

# Lecture 10: Part 2 - Design Principles

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**Summary and conclusions**

**"How the body shapes the way we think -  
principles and insights"**

**(slides for self-study)**

**Rolf Pfeifer**

**11 December 2012**



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# Agent design principles

<b>Agent design principles</b>	<b>Description</b>
Three constituents	ecological niche, behaviors/tasks, and agents
Complete agent	complete agent, not only isolated components
Parallel, loosely coupled processes	parallel, asynchronous, processes, largely coupled through interaction with environment
Sensory-motor coordination	behavior sensory-motor coordinated with respect to target; self-generated sensory stimulation
Cheap design	exploitation of niche and interaction; parsimony
Redundancy	partial overlap in functionality based on different physical processes
Ecological balance	Balance in complexity of sensory-motor and neural system; task-distribution: morphology, materials, control, environment
Value	driving forces, developmental mechanisms, self-organization

# Design principles for development

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<b>Principles for development</b>	<b>Description</b>
Integration of time scales	many time scales need to be integrated into one agent
Development as an incremental process	start simply, build successively on top of what has already been learned
Discover	agent must have ability to explore and evaluate, which implies that agent can discover through its own activities
Social interaction	sensory-motor coordination together with social interaction provides most powerful engine for development
Motivated complexity	why complexity increases during ontogenetic development (driving force) 3

# Design principles for evolution

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<b>Principles for evolution</b>	<b>Description</b>
Population	Population is the prerequisite for evolution to function
Cumulative selection and self-organization	Cumulative selection will produce interesting results only if evolutionary process exploits processes of self-organization
Brain-body co-evolution	“Brain” (neural control) and body must be evolved simultaneously
Scalable complexity	In order for complex organisms to emerge, the ontogenetic developmental processes must be encoded in the genome
Evolution as a fluid process	Agents should be modeled with a large number of cells: evolution should make only small modifications (at the genome)
Minimal designer bias	Design as little as possible and let evolution do as much work as possible

# Design principles for collective intelligence

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<b>Principles for collective systems</b>	<b>Description</b>
Level of abstraction	Proper level of abstraction must be chosen, and the implications (of abstraction) important
Design for emergence	Find local rules of interaction that lead to desired global behavioral patterns (holds also for individual agents)
From agent to group	Agent design principles often applicable to collective systems (e.g. parallel, loosely coupled processes)
Homogeneity-heterogeneity tradeoff	Find compromise between systems using only one type of module/robot and systems employing several specialized types

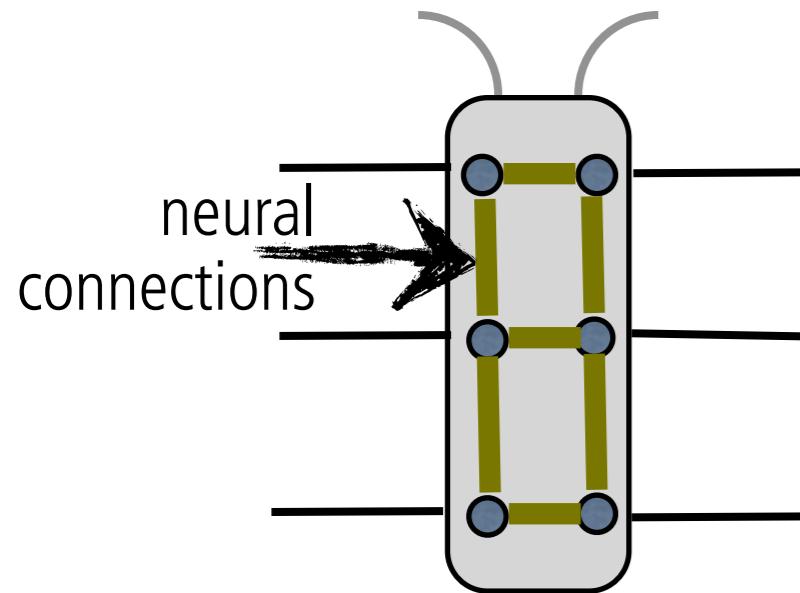
# Parallel, loosely coupled processes

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Holk Cruse, German biologist

- no central control for leg coordination
- only communication between neighboring legs
- global communication/coupling: through interaction with environment



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# Parallel, loosely coupled processes

## coupling through pheromone trails in environment



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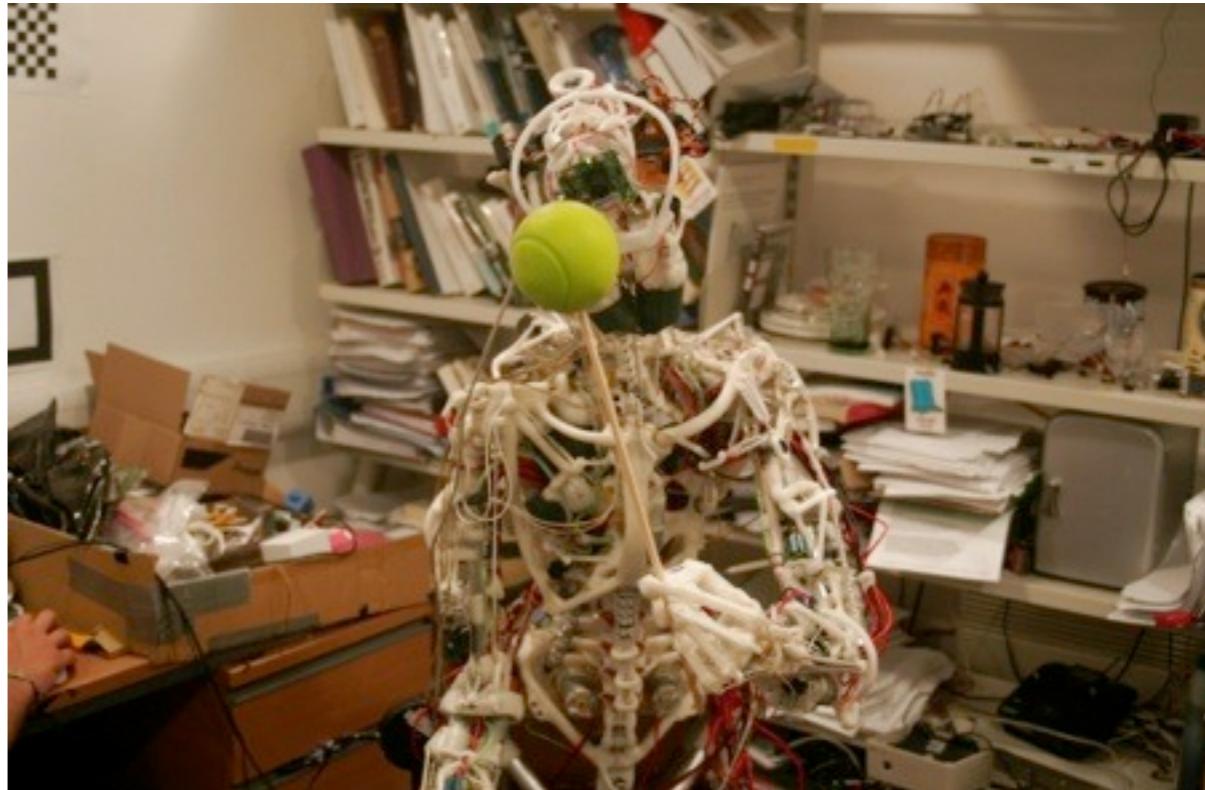
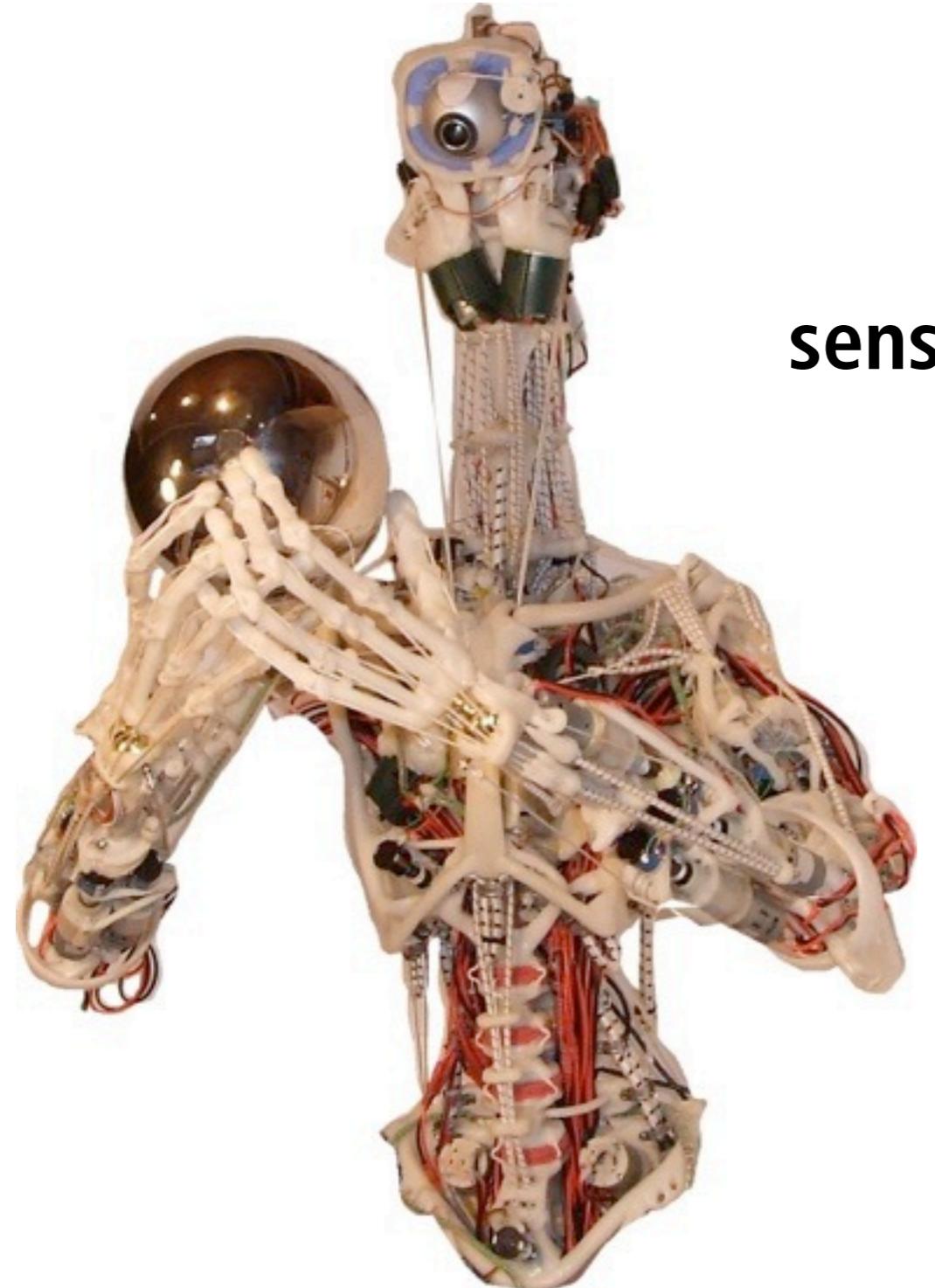
Pheromone trails enable ants to search for food efficiently: Two ants leave the nest at the same time (top), each taking a different path and marking it with pheromone. The ant that took the shorter path returns first (bottom). Because this trail is now marked with twice as much pheromone, it will attract other ants more than the longer route will.

WARNING: holds not only for ants, but also for humans!!!

# Sensory-motor coordination Information self-structuring

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induction of patterns of sensory stimulation containing information structure through sensory-motor coordinated interaction (e.g. grasping, foveating)



# The “story”: physical dynamics and information processing

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- cross-modal association, learning, concept formation

**brief explanation of “mutual information”  
what is it? example? importance?**

- categorization —> anyone



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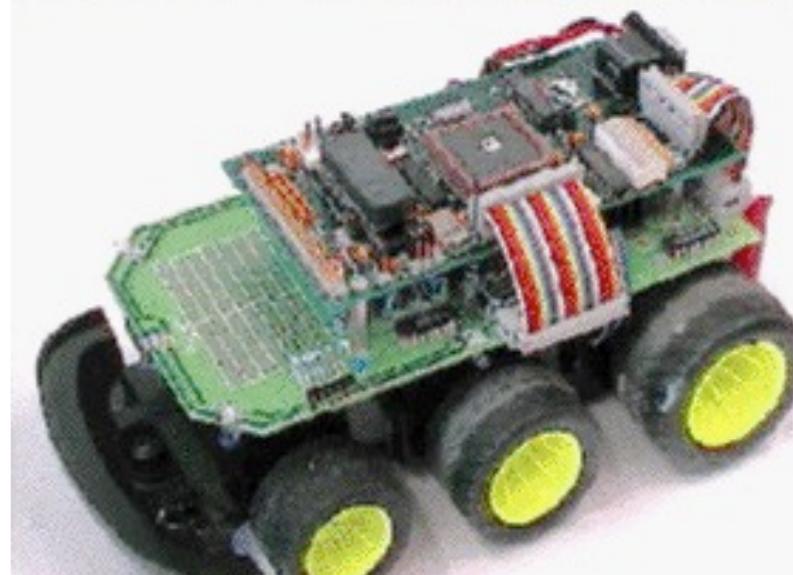
# Cheap design exploitation of niche



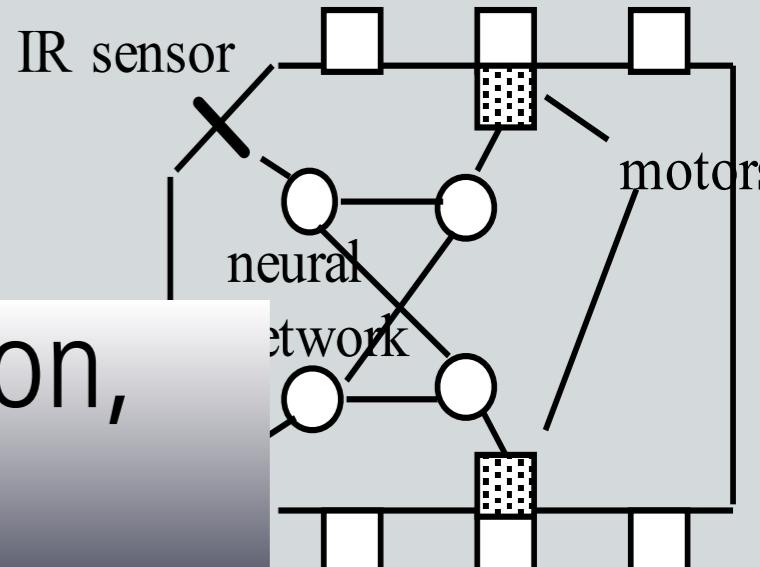
6x6m arena with Styrofoam cubes

exploiting: borders, shape, weight, friction,  
morphology, weight distribution

Passive Dynamic  
Walker



Didabot  
simple robot  
for didactical  
purposes



These robots are cleaning up, but that's not what they think they are doing (joke — can they think?)

