

人
工
智
能

The
ShanghaiAI

上海
Lectures
授課



The ShanghAI Lectures

An experiment in global teaching

Rolf Pfeifer and Nathan Labhart

National Competence Center Research in Robotics (NCCR Robotics)

Artificial Intelligence Laboratory

University of Zurich

Fabio Bonsignorio

University Carlos III of Madrid and Heron Robots

Today from the University Carlos III of Madrid

Spain

欢迎您参与

“来自上海的人工智能系列讲座”

Lecture 8

**Design principles for intelligent systems
(continued)**

From Locomotion to Cognition

5 December 2013

Successes and failures of the classical approach

successes

applications (e.g.
Google)

chess
manufacturing

("controlled" artificial
worlds)

failures

foundations of
behavior

natural forms of
intelligence

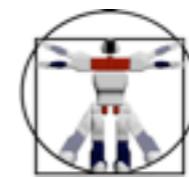
interaction with
real world



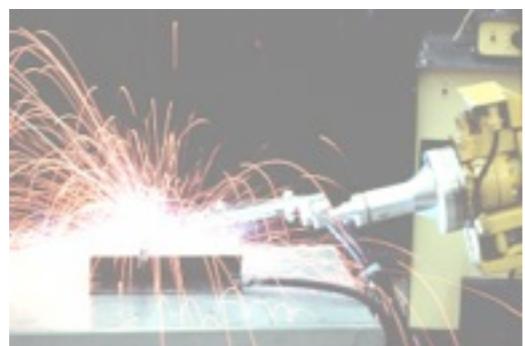
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Industrial robots vs. natural systems



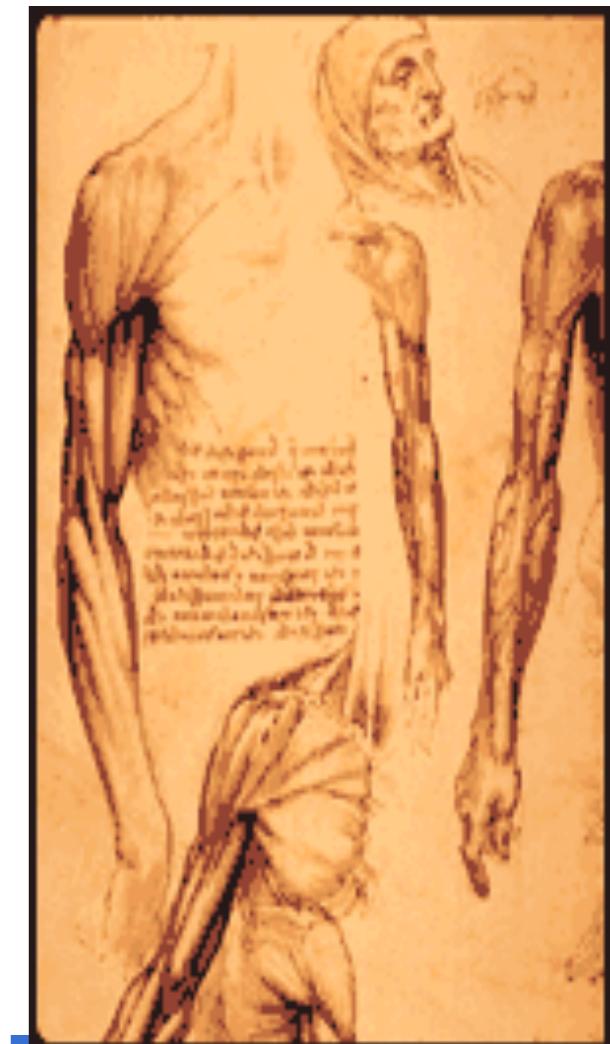
robots



no direct transfer of methods

- principles:
- low precision
 - compliant
 - reactive
 - coping with uncertainty

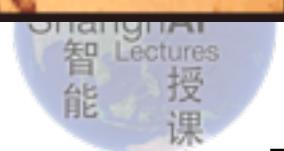
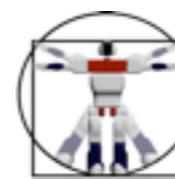
humans



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The “symbol grounding” problem

real world:
doesn't come
with labels ...

How to put the
labels??

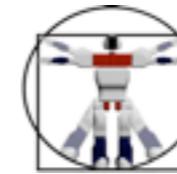
Gary Larson



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A step back: “Puppy” on a treadmill

Video “Puppy” on treadmill
slow motion

- no sensors
- no control



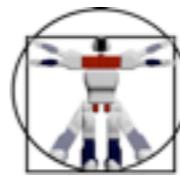
self-
stabilization



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Self-stabilization: “Puppy” on a treadmill

Video “Puppy” on treadmill
slow motion

- no sensors
- no control



principle of
“cheap
design”

self-
stabilization

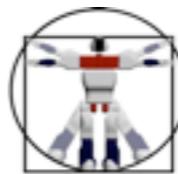


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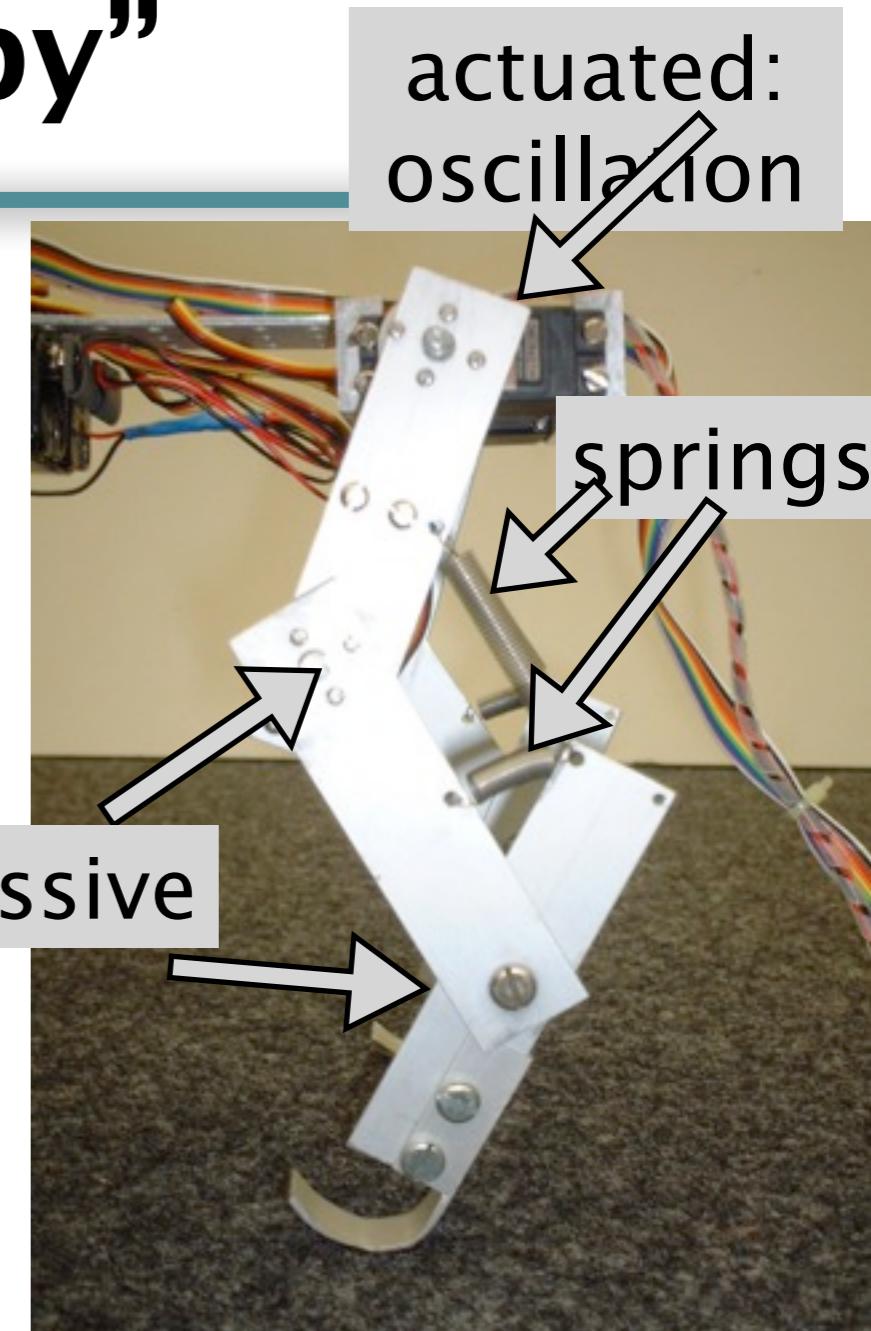


