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

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Abstract

Individual contributions

1 Introduction

Agent-based models can provide a easily implementable way to study complex systems. As Helbing et al. (1997) have shown, many aspects of pedestrian motion, such as the formation of trail systems in green areas, can be reproduced using a relatively simple “active walker” model that takes into account the attractiveness of terrain and feedback on the terrain as it is walked upon. In the current project, we plan to apply such an active walker model to real landscapes and compare the results to existing road systems.

We attempt to answer the question: is the active walker model able to predict reasonable pathways between neighboring villages in real landscapes? Here, “reasonable” will be evaluated first in a qualitative sense. Second, a energy function will be defined based on the distance traveled horizontally and vertically, where a minimal energy function is most reasonable.

In a second step, we will determine the influence of landscape slope on trail formation, under the assumption that modern roads are situated where historically trails used to go through. We will compare generated paths to current road networks at two test sites to answer the questions: How does trail formation change with increasing landscape slope? Do the formed paths fit to current road networks?

Theoretical work by Helbing et al. (1997) has previously been implemented in an agent-based model by Pfefferle & Pleschko (2010). We will base our investigation of the above research questions on this model, making adjustments where necessary. We will use topographical data from swisstopo.admin.ch with an emphasis on 1. determining reasonable model parameters and 2. comparing modeled trails to existing road systems. Two test sites are proposed, one in an mountainous region in St. Moritz, the other in the Swiss lowlands near Friburg (Figure 1). These two test sites provide very different types of terrain on which to study the problem of trail formation.

We expect to find that the smaller roads correspond more closely to results generated by the active walker model, while larger cantonal roads, being further removed from their trail origins, should correspond less with model results. We further expect increasingly mountainous terrain to tightly constrain possible routes: we expect closer correlation between road systems and generated model results in mountainous regions than in the lowlands, since there are less possibilities for taking a route with low associated energy cost.

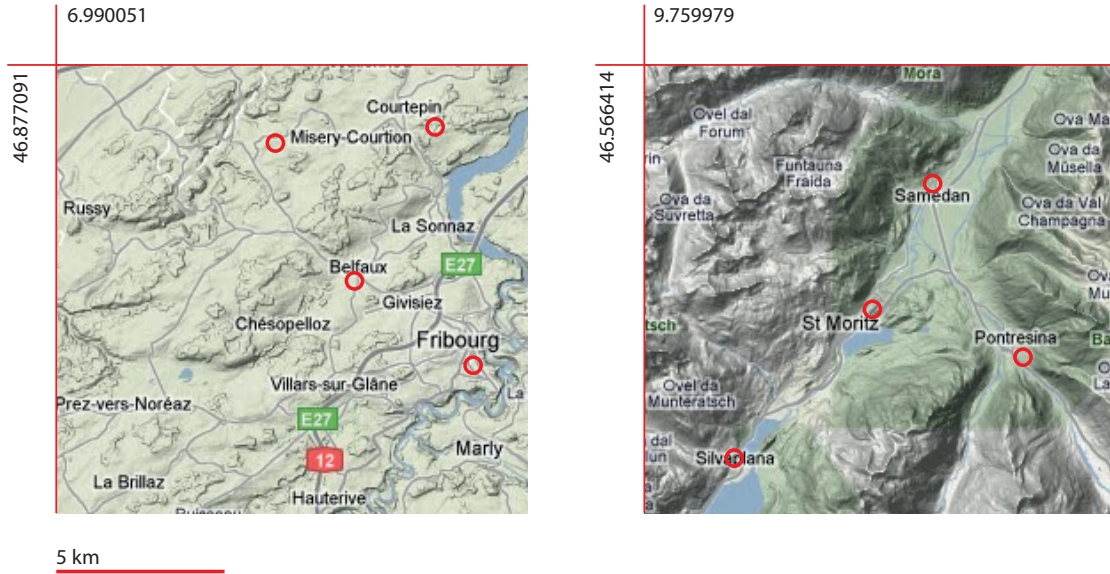


Figure 1. Two test sites, one mountainous one flat.

1.1 site

2 Description of the Model

Our studies are based on an active walker model developed by Helbing et al. (1997). Helbing et al. (1997) used this agent-based model to explain footpath formation in green areas in cities.

