

Collective Robotics

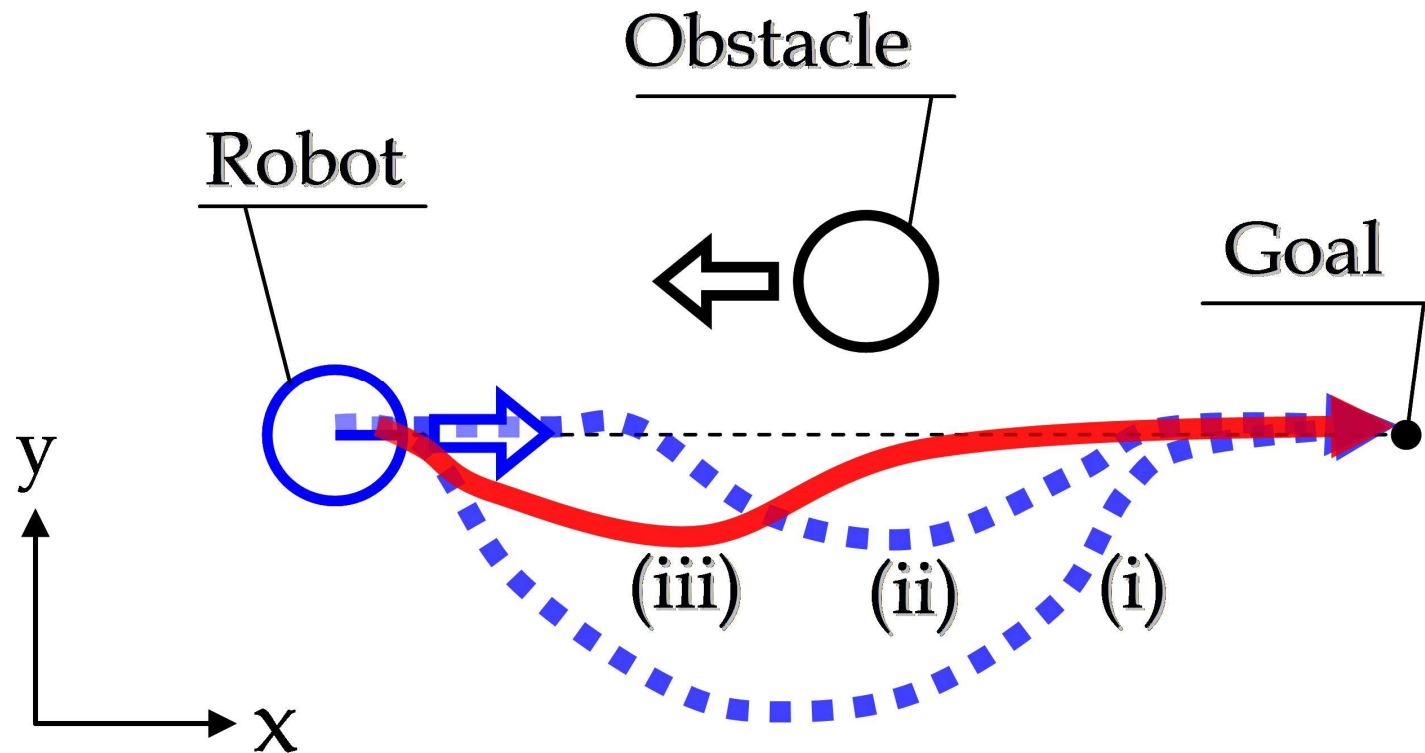
Part 2: Short Journey Through Nearly Everything

