

人
工
智能

The
ShanghaiAI

上
海
授
课

Lectures

The ShanghAI Lectures by the University of Zurich An experiment in global teaching

Today from the University of Zurich, Switzerland

Shanghai Jiao Tong University, China

Karlsruhe Institute of Technology, Germany

University of Reading, UK

Sussex University, UK

15 December 2011

欢迎您参与

“来自上海的人工智能系列讲座”

Future trends: Schedule

with world class guest lecturers - the super-cracks:

09.00 - 09.20 Rolf's introduction: Seeing things differently

09.20 - 09.50 Weidong Chen, Shanghai Jiao Tong University, China

09.55 - 10.20 Ruediger Dillmann, Karlsruhe Institute of Technology, Germany

10.25 - 10.55 Kevin Warwick, University of Reading, UK

11.00 - 11.30 Inman Harvey, University of Sussex, UK

11.30 Rolf: Final comments

moderator:

Rolf Pfeifer, University of Zurich, Switzerland

Seeing things differently a few illustrations



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Seeing things differently

- “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



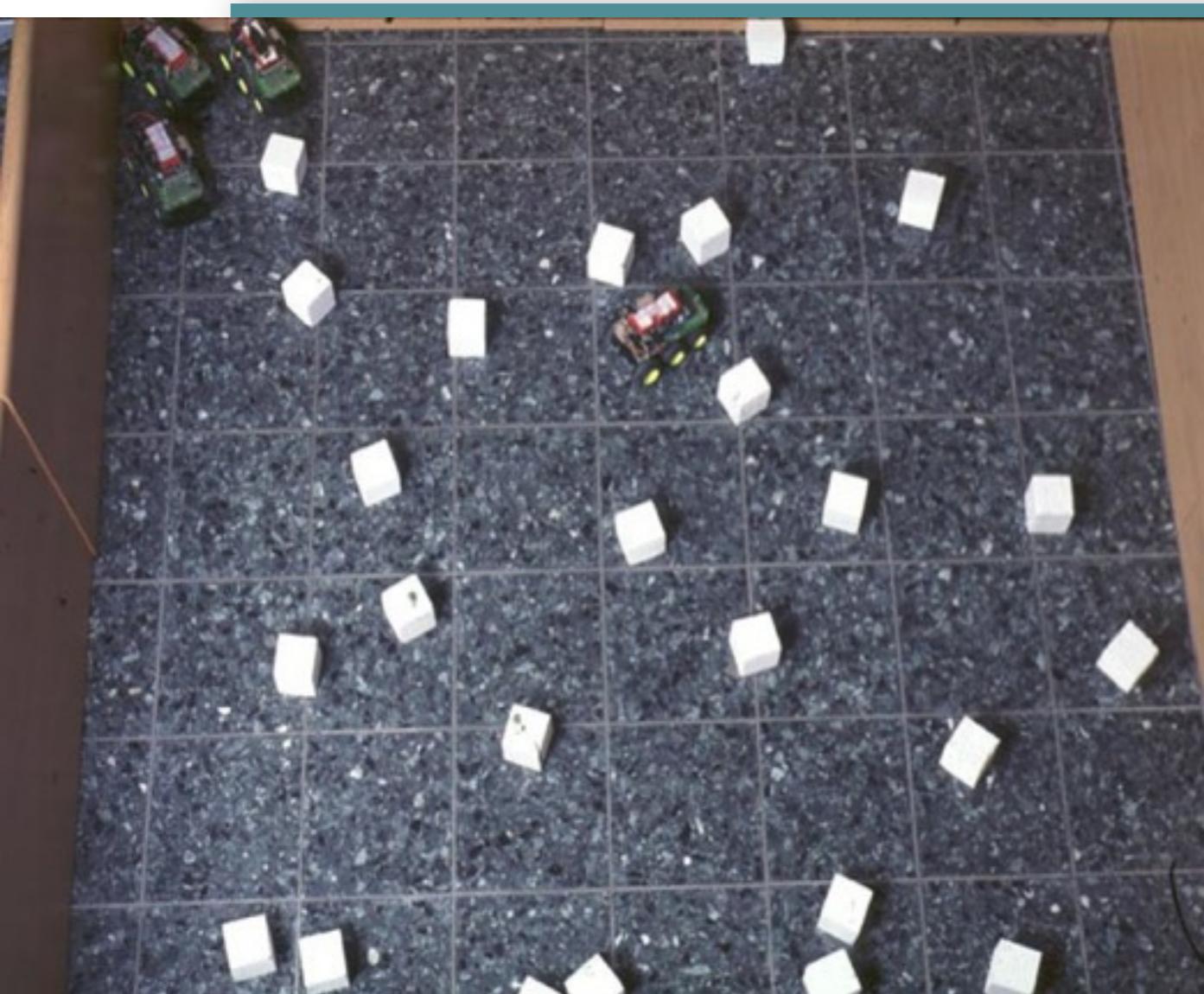
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Activities involved in clustering: standard solution



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

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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Design for emergence: cleaning an arena



entire process: \sim 20min
frames: 2-3min



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Walking without control?



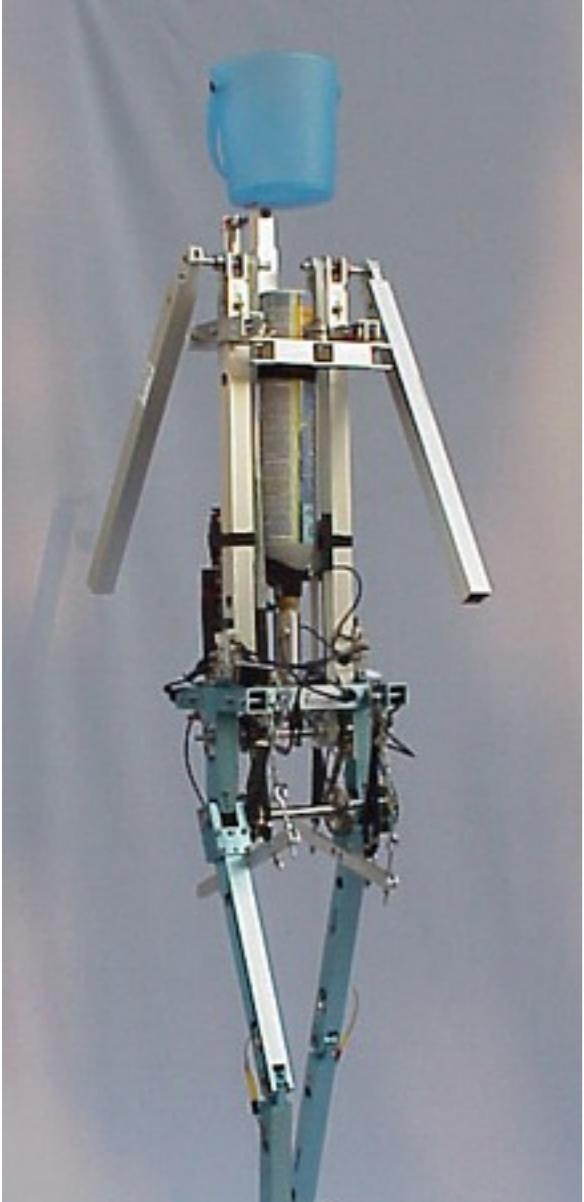
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The solution: The “Passive Dynamic Walker”



“Denise”
no control for balance



Rapid locomotion with slow electronics/brains?



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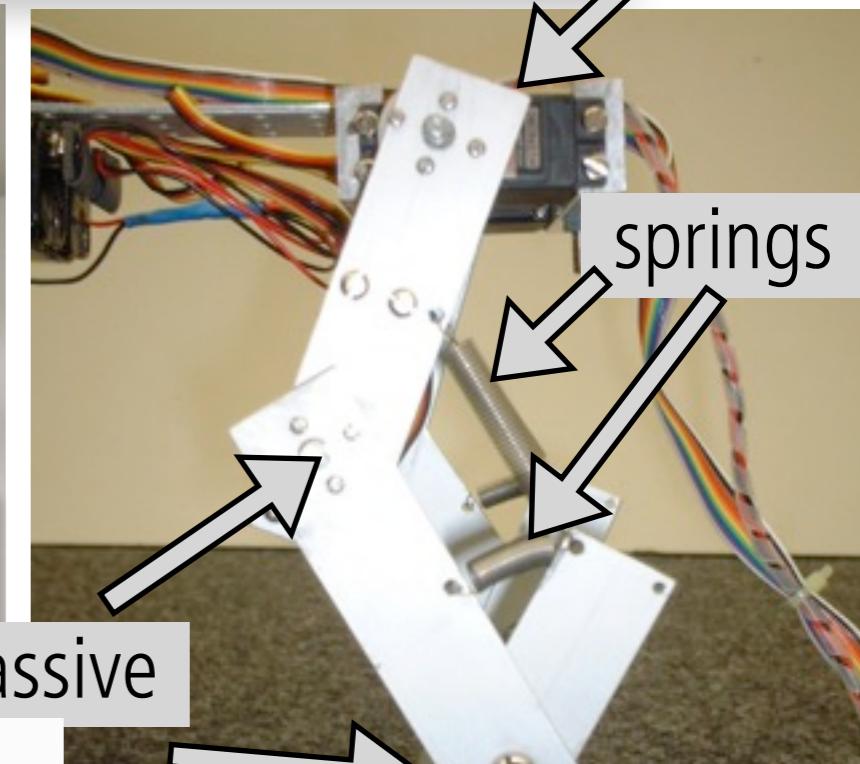
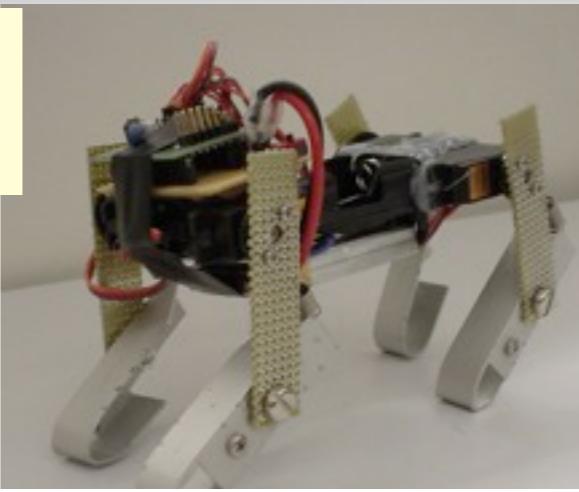
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The solution: “Exploitation of morphology and materials”

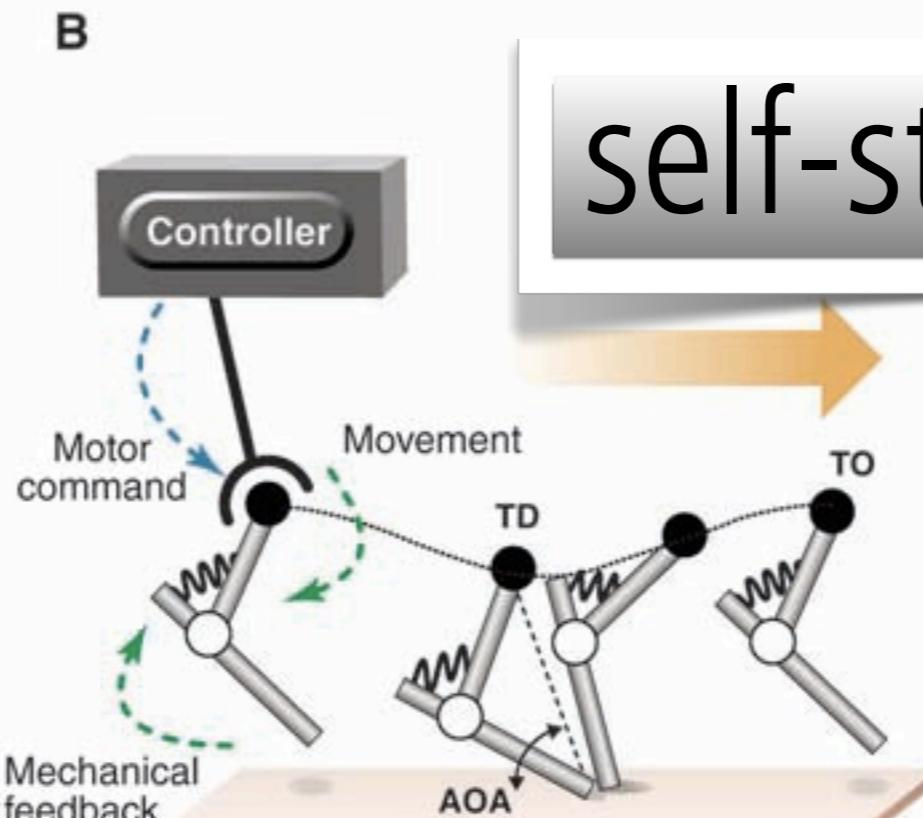
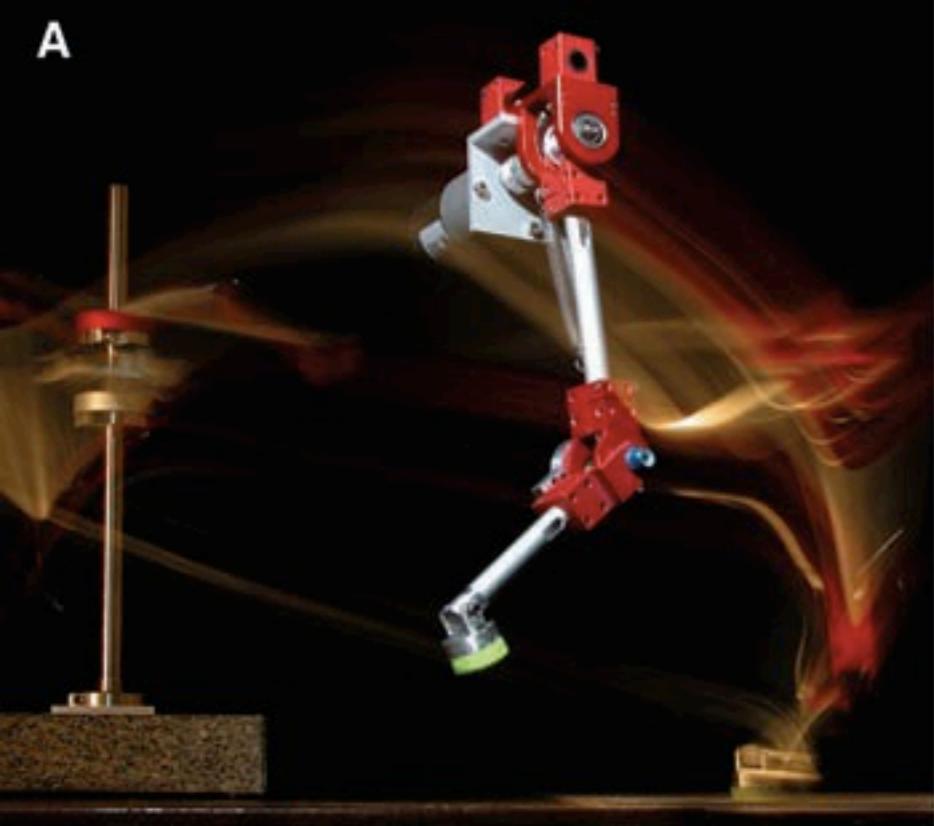
actuated:
oscillation

“Puppy”



morphological computation

passive



self-stabilization

Jena
monoped

Leg coordination without central control?