Emergence of behavior: the quadruped “Puppy”

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



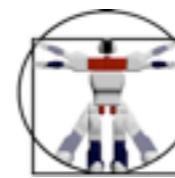
morphological
computation



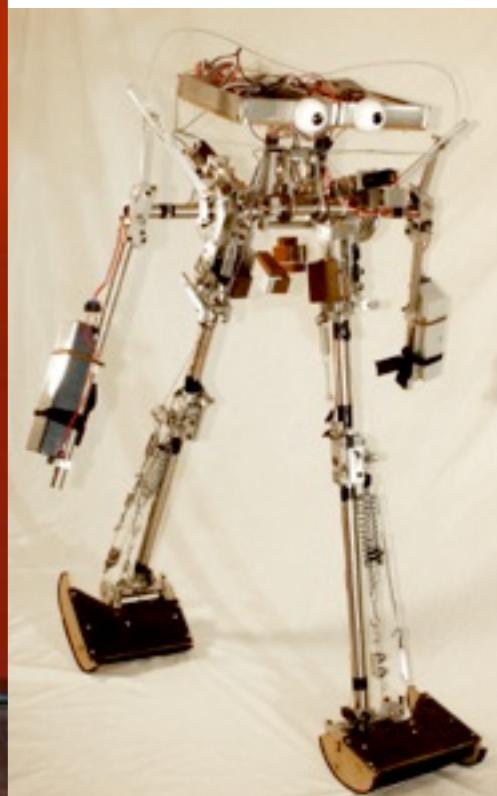
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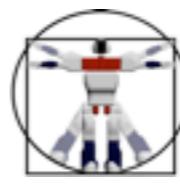
walking: GOF :-) and new designs



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Extreme case: The “Passive Dynamic

The “brainless” robot:
walking without control

Video “Passive Dynamic
Walker”

Design and construction:
**Ruina, Wisse, Collins: Cornell University
Ithaca, New York**



The Cornell Ranger



design and construction:
Andy Ruina
Cornell University

exploitation of passive dynamics

The Cornell Ranger



design and construction:
Andy Ruina
Cornell University

Video "Cornell Ranger"

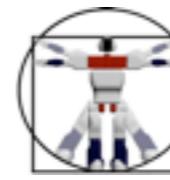
exploitation of passive dynamics



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The Cornell Ranger



conception et construction:
Andy Ruina
Cornell University

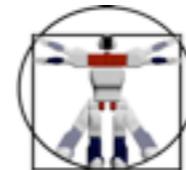
65km with one battery charge!



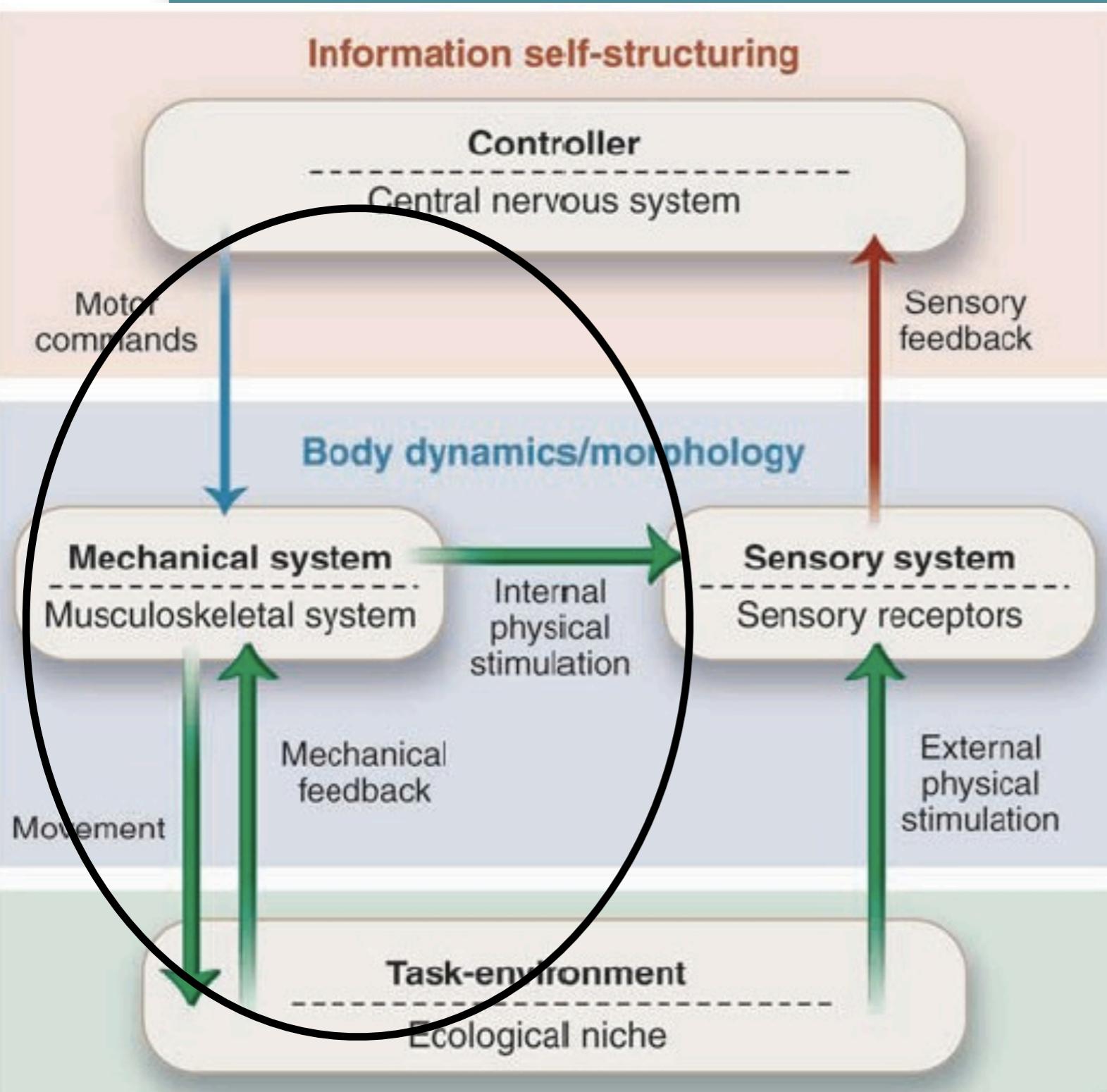
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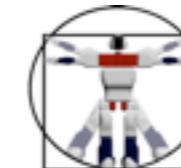
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Self-stabilization in Cornell Ranger (and Puppy!)