Prof. Dr. Javad Ghofrani



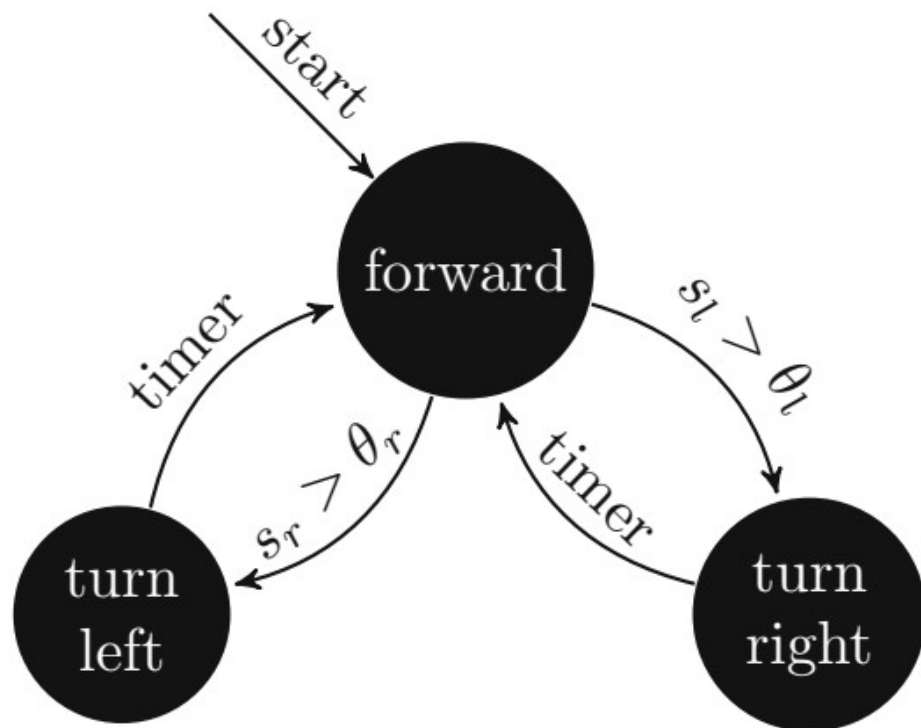
**Hochschule
Bonn-Rhein-Sieg**
University of Applied Sciences

Simple task: collision avoidance



(Takafumi Suzuki and Masaki Takahashi)

Finite state machines as robot controllers (1/2)



Left sensor s_l

Right sensor s_r

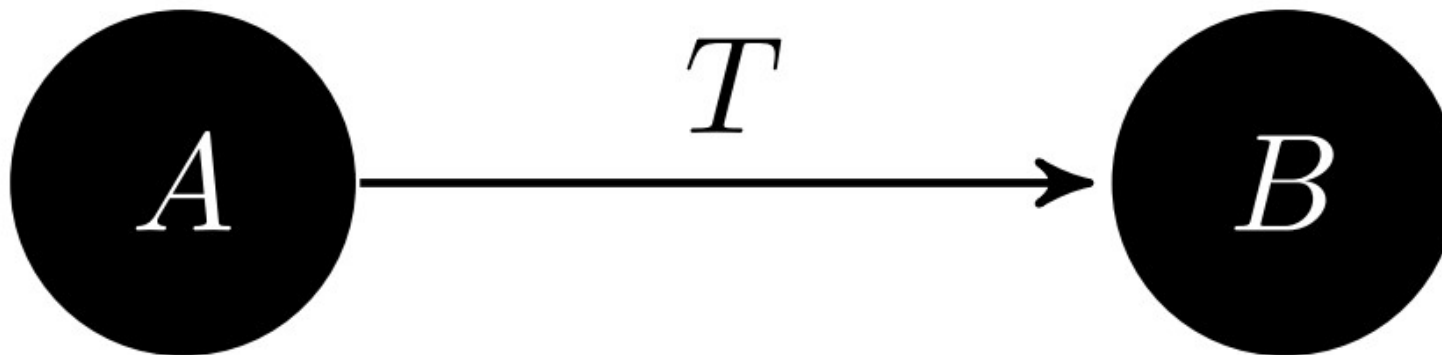
Threshold θ_r, θ_l

(big values mean a close-by object was detected)

