



Eidgenössische Technische Hochschule Zürich
Swiss Federal Institute of Technology Zurich

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

Project Report

Human Bees

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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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- 1 Abstract
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- 4 Description of the Model

The model describes the development of a bee hive. The colony consists of M bees, denoted with index i ($i = 1, 2, \dots, M$). Assume that there are N tasks, denoted by j ($j = 1, 2, \dots, N$). These tasks need to be performed in order for the colony to survive. They are associated with dynamic stimuli, denoted with s_j . The stimuli s_j , represent a measure of how urgent a specific task j needs to be performed. If a task is not pursued by enough workers of the hive or not in a sufficient rate, the need or the stimulus for the specific task will increase. Let θ_{ij} be the threshold for an individual i with respect to task j . Then the probability for an individual i to take up task j is given by

$$T_{\theta_{ij}} = \frac{s_j^2}{s_j^2 + \theta_{ij}^2} \quad (1)$$

One can see that for

$$\begin{aligned} s_j &\gg \theta_{ij}, T_{\theta_{ij}} \rightarrow 1; \\ s_j &= \theta_{ij}, T_{\theta_{ij}} = 0.5; \\ s_j &\ll \theta_{ij}, T_{\theta_{ij}} \rightarrow 0. \end{aligned}$$

So the higher the demand (or stimulus) for a task, the more probable it is for a bee to engage in it. Each individual bee i has a specific threshold θ_{ij} towards a given task j . Those bees i , that have a low θ_{ij} towards a task j will perform this task with a higher probability $T_{\theta_{ij}}$. Moreover, the thresholds vary over time as they reflect the ability of a human bee to perform a specific task. In full analogy to humans, a human bee i becomes better as it performs a specific task j during the time fraction x_{ij} . *Vice versa*, during time fraction $(1 - x_{ij})$, in which it does not perform this task, its ability to perform it, slowly decreases. Let ϕ be a fixed parameter associated with the learning process of a task and ζ fixed parameter associated with the forgetting process of a task. Then the temporal change of θ_{ij} is described by

$$\partial\theta_{ij} = [(1 - x_{ij})\phi - x_{ij}\zeta]\Theta(\theta_{ij} - \theta_{min})\Theta(\theta_{max} - \theta_{ij}) \quad (2)$$

where $\Theta(y)$ simply denotes a step-function:

$$\Theta(y) = \begin{cases} 0, & \text{if } y \leq 0 \\ 1, & \text{if } y > 0. \end{cases} \quad (3)$$

Θ is used to maintain θ_{ij} within the chosen boundaries $[\theta_{min}, \theta_{max}]$. Note that ϕ as well as ζ assume the same, constant value for each task. The temporal change of x_{ij} is given by

$$\partial x_{ij} = T_{\theta_{ij}}(s_j)(1 - \sum_{i=1}^M x_{ij}) - px_{ij} + \Psi(i, j, t) \quad (4)$$

The first term of the right hand side describes how the fraction of potentially free time $(1 - \sum_{i=1}^M x_{ij})$ is actually allocated to task performance j . The second term describes that a human bee stops performing a task and becomes inactive with probability p . p is identical and constant for all tasks and $1/p$ denotes the average time spent on task j . So our model requires a human bee to become inactive before it can continue to pursue a task in the next time interval, if the stimulus is high enough.

$\Psi(i, j, t)$ is.....???

Assuming that the demand for each task increases at a fixed rate δ per unit time, the temporal change of s is given by

$$\partial s_j = \delta - \frac{\alpha}{N} \left(\sum_{i=1}^N x_{ij} \right) \quad (5)$$

where α denotes a scale factor measuring the efficiency of task performance. Note that α assumes the same, constant value for all tasks.

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