# Cheap design simple mechanism

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behavioral rule:

**sensory stimulation on left: turn right**

**sensory stimulation on right: turn left**

(obstacle avoidance)



exploiting: borders, shape, weight,  
friction, morphology



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# Redundancy

## Different physical processes



airplane landing



jet engines  
(reverse) propulsion

braking systems  
partial overlap of functionality

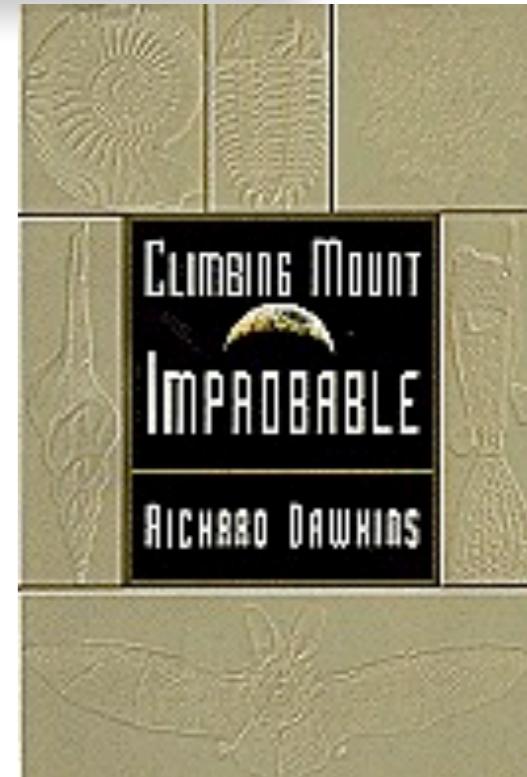
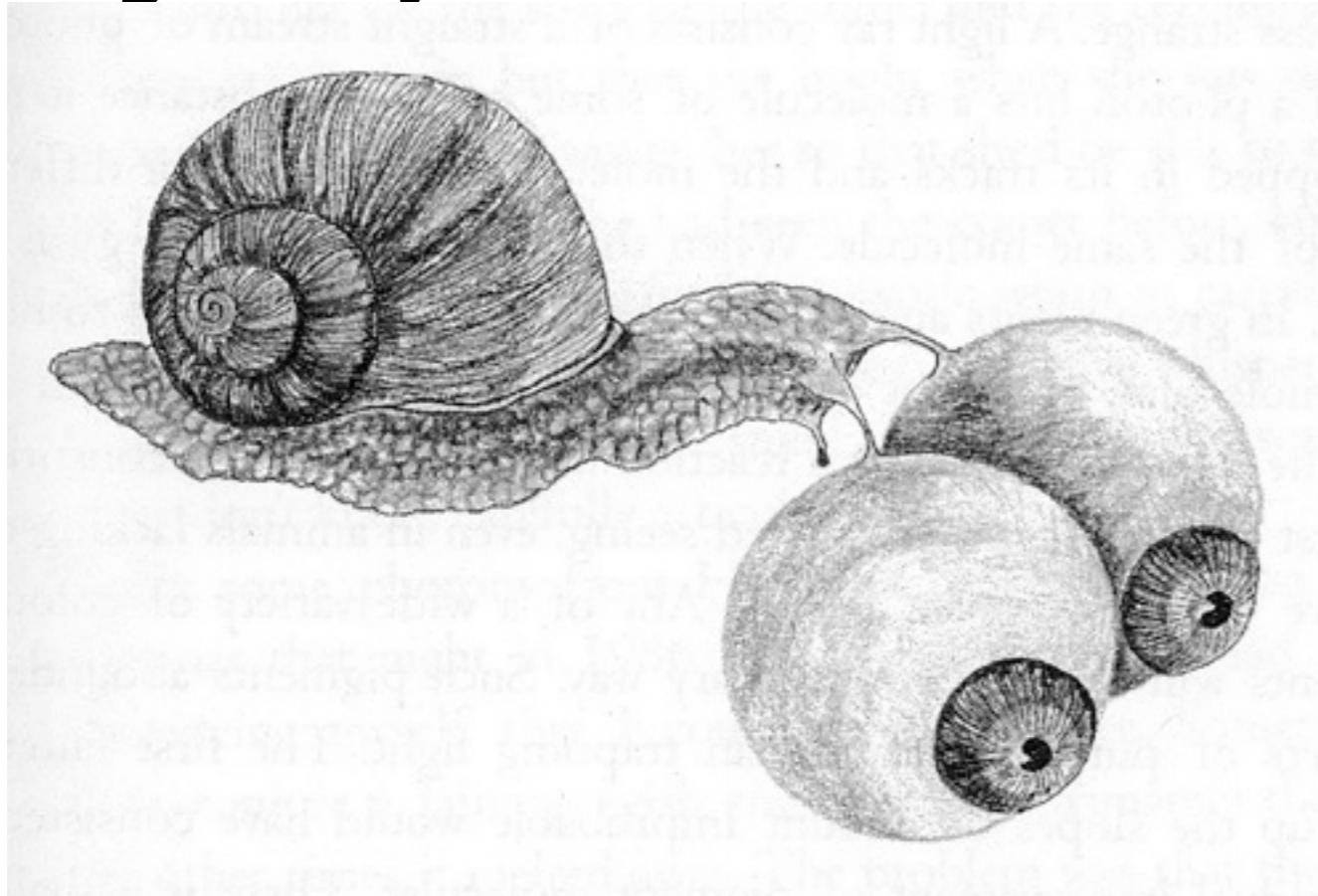


parachutes  
mechanical

wheels (friction)  
runway conditions

# Ecological balance matching complexity

**ecologically unbalanced system:  
Richard Dawkins's snail with  
giant eyes**



Author of:  
“The selfish gene” and  
“The blind watchmaker”



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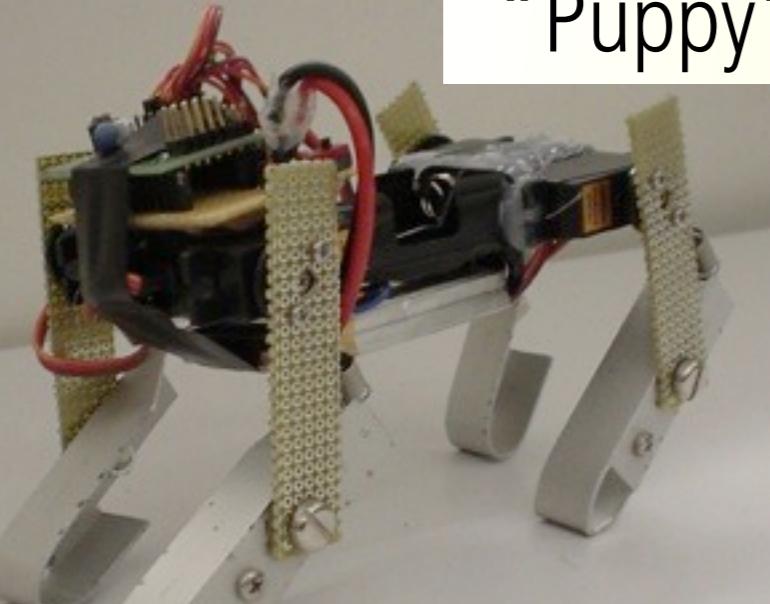
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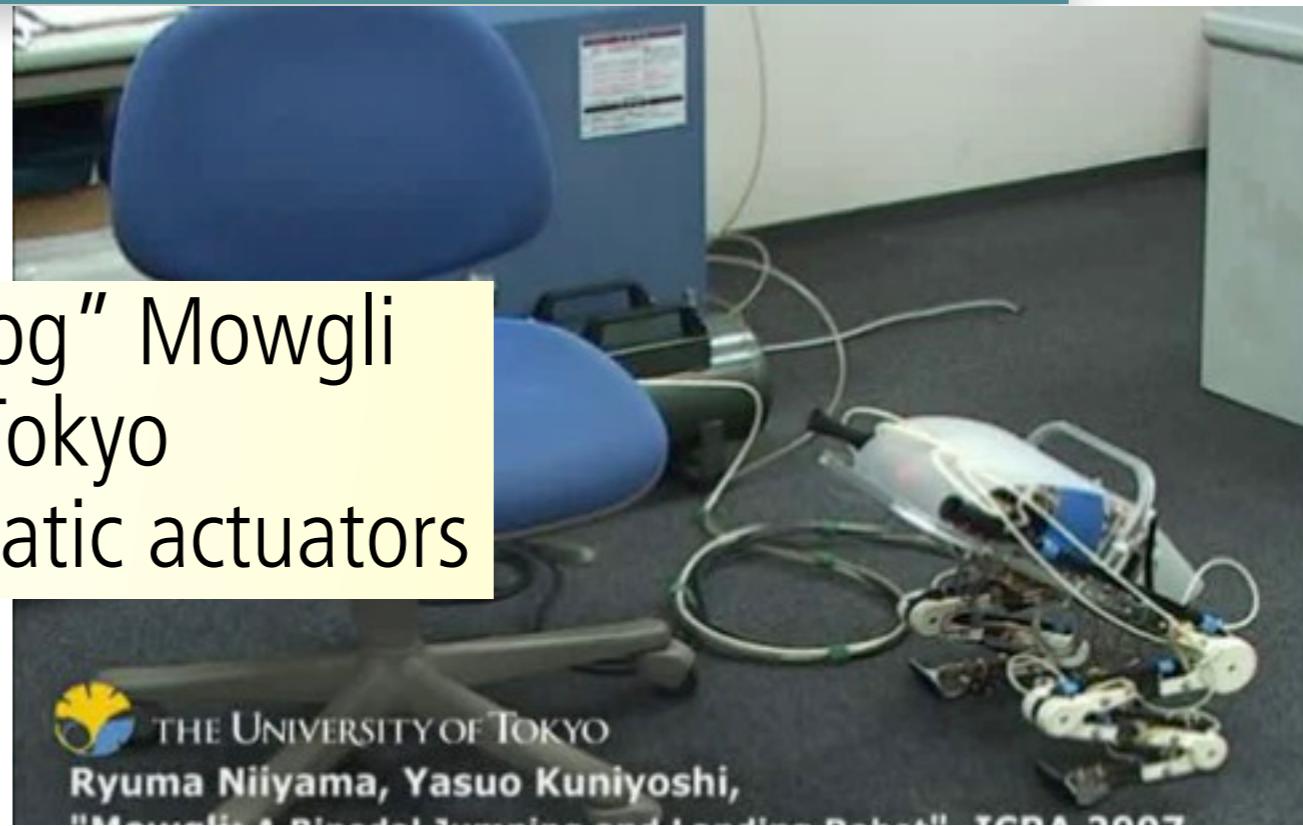


What are such huge eyes good for if the snail cannot move quickly? It's only useless additional weight.