In comparison to a normal pedestrian model, an active walker model also takes the interactions of the pedestrians and the terrain they walked upon into account. This means that the pedestrians change the landscape they walk upon and the changed landscape influences again the pedestrians movement. A second important characteristic of the model, which influences the direction of walking of the pedestrians, is the attractiveness of a trail segment. It is a property of each point in the terrain, which describes how interesting it is to go to this certain place. In other words, how good the prospects are on this place for further walking. It is the element of our model which handles the effect of human orientation and is described later on in more detail.

As the model consists of three major components (the ground, the attractiveness of a trail segment and the pedestrians), all three are described separately. The ground and the influence of the pedestrians on it is described in Sec. 2.1. Sec. 2.2 explains how the attractiveness of walking is computed and Sec. 2.3 illustrates how the pedestrians walking direction is determined.

2.1 Ground Structure

Our landscape is represented by a function $G(r, t)$, called comfort of walking, with r for the position in the plane and t for time. As it is a property of our plane it is defined on each point. High values of G stand for trails, i.e. places where many people walked upon, while low values of G stand for places where fewer people passed by. Every time a pedestrian walks on a certain point of the plain it changes the comfort of walking there. This is because pedestrians trample down the vegetation. This is described by

$$I(r)[1 - \frac{G(r, t)}{G_{max}(r)}], \quad (1)$$

where $I(r)$ stands for the intensity of the footprint and $G_{max}(r)$ for the maximal value of the comfort of walking at a certain place (i.e. the maximal value a place can be trampled "up"). The expression in the brackets of Eq. 1 account for the saturation effect, so that the impact of the footprints decreases when there are more people walking on a place until the maximal value is reached.

As the vegetation can be trampled down it can also regrow. This effect is expressed by

$$\frac{1}{T(r)}[G_0(r) - G(r, t)] \quad (2)$$

where $G_0(r)$ stands for the natural ground condition and $T(r)$ for the durability of the trails. The bigger the durability $T(r)$ the slower the ground goes back to natural conditions $G_0(r)$. Finally the change of the comfort of walking by time due to the walking pedestrians and the regrowth of the vegetation can be expressed as

$$\frac{dG(r, t)}{dt} = I(r)[1 - \frac{G(r, t)}{G_{max}(r)}] \sum_{\alpha} \delta(r - r_{\alpha}(t)) + \frac{1}{T(r)}[G_0(r) - G(r, t)] \quad (3)$$

with α for the set of pedestrians and $\delta(r - r_{\alpha}(t))$ standing for the Dirac delta function, which is 1 if $r = r_{\alpha}(t)$ and 0 in all other cases and therefore only contributes if a pedestrian is on the actual position.

2.2 Attractiveness of Trail Segment

As mentioned above the attractiveness of a trail segment is a measure of how interesting a place is in manner of later onward walking. It is defined for every place and depends on the comfort of walking of its surrounding, where the influence of the surround decreases with distance from the place. The attractiveness of a trail segment is called trail potential and is defined as

$$V_{tr}(r_t, t) = \int_P G(r, t) e^{\frac{-|r-r_t|}{\sigma(r_t)}} dP \quad (4)$$

where r_t stands for the position the trail potential is computed for, P for the plain and $\sigma(r_t)$ for the visibility. The visibility controls how fast the influence of the surrounding decreases. The higher the visibility the slower the influence of the surrounding decreases. Furthermore high values of the trail potential stand for a high attractiveness of a trail segment and vice versa.

2.3 Pedestrians Walking Direction

In the model every pedestrian has a starting point and a destination. When the pedestrians walking direction is determined two vectors decide about the final walking direction. One vector is the vector which points towards the pedestrians destination. It is given by the unit vector

$$e_\alpha^1(r_\alpha, t) = \frac{d_\alpha - r_\alpha}{|d_\alpha - r_\alpha|} \quad (5)$$

where d_α is the position of the pedestrians destination. The other vector which decides about the pedestrians walking direction is the vector which points into the direction of highest increase of the trail potential. It can be expressed in the normalized form as

$$e_\alpha^2(r_\alpha, t) = \frac{\nabla V_{tr}(r_\alpha, t)}{|\nabla V_{tr}(r_\alpha, t)|}. \quad (6)$$

Combining Eq. 5 and 6 and introducing a new variable ρ that controls the relative importance of the two vectors leads to the final walking direction

$$e_\alpha(r_\alpha, t) = \rho \cdot e_\alpha^1(r_\alpha, t) + e_\alpha^2(r_\alpha, t) = \rho \cdot \frac{d_\alpha - r_\alpha}{|d_\alpha - r_\alpha|} + \frac{\nabla V_{tr}(r_\alpha, t)}{|\nabla V_{tr}(r_\alpha, t)|}. \quad (7)$$

For value of $\rho > 1$ the destination vector gets more important and for value $\rho < 1$ the direction of the highest increase of the trail potential prevails.

