

Framsticks Fuzzy Control

Maciej Hapke and Maciej Komosinski. Evolutionary design of interpretable fuzzy controllers. *Foundations of Computing and Decision Sciences*, 33(4):351-367, 2008. [\[view pdf\]](#)

in cooperation with *Dawid Waclawski*

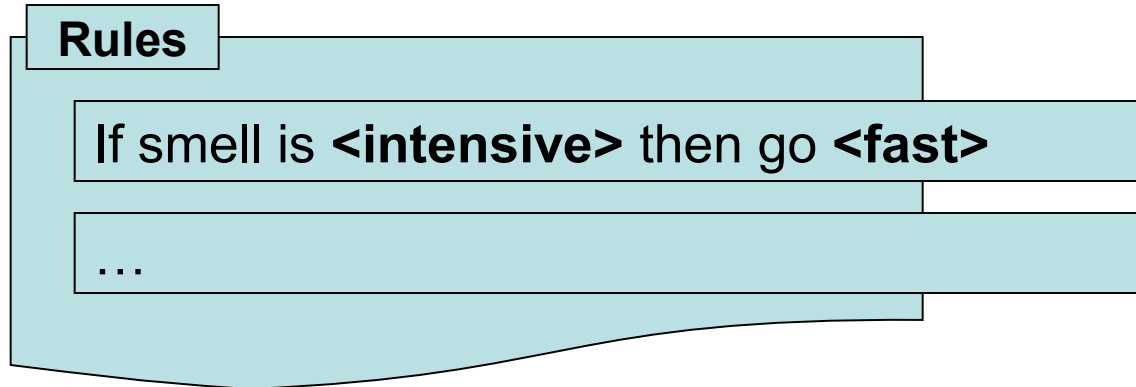
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Observations & motivations

- Observations:
 - framsticks move in a way similar to those evolved in the nature – e.g. virtual lizard, water snake
 - the simulation confirms that real evolution makes bodies of different structures move optimally
- Questions about the reasons:
 - why creatures behave in such a way
 - what caused such development of B&B
- This knowledge is hidden in the brain
- A trial to explain evolution