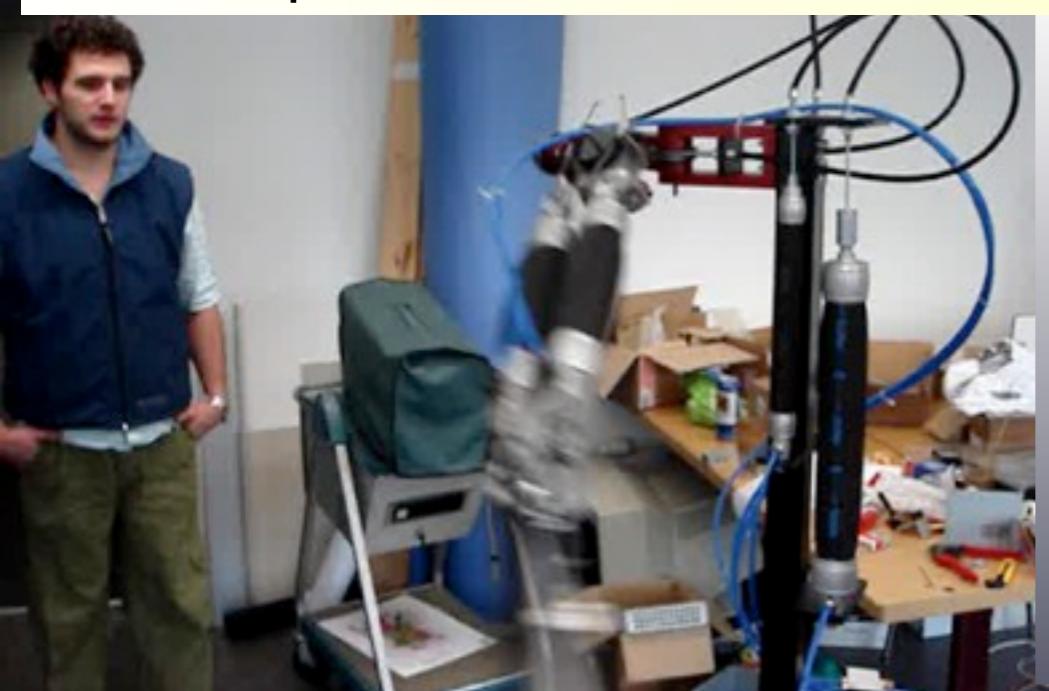
# Ecological balance: task distribution morphology, materials, control



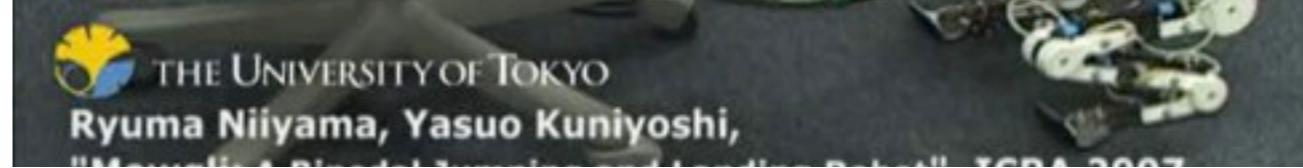
"Puppy" with springs



loosely swinging arm  
with pneumatic actuators



control partially  
“outsourced” to properties  
of springs, pneumatic  
actuators



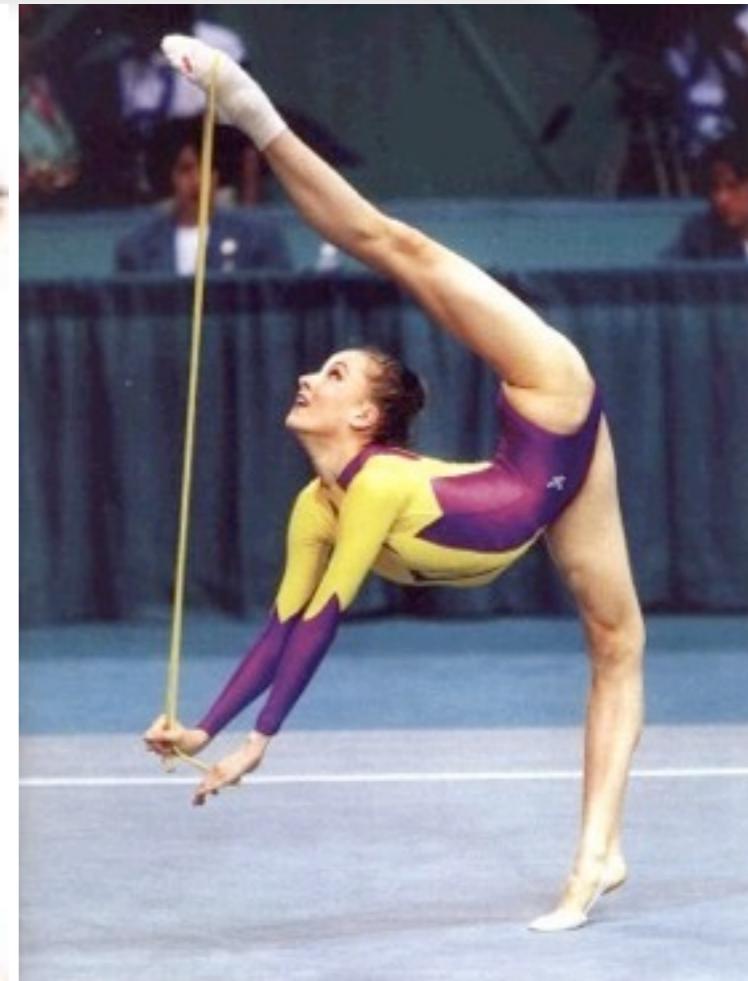
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# Bernstein's problem: Development as an incremental process

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learning to control high  
DOF body: freezing and  
freeing DOFs



# Discover, learn from real world The best exploration strategies

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only possible through embodiment

exploration: NOT random! —> preferred trajectories  
from biomechanical constraints

“loosely swinging arm”



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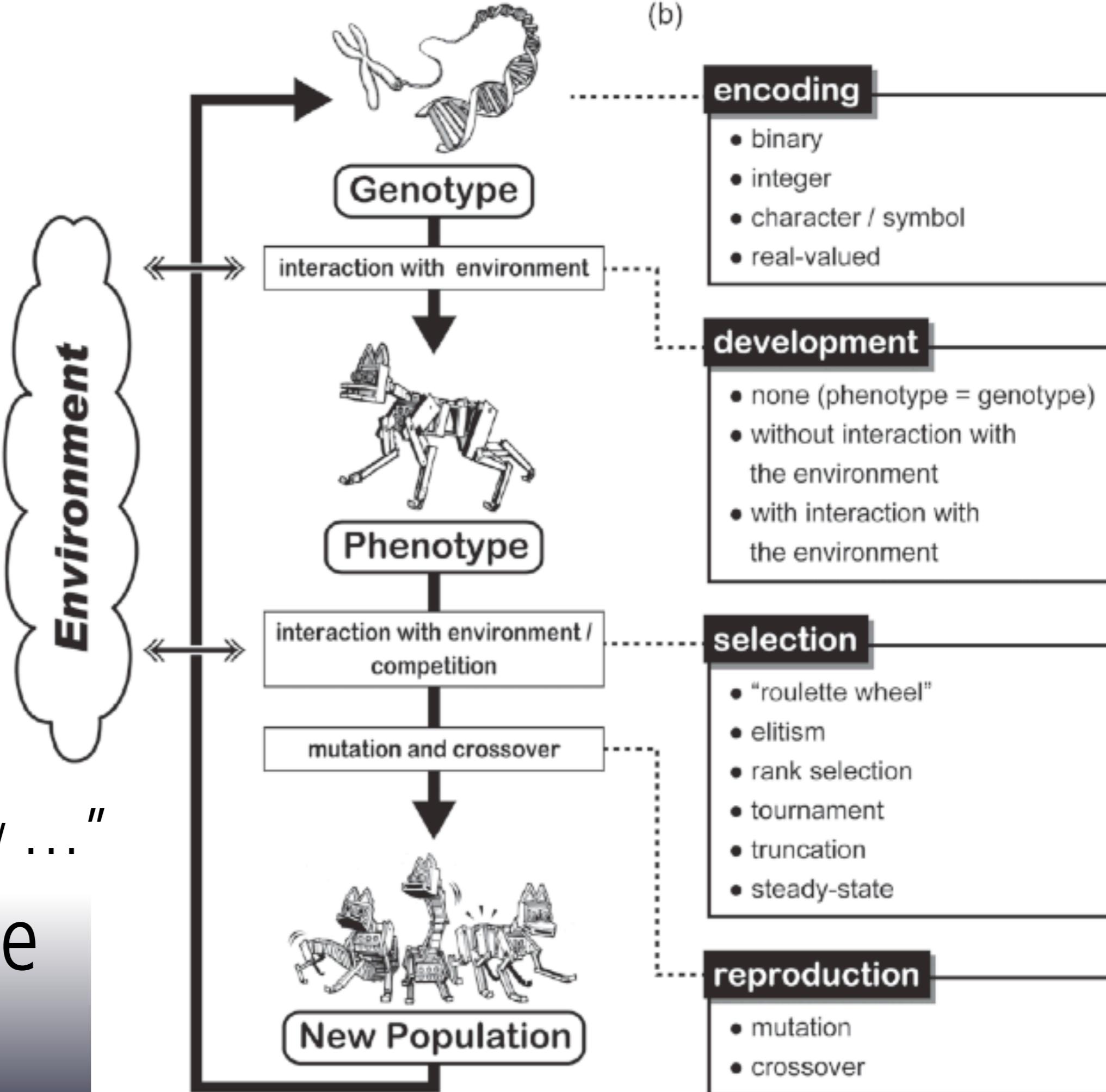
# Design principles for evolution

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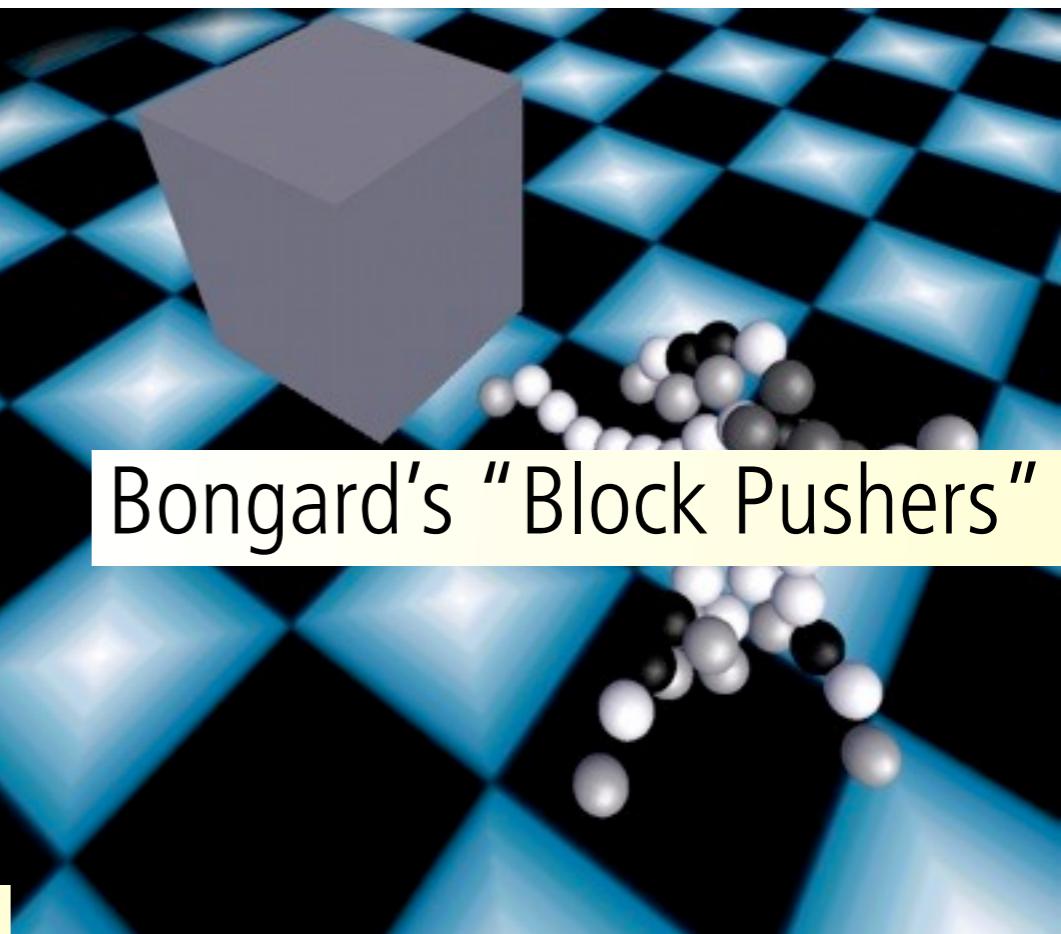
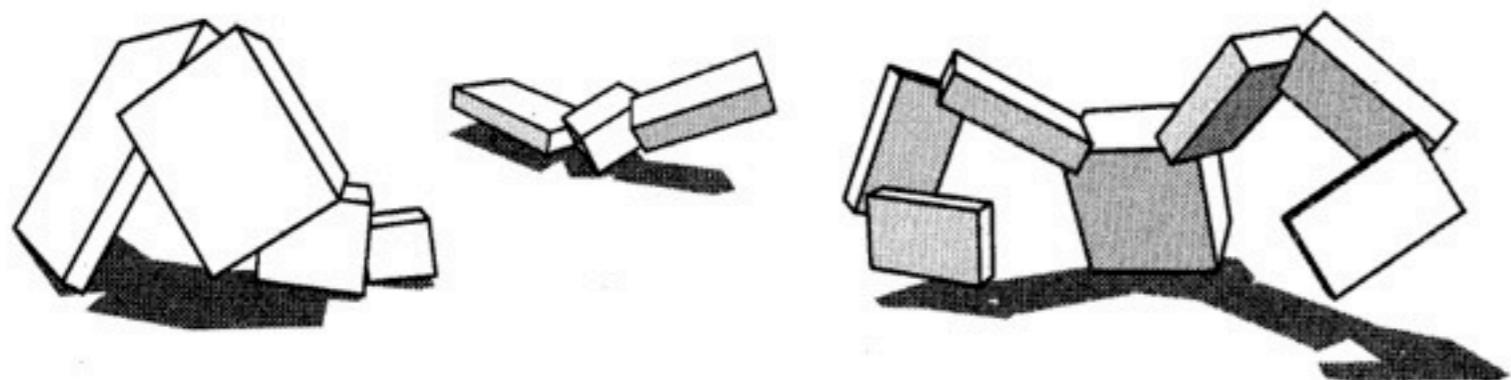
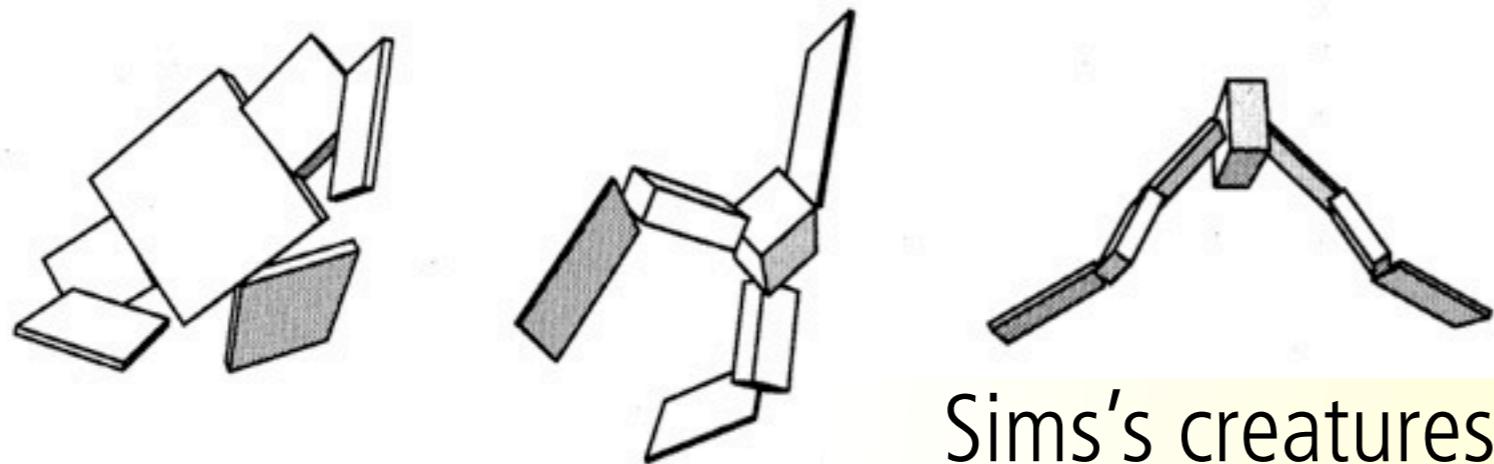
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# Basic cycle for artificial evolution

from  
“How the body . . .”  
cumulative selection



# Brain-body co-evolution



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Lipson's creatures  
"Golem" project