3 Description of the Path-Evaluation Function

To evaluate the paths taken by the pedestrians a function was defined to judge if a path is reasonable or not. As described by Kölbl & Helbing (2003) humans try to minimize their cost of travel. Cost of travel in our case is travel-time. Assuming a horizontal speed $u_{horiz}(G(r_\alpha, t))$ which scales with comfort of walking and a constant vertical speed u_{vert} the travel-time is given by

Figure 2. The road structure at the two study sites. Data taken from Swisstopo illustrates the differences. The weight of the lines is derived from road class order (highways and 1. order roads are heavier in weight than higher order roads) and does not necessarily correlate to throughput.

$$T = \frac{s_{horiz}}{u_{horiz}(G(r_\alpha, t))} + \frac{s_{vert}}{u_{vert}}, \quad (8)$$

where s_{vert} and s_{horiz} are the vertically and horizontally traveled distance. In more detail the horizontal speed u_{horiz} scales between a minimal horizontal speed $u_{horiz,min}$ and a maximal horizontal speed $u_{horiz,max}$ depending on how strongly the path is trampled down. This is expressed as

$$u_{horiz}(r_\alpha, t) = u_{horiz,min} + (u_{horiz,max} - u_{horiz,min}) \cdot \frac{G(r_\alpha, t) - G_0(r_\alpha, t)}{G_{max}(r_\alpha, t) - G_0(r_\alpha, t)}. \quad (9)$$

4 Implementation

The active walker model was previously implemented by Pfefferle & Pleschko (2010). We further developed this model to fit our requirements. Therefore first the implementation of Pfefferle & Pleschko (2010) is described followed by our additional contribution.

5 Simulation Results and Discussion

In our analysis of the path formation process using our active walker model, the biggest uncertainty concerned the values of the various independent parameters (durability and intensity of trails, visibility, relative importance of the destination vector). No literature values could be we couldn't find any literature references to this. Since our initial hypotheses cannot be conclusively addressed without first dealing with the issue of model parameters, we conducted a sensitivity analysis. we'll discuss first the results of a parameter variation, then look in detail at the generated path structures, and finally compare our results with existing road structures.

5.1 Parameter variation

5.2 Path structure

5.3 Existing road structure

qualitative differences between the two sites. Road structure at the St. Moritz site shows close similarity to the structure of alpine rivers,

6 Summary and Outlook

We found: -it's most important to determine correct parameter values -there are qualitative differences between the two study sites.

References

- Helbing, D., Keltsch, J., & Molnár, P. (1997). Modelling the evolution of human trail systems. *Nature*, 388(3).
- Kölbl, R. & Helbing, D. (2003). Energy laws in human travel behaviour. *New Journal of Physics*, 5(48).
- Pfefferle, J. & Pleschko, N. (2010). Simulation of human trail systems. Project Report for Lecture "Modelling and Simulating Social Systems with MATLAB".

7 Appendix A: Matlab Code

batch.m

```
% location of cities on original grid (25x25m)
stm_p = [454 620; 567 452; 674 638; 528 542; 328 785; 623 372];
fri_p = [667 503; 596 534; 693 607; 599 227; 612 459; 546 417; 255 223; 662 399; 412 655; 510 685];

% location of cities on reduced grid (500x500m)
stm_p = [23 31;29 23;34 32;27 28;17 40;32 19];
fri_p = [34 26;30 27;35 31;30 12;31 23;28 21;13 12;34 20;21 33;26 35;27 26;21 19;21 29;14 34];

%dur = 25; % Durability
%inten = 10; % Intensity
%vis = 0.3, 1, 4; % Visability
%importance = 1.6;
%location = 'fri'
%numframes = 100

for vis=[0.3 1 4]
    for dur = [5 25 50]
        for inten = [5 10 30]
            for importance = [0.5 1 1.6]
                smDriver(dur, inten, vis, importance, 'fri', 100);
                smDriver(dur, inten, vis, importance, 'stm', 100);
            end
        end
    end
end
end
```

