

# Multi-Agent Systems

Albert-Ludwigs-Universität Freiburg



Bernhard Nebel, Felix Lindner, and Thorsten Engesser

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## Definition (Wilensky & Rand, 2015)

Agent-based modeling is a form of computational modeling whereby a phenomenon is modeled in terms of agents and their interactions.

- Agents are entities that have state variables and values (e.g., position, velocity, age, wealth)
  - Gas molecule agent: mass, speed, heading
  - Sheep agent: speed, weight, fleece
- Agents also have rules of behavior
  - Gas molecule: Rule to collide with another molecule
  - Sheep: Rule to eat grass
- Universal clock: At each tick, all agents invoke their rules.

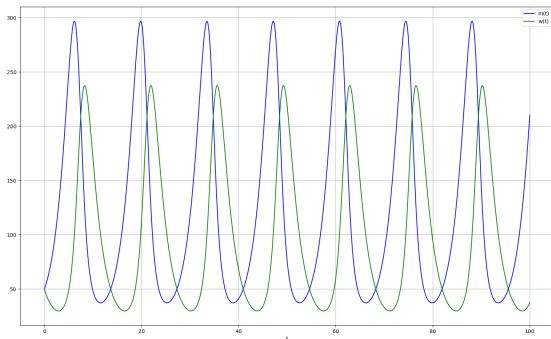
The populations of wolves and moose of Isle Royale have been observed for more than 50 years. Result: Dynamic variation rather than 'balance of nature'.

- More wolves
- ... leads to less moose
- ... leads to less wolves
- ... leads to more moose.

# Wolves and Moose: Classical Model

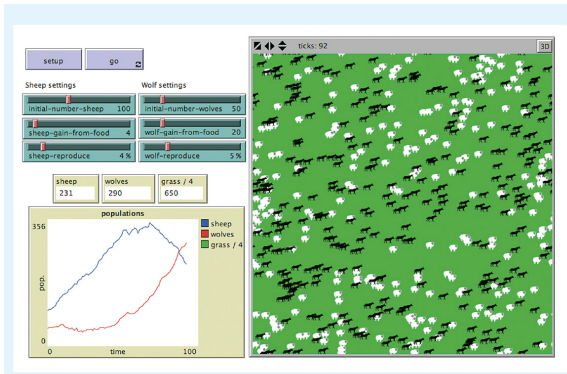
Lotka-Volterra model for wolf ( $w$ ) and moose ( $m$ ) populations:

$$\frac{\delta m}{\delta t} = k_1 m - k_2 w m, \quad \frac{\delta w}{\delta t} = -k_3 w + k_4 k_2 w m$$



# Wolves and Moose: Agent-Based Model

- Spawn  $m$  moose and  $w$  wolves and invoke each agent's behavior in each loop:
  - ask moose [move death reproduce-sheep]
  - ask wolves [move set energy energy - 1 catch-sheep death reproduce-wolves]



## Differential Equations

- Pro: Mathematically well understood, analytical inference by using calculus, many tools available (e.g., Matlab)
- Con: Hard to explain, models phenomenon rather than behavior, harder to extend

## Agent-Based Model

- Pro: Easy to understand and to explain to stakeholders, models individual behavior and observes emergent phenomenon, easy to extend
- Con: Tool support improves slowly, no analytical tools comparable to calculus

- Observation: Traffic on the motorway produces certain patterns.
- Question: Can similar patterns be algorithmically reproduced?
- Agent-Based Simulation approach:
  - Modeling traffic on the motorway as a multi-agent system
  - Cars (drivers) as agents
    - Percepts: Distance to next car in front
    - Internal State: Current Speed
    - Actions: Speeding, braking



- **Research Question:** How do traffic jams emerge?
- **Research Hypothesis:** Might be due to the local behaviour of individual agents.
- **Approach:** Model traffic as a MAS and study the resulting system's behavior. If the systems' behavior matches empirical phenomenon, then the model might be an acceptable explanation.