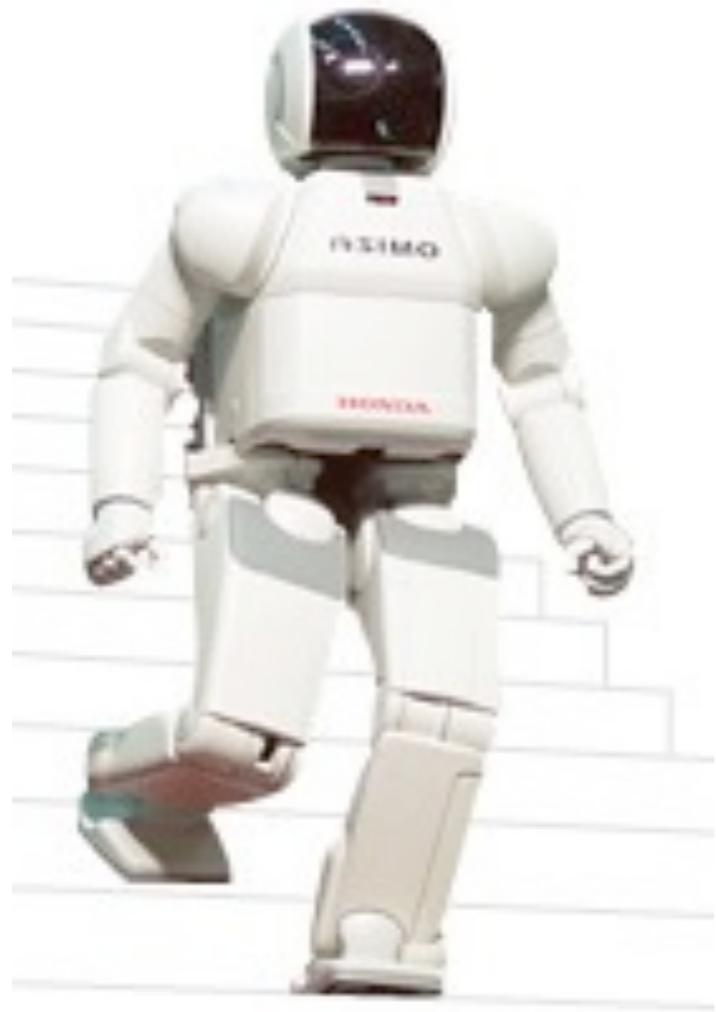


Pfeifer et al., Science,
2007



Contrast: Full control

Honda Asimo



Sony Qrio



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Redundancy principle

Redundancy/examples? →

further examples? →

Design principles for intelligent systems

Principle 1: Three-constituents principle (ecological niche, desired behaviors/tasks, agent's organization)

Principle 2: Complete-agent principle

Principle 3: Parallel, loosely coupled processes

Principle 4: Sensory–motor coordination/ information self-structuring

Principle 5: Cheap design

Principle 6: Redundancy

Principle 7: Ecological balance

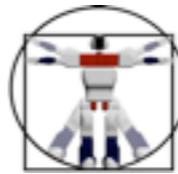
Principle 8: Value



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Principle of “cheap design”

The principle of “cheap design” states that if agents are built to exploit the properties of their ecological niche and the characteristics of the interaction with the environment, their design and construction will be much easier, or “cheaper”.

Principle of “cheap design”

The principle of “cheap design” states that if agents are built to exploit the properties of their ecological niche and the characteristics of the interaction with the environment, their design and construction will be much easier, or “cheaper”.

The value principle

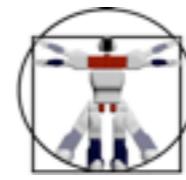
The value principle states that intelligent agents are equipped with a ‘value system’ which constitutes a basic set of assumptions about what is good for the agent.



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Principle of “ecological balance”

balance in complexity

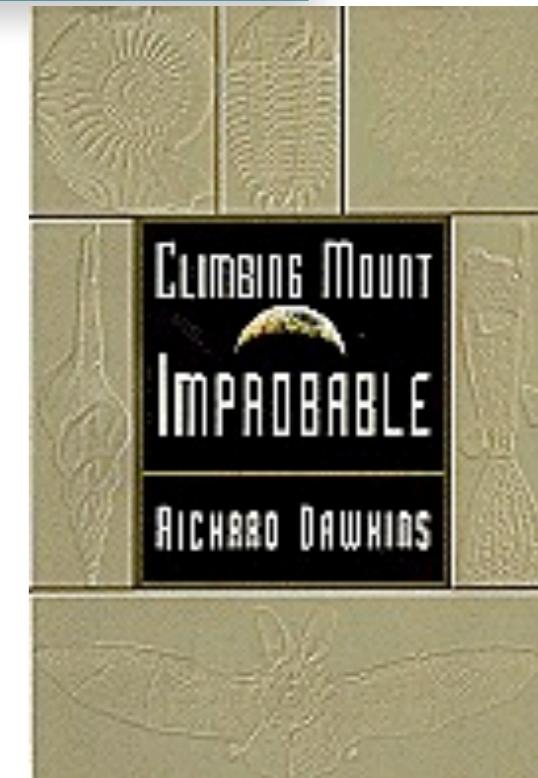
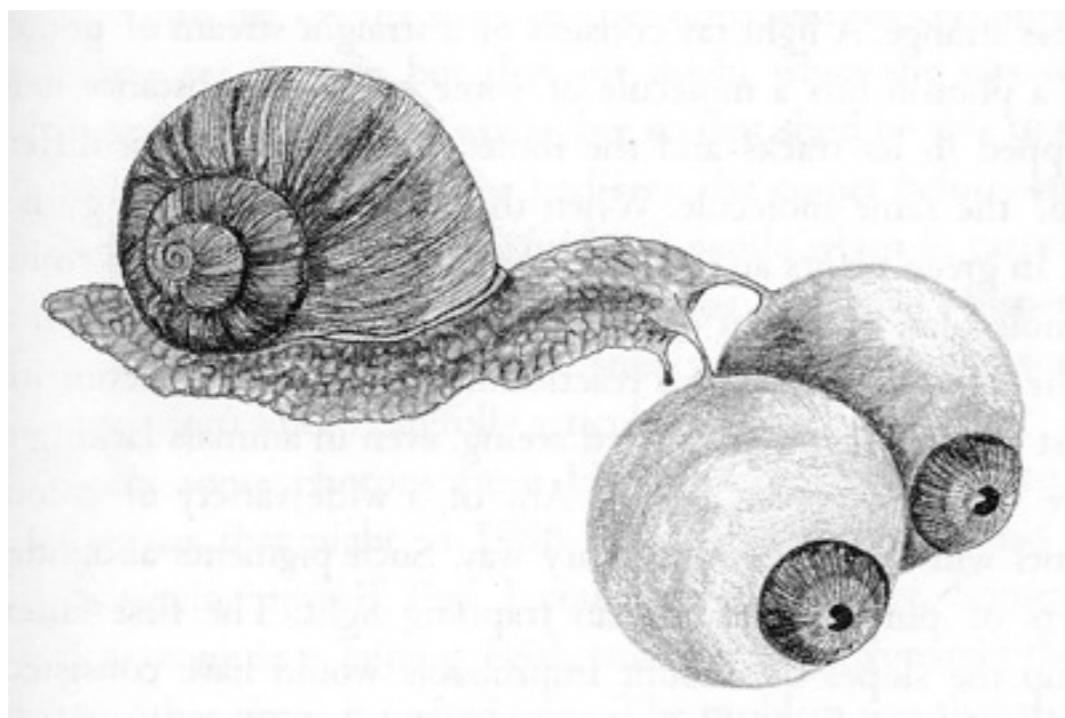
given task environment: match in complexity of sensory, motor, and neural system

balance / task distribution

brain (control), morphology, materials, and interaction with environment

Richard Dawkins's snail with giant eyes

**ecologically _unbalanced_
system**



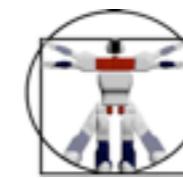
Author of:
“The selfish gene” and
“The blind watchmaker”