'timer' : the state is left when a certain time period has elapsed

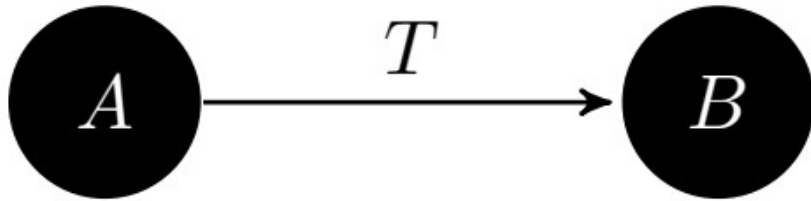
Finite state machines as robot controllers (2/2)

general case:



any condition T for the transition is possible

Simple population model



swarm size N

only two possible states: A and B

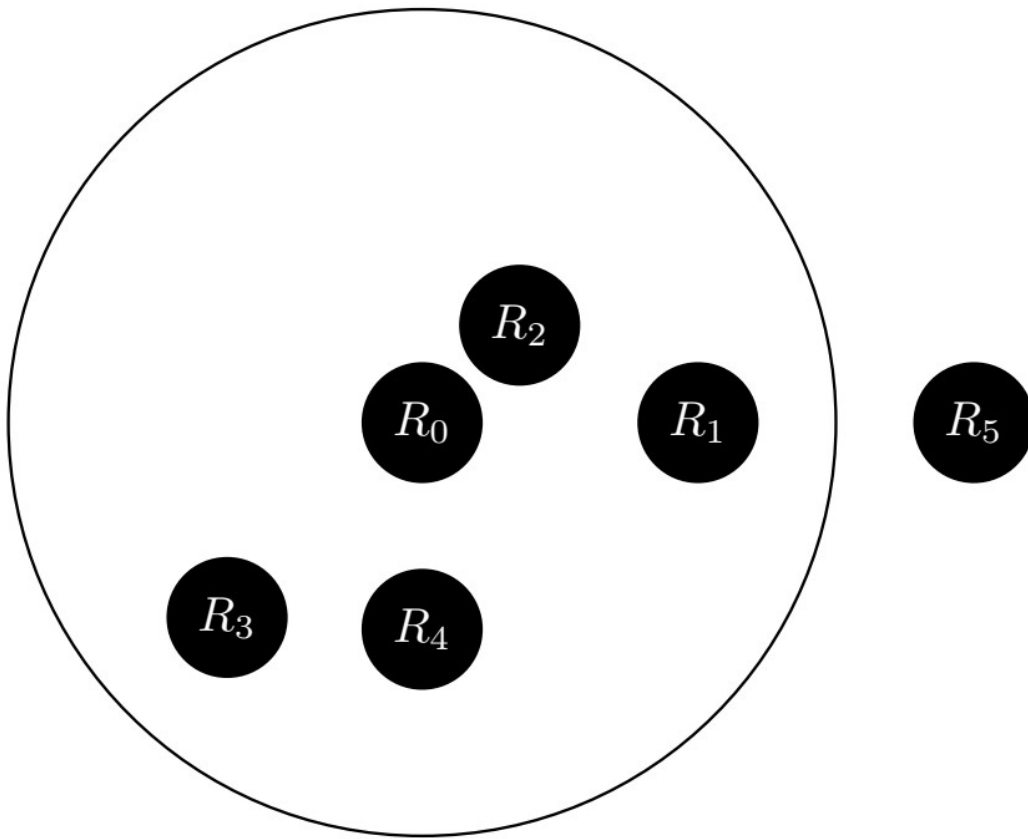
variable a counts robots currently in state A

variable b counts robots currently in state B

$$N = a + b$$

fraction of robots that are in state A : $\alpha = \frac{a}{N}$

State transitions & robot-robot interactions (1/2)



sensor range r

neighborhood of robot R_0 is:

$$N = \{R_1, R_2, R_3, R_4\}$$

State transitions & robot-robot interactions (2/2)

\hat{a} gives the number of neighboring robots of robot R_1 plus itself that are currently in state A

