# Modelling Crowd Behaviour in the Polymensa using the Social Force Model

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GESS Presentations, December 2011

#### **Outline**

Introduction and Motivation Introduction, Motivation and Fundamental Questions Fundamental Questions

Description of Model Sources Description

Implementation

Simulation and Discussion

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- General motivation: mensa dynamics predictable?
- ▶ Matlab simulation
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#### Variables of Interests

- ► Time it takes for agents to enter and exit mensa.
- ► The number of times agent paths are blocked by other agents.
- Queue locations.

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#### General

- ► All pedestrian interactions reduced to forces.
- ► Forces
  - Destination
  - Other Agents
  - 3. Boarders
  - Objects of Attraction

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  - 4. Objects of Attraction

#### **Destination Force**

#### Unit Vector towards destination:

$$lackbr{e}_lpha(t) := rac{ec{r}_lpha^k - ec{r}_lpha(t)}{||ec{r}_lpha^k - ec{r}_lpha(t)||}$$

Force towards destination:

$$\blacktriangleright \vec{F}^0_\alpha(\vec{v}_\alpha, \vec{v}^0_\alpha, \vec{e}_\alpha(t)) := \frac{1}{\tau_\alpha} (v^0_\alpha \vec{e}_\alpha - \vec{v}_\alpha)$$

#### **Destination Force**

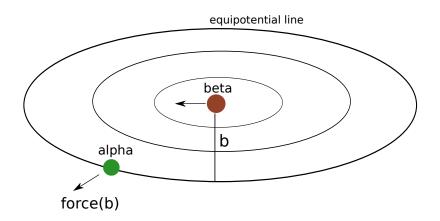
#### Unit Vector towards destination:

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# Other Agents Force



# Other Agents Force (alternative)

## according to more recent Social Force model:

$$ec{F}_{lphaeta}(t) = A_{lpha}^{1} \exp\left[(r_{lphaeta} - d_{lphaeta})/B_{lpha}^{1}
ight] ec{n}_{lphaeta} \ \cdot \left(\lambda_{lpha} + (1 - \lambda_{lpha}) rac{1 + \cos(arphi_{lphaeta})}{2}
ight) \ + A_{lpha}^{2} \exp\left[(r_{lphaeta} - d_{lphaeta})/B_{lpha}^{2}
ight] ec{n}_{lphaeta}$$

## **Boundary Forces**

#### **Boundary Forces:**

lacktriangle boundry Repulsion Force  $ec{F}_{lpha B} = - 
abla_{ec{r_{lpha B}}} U_{0}{}_{lpha B} e^{-||ec{r}_{lpha B}||/R}$ 

#### **Total Forces**

► Note: attractive object forces calculated in similar fashion.

#### Superposition of all Forces:

$$\vec{F}_{\alpha}(t) = \vec{F}_{\alpha}^{0}(\vec{v}_{\alpha}\vec{v}_{\alpha}^{0}\vec{e}_{\alpha}) + \sum_{\beta}\vec{F}_{\alpha\beta}(\vec{e}_{\alpha},\vec{r}_{\alpha} - \vec{r}_{\beta}) + \sum_{B}\vec{F}_{\alpha B}(\vec{e}_{\alpha},\vec{r}_{\alpha} - \vec{r}_{\beta}) + \sum_{B}\vec{F}_{\alpha B}(\vec{e}_{\alpha},\vec{r}_{\alpha} - \vec{r}_{i},t)$$

# Using the force



## Putting it all together

- $\frac{d\vec{w_{\alpha}}}{dt} := \vec{F}_{\alpha}(t) + fluctuations$
- Actual velocity  $\vec{v}_{\alpha} = \vec{w}_{\alpha} \cdot \begin{cases} 1 & , \text{ if } ||\vec{w}_{\alpha}|| < v_{0}^{\alpha} \\ \frac{v_{0}^{\alpha}}{||\vec{w}_{\alpha}||} & , \text{ otherwise} \end{cases}$



## Structure of our Program

- No objects
  - Objects need loops
  - Matrices are faster (usually)
- All agents stored inside one matrix

```
% | Agent 1 | Agent 2 | Agent 3
%------
% 1 position x |
% 2 position y |
% 3 speed x |
% 4 speed y |
% 5 desired speed v0|
% 6 goal |
% 7 last counter
% 8 ETR from FM
% 9 Red Carpet...
```

- Rather scripts than functions
  - Functions: pass by value
  - ► Global Variables: pass by "reference"

#### Simulation: Euler Method

#### Calculate forces

```
agents_f(:,agentID) = agents_force(A,agentID);
agents_p(:,agentID) =
potential_force(round(A(1,agentID)),round(A(2,agentID)));
```