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Task distribution

**between brain, morphology, materials,
and environment**

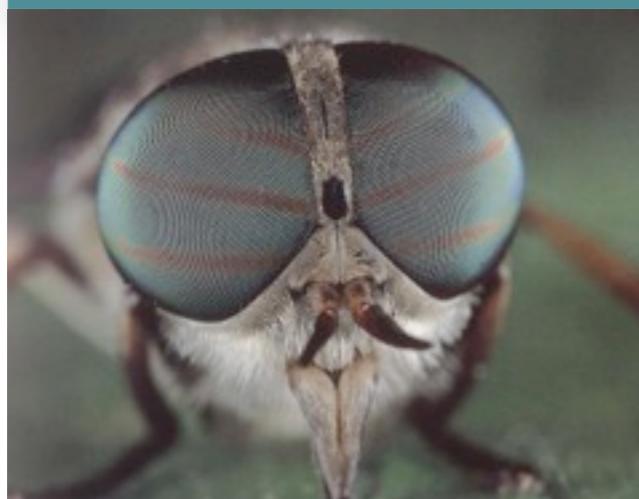
extreme case: Passive Dynamic Walker

Puppy, Stumpy

**Animals, humans: dynamic change of
muscle stiffness**

Loosely swinging arm (later today)

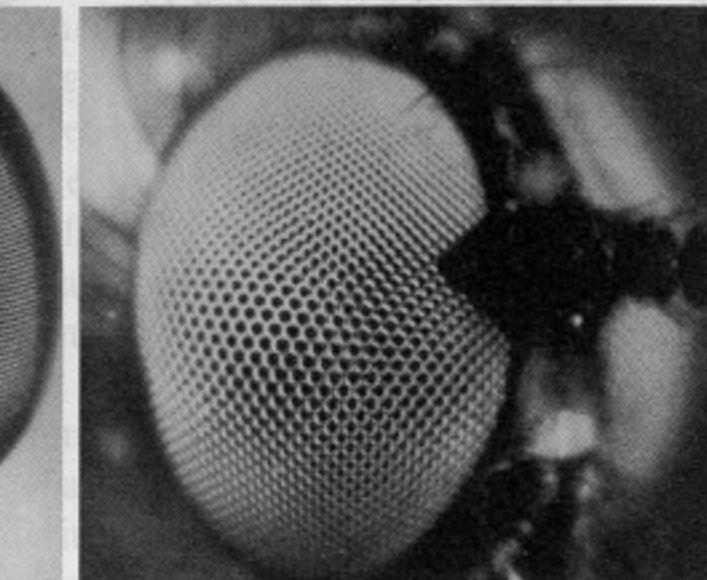
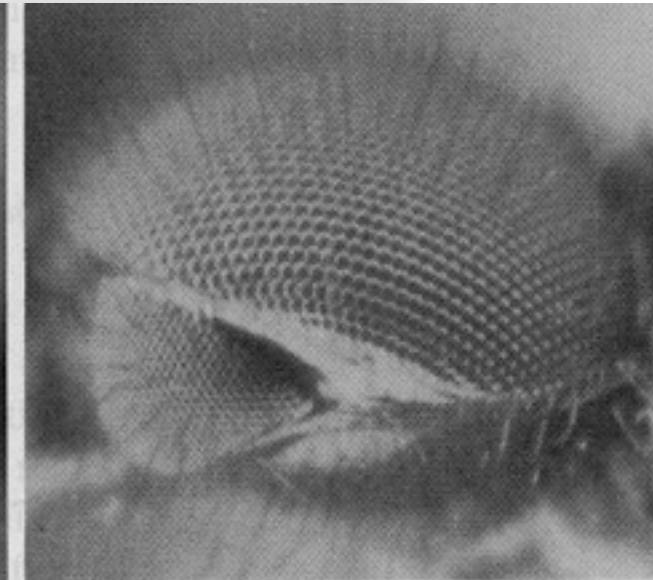
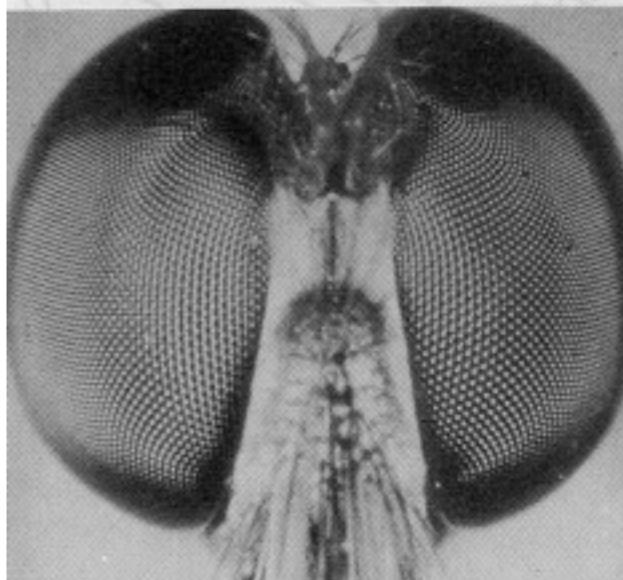
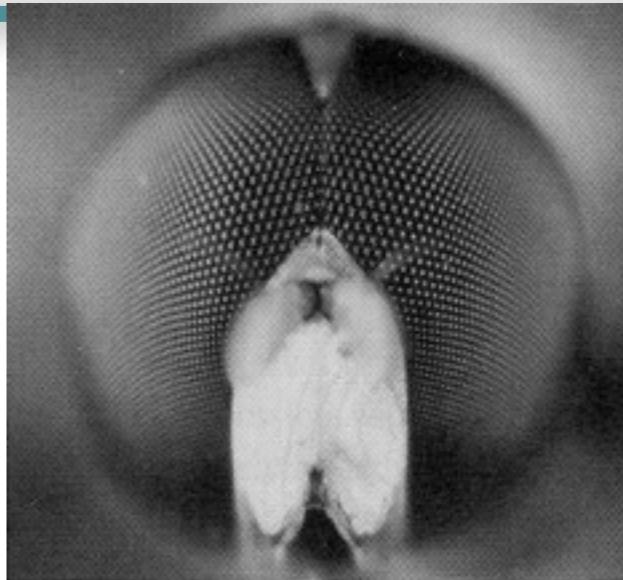
Different morphologies of insect eyes