Path.m

```
classdef Path < handle
    %PATH our path class

    properties(SetAccess = public)
        coordinates;
        type;
        time;
        timeOfArrival;
        relativeGround;
    end

    methods
```

```

function obj = Path(ped,entryP,speed,aPlain,aTime)
    % takes deleted pedestrian and generates its path with
    % properties
    obj.coordinates = [ped.way; ped.destination];
    obj.relativeGround = [ped.relativeGround;...
        aPlain.relativePath(ped.destination(1),ped.destination(2))];
    obj.timeOfArrival = aTime;
    PathType(obj,entryP);
    PathTime(obj,speed,aPlain);
end

function PathType(obj,entryP)
    % checks which from the possible paths the pedestrian went

    % generate possible paths
    PossPathIndex = nchoosek(1:length(entryP),2);

    PossPath.origin = zeros(size(PossPathIndex));
    PossPath.destination = zeros(size(PossPathIndex));

    for i=1:size(PossPathIndex,1)
        PossPath.origin(i,:) = entryP(PossPathIndex(i,1),:);
        PossPath.destination(i,:) = entryP(PossPathIndex(i,2),:);
    end

    % check which from the possible paths the path is
    for i=1:size(PossPathIndex,1)

        if ((obj.coordinates(1,:)==PossPath.origin(i,:)|...
            obj.coordinates(1,:)==PossPath.destination(i,:))&...
            (obj.coordinates(end,:)==PossPath.origin(i,:)|...
            obj.coordinates(end,:)==PossPath.destination(i,:)))
            obj.type=i;
        end
    end

end

function PathTime(obj,speed,aPlain)
    % calculates the time it takes the pedestrian to walk the path

    path.horiz = obj.coordinates;

    % extracting height info from elevation model
    path.vert = zeros(size(path.horiz,1),1);
    for i=1:size(path.horiz,1)
        path.vert(i) = aPlain.realGround(path.horiz(i,1),path.horiz(i,2));
    end
end

```

```

% calculating horizontal and vertical distance from path
% (vertical distance only taken if path goes uphill)

dist_horiz = zeros(size(path.horiz,1)-1,1);
dist_vert = zeros(size(path.horiz,1)-1,1);
relativeGroundMean = zeros(size(path.horiz,1)-1,1);

for i=1:size(path.horiz,1)-1

    % calculate horizontal distance
    delta_horiz = norm(path.horiz(i+1,:)-path.horiz(i,:));
    dist_horiz(i,1) = delta_horiz;

    % calculate horizontal distance
    delta_vert = path.vert(i+1)-path.vert(i);
    if delta_vert>=0
        dist_vert(i,1) = delta_vert;
    else
        dist_vert(i,1) = 0;
    end

    % calculate relative ground between the gridpoints
    relativeGroundMean(i,1) = (obj.relativeGround(i+1,1)-obj.relativeGround(i,1))/2;

end

% scale horizontal distance with grid size and calculate walking
% time

speed.horizontal.real = speed.horizontal.min + (speed.horizontal.max-speed.horizontal.min)*dist_horiz;

obj.time = sum(aPlain.gridSize*dist_horiz./speed.horizontal.real + dist_vert/speed.vertical);

end

end

end

```

Pedestrian.m

```

classdef Pedestrian < handle
    %PEDESTRIAN our pedestrian class

    properties(SetAccess = private )
        destination;
        way;
        relativeGround;
    end
end

```

```

properties(SetAccess = public)
    position;
end

methods
    function obj = Pedestrian(entryP,aPlain)
        % generate new pedestrian and randomly choose origin and
        % destination from given entry points

        r = randperm(length(entryP));
        orig = entryP(r(1),:);
        dest = entryP(r(2),:);
        obj.way = orig;
        obj.destination = dest;
        obj.position = orig;
        obj.relativeGround = aPlain.relativePath(obj.position(1),...
            obj.position(2));
    end

    function set.position(obj,pos)
        % put pedestrian to position
        obj.position = pos;
    end

    function saveWay(obj,aPlain)
        % save way of pedestrian and relative strength of ground
        obj.way = [obj.way; obj.position];
        obj.relativeGround = [obj.relativeGround;...
            aPlain.relativePath(obj.position(1),obj.position(2))];
    end

    function val = isAtDestination(obj)
        % check if pedestrian arrived at destination
        val = (norm(obj.position - obj.destination)<2);
    end
end
end

```

Plain.m

```

classdef Plain < handle
    %PLAIN Saves state of the plain

    properties(SetAccess = public)
        ground;           % The current ground structure
        initialGround;    % initial ground model
        groundMax;        % The maximum values of the walking comfort
        intensity;        % The footprint intensity
        durability;       % The durability of trails
        visibility;       % The visibility at each point
    end
end

