i,1), A.position(2) + ...
          A.move_radius(i,2), 2);
162     end
163     while move_vector(1) == 0 && move_vector(2) == 0
164         move_vector = A.move_radius(randi([1,8]));
165     end
166
167
168     % Maindirection and seconddirection are calculated from the
169     % direction given by the global veor. The seconddirection ...
        gets a
170     % Probability smaller than 0.5 based on the angle between
171     % maindirection and global vector.
172     maindir = round(...
173         move_vector/max(abs(move_vector))...
174     );
175     secdir = sign(...
176         move_vector - maindir * min(abs(move_vector))...
177     );
178     secprob = min(abs(move_vector)/max(abs(move_vector)));
179
180     % the following tests make sure no error is produced because of
181     % limit cases.
182     if secdir(1) == 0 && secdir(2) == 0
183         secdir = maindir;
184     end
185     if secprob == 0
186         secdir = maindir;
187     end
188     if secprob ≤ 0.5
189         tempdir = maindir;
190         maindir = secdir;
191         secdir = tempdir;
192         secprob = 1-secprob;
193     end
194
195
196     temp = maindir;
197     if rand < secprob
198         temp = secdir;
199     end
200
201     % If there is no obstacle near the ant the rotation-direction
202     % can change.

```

```

203         count = 0;
204         for i = 1:8
205             count = count + L.plant(A.position(2) + ...
                A.move_radius(i,2), A.position(1) + A.move_radius(i,1));
206         end
207         if count == 0
208             A.rotation = sign(rand-0.5);
209         end
210
211         phi = pi/4;
212         rot = [cos(phi), A.rotation*sin(phi); -A.rotation*sin(phi), ...
                cos(phi)];
213
214         % Obstacle-Avoiding: New maindirection until possible move ...
                is found!
215         % 180deg-Turn-Avoiding: New maindirection if ant tries to ...
                turn around
216         while L.plant(A.position(2) + temp(2), A.position(1) + ...
                temp(1)) ≠ 0 ...
217             || ( temp(1) == -A.move_direction(1) && temp(2) == ...
                -A.move_direction(2) )
218
219             % A obstacle_vector is created and helps the ant to ...
                avoid the wall
220             % and endless iterations.
221             A.obstacle_vector(A.position(1) + temp(1), A.position(2) ...
                + temp(2), 1) = ...
222                 A.obstacle_vector(A.position(1) + temp(1), ...
                A.position(2) + temp(2), 1) ...
223                 + 10*temp(1);
224             A.obstacle_vector(A.position(1) + temp(1), A.position(2) ...
                + temp(2), 2) = ...
225                 A.obstacle_vector(A.position(1) + temp(1), ...
                A.position(2) + temp(2), 2) ...
226                 + 10*temp(2);
227
228             % The ant "turns" in direction of secdir. New secdir is old
229             % maindirection rotated over old secdir. (mirror)
230             % rot rotates
231
232             temp = round(temp * rot);
233         end
234
235         A.move_direction = temp;
236         A.position = A.position + temp;
237         A.global_vector = A.global_vector - temp;
238
239     end % move
240 end % public methods
241 methods (Static)

```

```
242  
243     end % static methods  
244 end
```

C References