# Encoding of developmental processes in genome

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rather than the structure of the organism



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# Encoding developmental processes

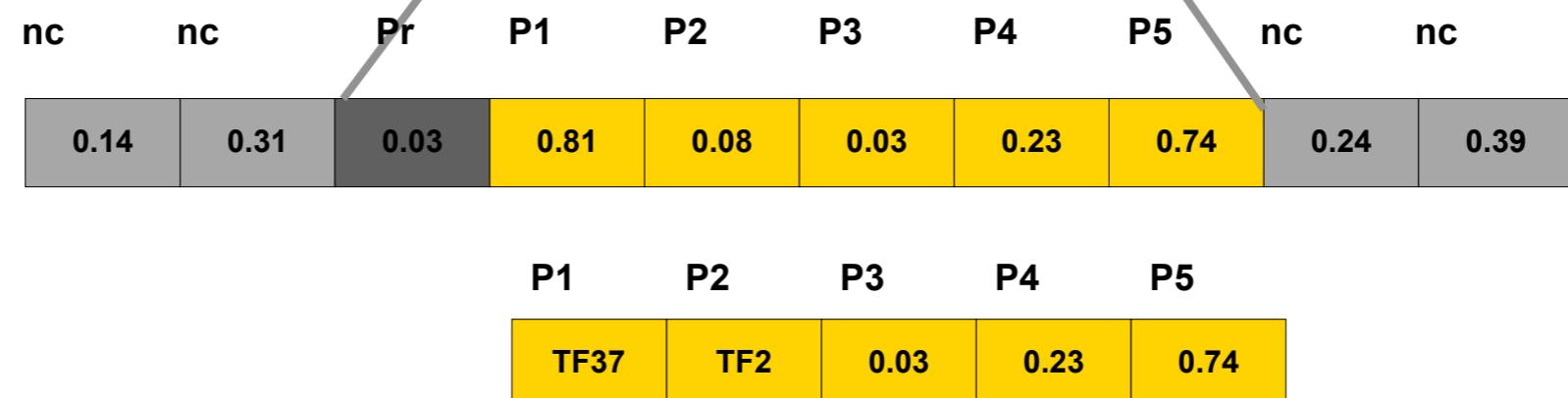
## Representation of “gene”

nc: “non-coding region”



G1, G2, ....:  
“genes” on “genome”

TF: “transcription factor”



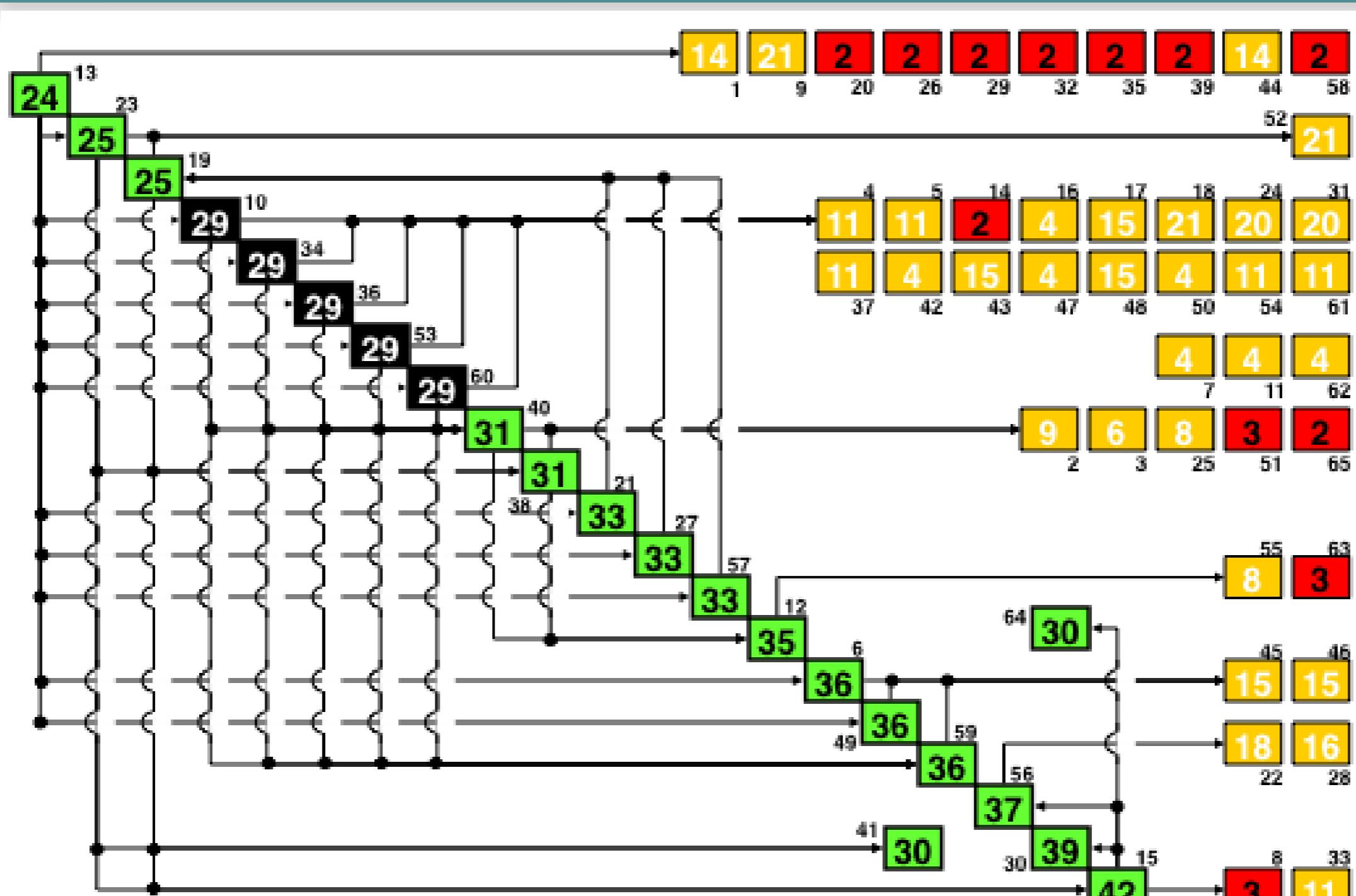
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# GRN for evolved creature (Josh Bongard)



# Design principles for collective intelligence

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# Design for emergence: cleaning an arena



entire process:  $\sim 20\text{min}$   
frames: 2-3min



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# Seeing things differently

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# Seeing things differently

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- “Cleaning up”, the “Swiss robots”
- Walking without control
- Rapid locomotion with slow electronics
- Coordination through interaction with world
- Social interaction as “reflexes”
- Optimization without cognition
- Creative computers



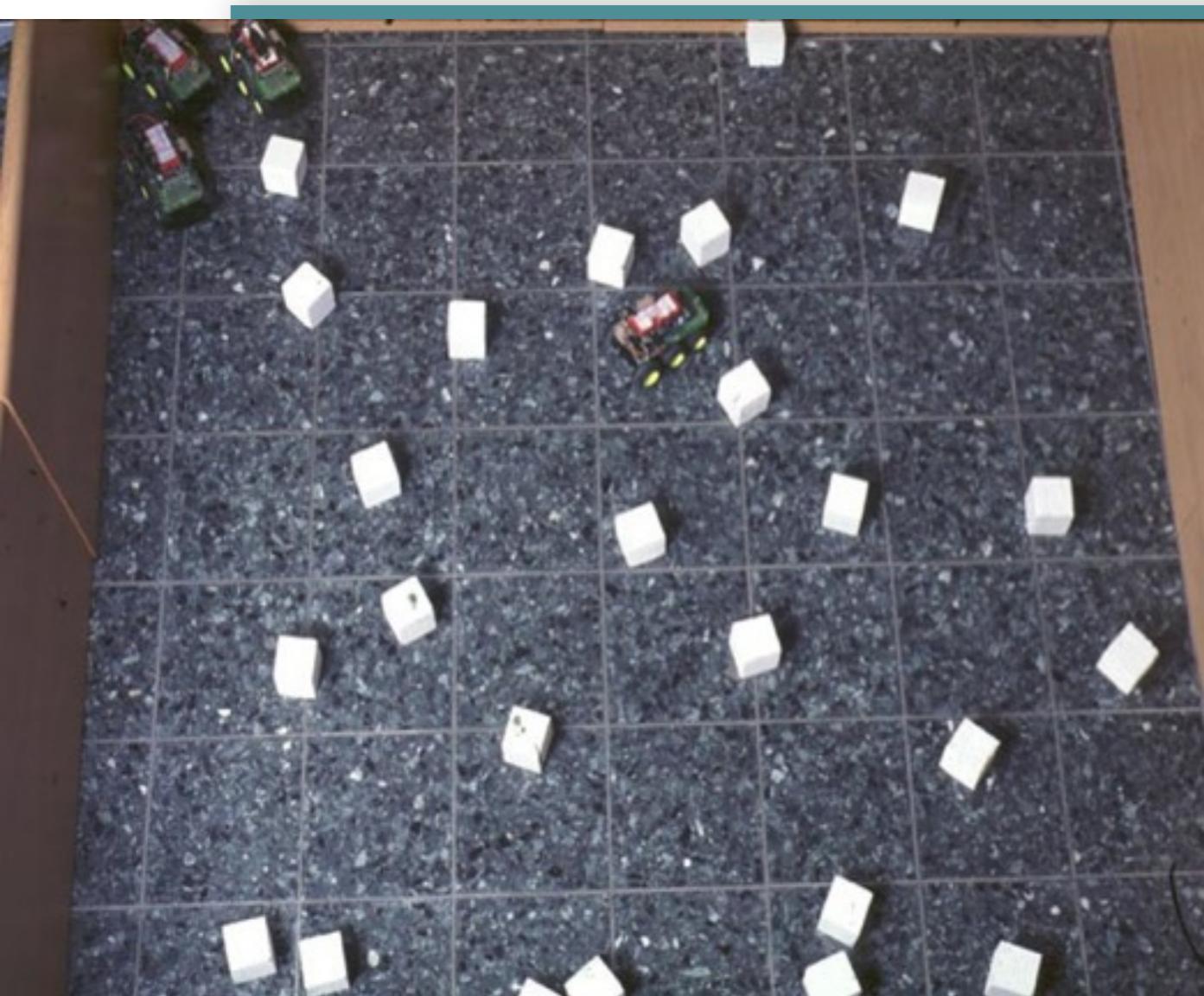
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# Activities involved in clustering: standard solution



6x6m arena with Styrofoam cubes

- Look for cube, if possible, the nearest one).
- Pick up cube (somehow).
- Look for nearest cluster.
- Go to cluster.
- Deposit cube.
- Look for new cube, etc.



sophisticated perceptual skills required



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Look for cube, if possible, the nearest one). Pick up cube (somehow). Look for nearest cluster. Go to cluster. Deposit cube. Look for new cube, etc.

Requires quite sophisticated perceptual skills on the part of the robots (recognizing a cube from different angles and distances, recognizing a cluster). Some elementary motor skills for grasping and dropping. locomotion skills.

# The solution: simple rules

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behavioral rule:

**sensory stimulation on left: turn right**

**sensory stimulation on right: turn left**

(obstacle avoidance)



**situated perspective (from the agent's point of view)**



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# Walking without control?