housefly



large variation of
shapes



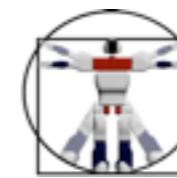
honey bee



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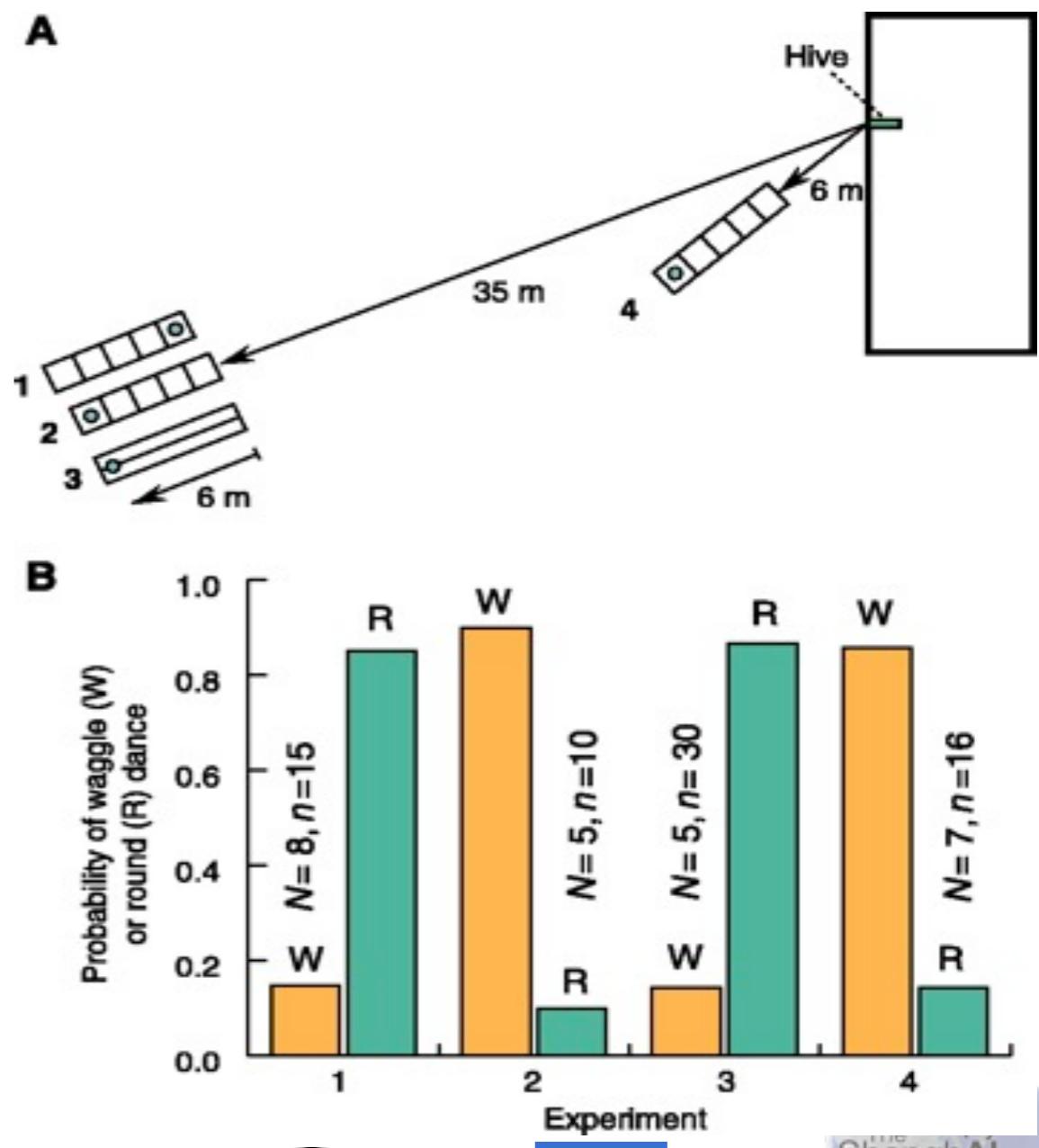
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Optic flow-based odometry in bees

Srinivasan's fascinating experiments (2000)

(A) Layout for experiments using tunnels. Each tunnel represents a separate experiment (1, 2, 3, or 4). The dot in the tunnel shows the position of the feeder in each case. (B) Probability of waggle (W) round (R) dance for experiments 1 to 4. N and n represent the numbers of bees and dances analyzed, respectively in each experiment. Science, 287, p. 852, 2000.



Design principles for intelligent systems

Principle 1: Three-constituents principle

Principle 2: Complete-agent principle

Principle 3: Parallel, loosely coupled processes

Principle 4: Sensory-motor coordination/
information self-structuring

Principle 5: Cheap design

Principle 6: Redundancy

Principle 7: Ecological balance

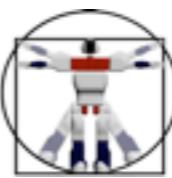
Principle 8: Value



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The principle of sensory-motor coordination

induction of structured sensory stimulation through sensory-motor coordinated action

principle of information self-structuring: effect

Grasping an object

- many ways
- winding spring (effort)
- release
- exploitation by brain
 - “cheap design”, exploitation of material properties, “free”
 - “ecological balance”: outsourcing of functionality to morph. and material

Grasping an object

- induction of sensory stimulation
- dependence on
 - morphology: high density of touch, temperature, vibration sensors in hand
 - actuation: sensory–motor coordination

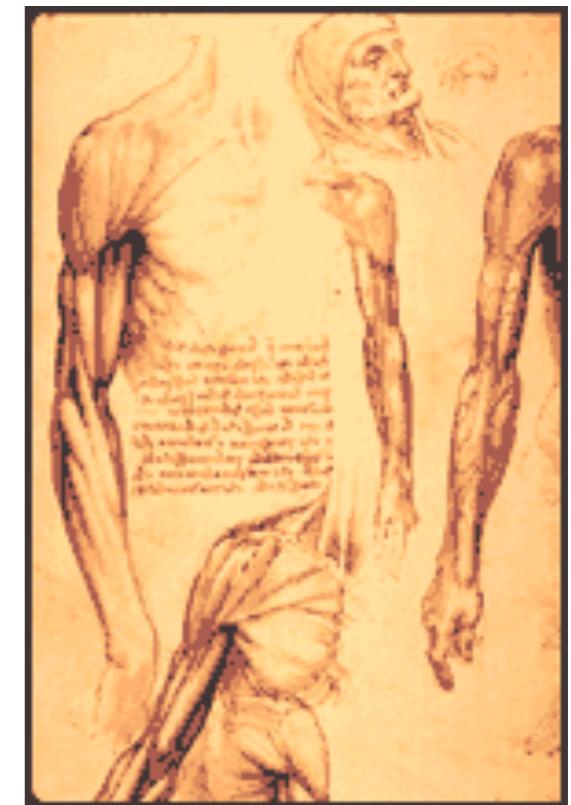


- induction of correlations

“raw material” for information processing of brain

Loosely swinging arm

- complex trajectory of hand
- simple control (“cheap design”, “ecological balance”)
- exploitation of morphology / materials (biomechanical constraints)



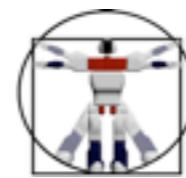
control
“decentralized”
“free”