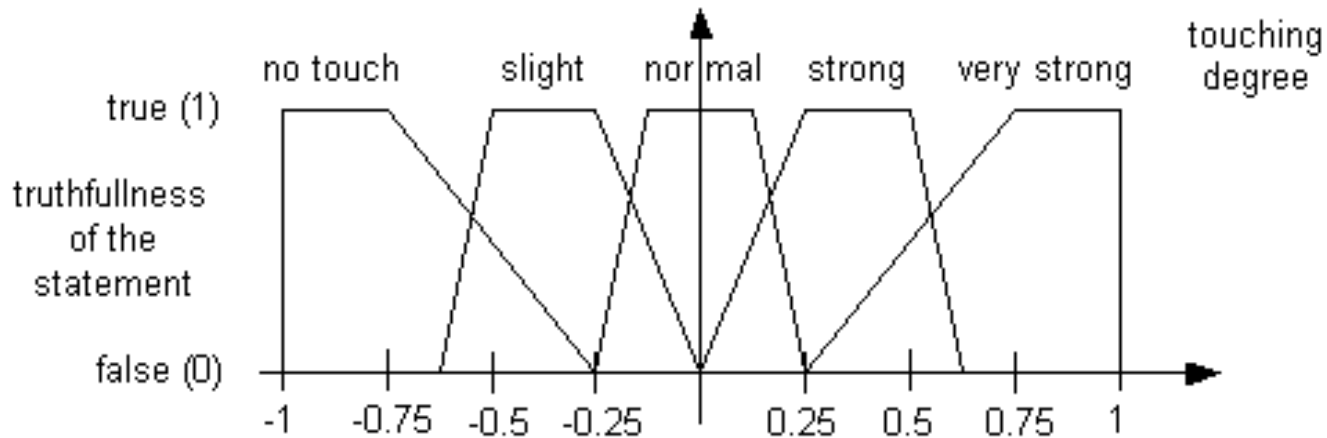
Knowledge representation

- ANN ?
- Fuzzy system



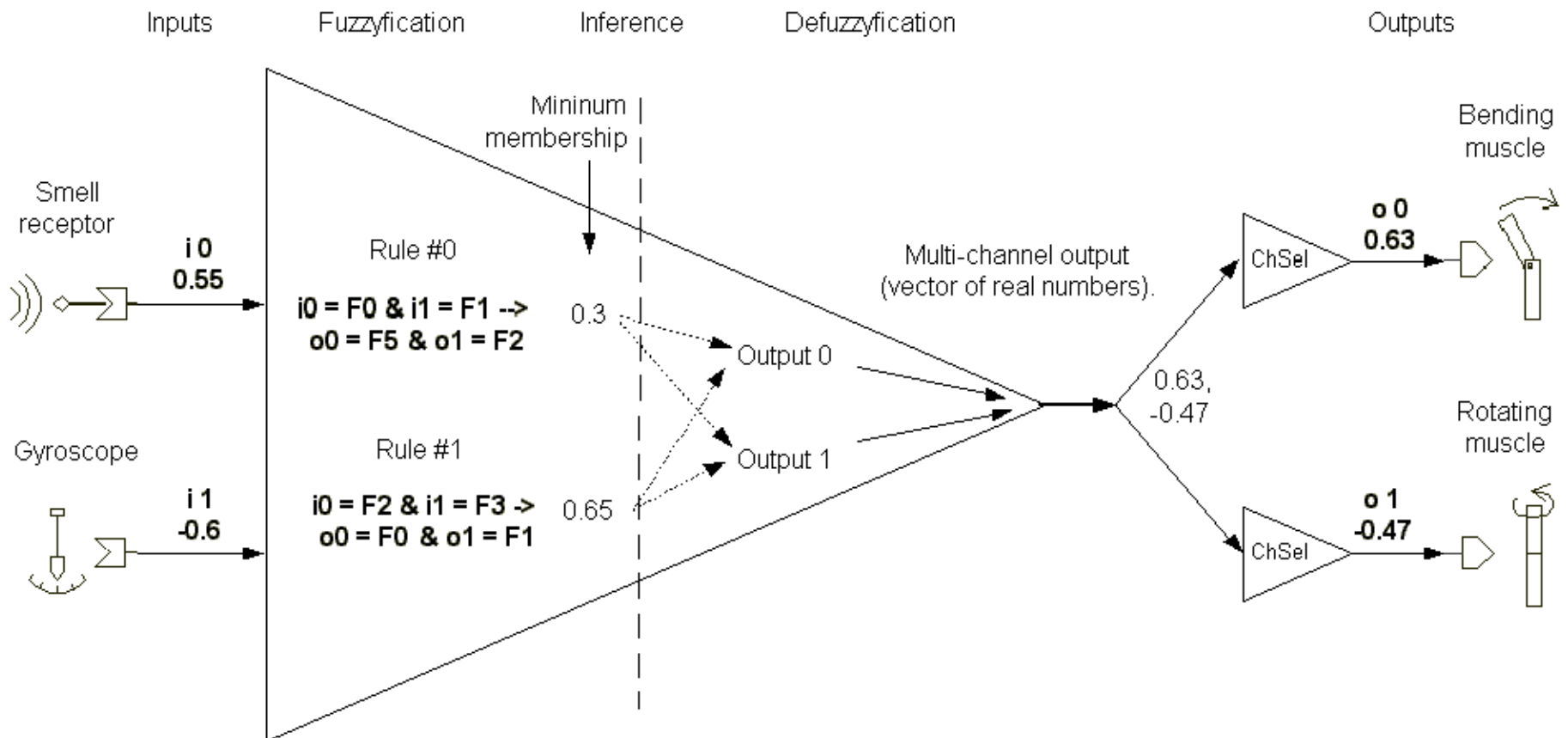
Fuzzy variables

- Example - touch sensor
- Normalized variable domain



Fuzzy “neuron”