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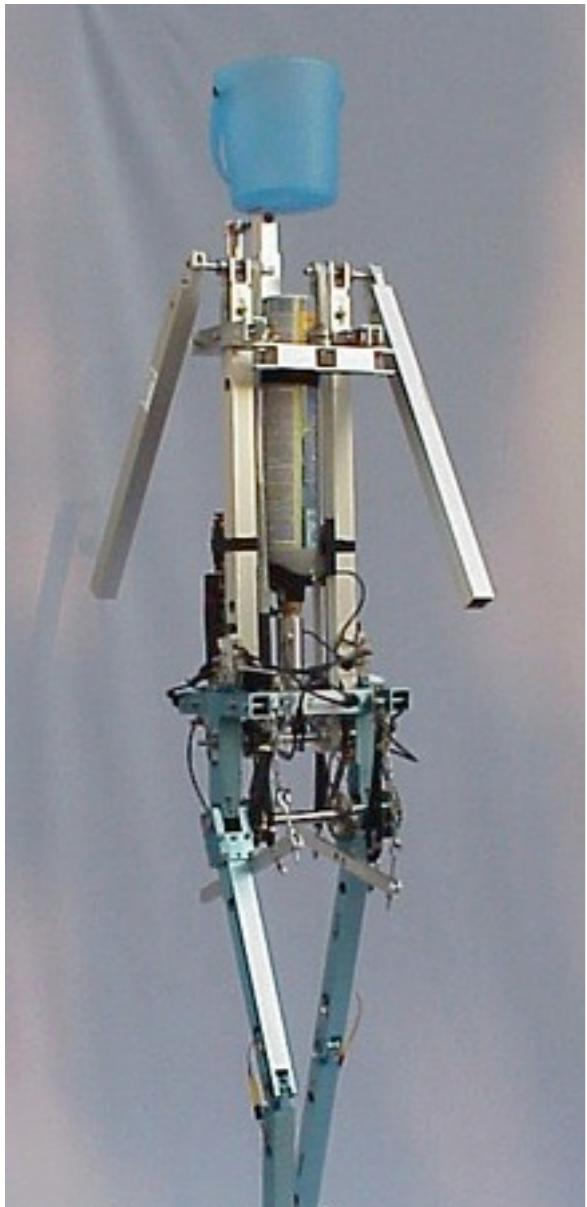


# The solution: The “Passive Dynamic Walker”

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**“Denise”**  
no control for balance



# Rapid locomotion with slow electronics/brains?

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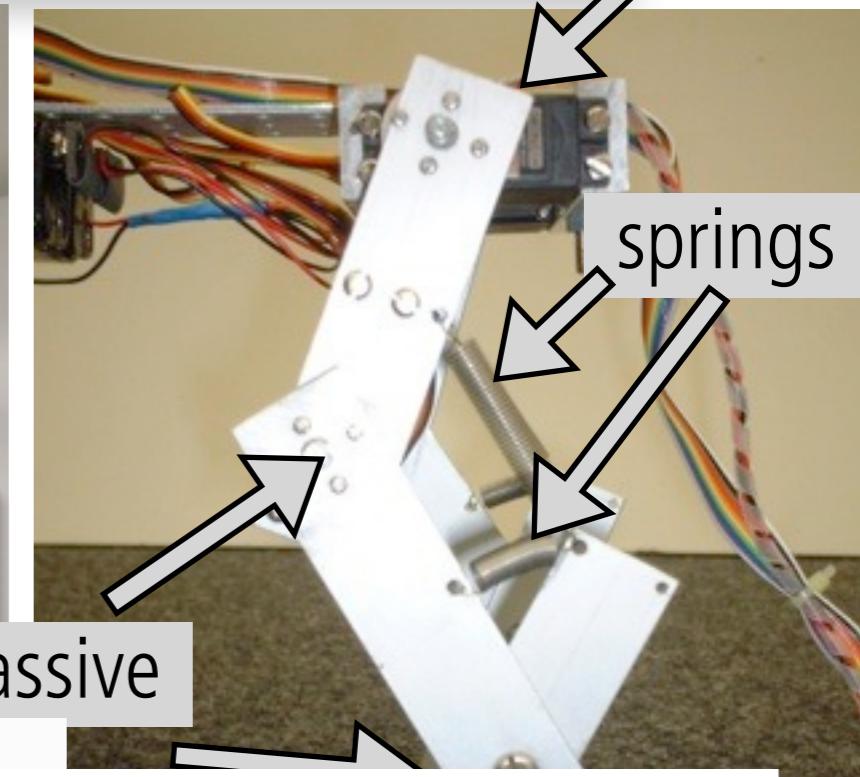
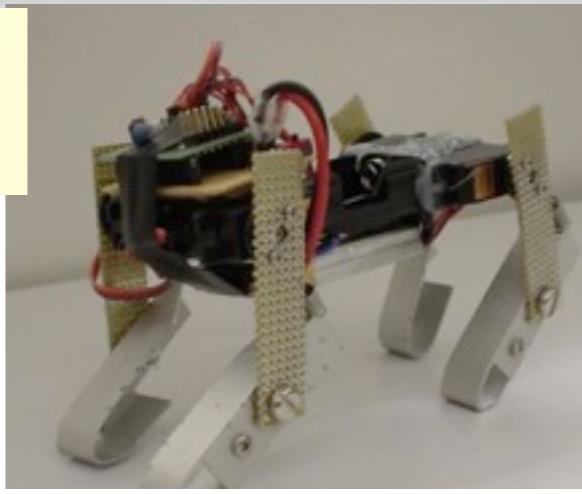
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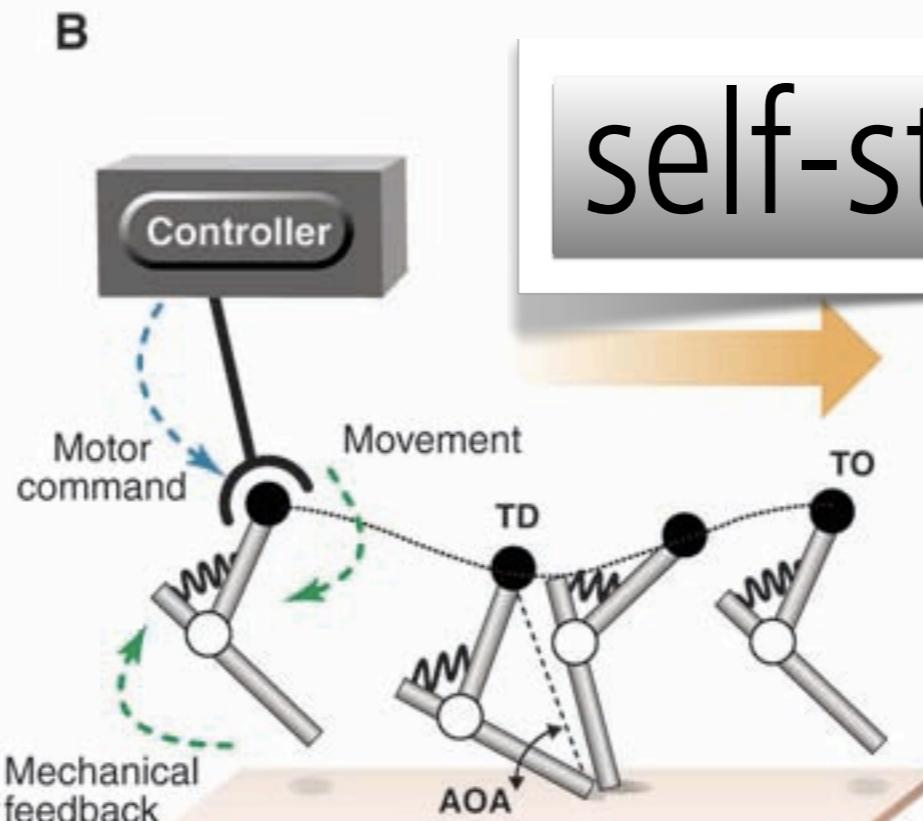
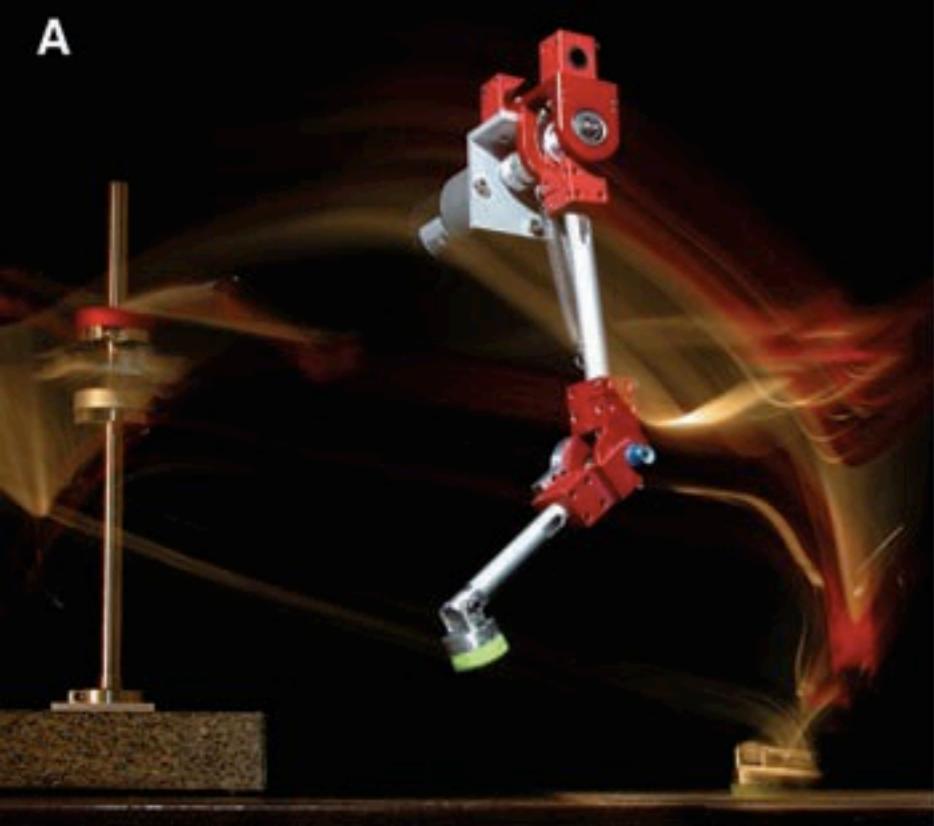
# The solution: “Exploitation of morphology and materials”

actuated:  
oscillation

“Puppy”



morphological computation



self-stabilization

Jena  
monoped

# Leg coordination without central control?

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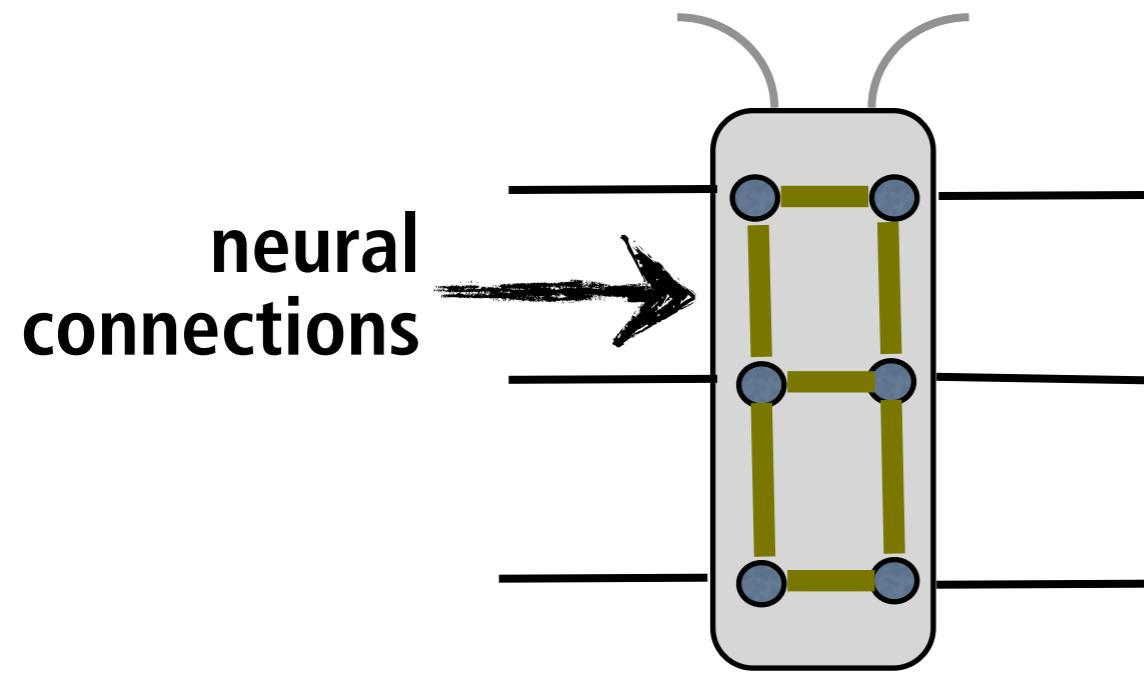
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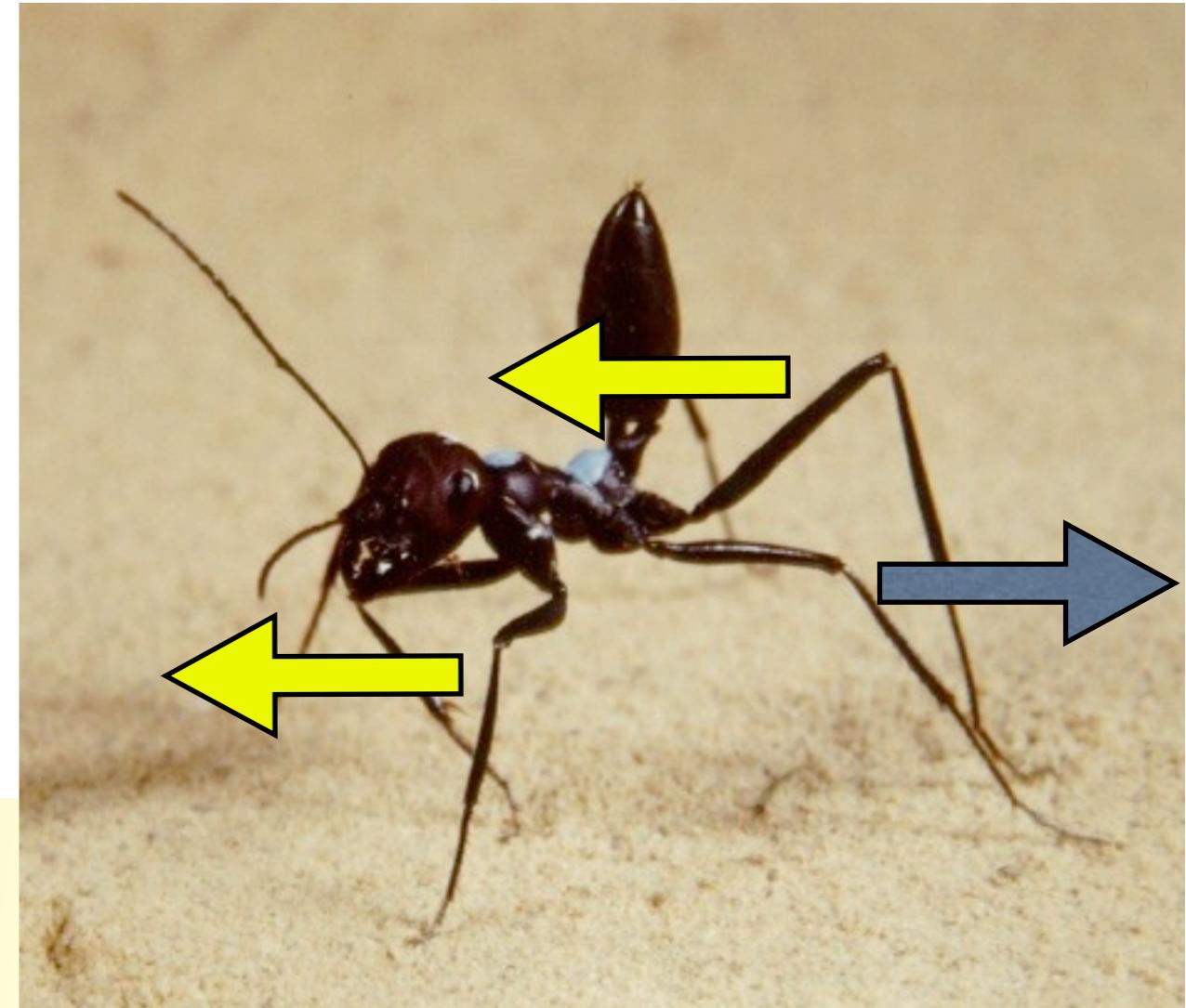