\hat{b} for state B

$|N|$ is size of neighborhood

$$|N| + 1 = \hat{b} + \hat{a}$$

fraction of robots in state A :

$$\hat{\alpha} = \frac{\hat{a}}{|N| + 1}$$

next:

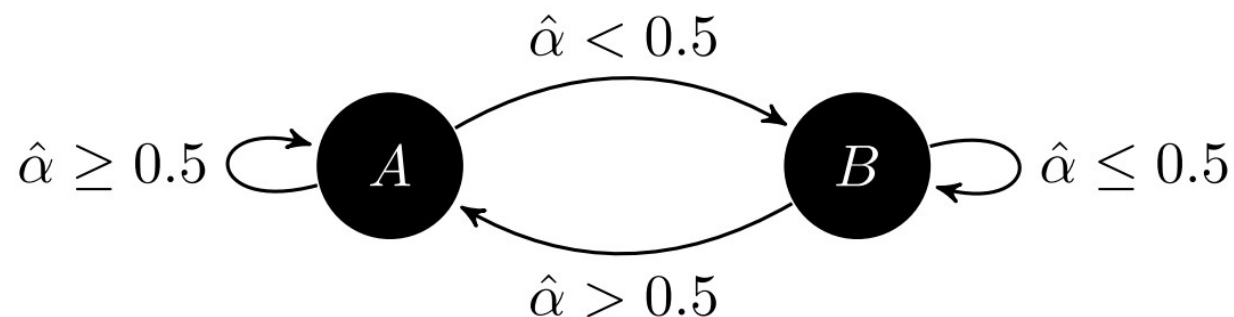
say, a robot's state transitions between A and B

depend exclusively on $\hat{\alpha}$

Minimal example: collective decision-making

task in collective decision-making: find a consensus
(100% of the robots in the swarm agree on a decision)

we use $\hat{\alpha}$ and a threshold of 0.5:



majority rule: reinforce the current majority

Towards a macroscopic perspective

What's the effect on the swarm level?

we can interpret $\hat{\alpha}$ as a local measurement of the global situation α

\Rightarrow local sampling (later)

generally: $\alpha \neq \hat{\alpha}$

could we hope for $\alpha \approx \hat{\alpha}$?

for a so-called 'well-mixed' system (without bias) we can assume in average: $\alpha \approx \hat{\alpha}$

but well-mixed assumption does often not hold!

this is a small taste of the \Rightarrow micro-macro problem (later)

A macroscopic perspective (1/2)

Measurement of $\hat{\alpha}$ is probabilistic

State transitions for a given $\hat{\alpha}$ are deterministic

We have a look into the combinatorics of all possible neighborhoods

For simplicity, small neighborhood of $|N| = 2$

probability $P_{B \rightarrow A}$ to switch from B to A is

$$P_{B \rightarrow A}(\alpha) = (1 - \alpha)\alpha^2 \quad (1)$$

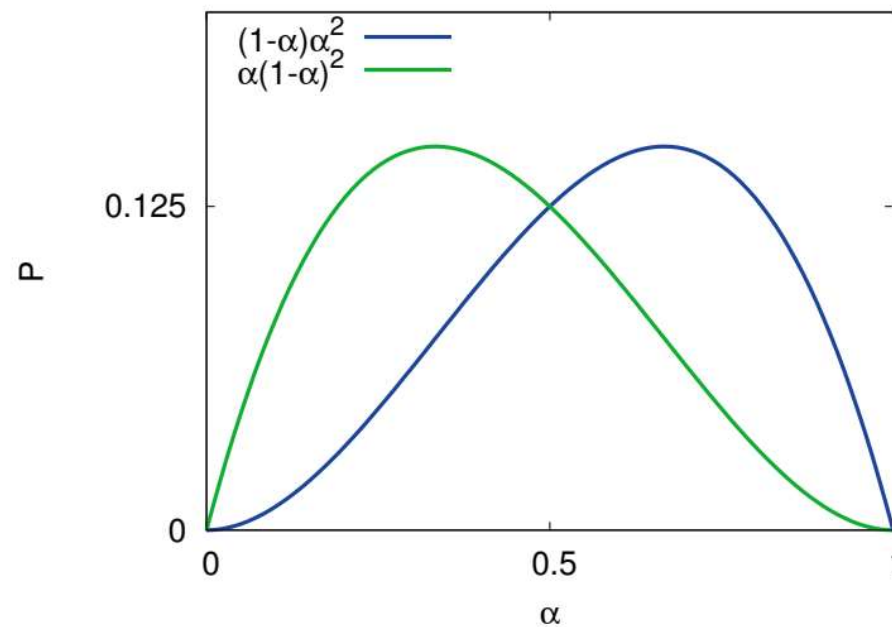
probability to switch from A to B is fully symmetric