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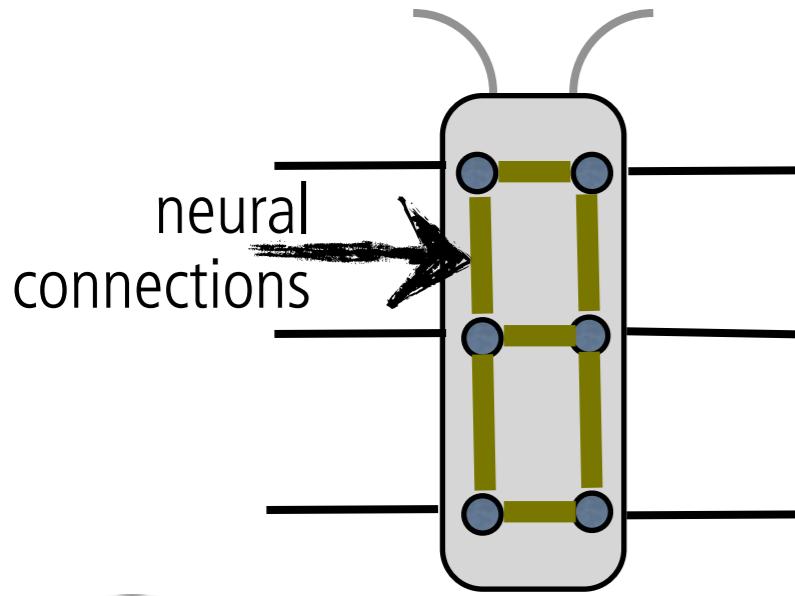


Parallel, loosely coupled processes



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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There are two aspects to this case study. First, because the insect is a physical system, if one leg pushes backwards, all the other legs will be moved as well, i.e. the joint angles will change. Second, because every action - in this case the pushing back of the leg - has a consequence in terms of patterns of sensory stimulation induced - which leads to a change of the joint angles - there is global communication between the legs, through the interaction with the environment. And this communication can be exploited for leg coordination.

Social competence without high-level cognition?



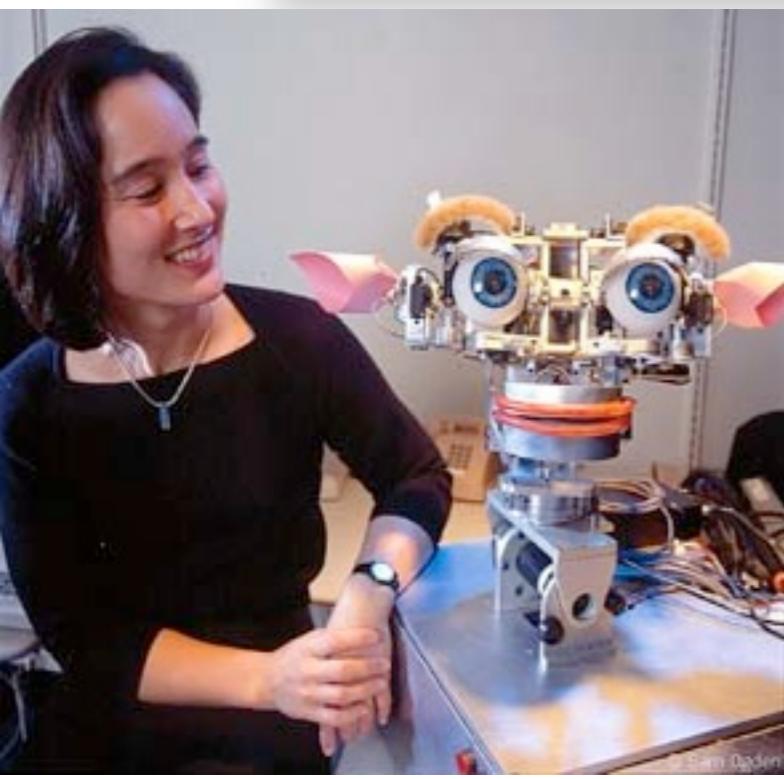
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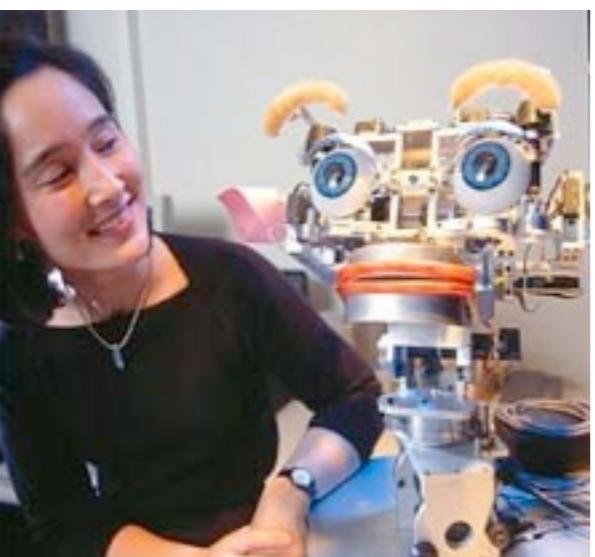
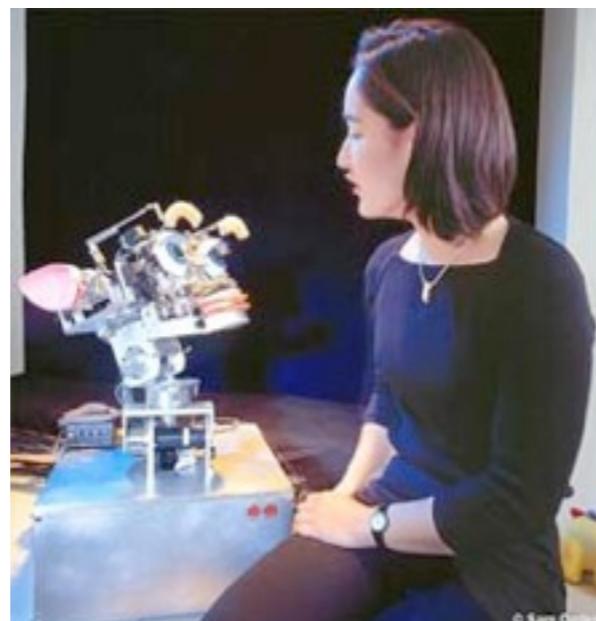
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Social competence without high-level cognition?



Cynthia Breazeal, MIT
Media Lab



of Zurich

reflexes:

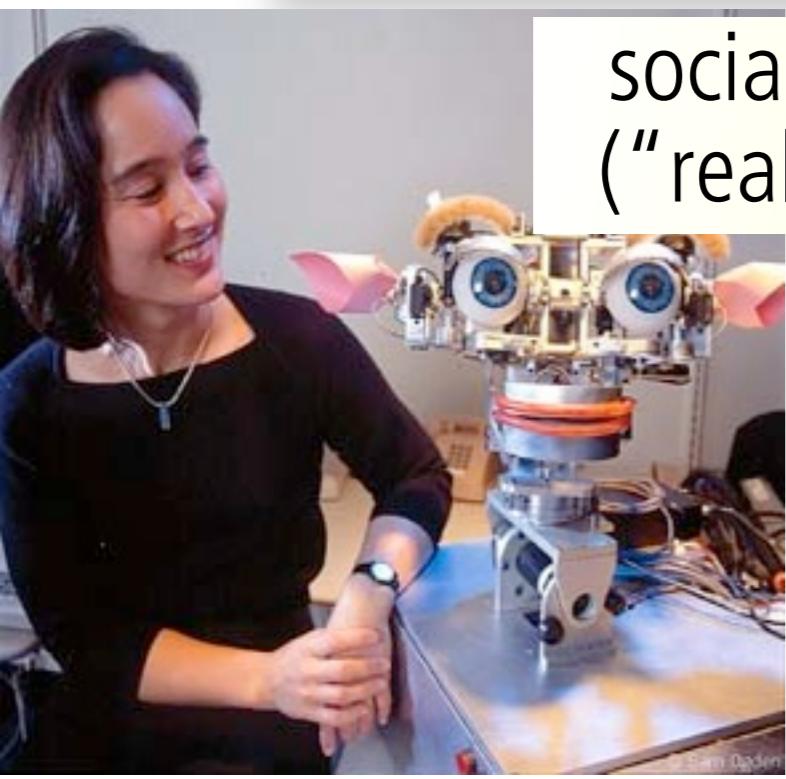
turn towards sound

turn towards moving objects

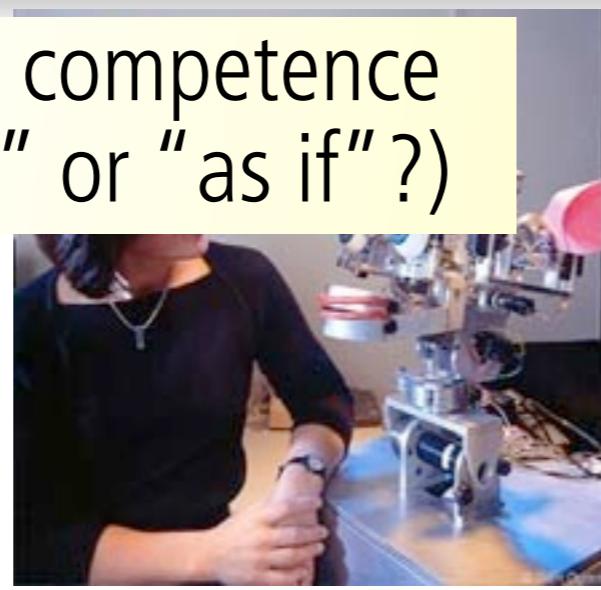
visual tracking of slow objects

habituation

Social competence without high-level cognition?



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

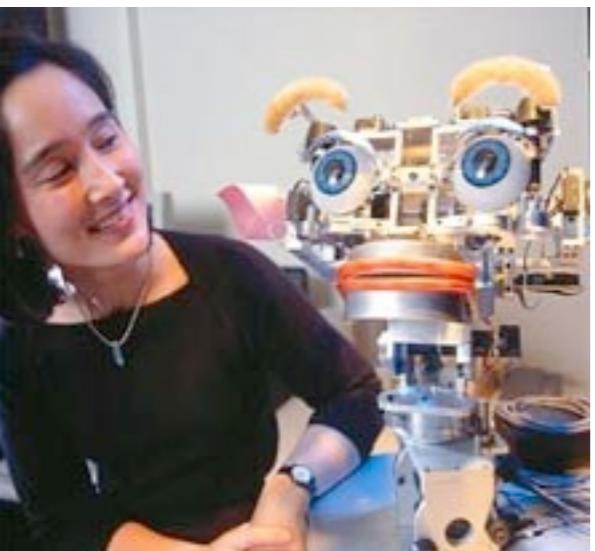


reflexes:
turn towards sound
turn towards moving objects
visual tracking of slow objects

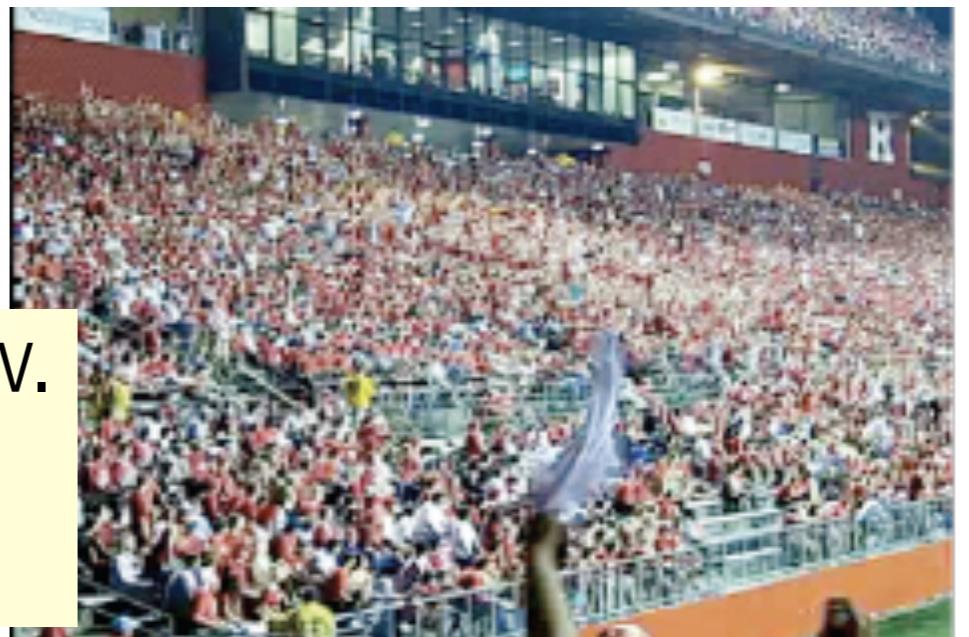
Cynthia Breazeal, MIT
Media Lab



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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Optimization without cognition 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!!!

Creative computers, robots?

- “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?”