- Fuzzy system representation
- Mamdani approach



Evolutionary encoding of FS

- Fuzzy “neuron” genotype sections

Def	Fuzzy sets	Fuzzy rules
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```
n:d="Fuzzy:ns=4, nr=2,  
fs=-0.1647;-0.1526;-0.0087;0.0631;  
-1.0000;-0.8774;-0.7725;-0.6767;  
0.0087;0.2308;0.3585;0.4806;  
0.0110;0.1664;0.2362;0.2718;  
fr=0;3;1;0;2;0:0;2;3;1;2;1;1;3/  
2;0;0;2;1;2:3;1;2;0;1;2;0;0/"
```

Example

The example of a fuzzy rule-based system with two inputs (x_0, x_1), two outputs (y_0, y_1), two rules (R_0, R_1) and five fuzzy sets ($F_0 \dots F_4$) can be described as follows:

$$F_0 = \{-0.35; 0.05; 0.4; 0.65\}$$

$$F_1 = \{-1; -0.8; -0.8; -0.35\}$$

$$F_2 = \{0.2; 0.5; 0.7; 0.8\}$$

$$F_3 = \{-0.65; -0.5; -0.3; 0.1\}$$

$$F_4 = \{0.4; 1; 1; 1\}$$

R_0 : IF x_0 is F_0 AND x_1 is F_1 THEN y_0 is F_5 AND y_1 is F_2

R_1 : IF x_0 is F_2 AND x_1 is F_3 THEN y_0 is F_0 AND y_1 is F_1

Evolutionary operators

Mutation

- Add/remove a fuzzy set
- Add/remove a fuzzy rule
- Add/remove an input/output