# The solution: global communication through interaction with real world

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"parallel, loosely coupled processes"



# Social competence without high-level cognition?

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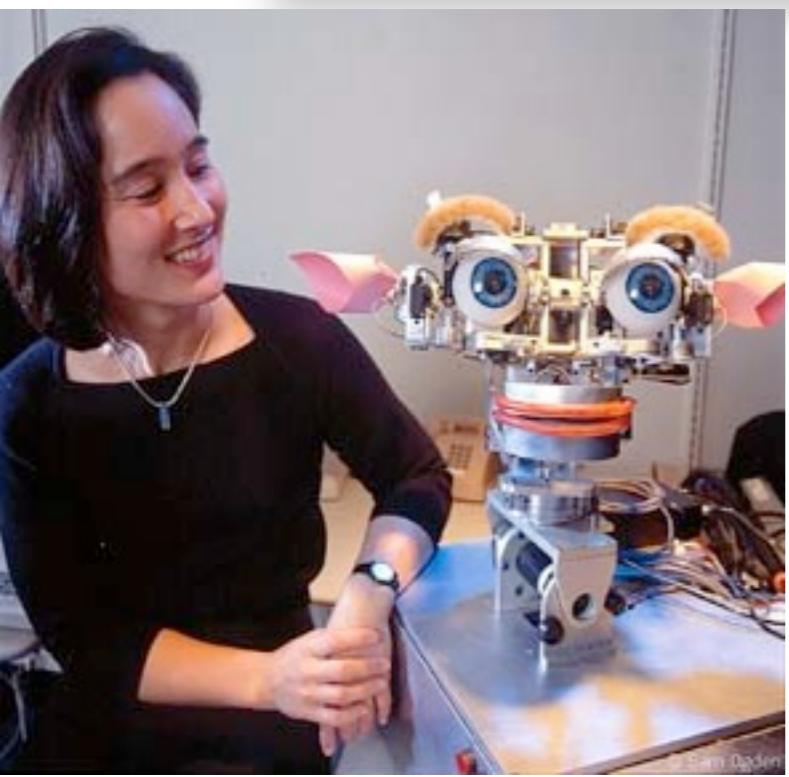
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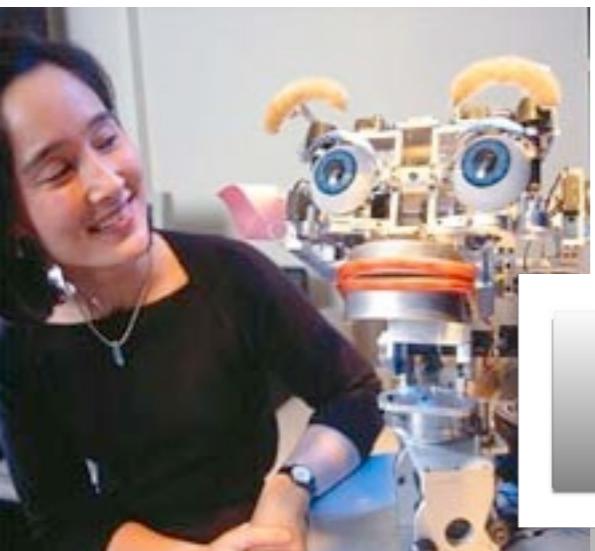
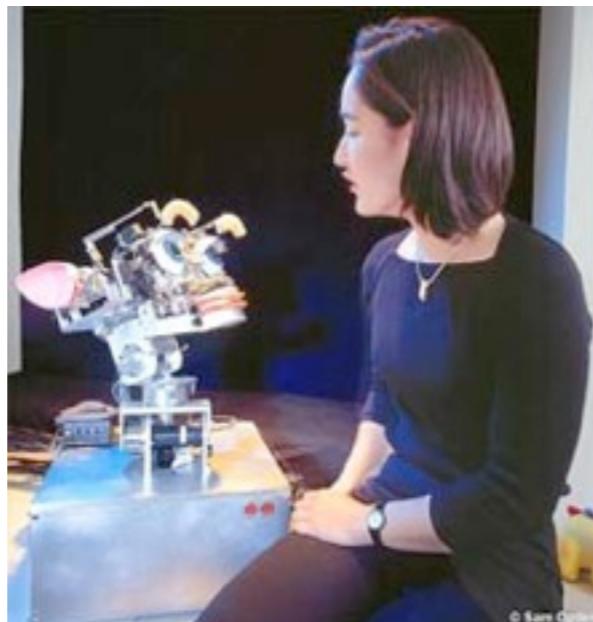
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# Social competence without high-level cognition?



Cynthia Breazeal, MIT  
Media Lab

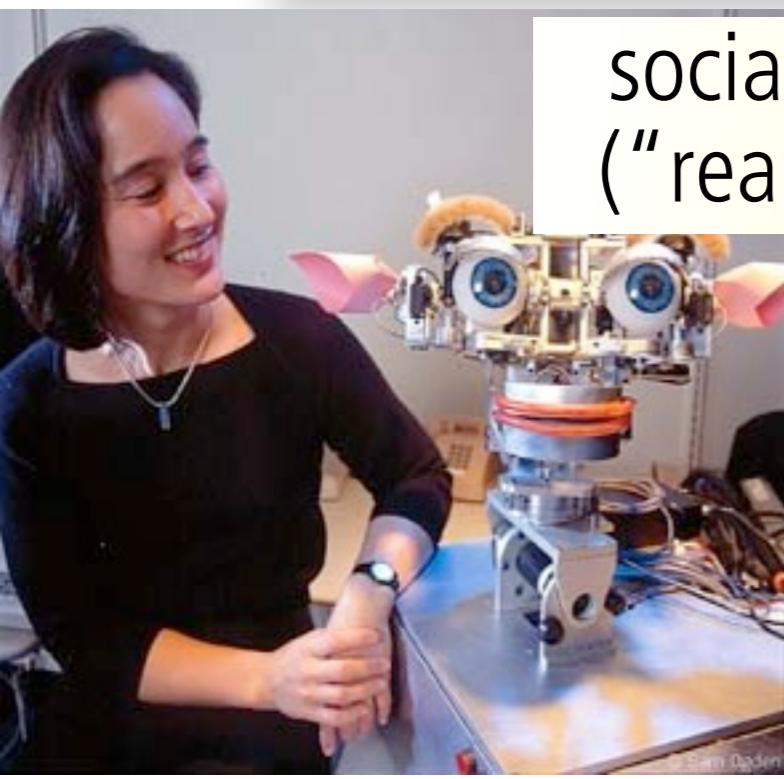


reflexes:

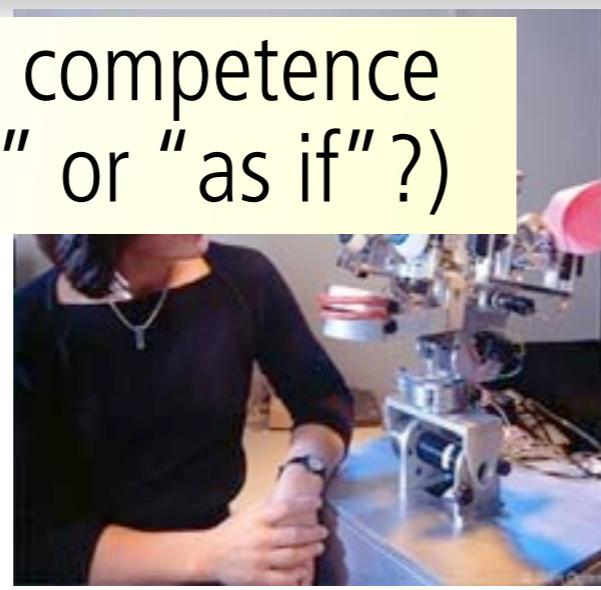
- turn towards sound**
- turn towards moving objects**
- visual tracking of slow objects**
- habituation**

Video "Kismet.mov"

# Social competence without high-level cognition?



social competence  
("real" or "as if"?)

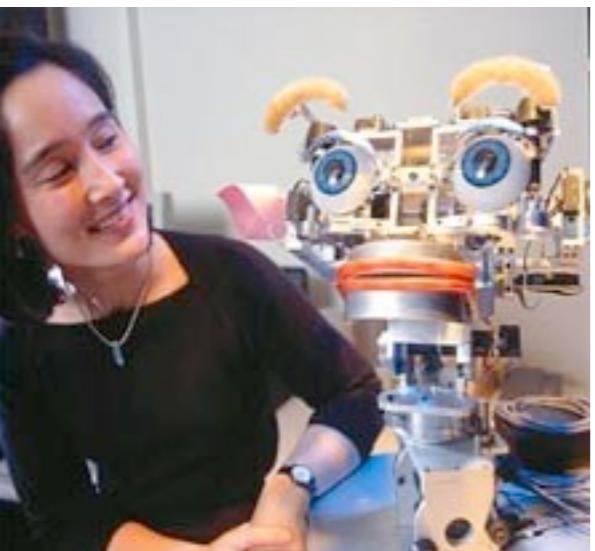


reflexes:  
**turn towards sound**  
**turn towards moving objects**  
**visual tracking of slow objects**

Cynthia Breazeal, MIT  
Media Lab



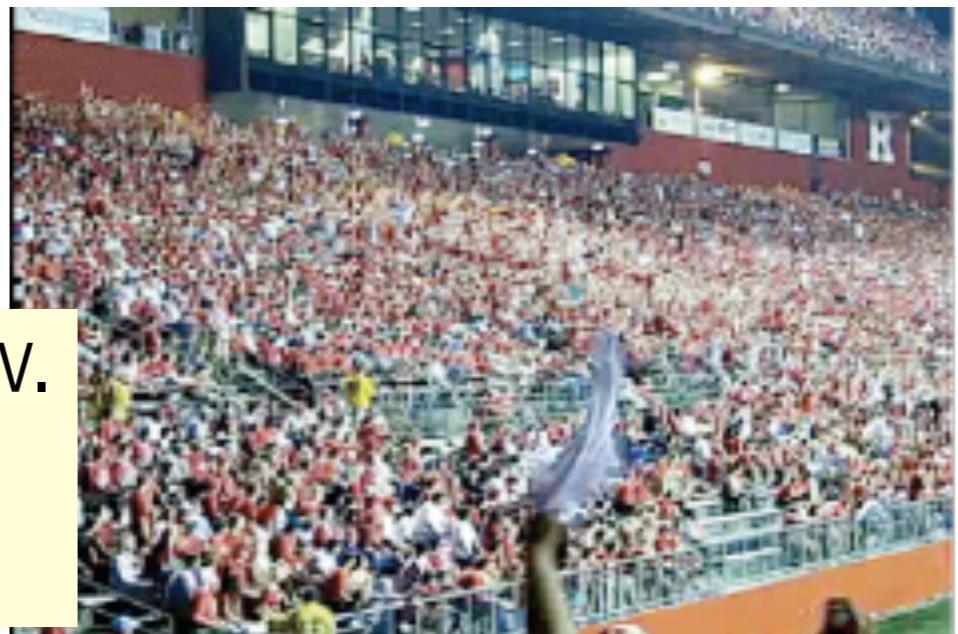
Rodney Brooks, MIT:  
"What do you mean — 'as if'? This **IS** social competence"



# Social behavior through simple reflexes?



John Bargh, Psychologist, Yale Univ.  
“The unbearable automaticity of being”, 1999



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# Finding shortest paths without measuring, storing, comparing?

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# Creative computers, robots?

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- “Computers can never be creative — they can only reproduce what you program into them”
- From the movie “I, Robot”. Human to robot (my transcription): “Robots could never be creative, they could never write a symphony like Beethoven!” Robot to human: “Could you?”