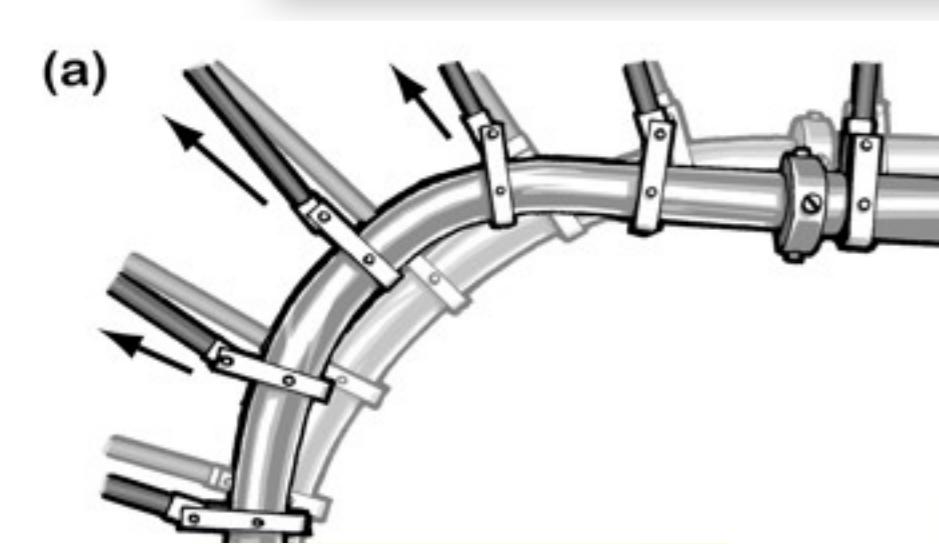
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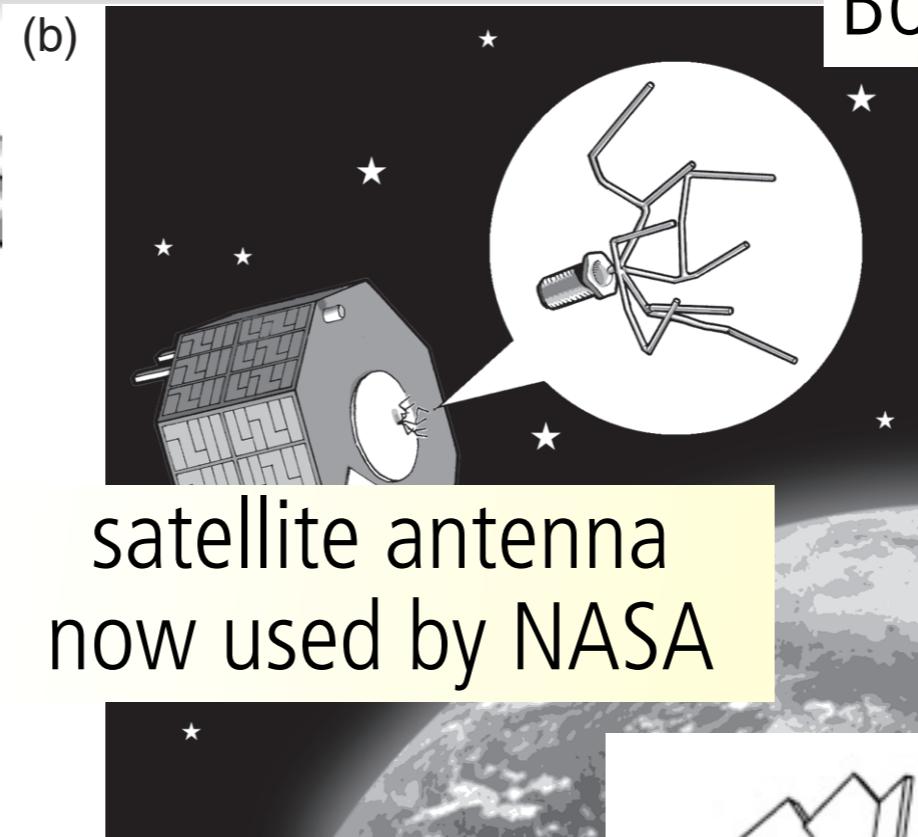
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Solution: Artificial evolution and morphogenesis

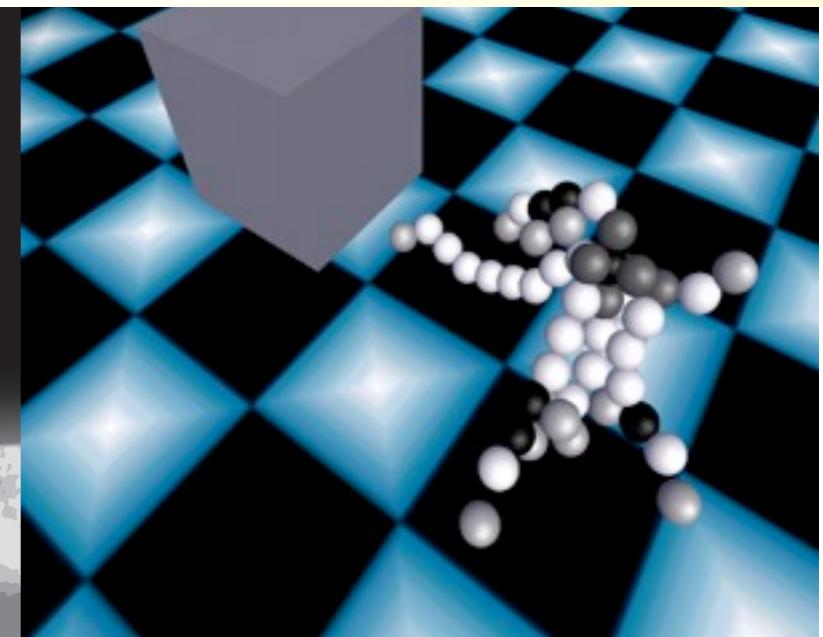


Rechenberg's
'hunched' fuel pipe
solution

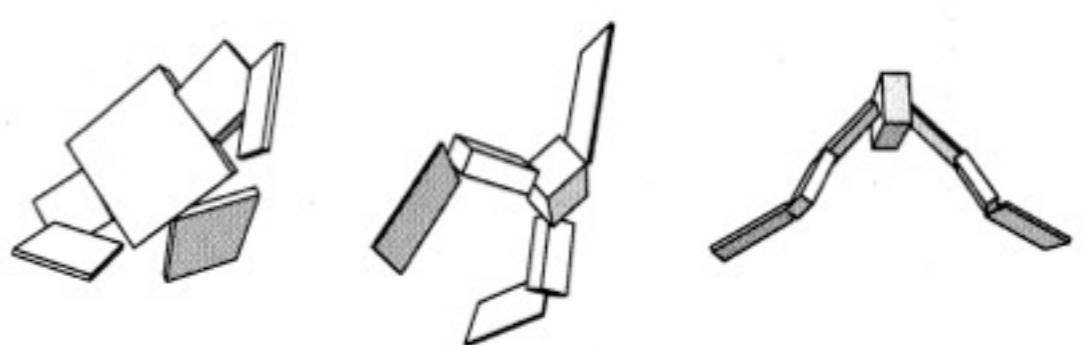


satellite antenna
now used by NASA

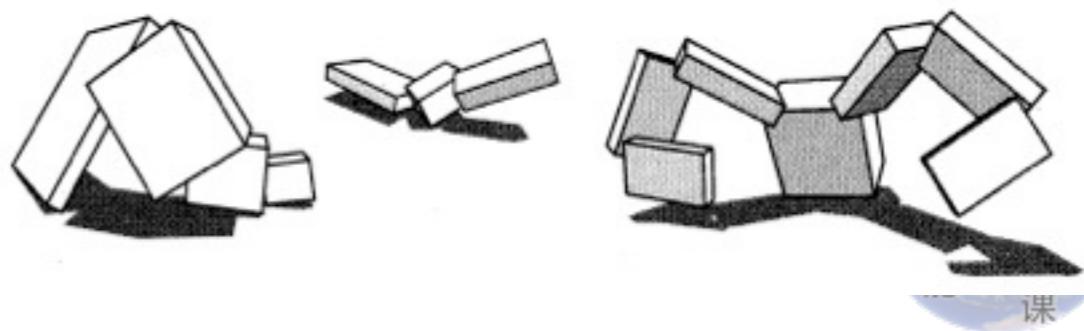
Bongard's "Block Pushers"



Lipson's creatures
"Golem" project



Sims's creatures



GAs come up with solutions that humans, at least until to date, have not been able to find.

That's it!

Thank you for your attention!

stay tuned for

Chen, Dillmann, Warwick, Harvey



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Weidong Chen: Guest speaker



Department of Automation
School of Electronic, Information, and
Electronic Engineering
Institute of Robotics and Intelligent
Information Processing
Shanghai Jiao Tong University
Shanghai, China

Prof. Weidong Chen
“Enhancing autonomy and safety of assistive robots”

First host of ShangAI Lectures, 2009



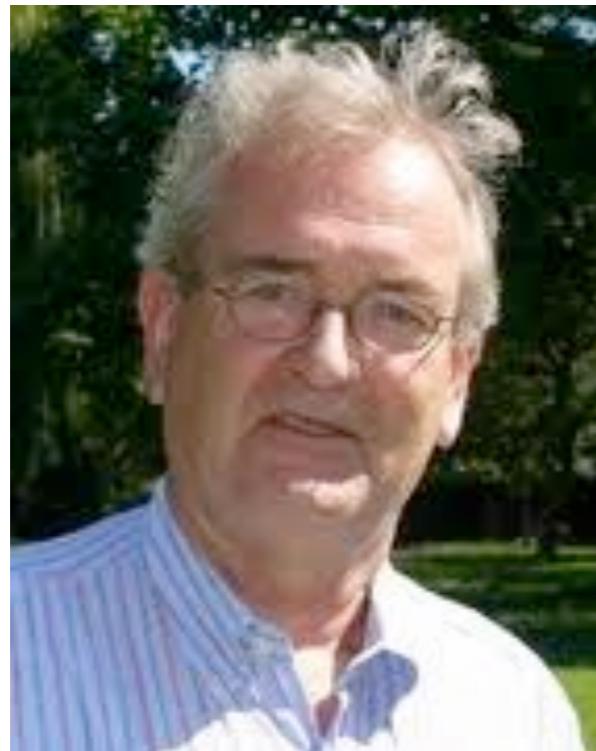
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Ruediger Dillmann: Guest speaker



Institute of Anthropomatics,
Karlsruhe Institute of Technology
Germany

Prof. Ruediger Dillmann
“Programming by demonstration”

“Mr. Robotics, Germany”



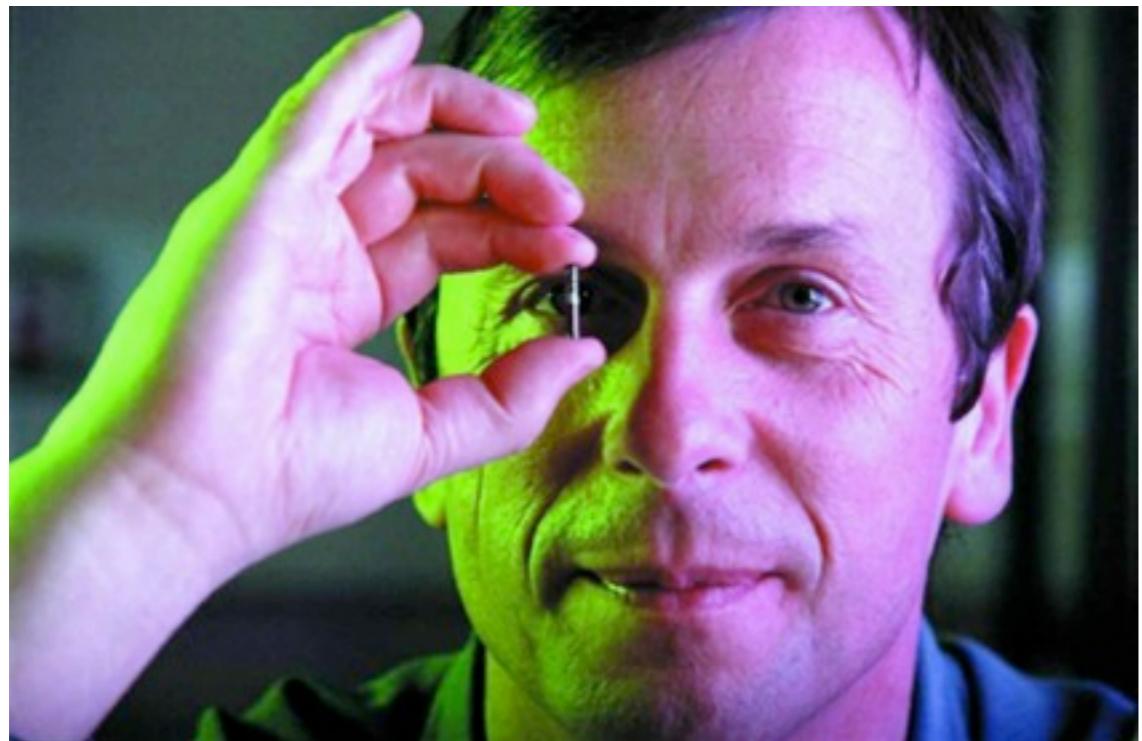
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Kevin Warwick: Guest speaker



Professor of Cybernetics
University of Reading
“Mr. Cyborg”

Prof. Kevin Warwick, University of Reading, UK
“AI with biological brains”



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Inman Harvey: Guest speaker



Sussex University, UK

Inman Harvey
“Metaphorical Homunculi: we don’t really have little men in our heads”



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Epilogue

David Payne *Confessions of a Taoist on Wall Street*

Sun I: little boy growing up in monestary

Wu: chef (cook) of monestary and Sun I's mentor



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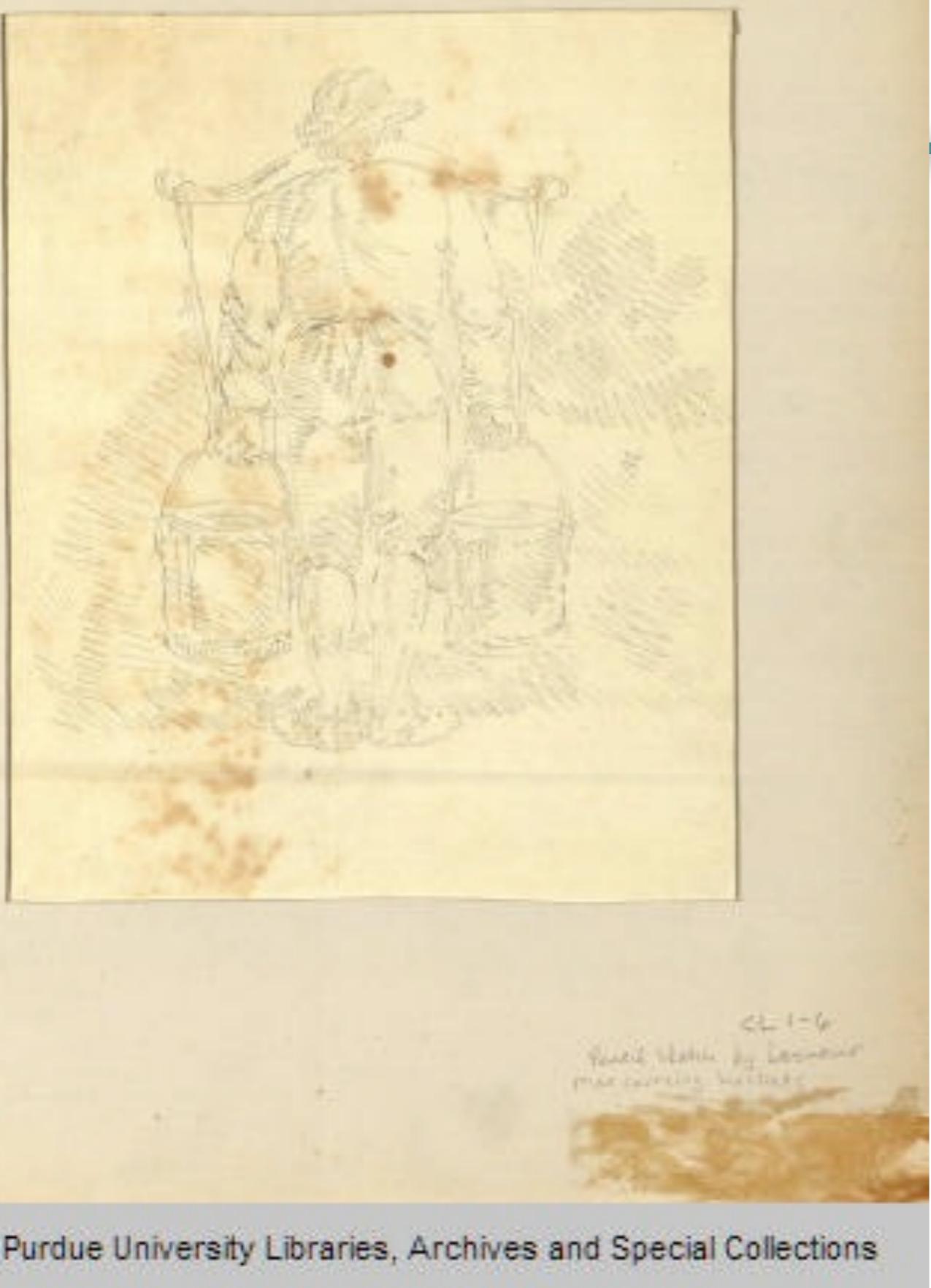
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The little boy, Sun I, is born of a Chinese mother and an American fighter pilot in China. His mother dies at birth and his father returns home to the United States; he is left alone and grows up in a monastery. His mentor and teacher is the chef, 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

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.” ...

