### Internal State Predictability as an Evolutionary Precursor of Self-Awareness and Agency

ICDL 2008

August 11, 2008

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#### **Motivation**

The concept of self (self-awareness, agency) is an important yet hard subject:

- It may lead to consciousness.
- It may be necessary for social interaction.
- It may play an important role in cognition (Block 1995).

#### **Research Question: Self-Awareness**

Why did self-awareness (or the sense of self) evolve?

- Self-awareness is an internal state that may be transparent to the process of evolution (cf. high-performance zombie).
- This is a hard question to answer without getting tangled in philosophical debate.

**Strategy**: Investigate the **necessary condition** of self-awareness that may be less controversial.

### Approach

Identify **necessary conditions** of self-awareness:

- Sense of self and agency are closely related.
- Authorship is a key ingredient: "I" prescribe my actions, and "I" own them.
- Important property of authorship: My actions are highly predictable while others' are not.

**Necessary condition identified:** Need to be able to **predict** one's own internal state (cf. Nolfi et al. 1994).

#### Method (Task): 2D Pole-Balancing



• Physical parameters of the pole balancing system: position (x, y); velocity  $(\dot{x}, \dot{y})$ ; pole angle  $(\theta_x, \theta y)$ ; angular velocity  $(\dot{\theta_x}, \dot{\theta_y})$ .

#### **Method: Neuroevolution Controller**



- Recurrent neural network for 2D pole balancing.
- Trained with standard neuroevolution.
- Investigate the internal state trajectories.



- 1. Evolve controllers to meet a fixed performance criterion (fitness does not measure predictability) in pole-balancing tasks.
- 2. Group high-performance individuals in to high- and low internal state predictability (ISP) groups.
- 3. Test the two groups in harder tasks.

### **Method: Experimental Setup**

- Neuroevolution:
  - population size 50
  - mutation rate 0.2; cross over rate 0.7.
- 2D pole balancing task:
  - Pole should be balanced within 15° within a 3 m  $\times$  3 m arena.
  - Force applied to cart every 0.1 second (= one step).
  - Success if pole balanced over 5,000 steps.

#### **Method: Measuring Predictability**



 $\hat{x}(t+1) = f(x(t), x(t-1), x(t-2), \cdots, x(t-N+1)).$ 



• Neural network predictor for a time series.

### **Method: Experimental Setup**

- Neural network predictor:
  - 2,000 training data.
  - 1,000 test data.
  - Back-propagation (learning rate 0.2).

# **Results: Internal State Predictability**

(ISP)



#### **Internal State Predictability**

Evolved agent sorted by the prediction rate

- Trained 130 pole balancing agents.
- Chose top 10 highest ISP agents and bottom 10 lowest ISP.
  - high ISPs:  $\mu=95.61\%$  and  $\sigma=5.55\%.$
  - low ISPs:  $\mu=31.74\%$  and  $\sigma=10.79\%.$

#### **Comparison High ISP and Low ISP**



- A comparison of the average predictability from two groups: high ISP and low ISP.
- The predictive success rate of the top 10 and the bottom 10 agents.

#### **Results: Learning Time**



• No significant difference in learning time

#### Performance and Int. State Dyn.



- Made the initial conditions in the 2D pole balancing task harsher.
- Performance of high- and low-ISP groups compared.
- High-ISP group outperforms the low-ISP group in the changed environment.

#### **Behavioral Predictability**



- Success of high-ISP group may simply be due to simpler behavioral trajectory.
- However, predictability in behavioral predictability is no different between high- and low-ISP groups.

## Examples of internal state dynamics

#### from the high ISP



• Internal state dynamics show smooth trajectories.

#### **Examples of internal state dynamics**

#### from the low ISP



• Internal state dynamics show abrupt and jittery trajectories.

## Examples of cart x and y position from high ISP



• Behavioral trajectories of x and y positions show complex trajectories.

## Examples of cart x and y position from low ISP



• Behavioral trajectories of x and y positions show complex trajectories.

#### **Related Work**

- Bayesian self-model (Gold and Scassellati 2007).
- Continuous self re-modeling for resilient machines (Bongard et al. 2006).
- Autonomous mental development (Weng et al. 2001; Han et al. 2002).
- Role of self-awareness in cognition (Block 1995).
- Emergence of self-awareness from self-representation (Menant 2007).

#### Conclusions

- Simpler (more predictable) internal dynamics can achieve higher levels of performance in harsher environmental conditions.
- The increased survival value is not always due to smoother behavior resulting from the simpler internal states.
- Initially evloution-transparent internal agent properties can affect external behavioral performance and fitness in a changing environment.
- An initial stepping stone in the evolutionary pathway leading to self-awareness and agency could have formed in such a way.

#### Why Do We Have a Brain?



Tree

(no Brain)



• Brain vs. no brain (cf. Llinás et al. 1994).

Sources: http://homepages.inf.ed.ac.uk/jbednar/ and http://bill.srnr.arizona.edu/classes/182/Lecture-9.htm

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