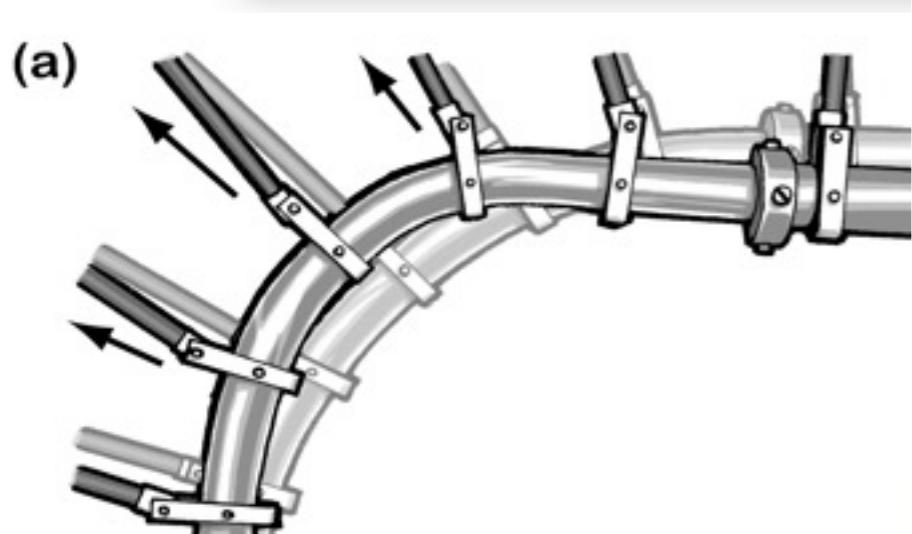
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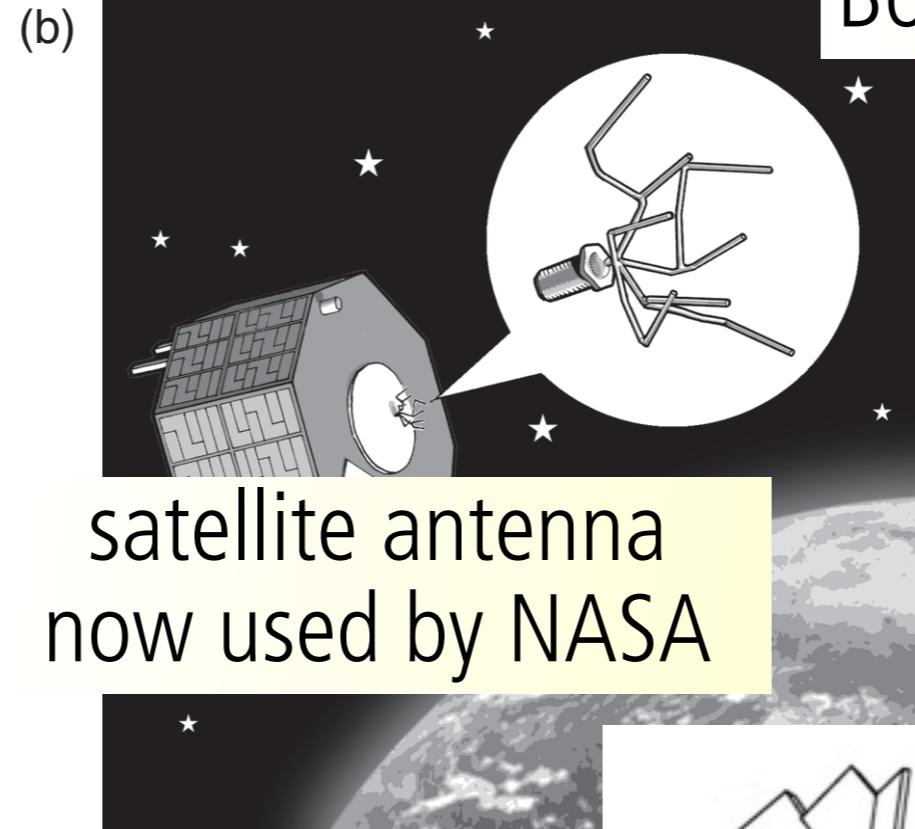
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# Solution: Artificial evolution and morphogenesis



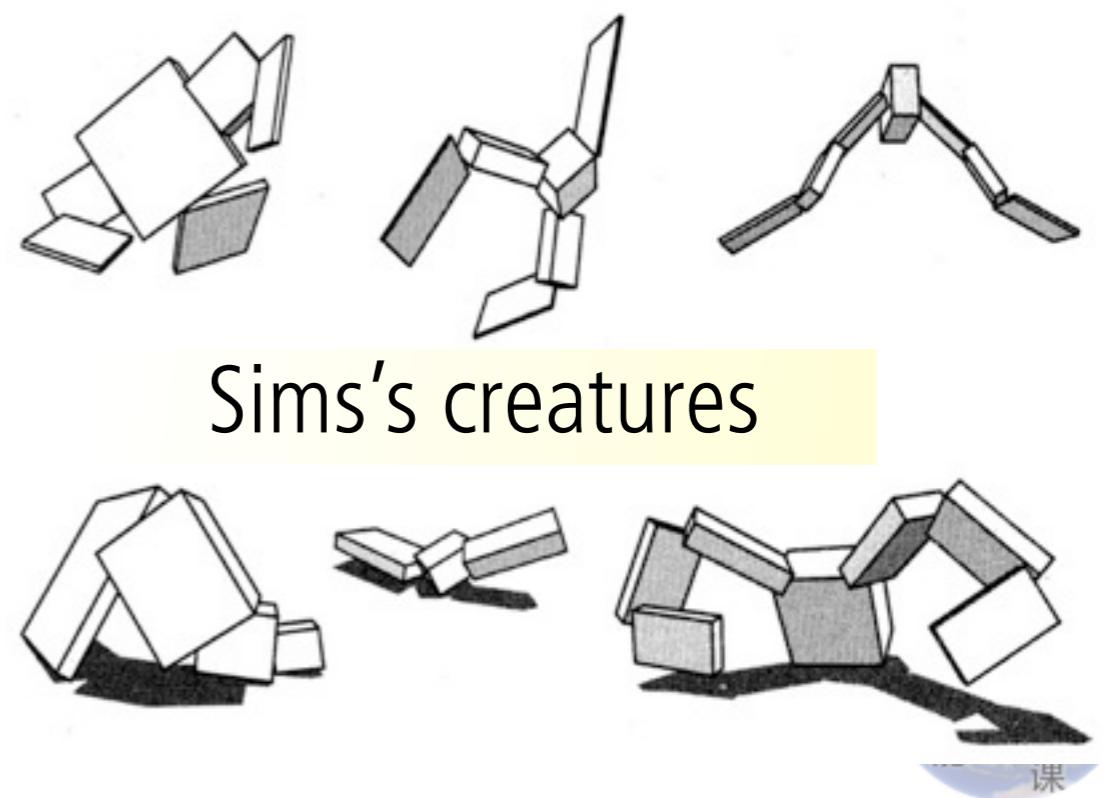
Rechenberg's  
'hunched' fuel pipe  
solution



Bongard's "Block Pushers"



Lipson's creatures  
"Golem" project



Sims's creatures

# Morphogenesis (Bongard and Bisig, 2001)

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Video "growth.mov"



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# Problem solved?

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# Epilogue

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David Payne *Confessions of a Taoist on Wall Street*  
Sun I: little boy growing up in monastery  
Wu: chef (cook) of monastery and Sun I's mentor



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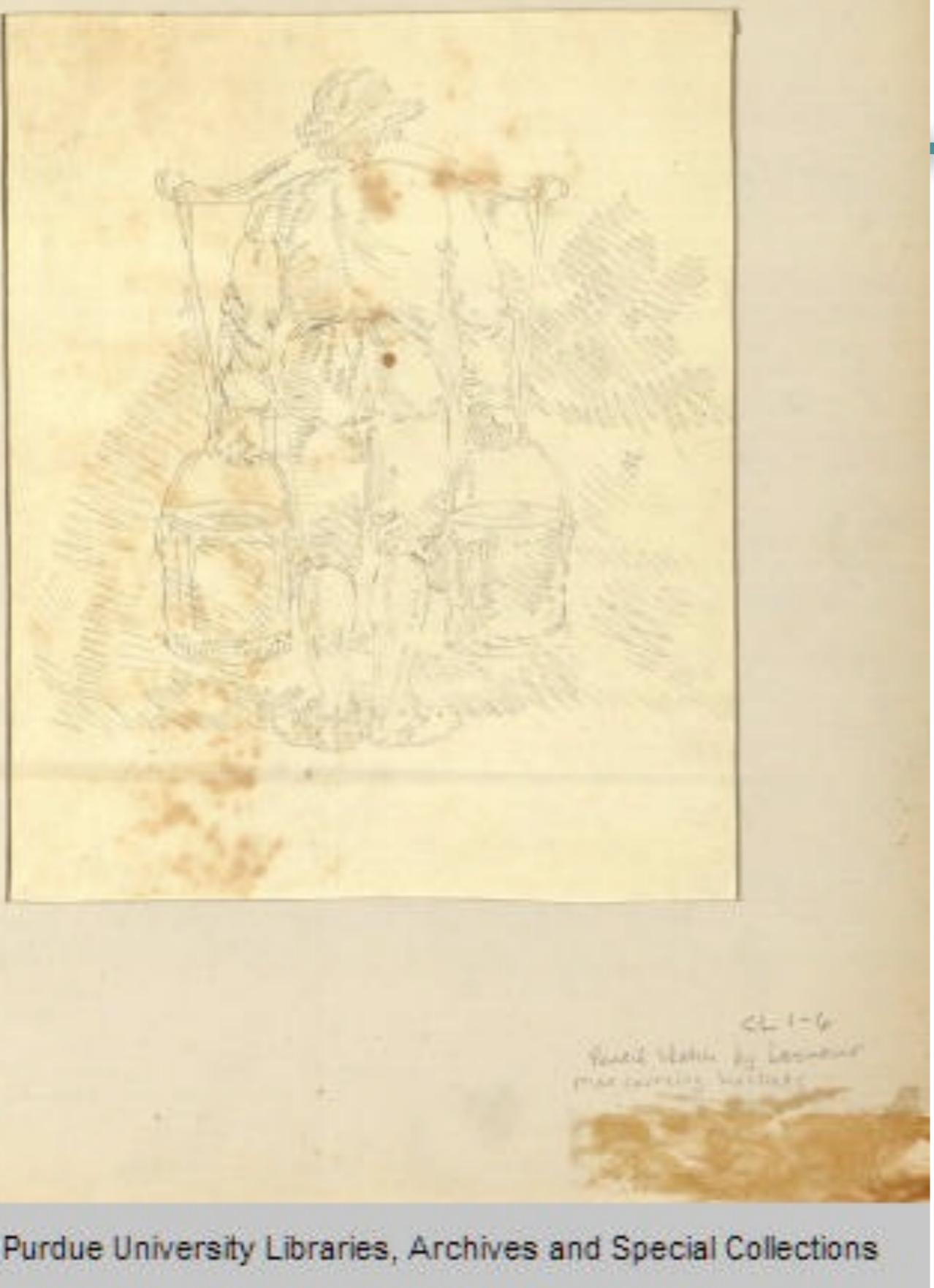
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The little boy, Sun I, is born of a Chinese mother and an American fighter pilot in China. His mother dies at birth and his father returns home to the United States; he is left alone and grows up in a monastery. His mentor and teacher is the chef, Wu, who takes good care of him. The monastery is on a high rock upon a river. One of their daily chores is to carry water from the river to the monastery up a rocky path. The boy remembers that whenever they arrived at the top of the rock his buckets were empty, all the water spilled, whereas Wu's were always full.

# Yoke with water buckets



Courtesy Purdue University Libraries, Archives and Special Collections



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# Conversation between Sun I and Wu

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It was true. By some extraordinary luck or skill Wu never seemed to lose a drop, though he hurried along the treacherous stair at twice my pace. (I tried to cut my losses by moving slowly, plotting my course in advance and picking each footrest with deliberate care.)

“I don’t understand it,” I confessed to him. “You must know some kind of trick. Explain your method.” ...

“You haven’t yet caught on. It’s precisely this—excess of method—that confounds you, leaves the buckets nearly empty ...”

“If you’re so smart, how do you do it then?”