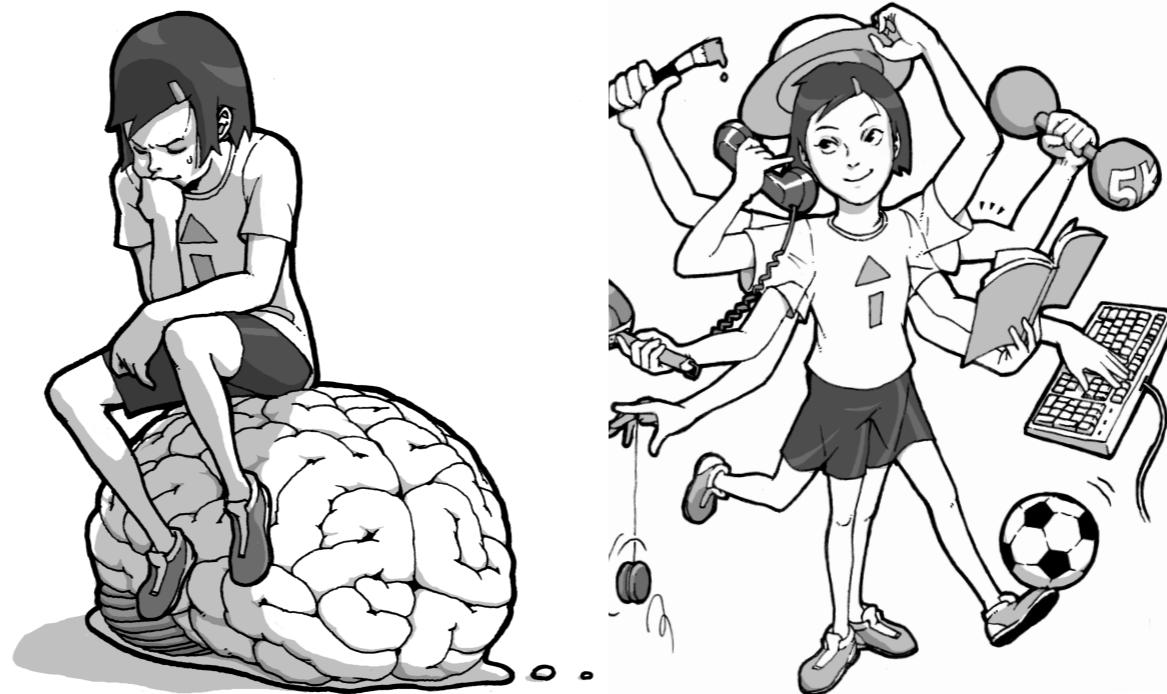
“You haven’t yet caught on. It’s precisely this—excess of method—that confounds you, leaves the buckets nearly empty ...”

“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)

End of Future Trends Session

Thank you for your attention and thanks to all the speakers!



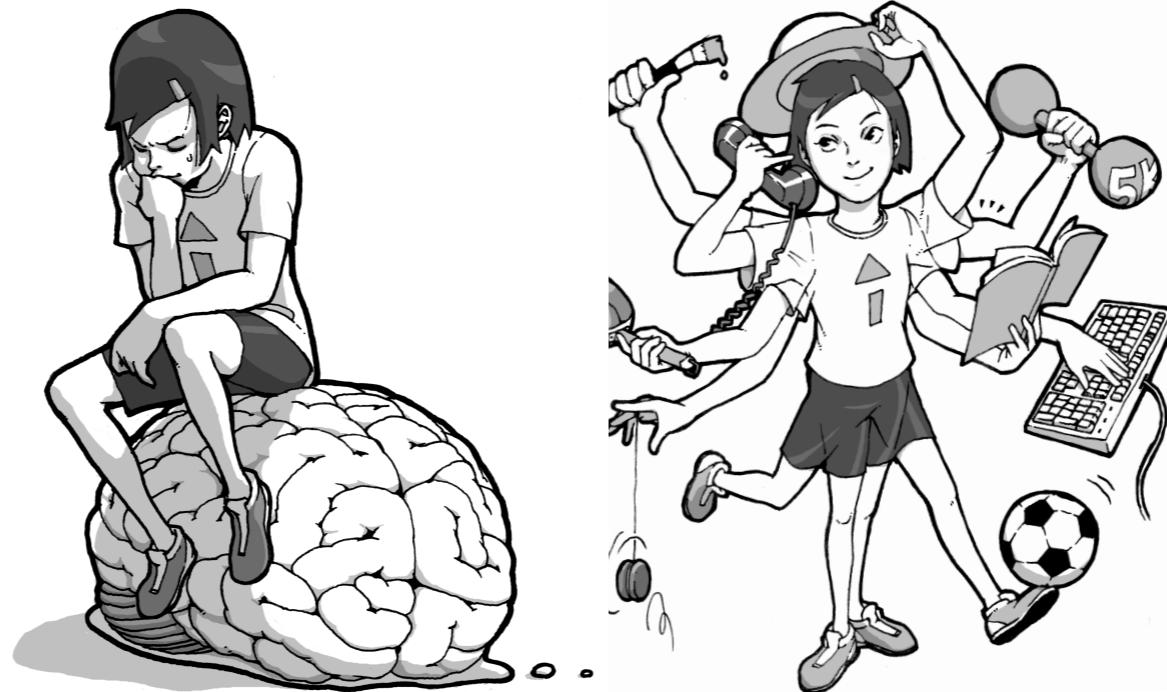
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End of Future Trends Session

Thank you, participants, technicians, all who have helped, and thank you Nathan!!

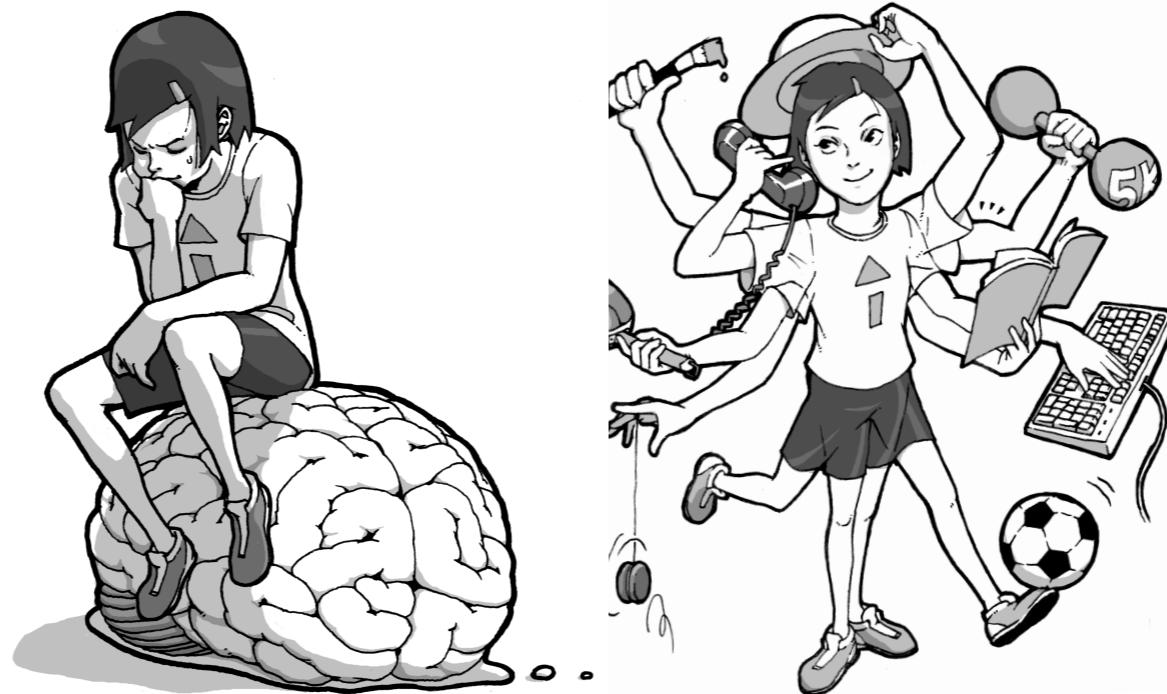


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The future of the ShanghAI Lectures

Join our discussion about the future of the ShanghAI Lectures - your input is extremely valuable and warmly appreciated!



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Additional materials for self-study

- Steps toward a theory of intelligence
 - “Mind set”
 - Properties of real world and embodied agents
 - Design principles
- Design principles for intelligent systems
- Epilogue



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Steps toward a theory of intelligence

1. Meta-considerations: “Mind Set”
2. Properties of physically embodied agents and of real world
3. Design principles



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Steps toward a theory of intelligence

1. Meta-considerations: “Mind Set”
2. Properties of physically embodied agents and of real world
3. Design principles



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Steps toward a theory of intelligence

- level of generality and form of theory
- diversity-compliance
- frame-of-reference
- synthetic methodology
- time perspectives
- emergence



Steps toward a theory of intelligence

- level of generality and form of theory
 - diversity-compliance

design principles

theory of complex dynamical systems

- time perspectives
- emergence



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Steps toward a theory of intelligence

- level of generality and form of theory
- diversity-compliance

here-and-now: comply/exploit—diversity
learning: plasticity—stability trade-off
evolution: exploration—exploitation



Steps toward a theory of intelligence

- level of generality and form of theory
- diversity-compliance
- **frame-of-reference**
- synthetic methodology
- time perspectives
- emergence



Steps toward a theory of intelligence

- level of generality and form of theory
- diversity-compliance
- **frame-of-reference**
- **s**ystematicity
- **t**heoretical perspective
- **e**xplanatory behavior vs. mechanism complexity



Steps toward a theory of intelligence

- level of generality and form of theory
- diversity-compliance
- frame-of-reference
- **synthetic methodology**
- time perspectives
- emergence



Steps toward a theory of intelligence

- level of generality and form of theory
- diversity-compliance
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- **synthetic methodology**
- time perspectives
- emergence

“understanding by building”

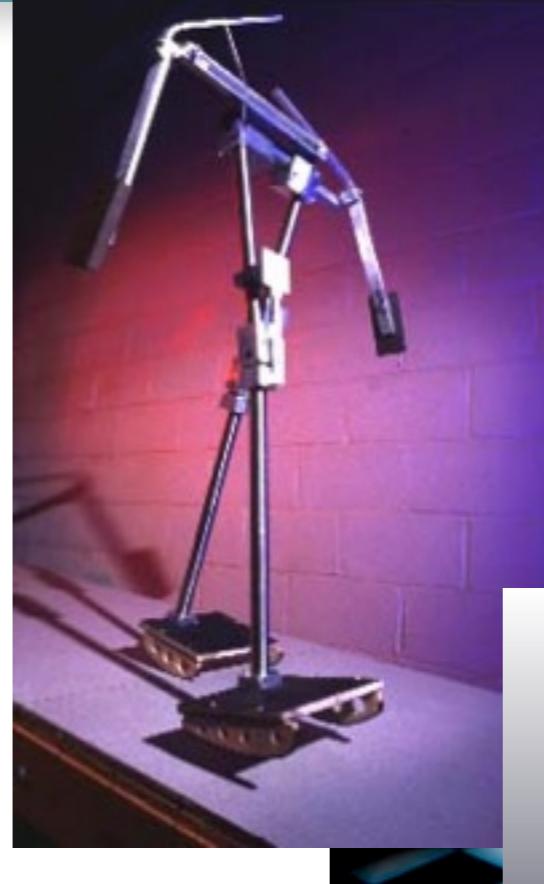




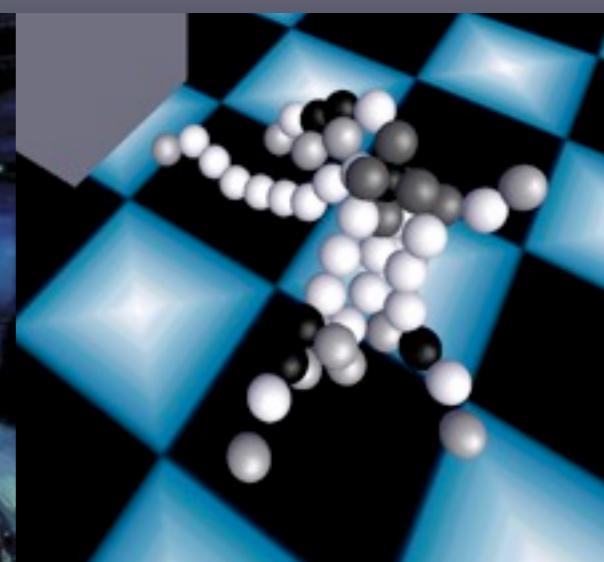
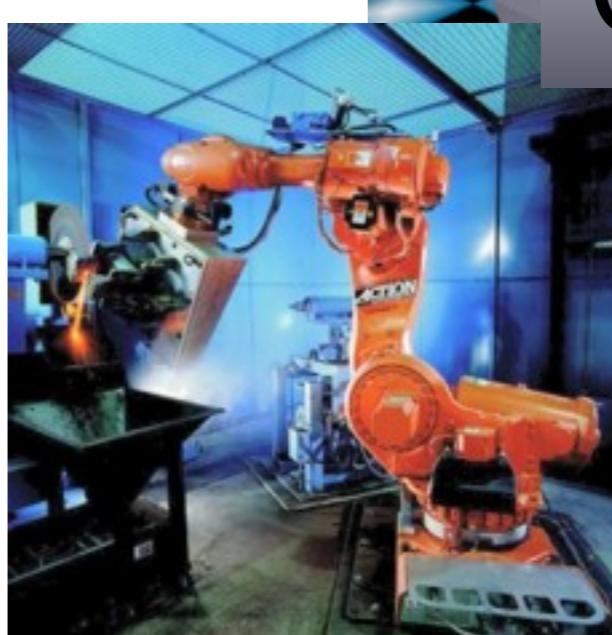
Intelligence/life as it is — as it could be



Intelligence/life as it is — as it could be



life as it
could be



life as it is



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Steps toward a theory of intelligence

- level of generality and form of theory
- diversity-compliance
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- emergence