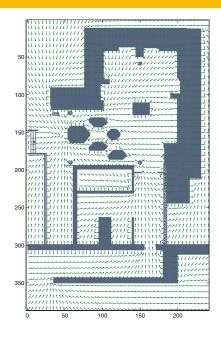
#### Resulting Velocities

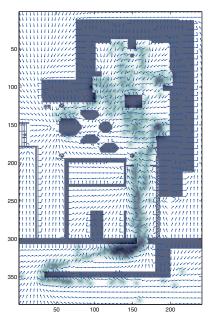
```
A(3:4,agentID) = (agents_p(:,agentID)...
*agents_f(:,agentID)...
+10 *[(rand(1)-.5);(rand(1)-.5)])...
*timestep;
```

## ▶ New positions: $\vec{x_{n+1}} = \vec{x_n} + \vec{v} \cdot \Delta t$

```
▶ deltaPos=A(3:4,:)*timestep;
A(1:2,:)=A(1:2,:)+deltaPos;
```

#### **Desired Direction**





## **Agent Forces**

- Distances to other Agents: A single matrix multiplication
  - ▶ distances=A(1:2,:)-A(1:2,alpha)\*ones(1,size(A,2))
- Neglect agents outside 1 meter

```
close_agents =
  sqrt(sum(r_alphabeta_matrix.^2))<sight;
  agent_others = A(:,close_agents);</pre>
```

Calculate resulting interaction forces

```
F_agents = A2*sum(...
  (ones(2,1)...
  *exp((2+tray_factor*(agent_alpha(6)==1)...
  *sigma*ones(1,agent_number_back-1)-sum(r_alphabeta_matrity)
  +sigma*tray_factor*(agent_others(6,:)==1))/B2)...
).*e_beta_matrix,2);
```

## Simple Queueing

#### Algorithm

- Look around. Who is headed towards the same goal?
- Who ist last in line?
- Join him/her

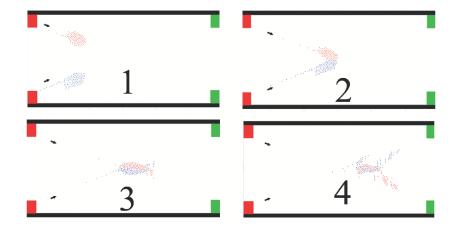
#### Implementation

- Fast Marching offers the expected time until reaching a certain goal
- Take those closer to the goal than you (less time expected)
- Of those take the last (max time)

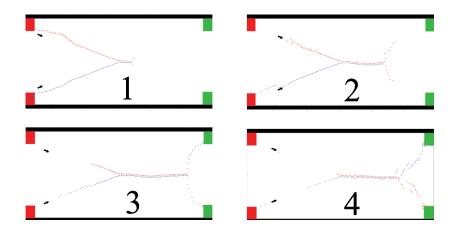
# **Testing Heuristics**

- ► Basic social force model
- additional Queuing
- additional real-time-Path-planning

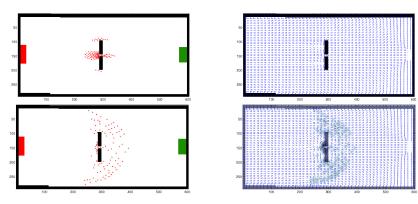
## **Basic Social Force Model**



# Queuing

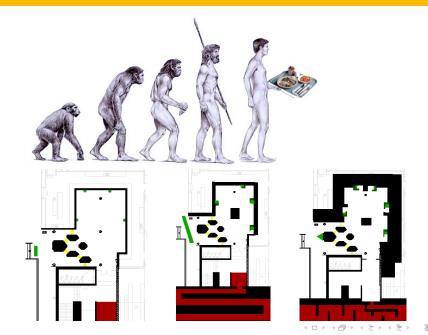


# Real-time Path Planning

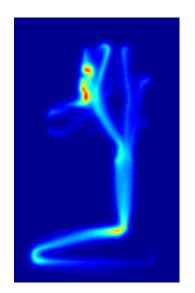


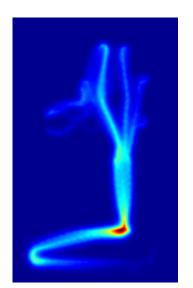
► Path planning takes 250 ms

## The Real Mensa

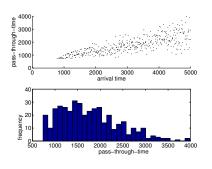


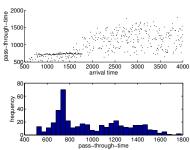
# Finding The Balance



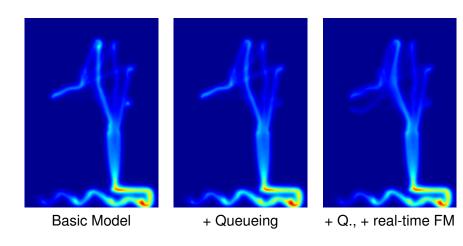


## Measuring The Balance





# Comparison In The Mensa



#### Video demo

- Struggling to get through waiting lines
- No waiting at the serving counters and checkouts

- Fetching food is way more than bare queueing
- We were able to reproduce some characteristics of the polymensa-Crowd
- Simulating 2 Frames/second is tedious

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