Evolutionary operators

Crossover

- One/multiple crossing points
- Inheritance
 - Two parents/one descendant
 - Parents may be of different length
 - Crossover
 - Draws # of rules
 - For each rule
 - Randomly chooses a pair of rules from p1 and p2
 - Draws # of inputs and outputs
 - Copies inputs and outputs

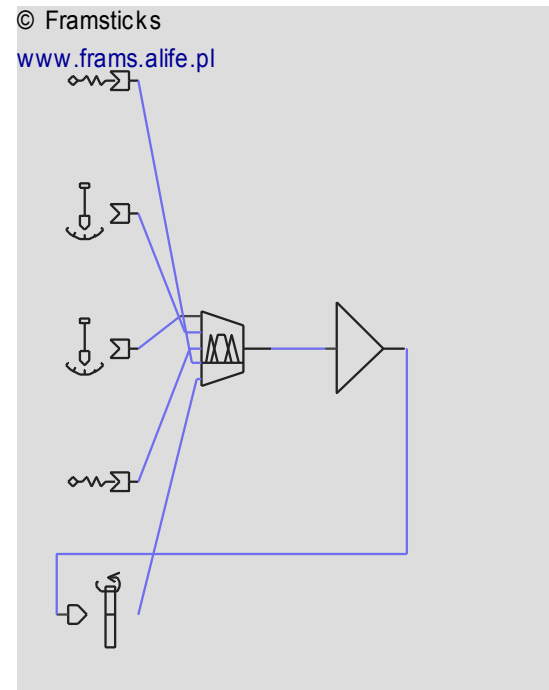
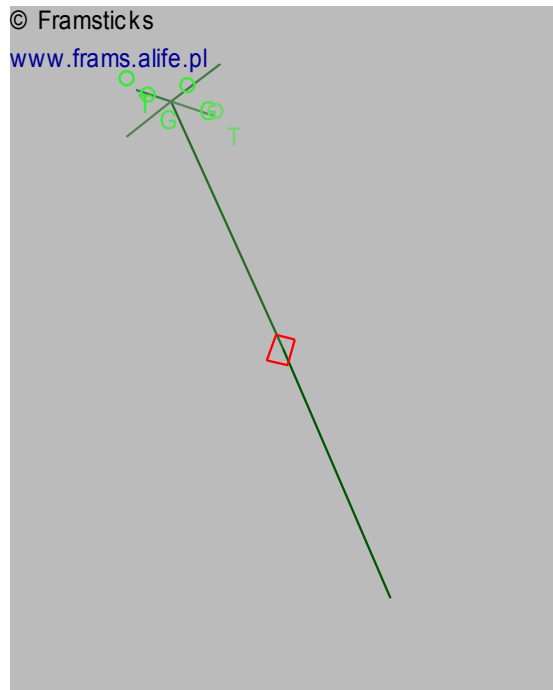
Experiment design

- Goal: to evolve only “fuzzy brain”
- Fixed body structure (parts, joints)
- Fixed # of receptors
- Variable # of fuzzy sets
- Variable # of fuzzy rules

Experiments

Stand-up agent

- Inputs: 2 gyroscopes, 2 touch sens.
- Output: muscle



Experiment

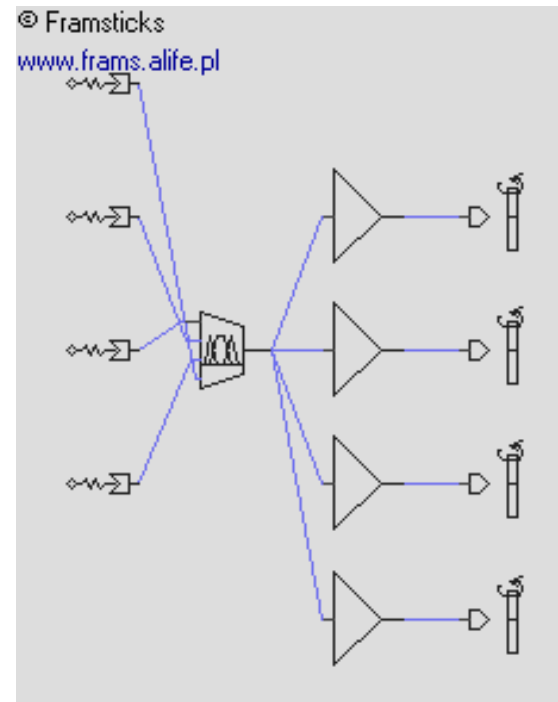
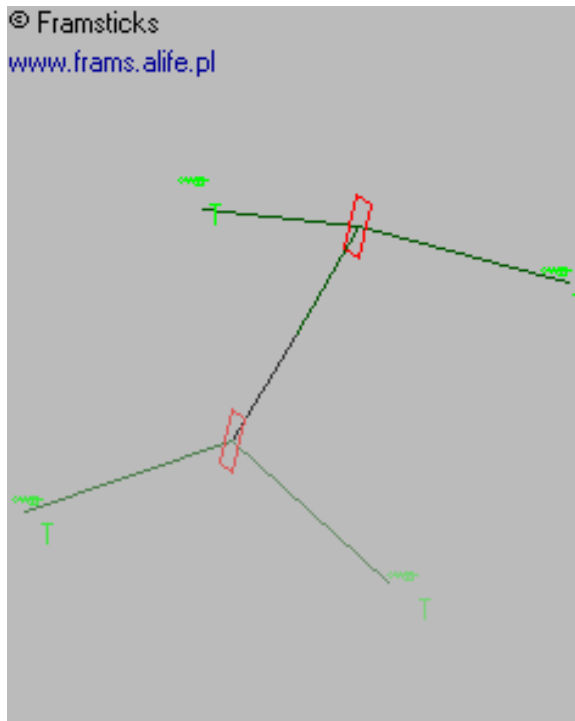
Stand-up agent

- Fitness function: maximize the average height
- The goal of a FS: to force the agent to stand up
- Two example creatures chosen from the population

Experiments

Walker

- 4 inputs: touch receptors,
- 4 outputs: rotating muscles
- Fitness function: velocity