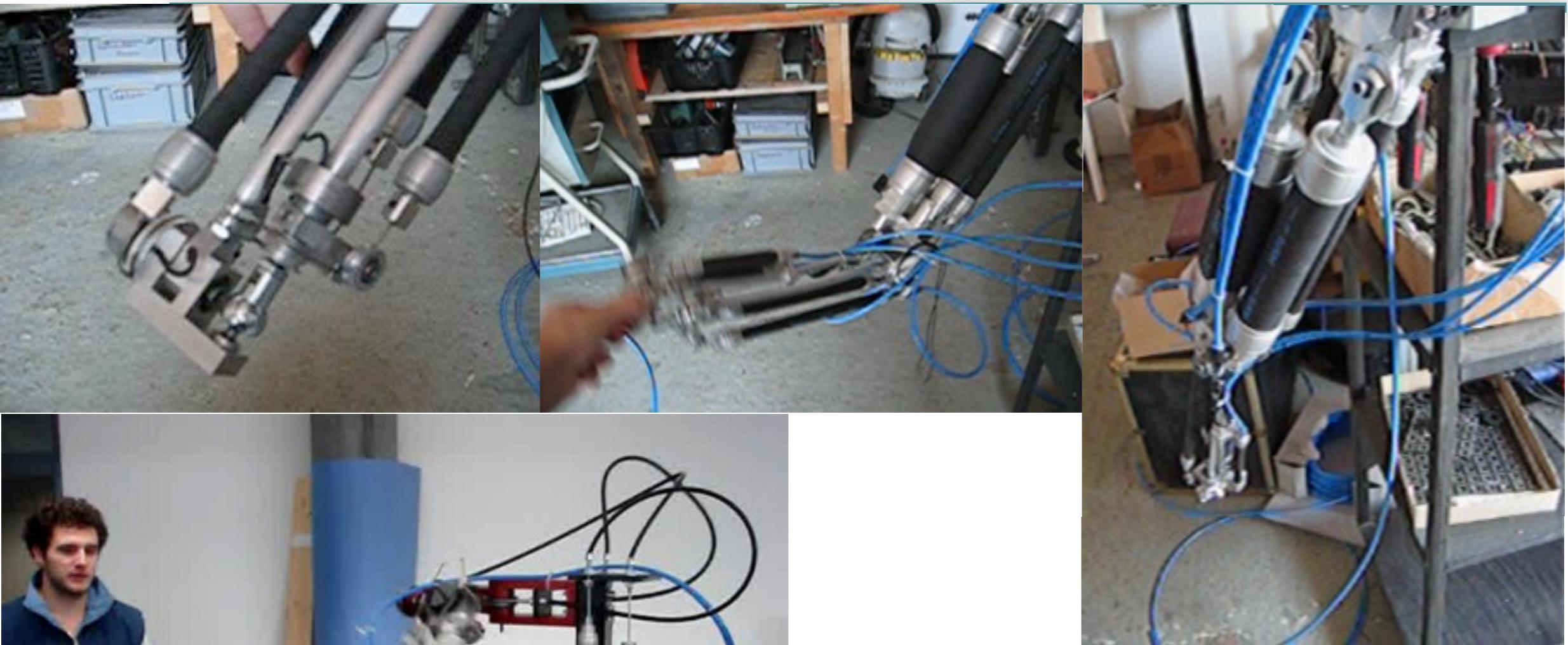
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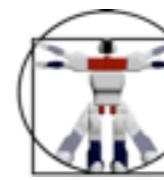
Anthropomorphic arm with pneumatic



Design and construction:
Raja Dravid, Max Lungarella, Juan
Pablo Carbajal, AI Lab, Zurich



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Anthropomorphic arm with pneumatic

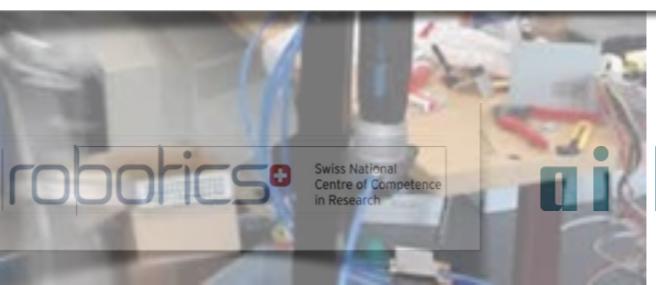
Video “Heavily
swinging arm”

Video “Loosely
swinging arm”

Video “Passive
compliance”

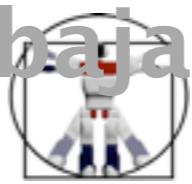


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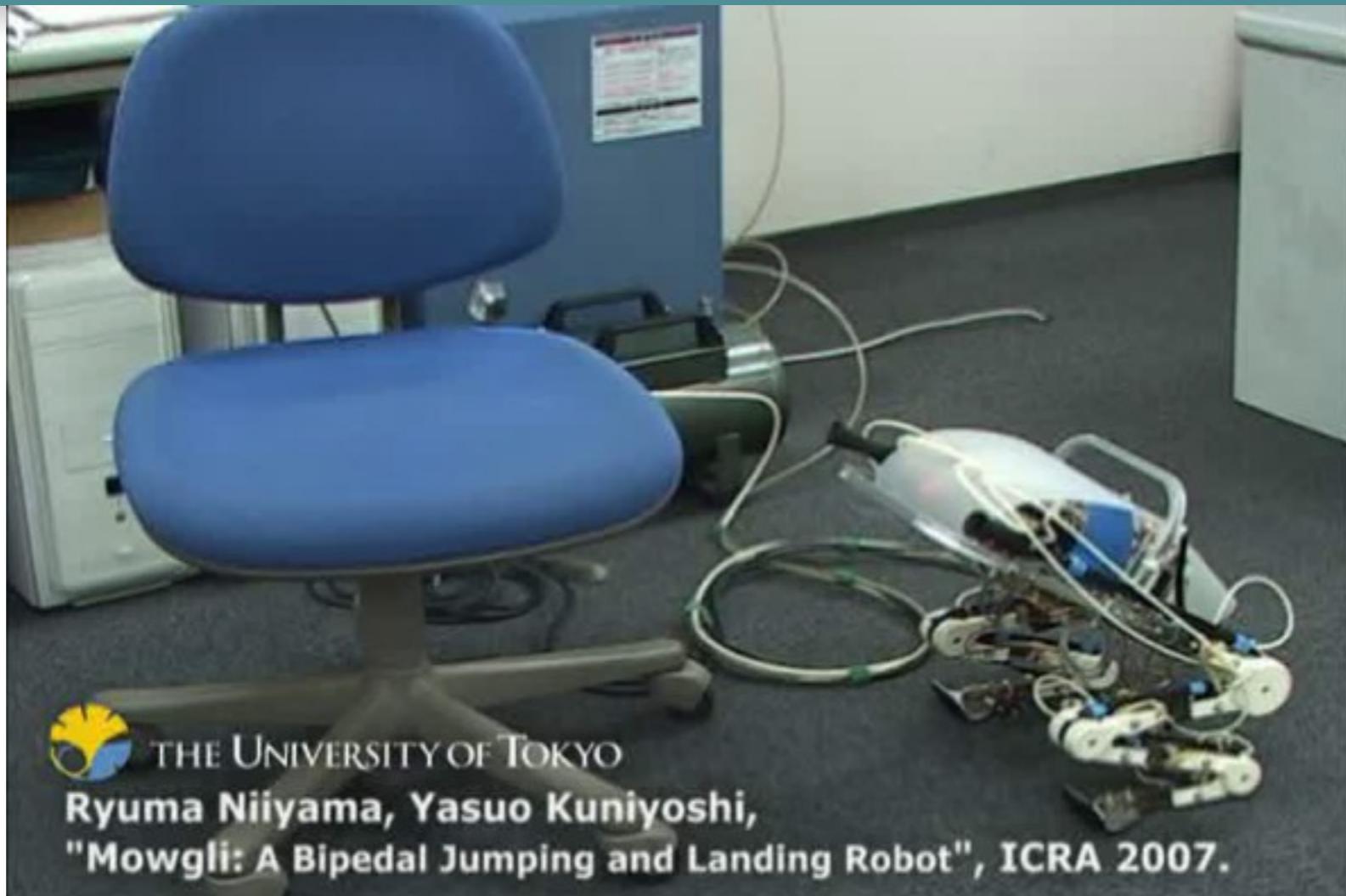


JUAN
Pablo Carballo, AI Lab, Zurich

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Robot Frog "Mowgli" driven by pneumatic



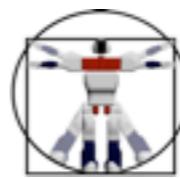
Design and construction:
Ryuma Niiyama, Yasuo Kuniyoshi, The University of
Tokyo



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Robot Frog "Mowgli" driven by pneumatic



Video "Mowgli"