```

```

    realGround;      % real, not inverted, ground
    gridSize;       % grid size of plain in m
    relativePath;    % relative strength of path
end

methods
    function obj=Plain(aInitialGround,aGroundMax,aIntensity,...
        aDurability,aVisibility,aRealGround,aGridSize)
        initSize = size(aInitialGround);
        % Check if initialGround has same size as intensity and
        % durability matrix

        if((nnz(initSize == size(aIntensity)) == 2) &&...
            (nnz(initSize == size(aDurability))==2) &&...
            (nnz(initSize == size(aGroundMax))==2) &&...
            (nnz(initSize == size(aVisibility))==2) &&...
            (nnz(initSize == size(aRealGround))==2))

            obj.ground = aInitialGround;
            obj.groundMax = aGroundMax;
            obj.initialGround = aInitialGround;
            obj.intensity = aIntensity;
            obj.durability = aDurability;
            obj.visibility = aVisibility;
            obj.realGround = aRealGround;
            obj.gridSize = aGridSize;
        else
            error('PLAIN(): initialGround must be same size as intensity and durability');
        end
    end

    function changeEnvironment(obj,pedestrians)
        % Changes the environment according to the positions of the
        % pedestrians
        [n m] = size(obj.ground);
        pedAt = sparse(n,m);

        for i=1:length(pedestrians)
            ped = pedestrians(i);
            pedAt(ped.position(1),ped.position(2)) = ...
                pedAt(ped.position(1),ped.position(2)) + 1;
        end

        % Change the environment on each square of the plain
        for i=1:n
            for j=1:m
                % Change the ground according to the formula
                obj.ground(i,j) = obj.ground(i,j) + ...
                    1/obj.durability(i,j) * (obj.initialGround(i,j)-...
                    obj.ground(i,j)) + obj.intensity(i,j) * ...
                    (1-(obj.ground(i,j)/obj.groundMax(i,j))) * ...

```

```

        pedAt(i,j);

        % Check for the boundaries of the ground values
        if(obj.ground(i,j) > obj.groundMax(i,j))
            obj.ground(i,j) = obj.groundMax(i,j);
        elseif(obj.ground(i,j) < obj.initialGround(i,j))
            obj.ground(i,j) = obj.initialGround(i,j);
        end
    end
end

end

function val = isPointInPlain(obj,y,x)
    % Returns wheter or not a point (x,y) is in this plain
    val = (y>0 && x>0);
    val = val && y<=size(obj.ground,1) && x<=size(obj.ground,2);

end

function MakeRelativePath(obj)
    % Calculates relative path strength (1 = maximal path)
    obj.relativePath = (obj.ground-obj.initialGround)./...
        (obj.groundMax-obj.initialGround);

end

end
end

```

smDriver.m

```

function smDriver(dur, inten, vis, importance, site, numframes)
%SMDDRIVER Sets up a simulation

f1 = figure('OuterPosition',[0 0 700 600]);
winSize = get(f1,'Position');
%numframes = 200;

% selecting elevation model
load elevation
if(strcmp(site,'stm'))
    elevation = stmoritz(1:1140,:); % for St. Moritz
    entryPoints = [23 31;29 23;34 32;27 28;17 40;32 19];
elseif(strcmp(site,'fri'))
    elevation = friburg(1:1100,1:1260); % for Freiburg
    entryPoints = [34 26;30 27;35 31;30 12;31 23;28 21;13 12;34 20;21 33;26 35;27 26;21 19;21 29];
end

```

```

% resizing elevation model (original elevation dim must be multiple of 20)
elevation_re = zeros(size(elevation)/20);
[m n] = size(elevation);

for i=20:20:m
    for j=20:20:n
        elevation_re(i/20,j/20) = mean2(elevation(i-19:i,j-19:j));
    end
end

[m n] = size(elevation_re);

% Set the parameters
%dur = 25; % Durability
%inten = 10; % Intensity
%vis = 4; % Visability
%importance = 1.6; % Weight of the destination vector
speed.horizontal.min = 4000; % min horizontal speed in m/h
speed.horizontal.max = 6000; % max horizontal speed in m/h
speed.vertical = 500; % vertical speed in m/h
gridSize = 500; % grid size of plain in m
pathMax = 100; % maximal value of a path

initialGround = max(max(elevation_re))-elevation_re; % inverting elevation
groundMax = initialGround + ones(m,n)*pathMax;
intensity = ones(m,n) * inten;
durability = ones(m,n) * dur;
visibility = ones(m,n) * vis;
elevation = elevation_re;

% create new plain with the specified values
myplain = Plain(initialGround,groundMax,intensity,durability,visibility,...
    elevation,gridSize);

% show the plain for input of the entry points
%pcolor(myplain.realGround);
%colormap(gray);
%shading interp;
%axis ij;
%entryPoints = ginput;
%entryPoints = floor([entryPoints(:,2) entryPoints(:,1)]);

% create a state machine with the specified plain
mysm = StateMachine(myplain);
mysm.importance = importance;
mysm.entryPoints = entryPoints;
mysm.speed = speed;
% make cell, where possible paths on Plain are later saved
noPossPaths = length(nchoosek(1:length(mysm.entryPoints),2));