“How do I do it? . . . I close my eyes and think of nothing. My mind is somewhere else. My legs find their way without me, even over the most uneven ground. How can I tell you how I do it? . . . I can’t even remember myself!” (Payne, 1984, pp. 18–19)

# Conversation between Sun I and Wu

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Cartoon: **Shun Iwasawa, Studio Ghibli, Tokyo, Japan**

over the most uneven ground. How can I tell you how I do it? . . . I  
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air at twice  
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it me, even

# That's it!

Thank you for your  
attention!

stay tuned for

Tamas Haidegger  
Aude Billard  
Jamie Paik



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# End of week 11 Q&A session

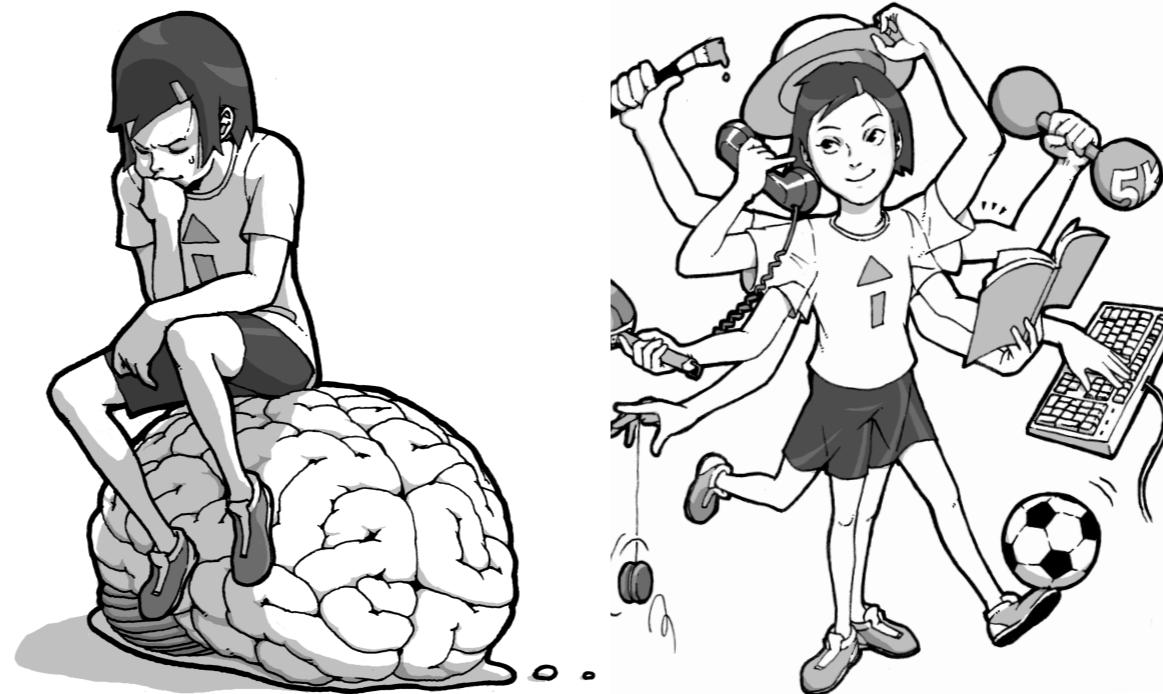
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**Thank you for your attention!**

**Now, enjoy: “Future Trends”  
(with various guest speakers)**



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# Selected highlights

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# Through engineering to science

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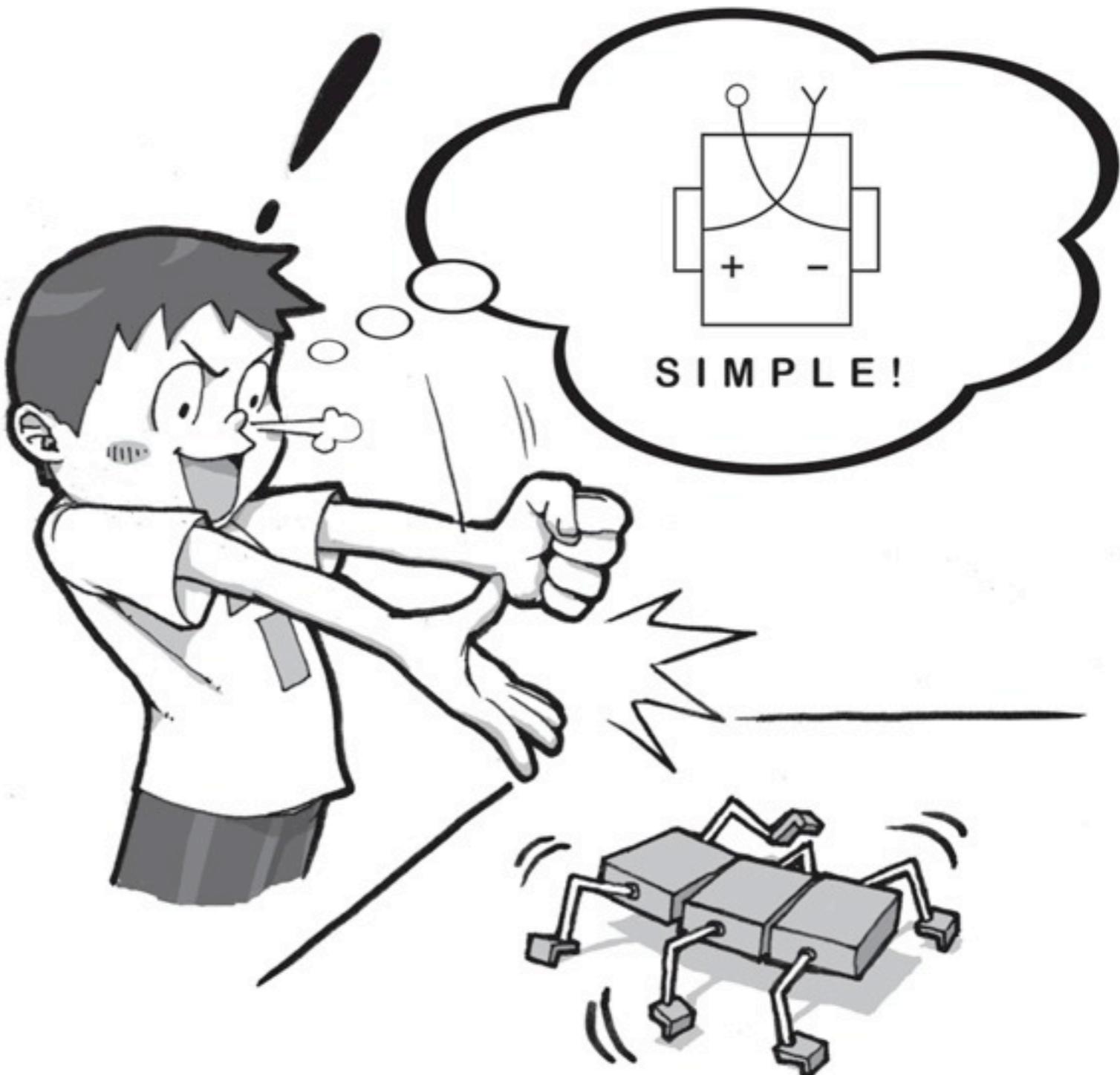
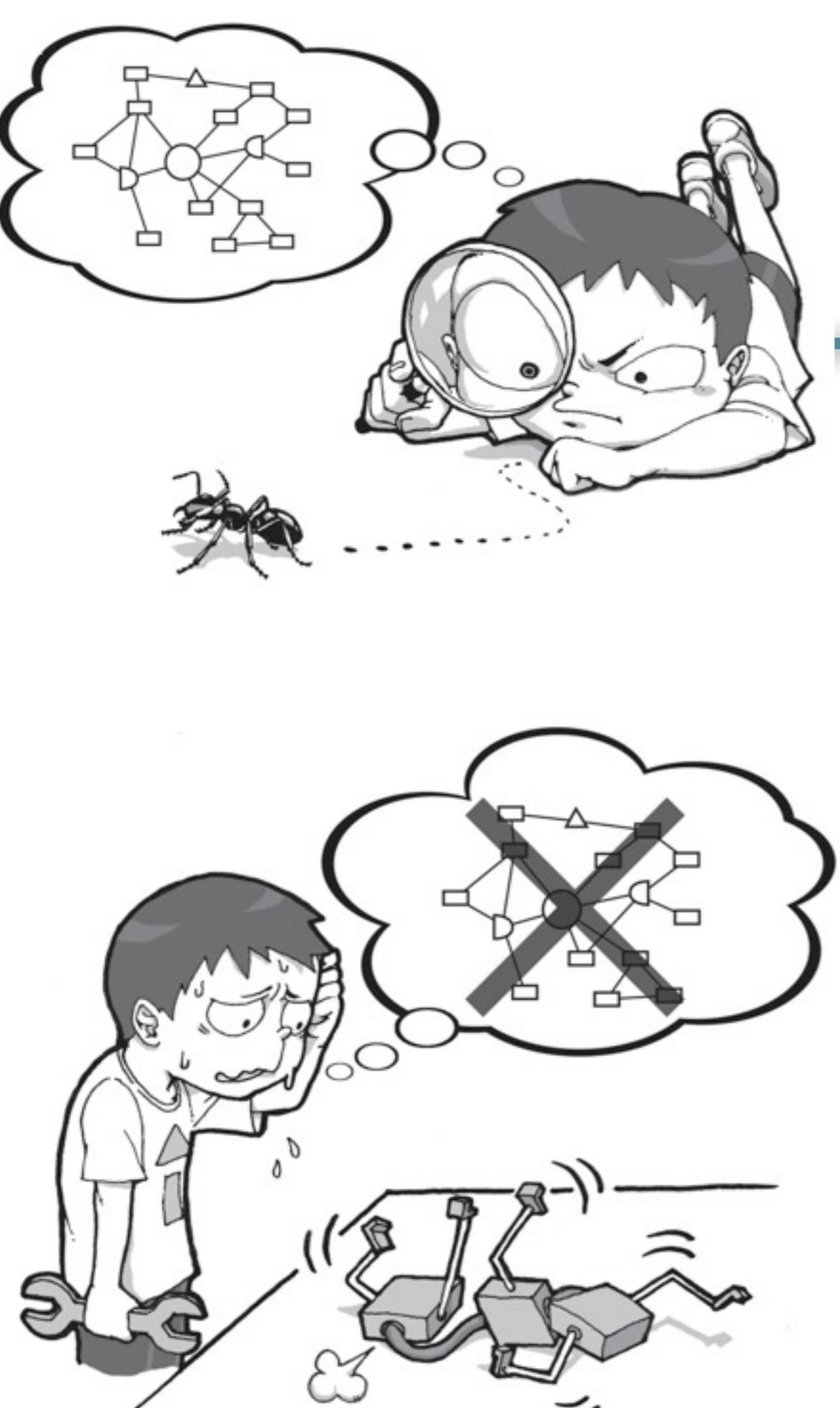
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# The synthetic methodology

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# Information self-structuring through sensory-motor coordination

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# The “story”: physical dynamics and information processing

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- morphology and materials
- exploration
- preferred trajectories from biomechanical constraints
- induction of patterns of sensory stimulation
- sensory-motor coordination → induction of information structure
- good “raw material” for brain



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# The “story”: physical dynamics and information processing

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- cross-modal association, learning, concept formation
- extraction of mutual information → prediction
- categorization (fundamental for cognition)



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# Exploiting morphological computation

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“outsourcing” control functions into  
morphological and material properties



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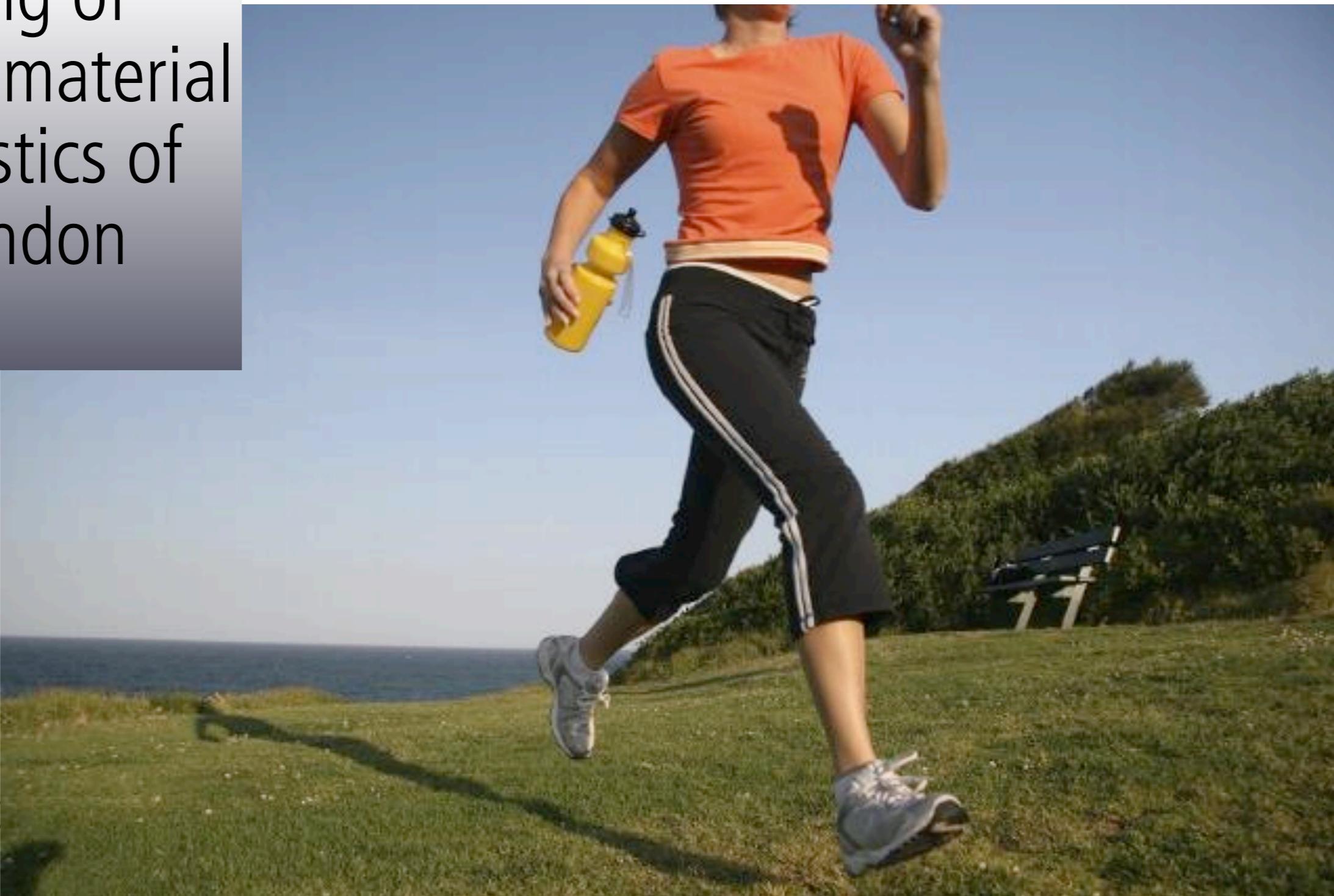
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# Dealing with impact while running

Outsourcing of control to material characteristics of muscle-tendon system



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# The big humiliations of mankind forcing a different view

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- The Copernican turn (1473 - 1543)
- Charles Darwin's theory of natural selection (19th Century)
- Sigmund Freud's psychoanalytic theory of neurosis —> behavior not under conscious control; free will?
- Watson-Crick: DNA — mechanism for life's hereditary information (Nobel Prize, 1962) (just as flatworm and yeast)
- Human Genome Project: 20'000 to 25'000 genes, instead of expected over 100'000, similar to flatworm C.elegans
- Humans as machines? as robots?



# End of selected highlights

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