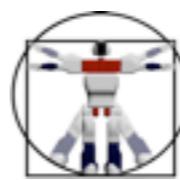
Design and construction:
Ryuma Niiyama, Yasuo Kuniyoshi, The University of



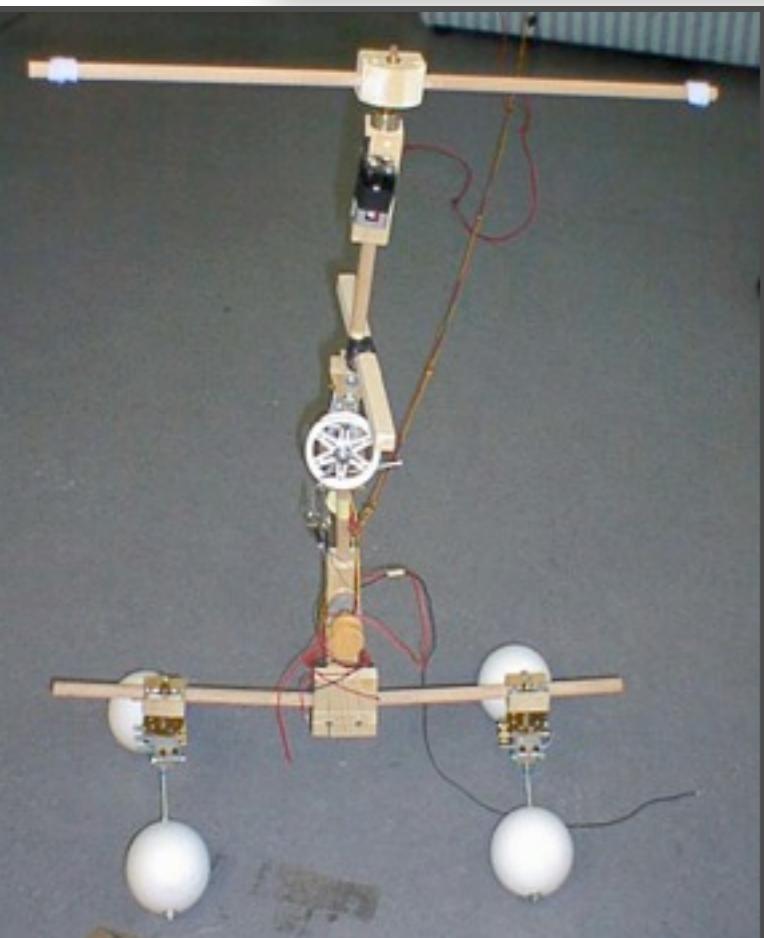
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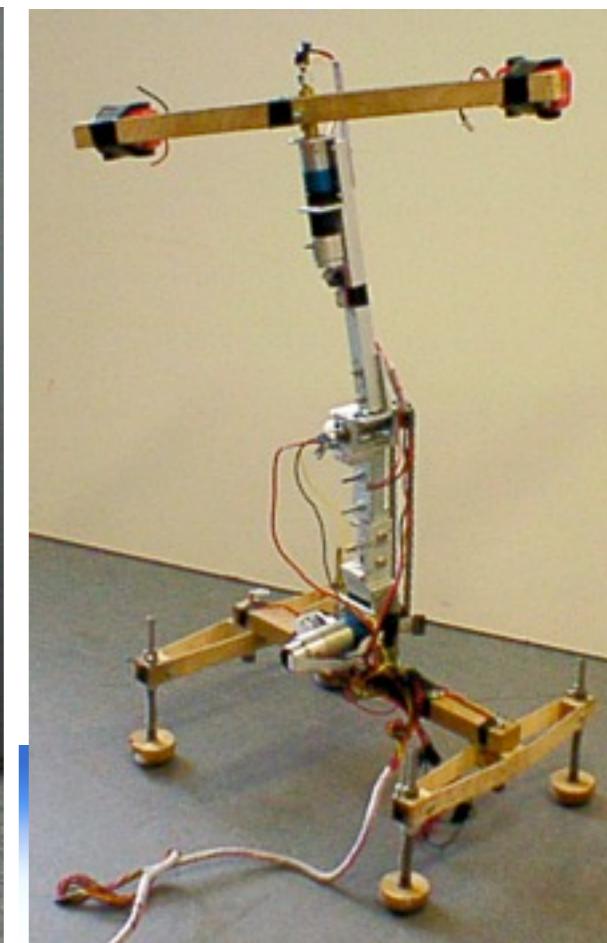
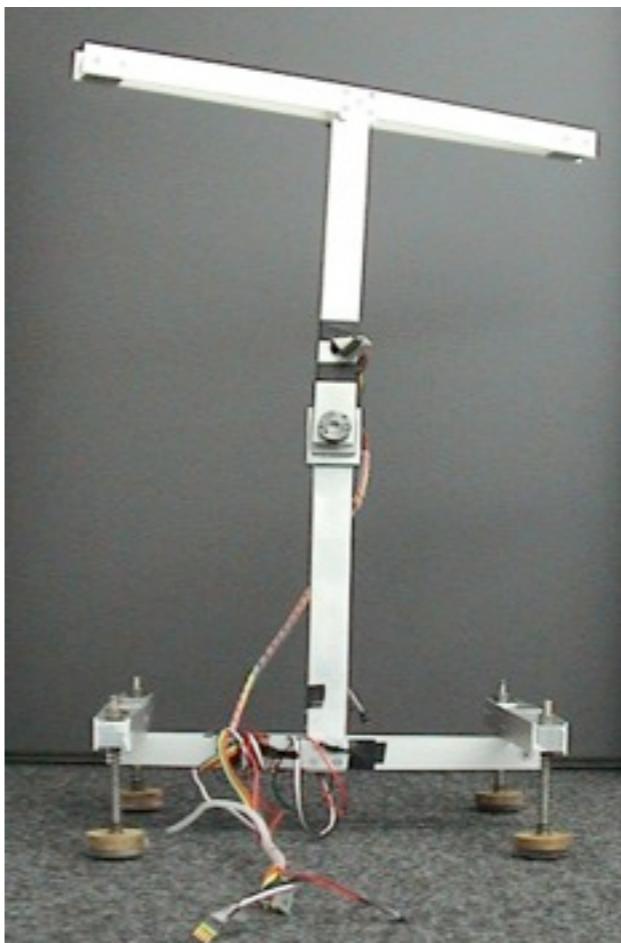
“Stumpy”: task distribution



almost brainless: 2 actuated joints
springy materials
surface properties of feet

Design and construction: **Raja Dravid,**
Chandana Paul, Fumiya Iida

self-stabilization



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Cognition: Memory: Are 'symbols' always needed?

which part of diagram relevant?

→

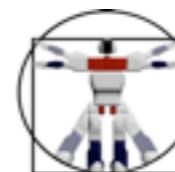
memory for walking?



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Water fountain Where is the memory for shape?

clear structure visible
underlying mechanism?



Where is the “structure” stored?
– what can we learn for human
memory?



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Ashby's concept of “memory as a theoretical construct”

W. Ross Ashby (1956). An introduction to cybernetics.



a.

copyright: Isabelle Follath, Zurich

b.