```



```

mysm.pathsSorted = cell(1,noPossPaths);

% Do 'numframes' timesteps
C(1) = getframe(gcf);
for i=1:numframes

    % print every 20th timestep into a .png file
    if(mod(i,20)==0 && i >0)
        str = sprintf('images/im_%d_d%d_i%d_v%d_%d.png',...
            importance,dur,inten,vis,i);
        saveas(f1,str);
    end

    % compute a new transition in the state machine
    vtr = mysm.transition;

    % tell the StateMachine in which state it is in
    mysm.time = i;

    % display how many pedestrians are on the plane
    pedestrians = mysm.pedestrians;
    fprintf('Number of pedestrians: %d\n',length(pedestrians));

    % making the 3 subplots
    clf(f1);
    suptitle({[],[];['Grid:' num2str(m) 'x' num2str(n)];['Durability:'...
        num2str(dur) ' Visibility:' num2str(vis) ' '];[ 'Intensity:' ...
        num2str(inten) ' Importance:' num2str(importance) ' '];['After '...
        num2str(i) ' timesteps']});

    % subplot 1
    subplot(1,3,1);
    title('Initial Ground Structure');
    pcolor(myplain.realGround);
    shading interp;
    axis equal tight off ij;
    colormap(gray)
    freezeColors

    % subplot 2
    subplot(1,3,2);
    title('Evolving Trails');

    A=myplain.realGround;
    B=myplain.ground-myplain.initialGround;

    % shift B above the maximum of A
    B_shifted = B-min(B(:))+max(A(:))+1;

    % create fitted colormap out of two colormaps
    range_A = max(A(:))-min(A(:));

```

```

range_B = max(B(:))-min(B(:));

% to adjust caxis and colormap
range_b = range_B;

for i=1:10
    if (range_b>=pathMax/10*(i-1))&&(range_b<=pathMax/10*i)
        range_B = pathMax/10*i;
    end
end

% adjusting colormap

cm = [gray(ceil(64*range_A/range_B));flipud(summer(64))];

% plotting
pcolor(B_shifted)
shading interp
hold on
contour(A)
axis equal tight off ij;
colormap(cm)
caxis([min(A(:)) max(A(:))+range_B])

freezeColors % http://www.mathworks.com/matlabcentral/fileexchange/7943

% subplot 3
subplot(1,3,3);
title('Attractiveness');
pcolor(vtr);
shading interp;
axis equal tight off ij;
colormap(jet)
freezeColors

% plotting the pedestrians into the subplots
for j=1:length(pedestrians)
    ped = pedestrians(j);

    subplot(1,3,1);
    title('Initial Ground Structure');

    hold on;
    plot(ped.position(2),ped.position(1),'wo');

    subplot(1,3,2);
    title('Evolving Trails');

    hold on;
    plot(ped.position(2),ped.position(1),'wo');

```

```

        subplot(1,3,3);
        title('Attractiveness');

        hold on;
        plot(ped.position(2),ped.position(1),'wo');
    end

    drawnow;
    C(i)=getframe(gcf);
end

% save data
savefile = sprintf('data/d%d_i%d_v%d_i%d_%s.mat',...
    dur, inten, vis, importance, site);
save(savefile, 'myplain', 'mysm');

%i = 1;
%str = sprintf('movie%d.avi',i);

%while(exist(str)>0)
%    i = i+1;
%    str = sprintf('movie%d.avi',i);
%end

%save movie to file
%movie2avi(C,str,'fps',3);

end