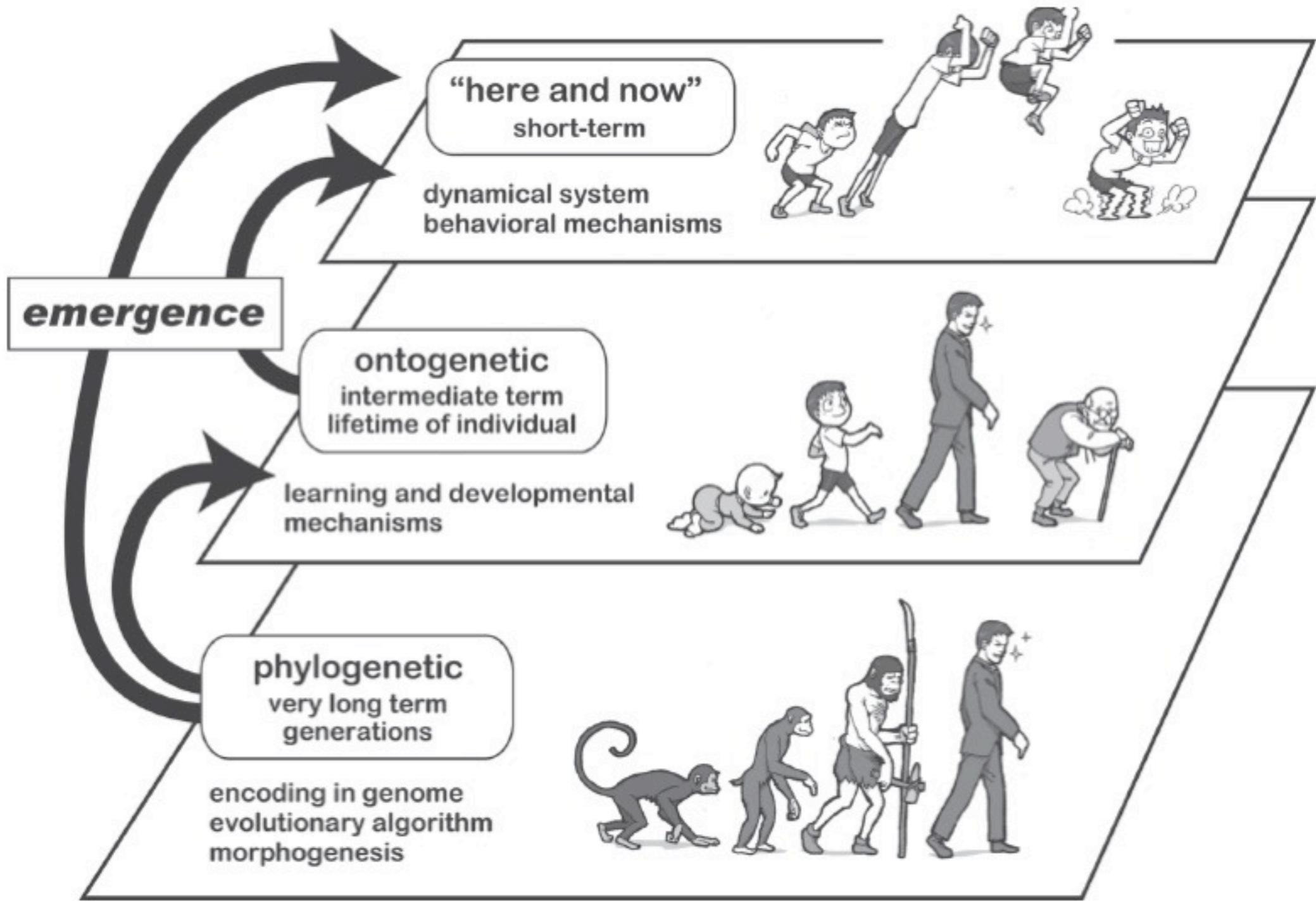
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Time perspectives



Steps toward a theory of intelligence

- level of generality and form of theory
- diversity-compliance
- frame-of-reference
- synthetic methodology
- time perspectives
- emergence



Emergence

- behavior of individual
- global patterns from local interactions
- from time scales



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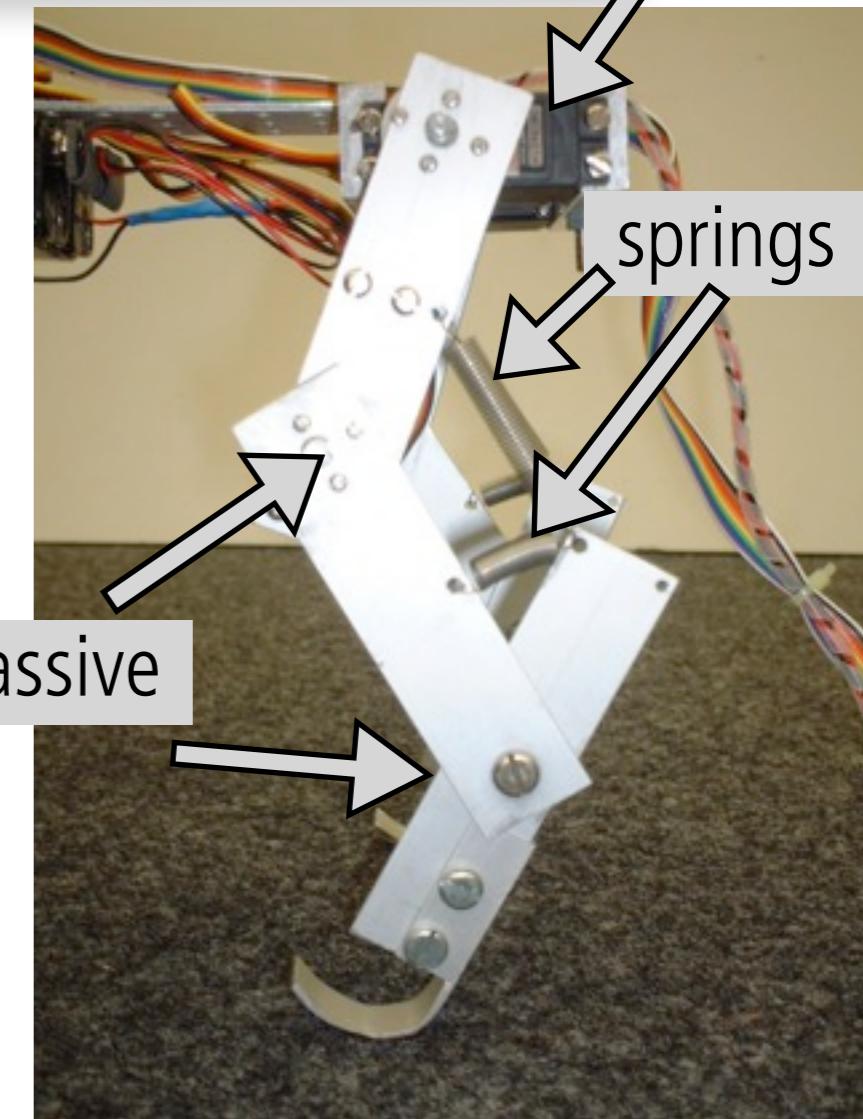
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Emergence of stable gait patterns: the quadruped “Puppy”

actuated:
oscillation

- simple control (oscillations of “hip” joints)
- spring-like material properties (“under-actuated” system)
- self-stabilization, no sensors
- “outsourcing” of functionality



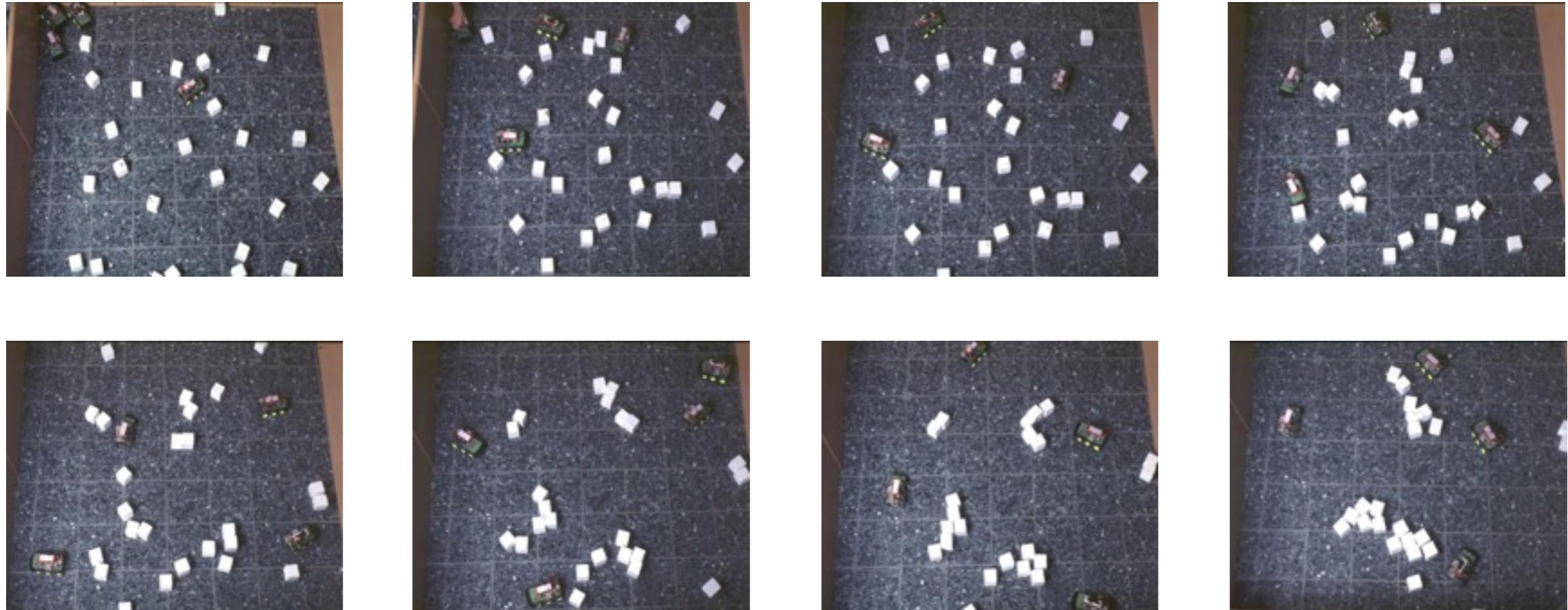
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Emergence of clustering



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



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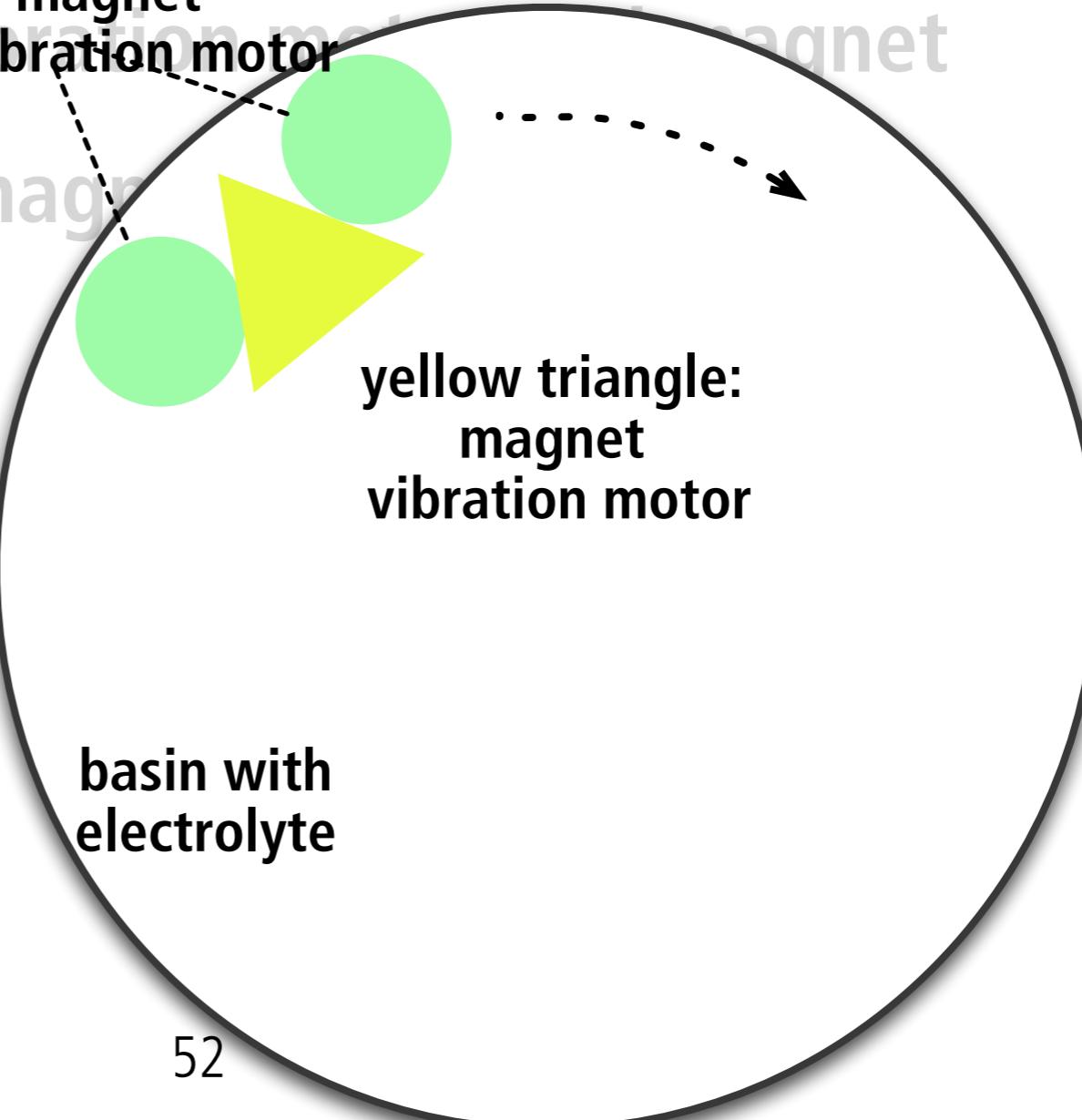


Design for emergence: a “bicycle” following the wall

emergent functionality (not self-assembly)

green discs: magnet
yellow triangle: no vibration motor

green discs: only magnet



how does it work?

basin with
electrolyte

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The vibrations of the triangle are propagated to the passive green discs, which then cause the entire structure (held together by the magnets) to follow the wall of the container. This clearly demonstrates that we cannot only work in simulation (at least up to the point where this experiment was demonstrated, this could not have been achieved in simulation).