A movie

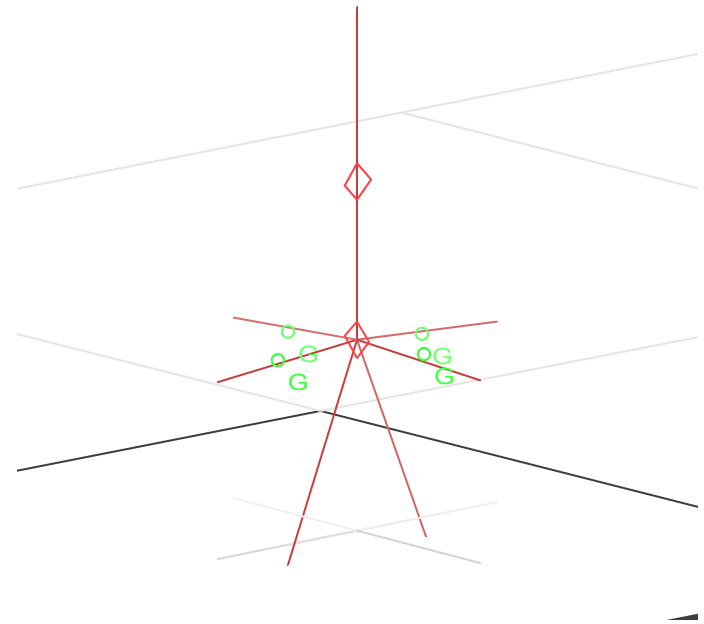
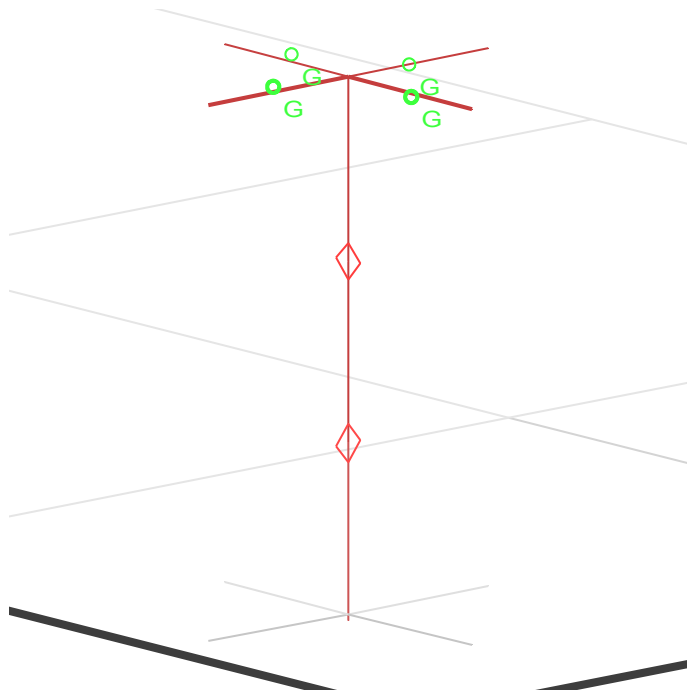
Walker

Conclusions

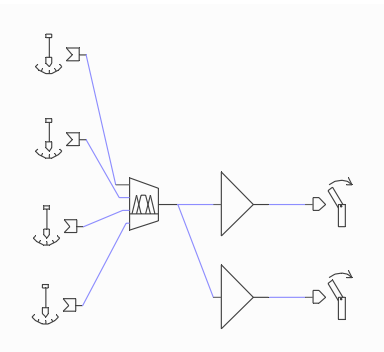
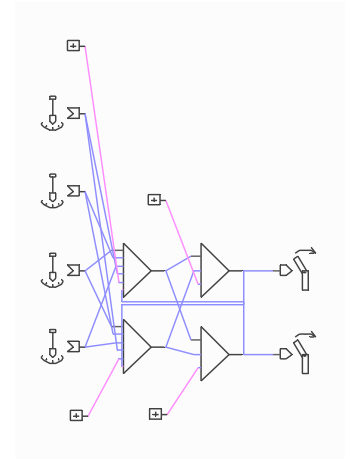
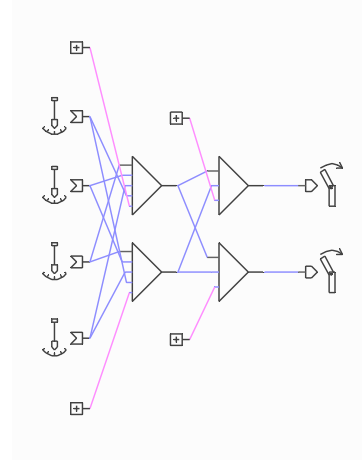
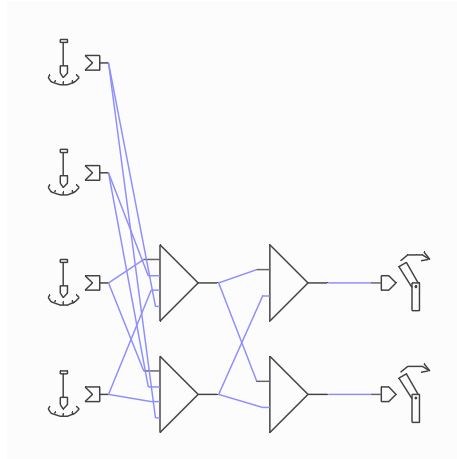
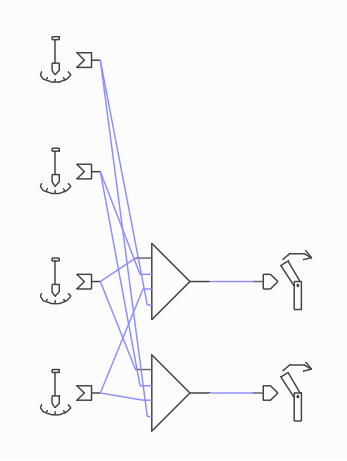
- Walker's behavior
 - Slightly jumps by means of back legs
 - Runs all around, fitness function does not imply straight running
- Two fuzzy rules are enough

Inverted pendula problem

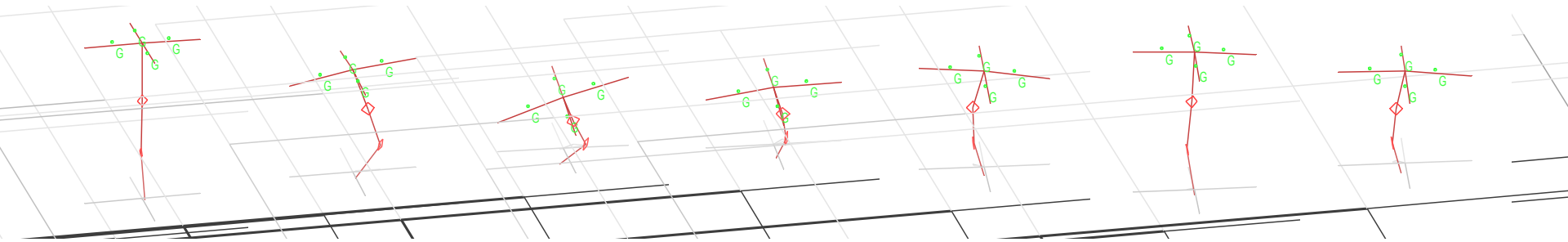
- Modified: active and elastic



Comparison with NN control



Evolved balancing behavior



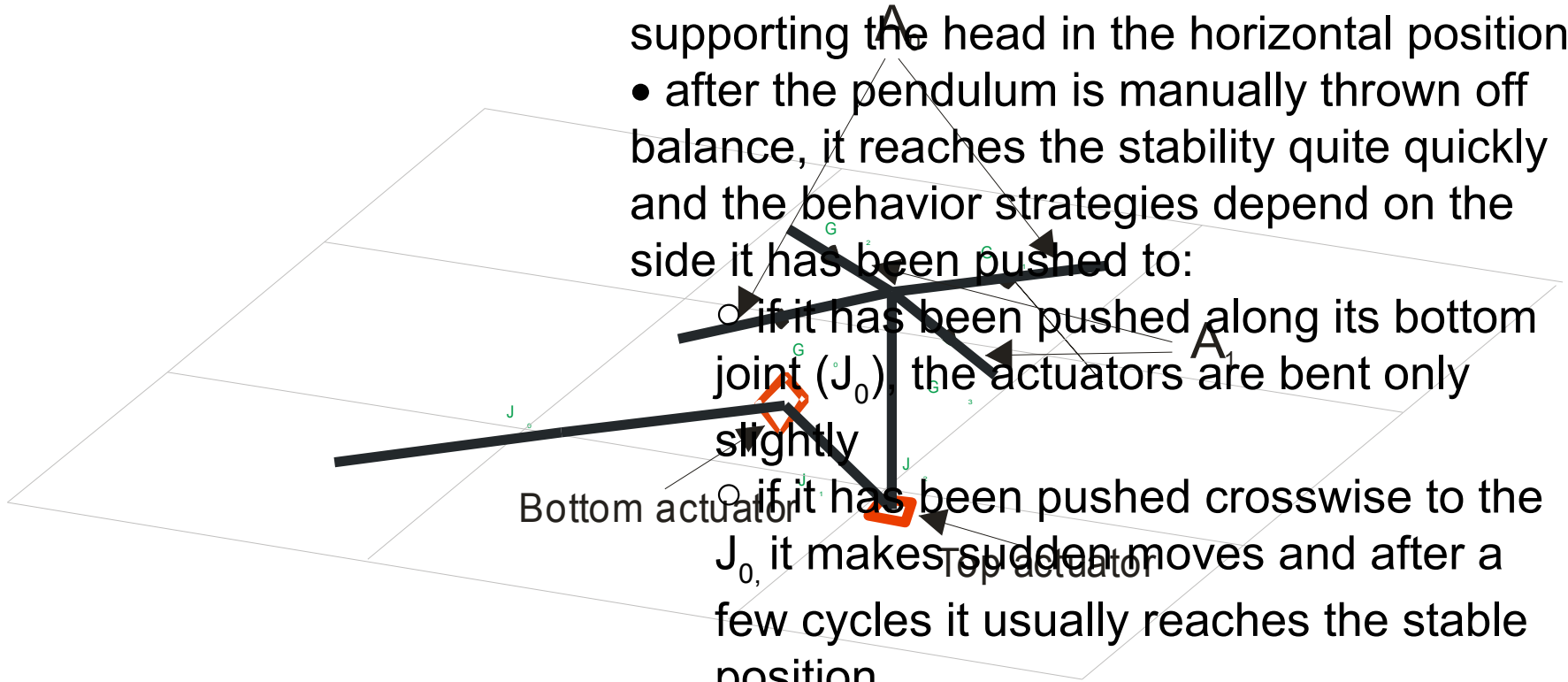
Understanding evolved fuzzy rules

- in the stable position, J_0 and J_1 lie down on the ground, while J_2 stands upright supporting the head in the horizontal position
- after the pendulum is manually thrown off balance, it reaches the stability quite quickly and the behavior strategies depend on the side it has been pushed to:

- if it has been pushed along its bottom joint (J_0), the actuators are bent only slightly

- if it has been pushed crosswise to the J_0 , it makes sudden moves and after a few cycles it usually reaches the stable position

- if the pendulum falls upside down, the fuzzy system is unable to make it stand straight.



Understanding evolved fuzzy rules

Each fuzzy system has four inputs and two outputs. Input signals s_0, s_1, s_2, s_3 come from four sensors. Based on their values, the fuzzy system sends two outputs signals for actuators: $bend_bottom$ and $bend_top$. Input and output fuzzy variables are defined in the normalized domain $[-1, 1]$. Input linguistic variables $upright$, $leveled$ and $upside_down$ are defined as follows: $(-1, -1, -1, 0)$, $(-1, 0, 0, 1)$ and $(0, 1, 1, 1)$, while the outputs characterizing bending directions are expressed by linguistic variables $right$ $(-1, -1, -1, 0)$, $none$ $(-1, 0, 0, 1)$ and $left$ $(0, 1, 1, 1)$.

Understanding evolved fuzzy rules

1. `s2=leveled and s0=leveled` \Rightarrow `bend_bottom=left and bend_top=left`
2. `s3=leveled and s1=upside_down` \Rightarrow `bend_top=left`
3. `s1=upright` \Rightarrow `bend_bottom=left and bend_top=left`
4. `s3=upside_down` \Rightarrow `bend_bottom=right and bend_top=left`
5. `s1=upside_down` \Rightarrow `bend_bottom=left and bend_top=none`

- The pairs of sensor signals (s_0, s_1) and (s_2, s_3) never come together in a single premise of the rule. It is because the optimization process discovered a property of the pendulum structure: the signals from these equilibrium sensor pairs are almost the same. This is the consequence of placing sensors (G_0, G_1) and (G_2, G_3) on the same arms, respectively.
- See text for detailed explanation and analysis of each rule.

Conclusions

- Successful (evolutionary) simplification of the fuzzy system (from 20 rules to 5 rules)
- both evolution of neural and fuzzy controllers for active inverted pendulum lead to similar pendulum behaviors
- NNs easier to optimize
- verified ability to extract knowledge from the fuzzy control system