```

StateMachine.m

```

classdef StateMachine < handle
    % STATEMACHINE Handles the state changes in the simulation
    % Computes the change of the environment and moves all the pedestrians

    properties(SetAccess = public)
        plain;           % G ... the current plain
        pedestrians;     % Array of pedestrians which are currently walking
        importance;      % How to weight the vector to the destination
        entryPoints;     % entry points as specified by ginput
        paths;           % paths walked by pedestrians
        pathsSorted;     % paths sorted by what way they went
        speed;           % horizontal and vertical speed
        time;            % time in which the state machine is in
    end
end

```

```

end

methods
function obj = StateMachine(aPlain)
    % Constructor: set the plain
    obj.plain = aPlain;
end

function [Vtr] = transition(obj)
    % Does a transition in the state machine according to the plain
    % and the pedestrians.

    [n m] = size(obj.plain.ground);
    Vtr = zeros(n,m);

    % Make relative strength of path
    MakeRelativePath(obj.plain);

    % Generate new pedestrians and put it to the other
    % pedestrians

    newPed = Pedestrian(obj.entryPoints,obj.plain);
    obj.pedestrians = [obj.pedestrians,newPed];

    % Change the environment according to the pedestrian positions
    obj.plain.changeEnvironment(obj.pedestrians);

    % Compute the attractiveness for each point in the plain
    for i=1:n
        for j=1:m
            Vtr(i,j) = obj.computeAttractiveness([i;j]);
        end
    end

    % Delete pedestrians which are at their destination or near
    ToDelete = false(1,length(obj.pedestrians));

    for i=1:length(obj.pedestrians)
        ToDelete(i) = isAtDestination(obj.pedestrians(i));
    end

    DeletedPed = obj.pedestrians(ToDelete);
    obj.pedestrians = obj.pedestrians(~ToDelete);

    % Save path of delted pedestrians and sort them
    for i=1:length(DeletedPed)
        % sort path
        newPath = Path(DeletedPed(i),obj.entryPoints,obj.speed,...
            obj.plain,obj.time);
        obj.pathsSorted{newPath.type} = ...

```

```

        [obj.pathsSorted{newPath.type}; ...
        [newPath.type newPath.time newPath.timeOfArrival]];
    % save path to other paths
    obj.paths = [obj.paths, newPath];
end

% move and save way of pedestrians
for i=1:length(obj.pedestrians)
    movePedestrian(obj,i,Vtr);
    saveWay(obj.pedestrians(i),obj.plain);
end

end

function movePedestrian(obj,pedestNum,vtr)
    % Moves a pedestrian according to the attractiveness of the
    % neighbourhood and its destination

    pedest = obj.pedestrians(pedestNum);

    maxvtr = -inf;
    maxcoords = [0;0];

    % compute the maximum value of vtr in the neighbourhood and
    % save the direction to it
    for i = -1:1
        for j = -1:1
            y = pedest.position(1)+i;
            x = pedest.position(2)+j;
            if(obj.plain.isPointInPlain(y,x))
                if maxvtr < vtr(y,x)
                    maxvtr = vtr(y,x);
                    maxcoords = [i j];
                end
            end
        end
    end

    % normalize the gradient vector (but check for zero division)
    if(norm(maxcoords)>0)
        maxcoords = maxcoords / norm(maxcoords);
    end

    % compute the vector to the destination and normalize it
    toDest = pedest.destination - pedest.position;
    toDest = toDest ./ norm(toDest);

    % add both vectors, but multiply the toDest vector with
    % importance to get better results

```

```

moveDir = obj.importance * toDest + maxcoords;

% compute the angle of the directional vector
alpha = atan(moveDir(1)/moveDir(2));

% Because tan is pi periodic we have to add pi to the angle
% if x is less than zero
if moveDir(2) < 0
    alpha = alpha + pi;
end

% Define the direction vectors
up = [-1 0];
down = [1 0];
left = [0 -1];
right = [0 1];

% Initialize the move vector
move = [0 0];

% Shortcut for pi/8
piEi = pi/8;

% Check the angle of the resulting vector and choose
% the moving direction accordingly

if (alpha < -3*piEi) || (alpha > 11*piEi)
    % move up
    move = up;

elseif (alpha >= -3*piEi) && (alpha < -piEi)
    % move right up
    move = up + right;

elseif (alpha >= -piEi) && (alpha < piEi)
    % move right
    move = right;

elseif (alpha >= piEi) && (alpha < 3*piEi)
    % move down right
    move = down + right;

elseif (alpha >= 3*piEi) && (alpha < 5*piEi)
    % move down
    move = down;

elseif (alpha >= 5*piEi) && (alpha < 7*piEi)
    % move down left
    move = down + left;

elseif (alpha >= 7*piEi) && (alpha < 9*piEi)