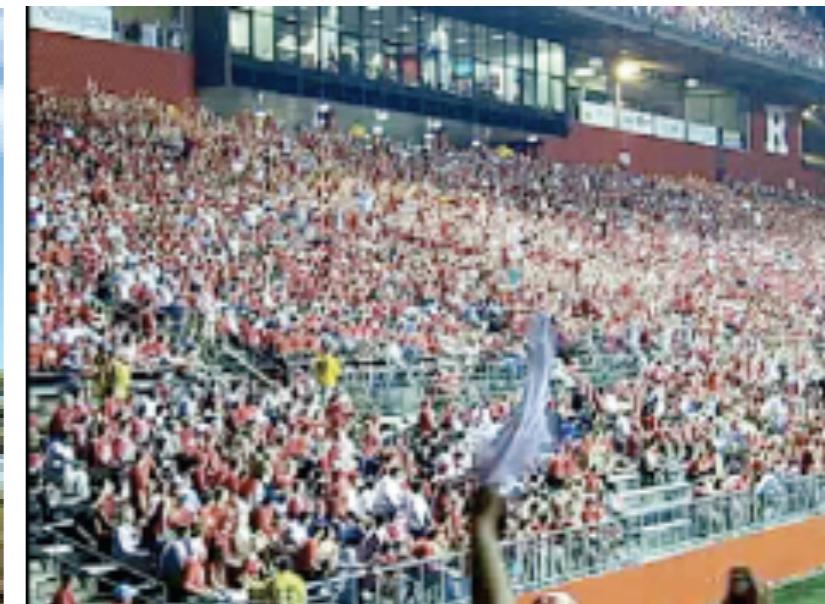
Emergence of global patterns from local rules — self-organization



bee
hive



termite mound



“wave” in stadium

open source development community



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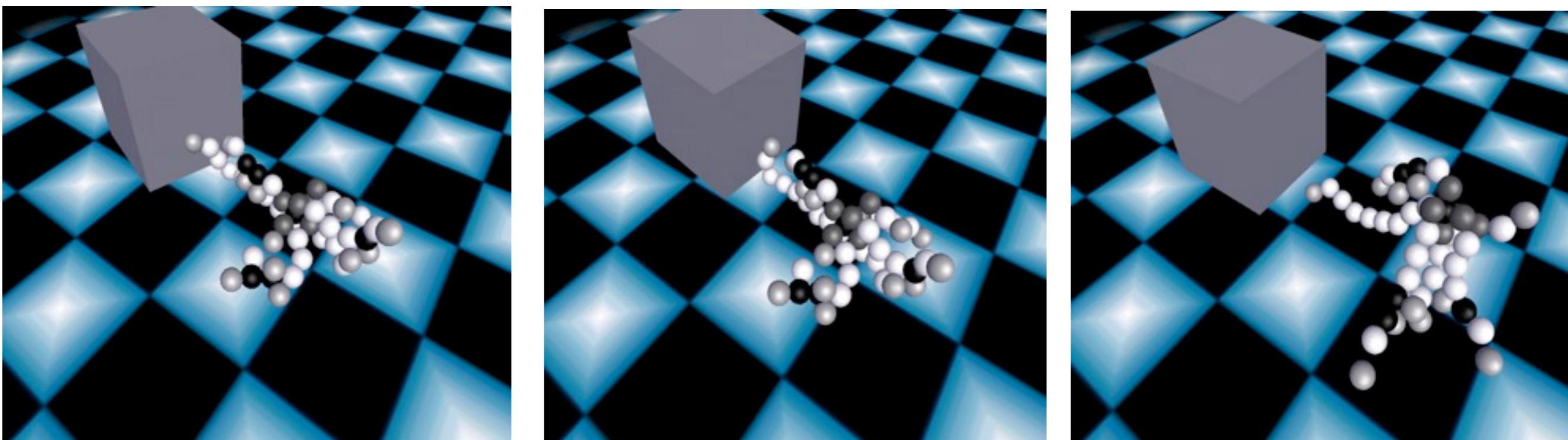
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Emergence of behavior from time scales: locomotion and pushing

Bongard's "Block Pushers"

- development (morphogenesis) embedded into evolutionary process, based on GRNs
- testing of phenotypes in physically realistic simulation



Emergence as a continuum



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Steps toward a theory of intelligence

1. Meta-considerations: “Mind Set”
2. Properties of physically embodied agents and of real world
3. Design principles



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Properties of physically embodied agents, e.g. “Puppy”

- subject to the laws of physics — gravity, friction
- generation of sensory stimulation through interaction with real world — angle, pressure sensors
- affect environment through behavior — exerting pressure on ground, crushing grass underfoot
- complex dynamical systems — gait patterns as attractor states
- perform morphological computation — materials (springs) take over part of control function



Characteristics of real-world environments

- information acquisition takes time
- information always limited
- noise and malfunction
- no clearly defined states
- multiple tasks
- rapid changes — time pressure
- non-linearity: intrinsic uncertainty



Characteristics of real-world environments

- information acquisition takes time

**Herbert Simon's concept of
“bounded rationality”**

- noise and malfunction
- no clearly defined states
- multiple tasks
- rapid changes — time pressure
- non-linearity: intrinsic uncertainty



Agent design principles

Agent design principles	Description
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

Principles for development	Description
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) 61

Design principles for evolution

Principles for evolution	Description
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 62

Design principles for collective intelligence

Principles for collective systems	Description
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 63

Illustrations of design principles

Selected highlights

(see additional slides - study materials)



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Agent design principles

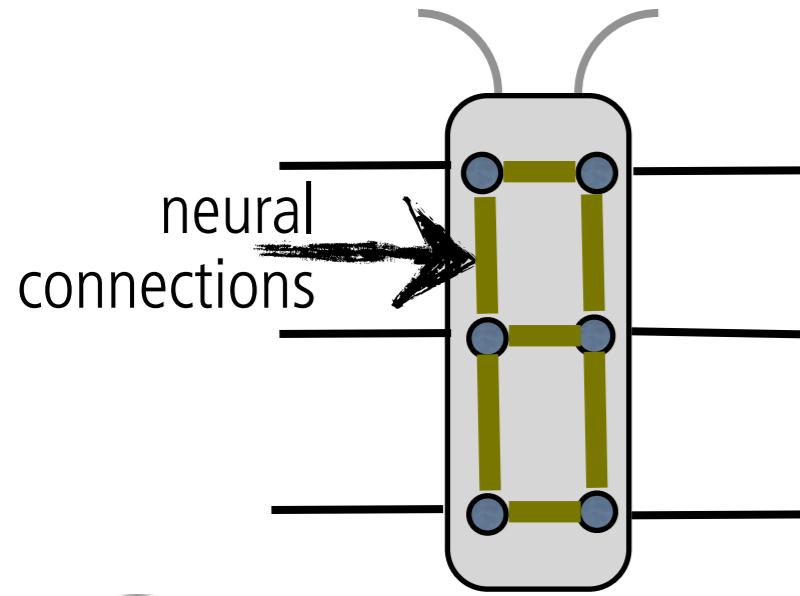
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Parallel, loosely coupled processes



Holk Cruse, German biologist

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Parallel, loosely coupled processes

coupling through pheromone trails in environment



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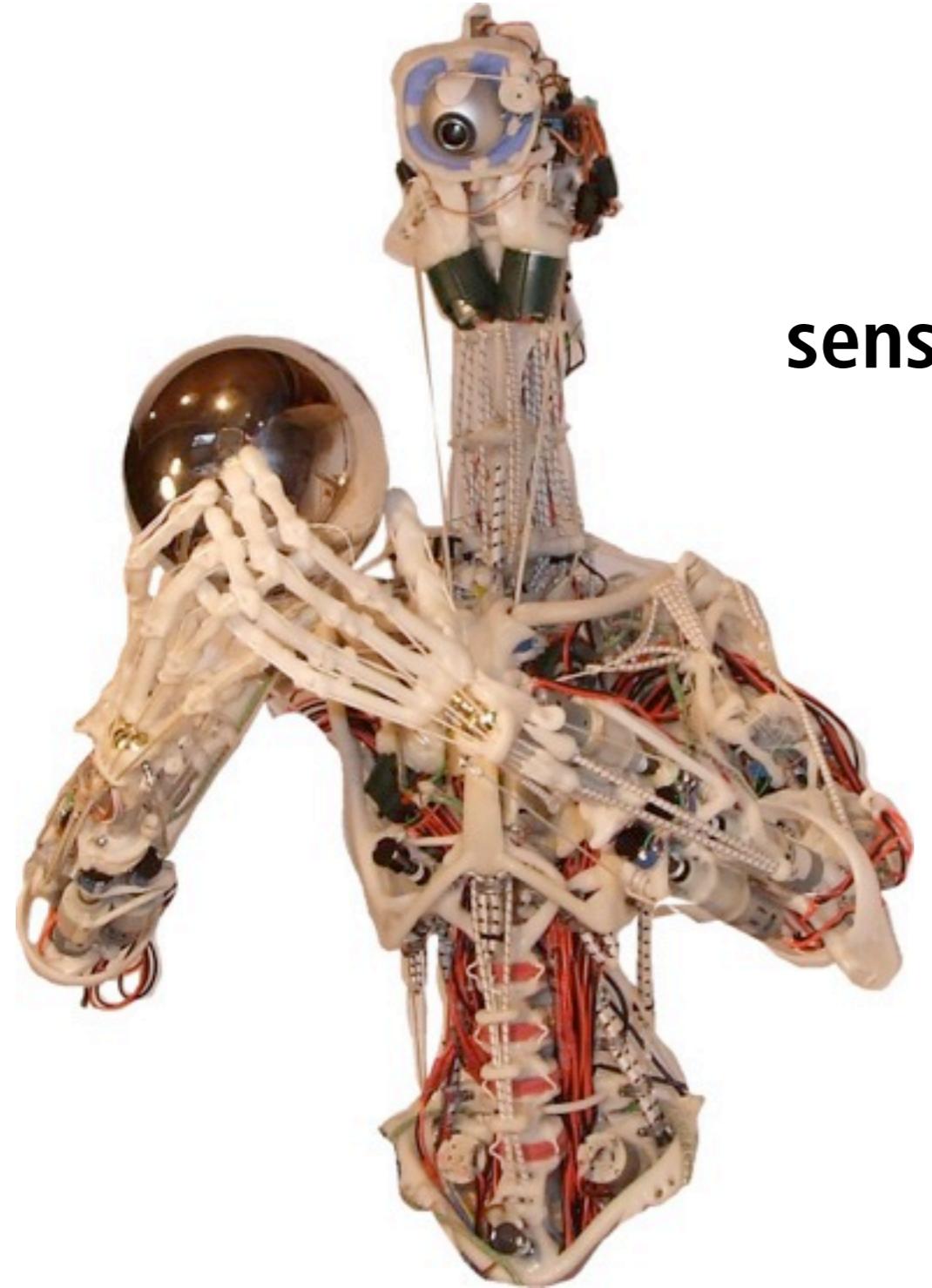


Pheromone trails enable ants to search for food efficiently: Two ants leave the nest at the same time (top), each taking a different path and marking it with pheromone. The ant that took the shorter path returns first (bottom). Because this trail is now marked with twice as much pheromone, it will attract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

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

Sensory-motor coordination Information self-structuring

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



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As pointed out many times, through sensory-motor coordinated interaction with the environment, information structure is induced.

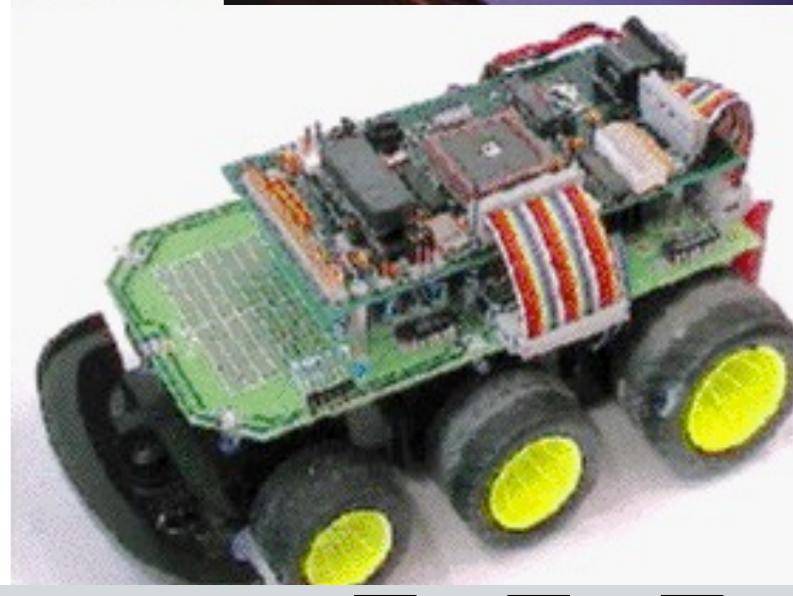
Cheap design exploitation of niche



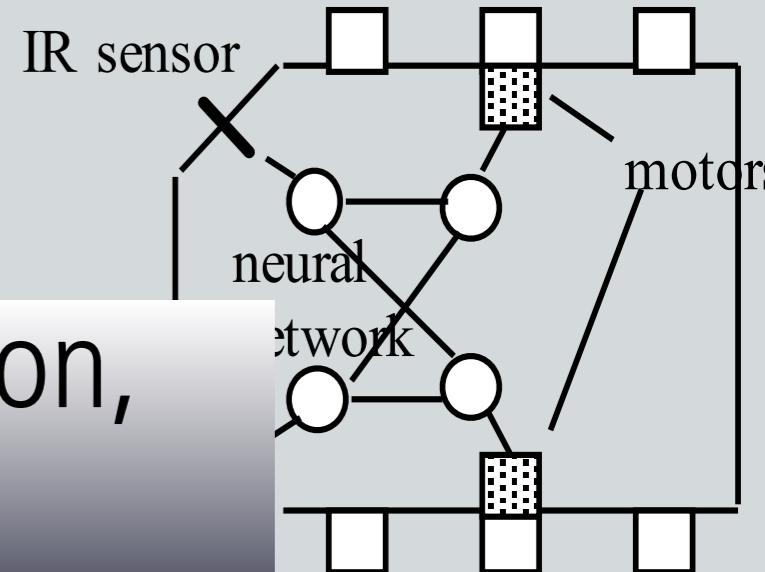
6x6m arena with Styrofoam cubes

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

Passive Dynamic
Walker



Didabot
simple robot
for didactical
purposes



Daniel Dennett: "These robots are cleaning up, but that's not what they think they are doing" (joke – can they think?)