- A **cellular automaton** is a quad-tuple  $A = \langle R, Q, N, \delta \rangle$
- A **cell space**  $R$
- A set  $Q$  of **states** each cell can be in
- A **neighborhood**  $N : R \rightarrow 2^R$
- A **transition function**  $\delta : Q^{|N|} \rightarrow Q$ 
  - For a probabilistic cellular automaton,  $\delta$  is a probability distribution  $P(r = q | N(r))$
- The **configuration** of  $A$  can be written as  $x_1 x_2 \dots x_n$  with  $x_i$  being the state of the cell  $r_i$ .

- Traffic is modeled as  $A = \langle R, Q, N, \delta \rangle$
- Entities of  $R = \{c_1, c_2, \dots\}$  stand for parts of the lane
  - Each cell corresponds to a discrete part of the lane (roughly the space needed by a car)
- $Q = \{0, \dots, v_{max}, free\}$ : Each cell is either occupied by one car with velocity  $v \leq v_{max}$ , or it is empty.
- $N(c_i) = \{c_{i-v_{max}}, \dots, c_{i+1}\}$
- $\delta$  is realized by a set of four rules executed by each driver

- Each car at cell  $c_i$  with velocity  $v$  performs four consecutive steps:
  - **Acceleration:** If  $v < v_{max}$  and gap to next car is larger than  $v + 1$ , then increment speed by 1.
  - **Slowing down:** If the next car is at cell  $i + j$  with  $j \leq v$ , then reduce speed to  $j - 1$ .
  - **Randomization:** If  $v > 0$ , then decrement  $v$  by 1 with probability  $p$ .
    - Car does not accelerate although it could (takes back **Acceleration**)
    - Car reached maximal velocity but slows down again
    - Overreaction when braking
  - **Car motion:** Move forward  $v$  cells.

# Nagel-Schreckenberg: Example

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# Nagel-Schreckenberg: Example

2	—	—	2	—	—	2	—	—
—	—	2	—	—	2	—	—	2
—	2	—	—	2	—	1	—	—

# Nagel-Schreckenberg: Example

2	_	_	2	_	_	2	_	_
_	_	2	_	_	2	_	_	2
_	2	_	_	2	1	_	_	_
_	_	_	2	0	_	_	_	2

# Nagel-Schreckenberg: Example

2	—	—	2	—	—	2	—	—
—	—	2	—	—	2	—	—	2
—	2	—	—	2	—	1	—	—
—	—	—	2	0	—	—	—	2
—	2	—	0	—	1	—	—	—



- Assume constant **system density**:  $\rho = \frac{|Ag|}{|R|}$
- For a fixed cell  $c_i$ , **time-averaged density** over time interval  $T$ :

$$\bar{\rho}^T = \frac{1}{T} \sum_{t=t_0+1}^{t_0+T} n_i(t)$$

- ...with  $n_i(t) = 1$  if  $i$  is occupied, else  $n_i(t) = 0$
- **Time-averaged flow**  $\bar{q}$  between  $i$  and  $i+1$ :

$$\bar{q}^T = \frac{1}{T} \sum_{t=t_0+1}^{t_0+T} n_{i,i+1}(t)$$

- ...with  $n_{i,i+1}(t) = 1$  if some car moved between  $i$  and  $i+1$  at  $t$ , else  $n_{i,i+1}(t) = 0$

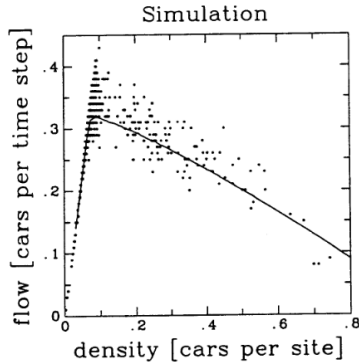


Fig. : Source: [2]

- **Exercises:** You will implement the Nagel-Schreckenberg Simulation by yourself and thereby become familiar with our multi-agent simulation framework.
- **Next Week:** On the diversity of agent architectures



U. Wilensky, W. Rand, An Introduction to Agent-Based Modeling, MIT Press, ISBN: 9780262731898, 2015.



K. Nagel, M. Schreckenberg (1992), A cellular automaton model for freeway traffic, J. Phys. I France 2, pp. 2221–2229.