$$P_{A \rightarrow B}(\alpha) = \alpha(1 - \alpha)^2 \quad (2)$$

A macroscopic perspective (2/2)

$$P_{B \rightarrow A}(\alpha) = (1 - \alpha)\alpha^2$$

$$P_{A \rightarrow B}(\alpha) = \alpha(1 - \alpha)^2$$



Expected macroscopic dynamics

How does the swarm fraction α develop over time?
normalized by a considered time interval Δt
expected change $\Delta\alpha$ of α :

$$\frac{\Delta\alpha(\alpha)}{\Delta t} = \frac{1}{N} \left((1 - \alpha)\alpha^2 \right) - \frac{1}{N} \left(\alpha(1 - \alpha)^2 \right) \quad (3)$$

Dynamics & feedbacks

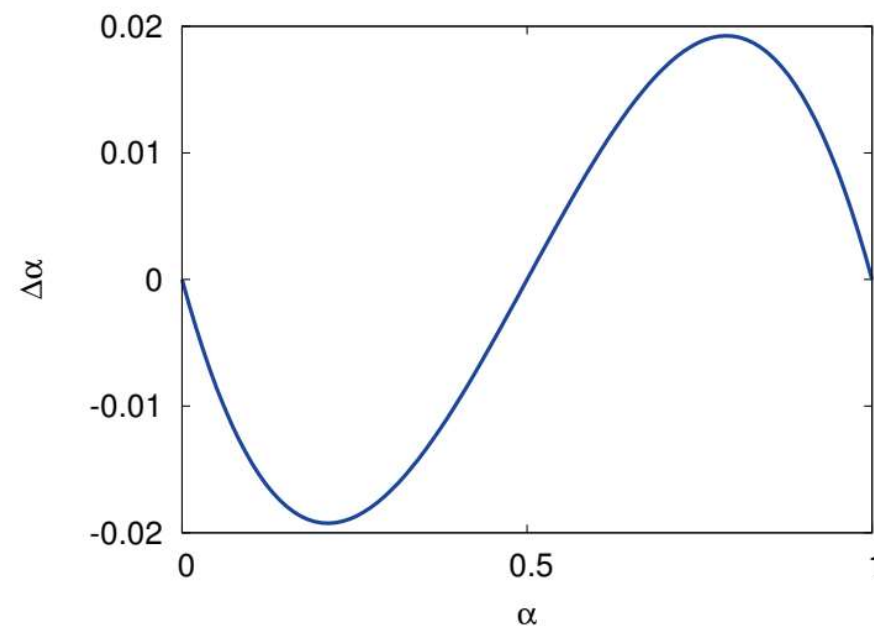
$$\frac{\Delta\alpha(\alpha)}{\Delta t} = \frac{1}{N}((1-\alpha)\alpha^2) - \frac{1}{N}(\alpha(1-\alpha)^2)$$

here for $\Delta t = 1$, $N = 10$

$1/N$: assumed switching rate of one robot per per time step

What kind of feedback do we see here?

positive feedback only!



Self-organizing system

Does this robot swarm qualify as self-organizing system?

positive feedback ✓

negative feedback ✗

initial fluctuations ✓

multiple interactions ✓

balance between exploitation and exploration ✗

Discussion : How could exploration be included?