Cheap design simple mechanism

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



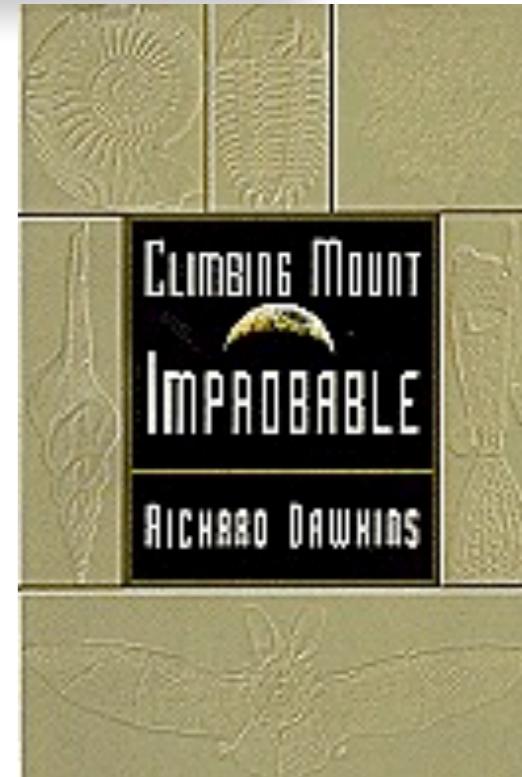
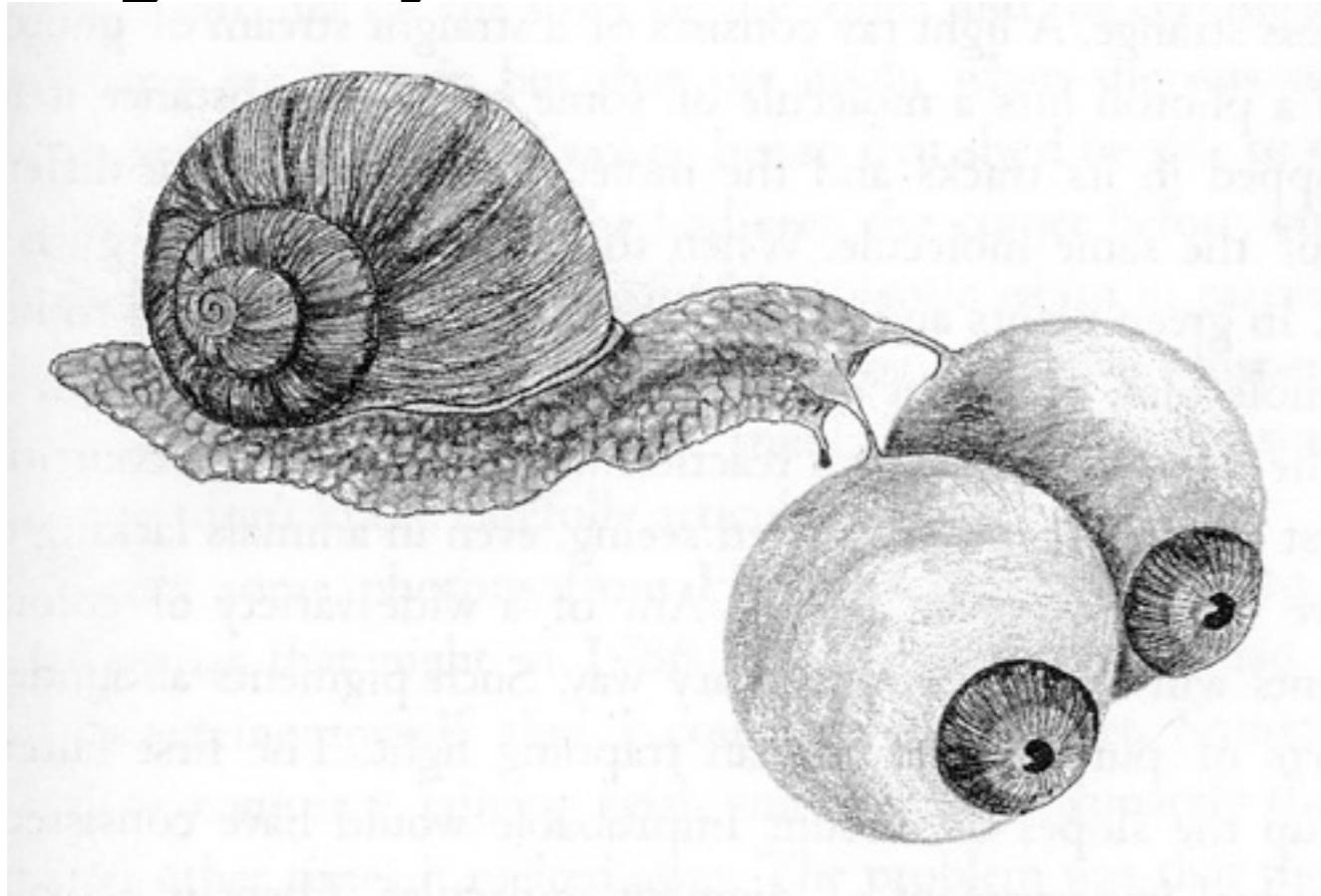
parachutes
mechanical

wheels (friction)
runway conditions

There is a partial overlap of functionality in the different sensory patterns.

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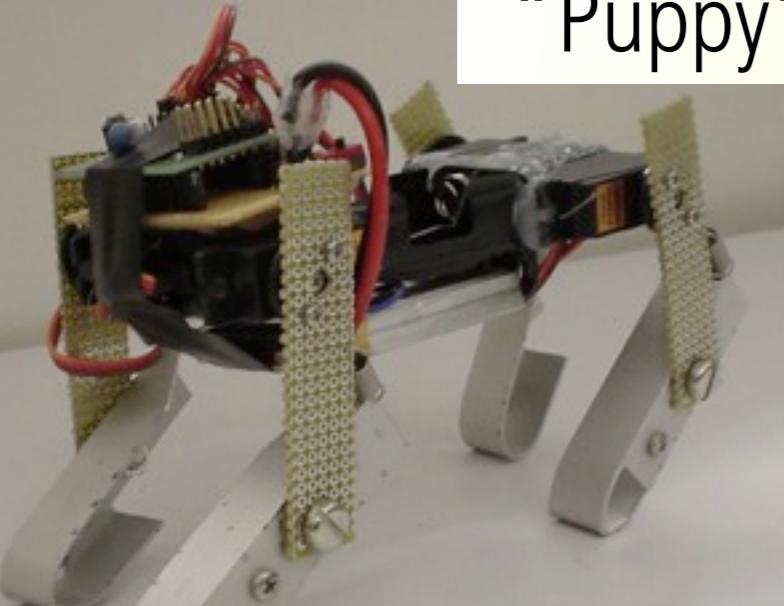
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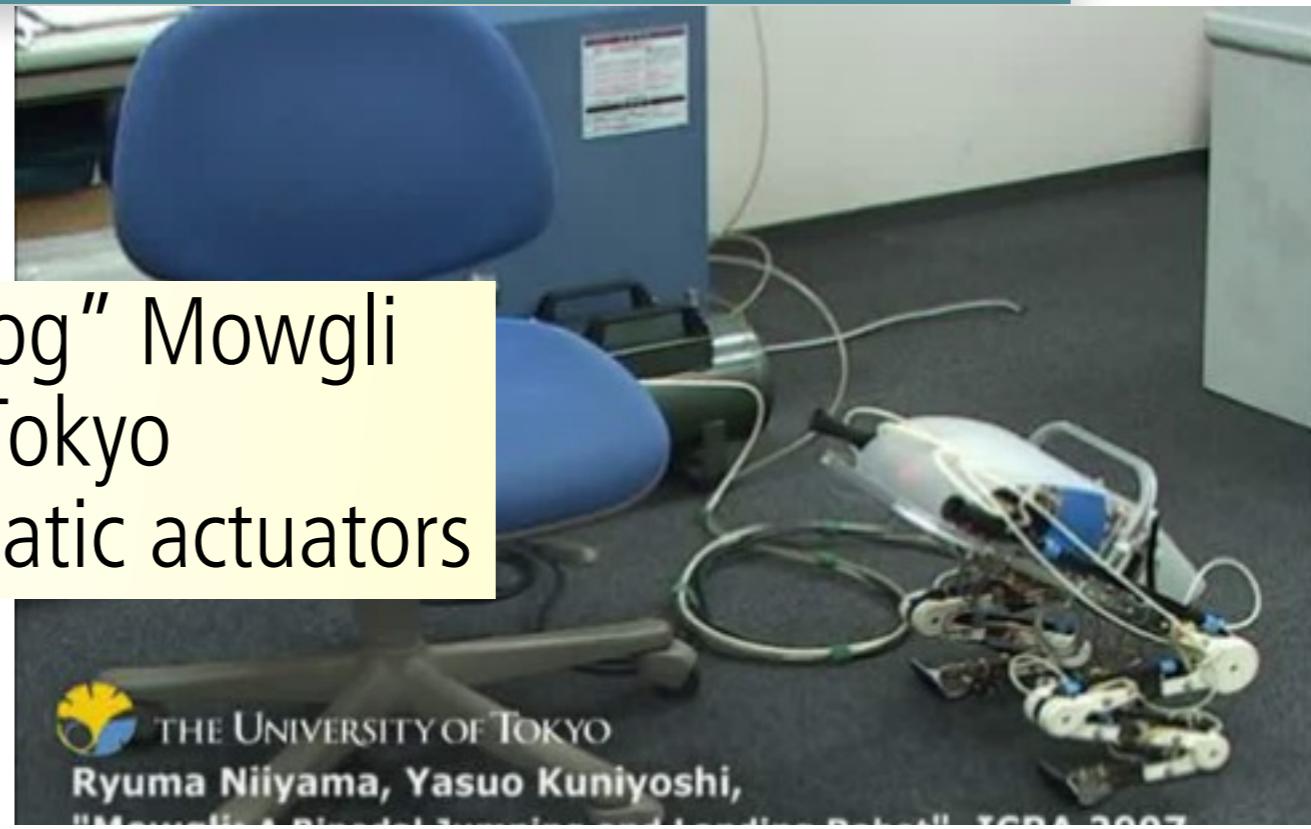
What are such huge eyes good for if the snail cannot move quickly? It's only useless additional weight.

Ecological balance: task distribution morphology, materials, control

"Puppy" with springs



Robot "frog" Mowgli
U-Tokyo
with pneumatic actuators



loosely swinging arm
with pneumatic actuators

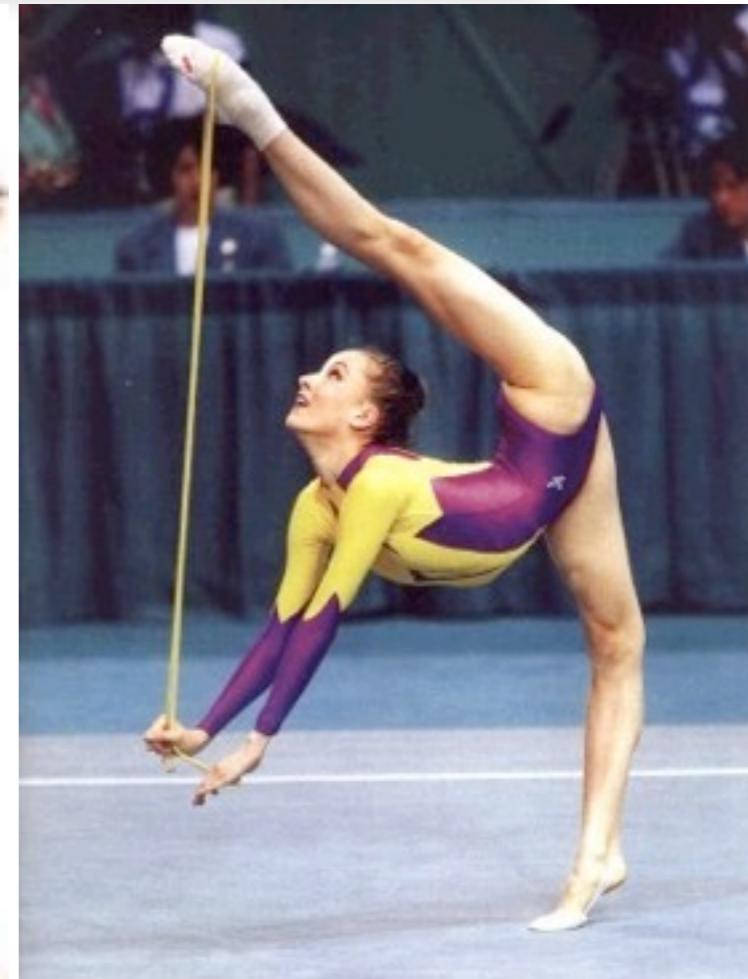


control partially
"outsourced" to properties
of springs, pneumatic
actuators

Design principles for development

Principles for development	Description
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) 74

Bernstein's problem: Development as an incremental process



learning to control high
DOF body: freezing and
freeing DOFs



Not all degrees of freedom are used at the outset, but some are “frozen” and only freed up once a basic set of movements is in place.

Discover, learn from real world

The best exploration strategies

only possible through embodiment

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

“loosely swinging arm”



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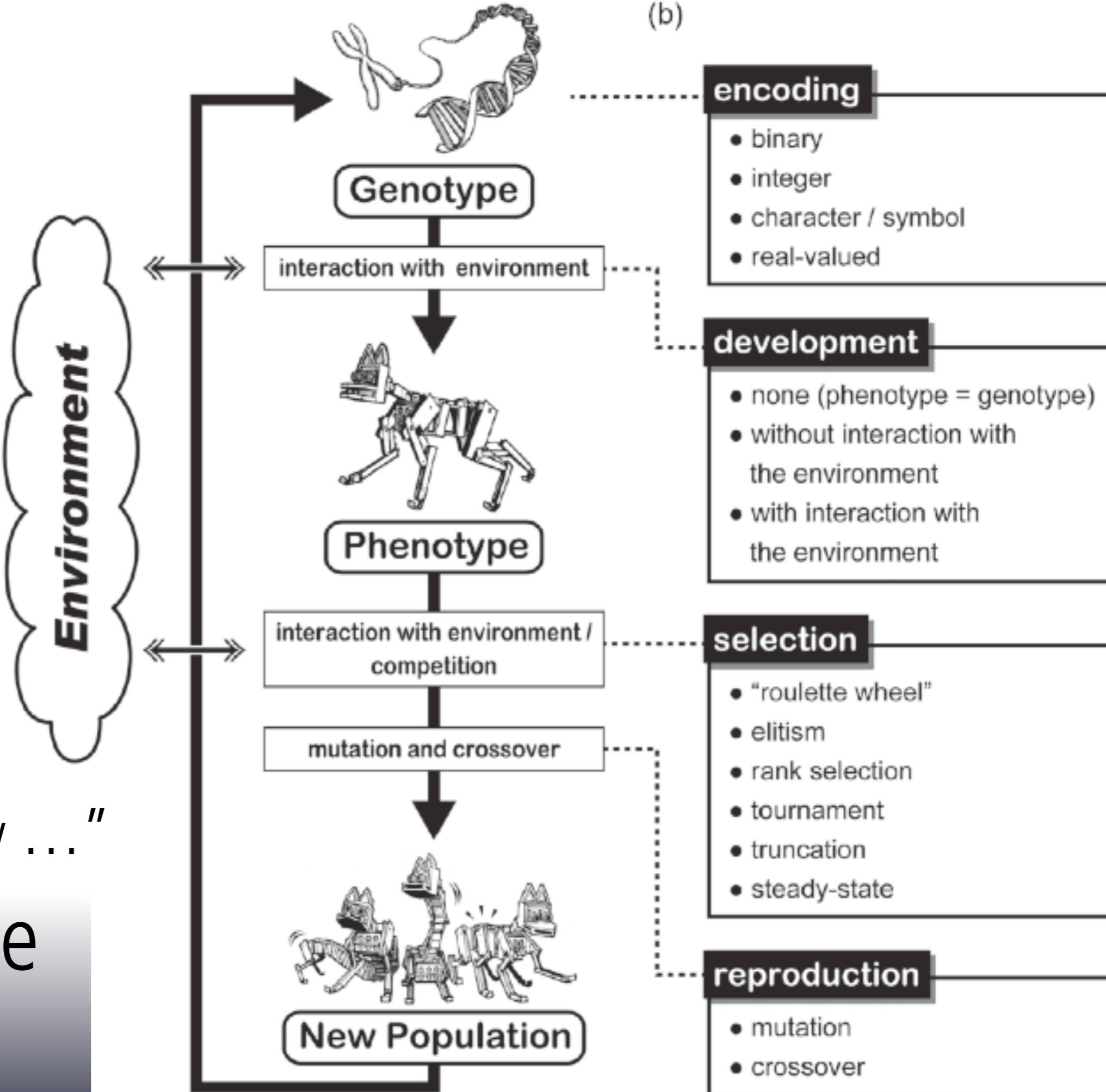
Biomechanical constraints solve one of the fundamental problems in learning, the reduction of the huge search spaces. Following the preferred trajectories increases the probability of an “interesting” interaction with the real world (i.e. where something can be learned).