```

```

        % move left
        move = left;

elseif (alpha >= 9*piEi) && (alpha < 11*piEi)
    % move up left
    move = up + left;
end

% Actually move the pedestrian
pedest.position = pedest.position + move;
end

function Vtr = computeAttractiveness(obj,coords)
    % This function computes the sum of all attractivenesses
    % of the whole area from the viewpoint of coords

    % Get the visibility at point coords
    visibility = obj.plain.visibility(coords(1),coords(2));

    % Get the current ground structure
    G = obj.plain.ground;
    [n m] = size(G);

    % Efficient implementation for the sum
    [A,B]=meshgrid((1:m)-coords(2)).^2,((1:n)-coords(1)).^2);
    S=-sqrt(A+B);
    S = exp(S/visibility);
    S = S.*G;
    Vtr = sum(sum(S));

    % Average the sum over the number of squares in the plain
    Vtr = Vtr/(m*n);
end

end

end

```

visualization.m

```

function visualization

global f1;
f1 = figure('OuterPosition',[0 0 700 600]);

files = dir('data/*stm.mat');

for file=1:numel(files)

```

```

clear myplain mysm;
clf;

filename = files(file).name;

% extract parameter values
params = sscanf(filename, 'd%d_i%d_v%f_i%f_.mat');
dur = params(1);
inten = params(2);
vis = params(3);
importance = params(4);
location = 'fri';

plot_paths(filename);
[n t dist dist_std traveltime] = completed_paths(filename);

data = sprintf('%i,%i,%f,%f,%i,%f,%f,%f,%f',dur,inten,vis,importance,n,t,dist,dist_std,traveltime);
disp(data);

end

end

%-----
function [n t_norm dist dist_std traveltime] = completed_paths(filename)
    load(strcat('data/',filename));

    % number of paths completed
    n = length(mysm.paths);

    % travel time
    t = zeros(3,n);

    for i=1:n
        path=mysm.paths(i);
        start = path.coordinates(1,:);
        dest = path.coordinates(end,:);
        dist = sqrt(sum((dest-start).^2));

        traveltime = path.time;

        t(:,i) = [dist;traveltime;dist/traveltime];
    end

    t_norm = mean(t(3,:));
    dist = mean(t(1,:));
    dist_std = std(t(1,:));
    traveltime = mean(t(2,:));

```


end

```
%-----  
function plot_paths(filename)  
    global f1;  
    load(strcat('data/',filename));  
  
    % plot ground structure  
    subplot(1,2,1);  
    pathMax = 100;  
    title('Evolving Trails');  
  
    A=myplain.realGround;  
    B=myplain.ground-myplain.initialGround;  
  
    % shift B above the maximum of A  
    B_shifted = B-min(B(:))+max(A(:))+1;  
  
    % create fitted colormap out of two colormaps  
    range_A = max(A(:))-min(A(:));  
    range_B = max(B(:))-min(B(:));  
  
    % to adjust caxis and colormap  
    range_b = range_B;  
  
    for i=1:10  
        if (range_b>=pathMax/10*(i-1))&&(range_b<=pathMax/10*i)  
            range_B = pathMax/10*i;  
        end  
    end  
  
    % adjusting colormap  
  
    cm = [gray(ceil(64*range_A/range_B));flipud(summer(64))];  
  
    % plotting  
    pcolor(B_shifted)  
    shading interp  
    hold on  
    contour(A)  
    axis equal tight off ij;  
    colormap(cm)  
    caxis([min(A(:)) max(A(:))+range_B])  
  
    freezeColors % http://www.mathworks.com/matlabcentral/fileexchange/7943
```

```

% plot vector paths
for ped=mysm.pedestrians
    plot(ped.way(:,1), ped.way(:,2));
end

% plot cities
if(strfind(filename,'fri'))
    p = [34 26;30 27;35 31;30 12;31 23;28 21;13 12;34 20;21 33;26 35;27 26;21 19;21 29;14 34];
else
    p = [23 31;29 23;34 32;27 28;17 40;32 19];
end
plot(p(:,1),p(:,2),'wo');

hold off;

% plot travel time
subplot(1,2,2);
hold on;
for path=mysm.paths
    plot(path.timeOfArrival,path.time,'o');
    text(path.timeOfArrival,path.time,sprintf('%i',path.type));
end

axis square;
hold off;

saveas(f1,strcat('images2/',filename,'.png'));

end

```