Design principles for evolution

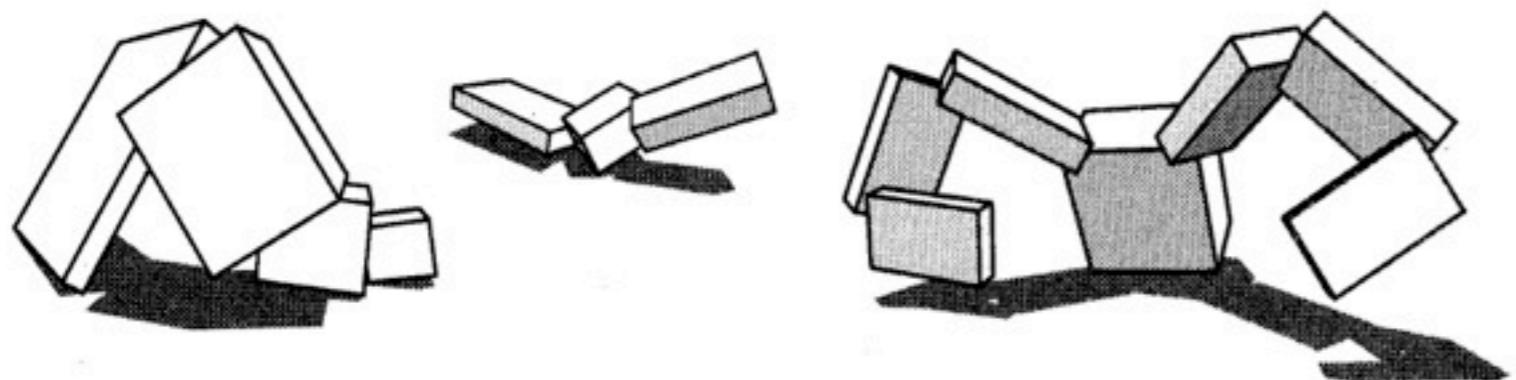
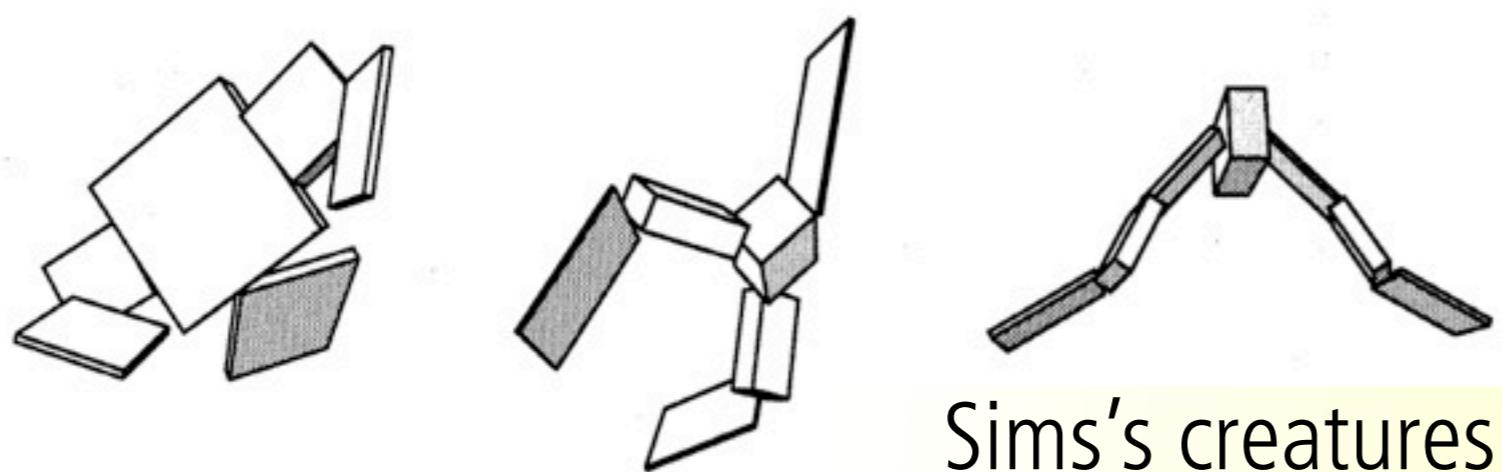
Principles for evolution	Description
Population	Population is the prerequisite for evolution to function
Cumulative selection and self-organization	Cumulative selection will produce interesting results only if evolutionary process exploits processes of self-organization
Brain-body co-evolution	“Brain” (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 77

Basic cycle for artificial evolution

from
“How the body . . .”
cumulative selection

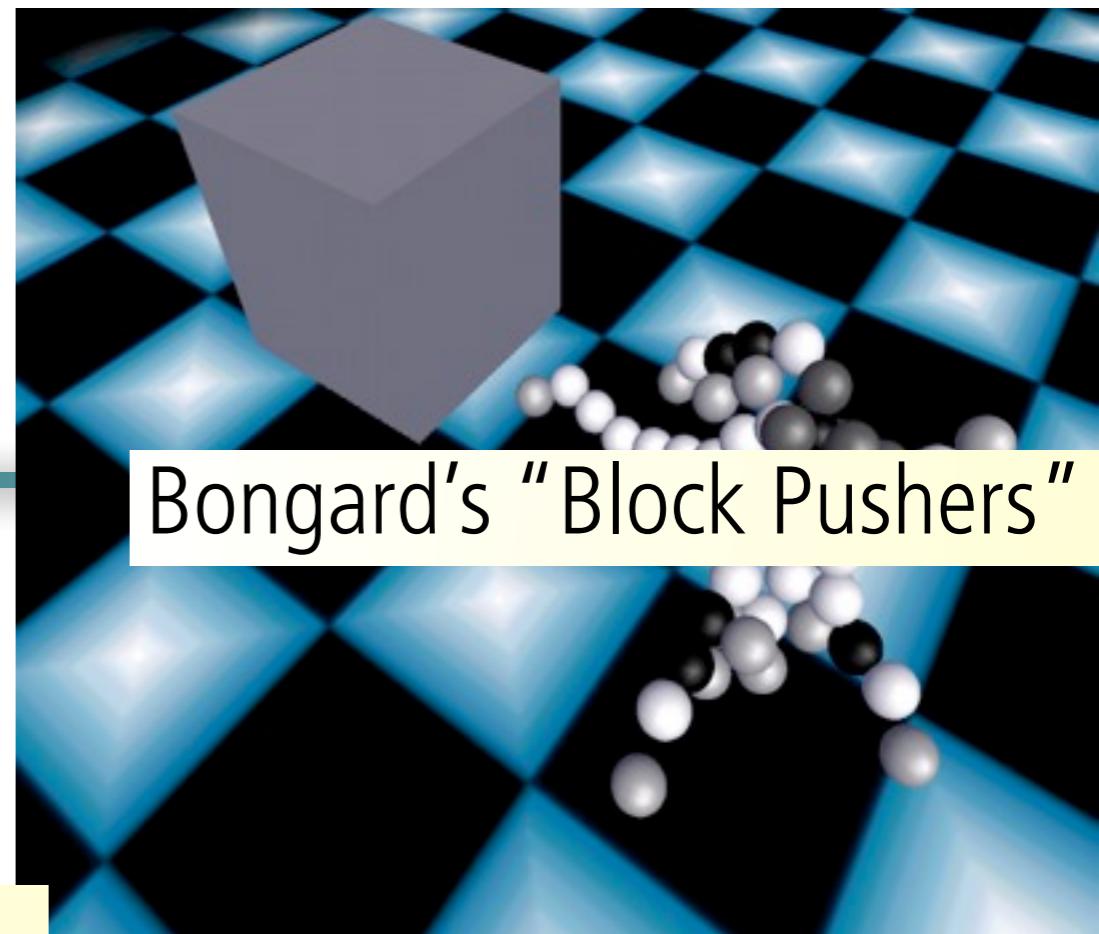


Brain-body co-evolution



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Sims and Golem: parameterization of body structure
Bongard: GRNs (Genetic Regulatory Networks)

Encoding of developmental processes in genome

rather than the structure of the organism



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Encoding developmental processes

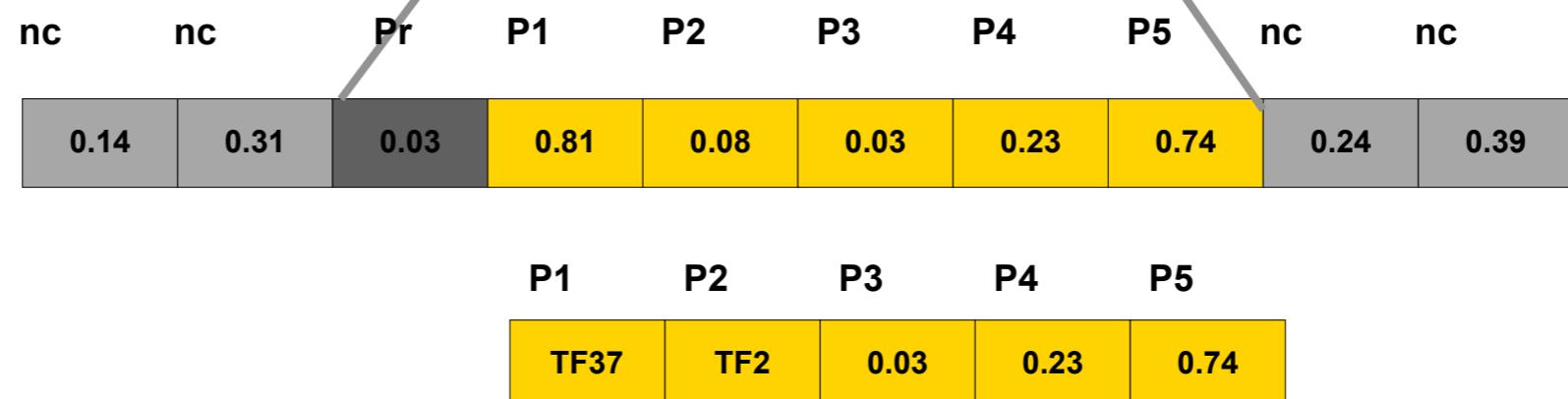
Representation of “gene”

nc: “non-coding region”



G1, G2,:
“genes” on “genome”

TF: “transcription factor”



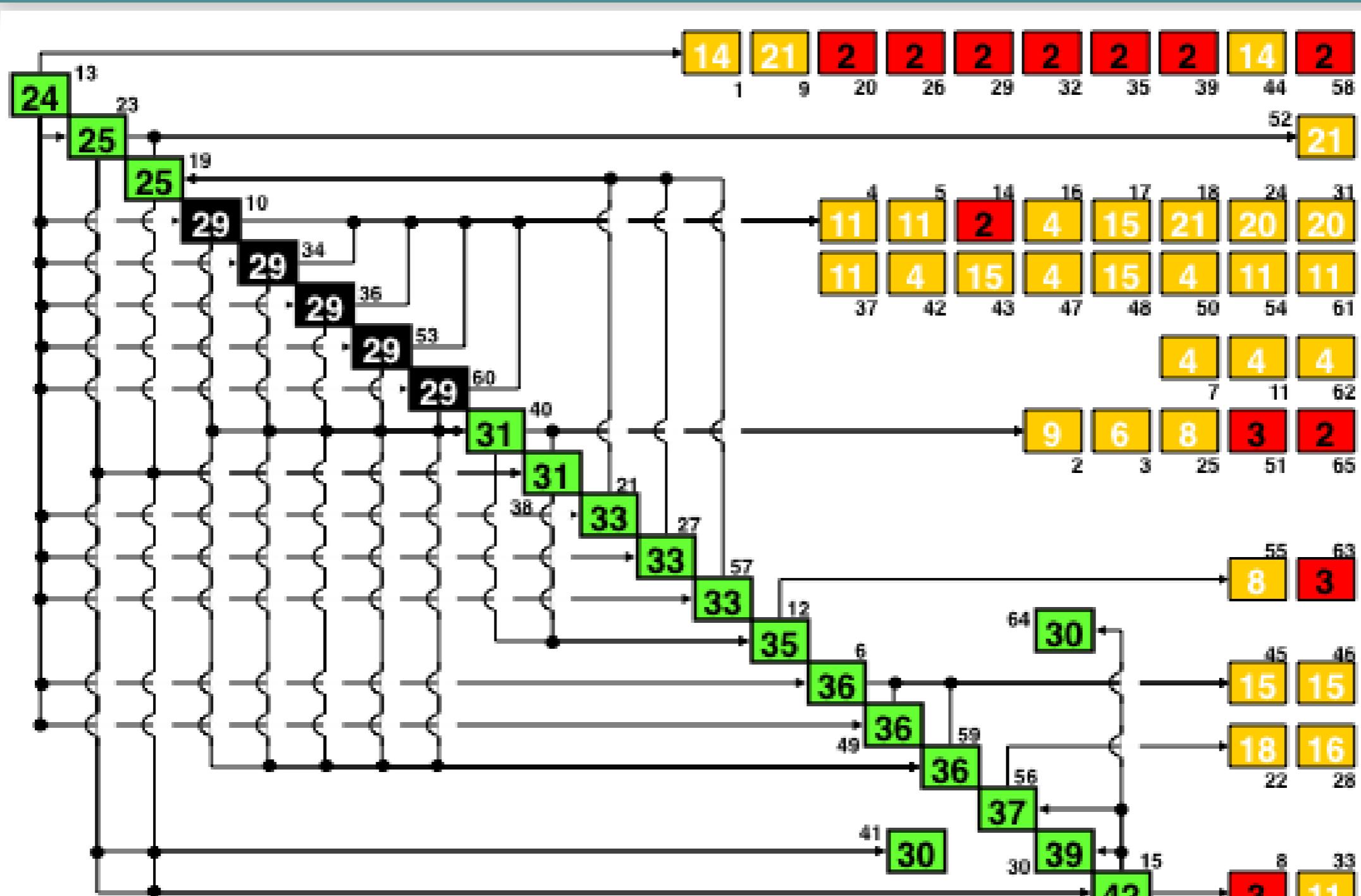
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GRN for evolved creature (Josh Bongard)



Design principles for collective intelligence

Principles for collective systems	Description
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 83

End of additional materials for self-study



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