Where does ‘symbols’ come from?: physical dynamics and information processing

- morphology and materials
- orchestration control
- exploration
- preferred trajectories from biomechanical constraints
- induction of patterns of sensory stimulation in different sensory channels
- sensory–motor coordination → induction of information structure

The “story”: physical dynamics and information

- good “raw material” for brain
- cross-modal association, learning, concept formation
- extraction of mutual information → prediction
(expectations: crucial for motor control)
- categorization (fundamental for cognition)

Sensory-motor coordination ("active perception")

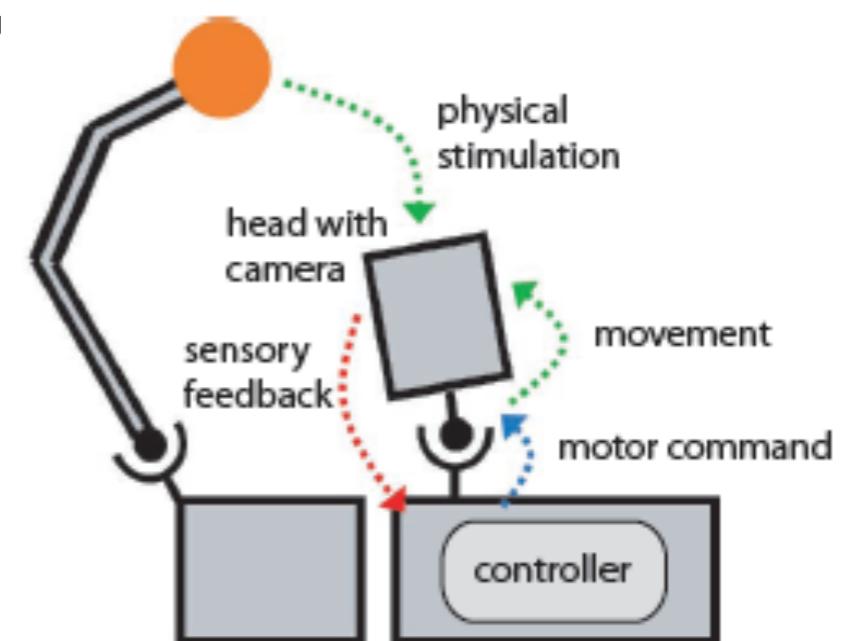
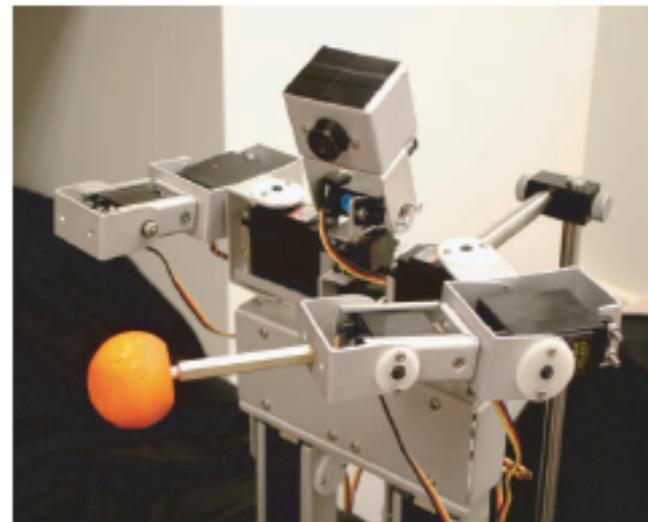
"We begin not with a sensory stimulus, but with a sensory-motor coordination [...] In a certain sense it is the movement which is primary, and the sensation which is secondary, the movement of the body, head, and eye muscles determining the quality of what is experienced. In other words, the real beginning is with the act of seeing; it is looking, and not a sensation of light." ("The reflex arc concept in psychology," John Dewey, 1896)

"Since all the stimulations which the organism receives have in turn been possible only by its preceding movements which have culminated in exposing the receptor organ to external influences, one could also say that behavior is the first cause of all the stimulations." ("The structure of Behavior," Maurice Merleau-Ponty, 1963)

Information self-structuring

Experiments:

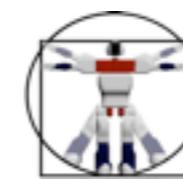
Lungarella and Sporns, 2006
Mapping information flow
in sensorimotor networks
PLoS Computational Biology



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Quantitative measures

entropy: disorder, information

$$H(X) = - \sum_i p(x_i) \log p(x_i)$$

mutual information: statistical dependency

$$MI(X, Y) = H(X) + H(Y) - H(X|Y) = - \sum_i \sum_j p(x_i, y_j) \log \frac{p(x_i)p(y_j)}{p(x_i, y_j)}$$

integration: global statistical dependence

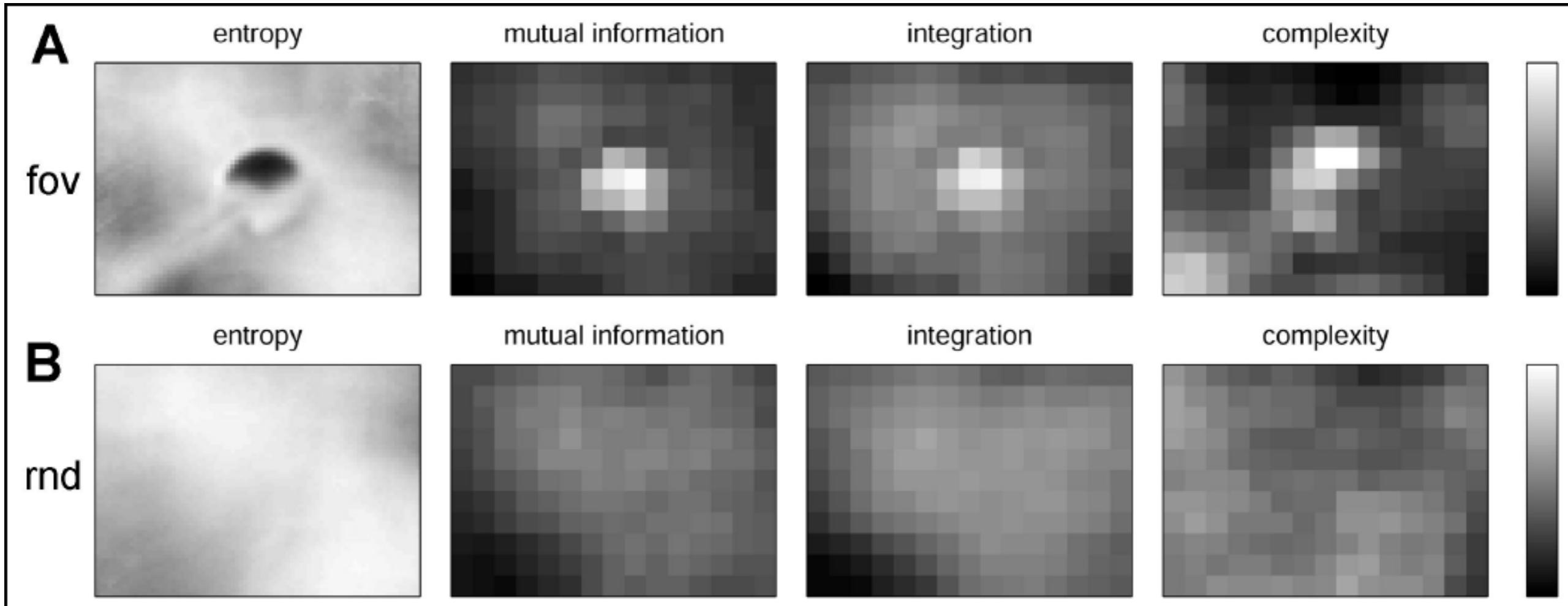
$$I(X) = \sum_i H(x_i) - H(X)$$

complexity: co-existence of local and global structure

$$C(X) = H(X) - \sum_i H(x_i | X - x_i).$$

from: **Tononi, Sporns, and Edelman, PNAS, 1994, 1996**

Results: foveation vs. random



entropy

mutual
information

integration
(over patch)

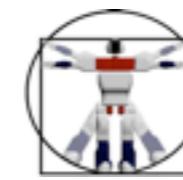
complexity
(over patch)



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Information Driven Self Organization

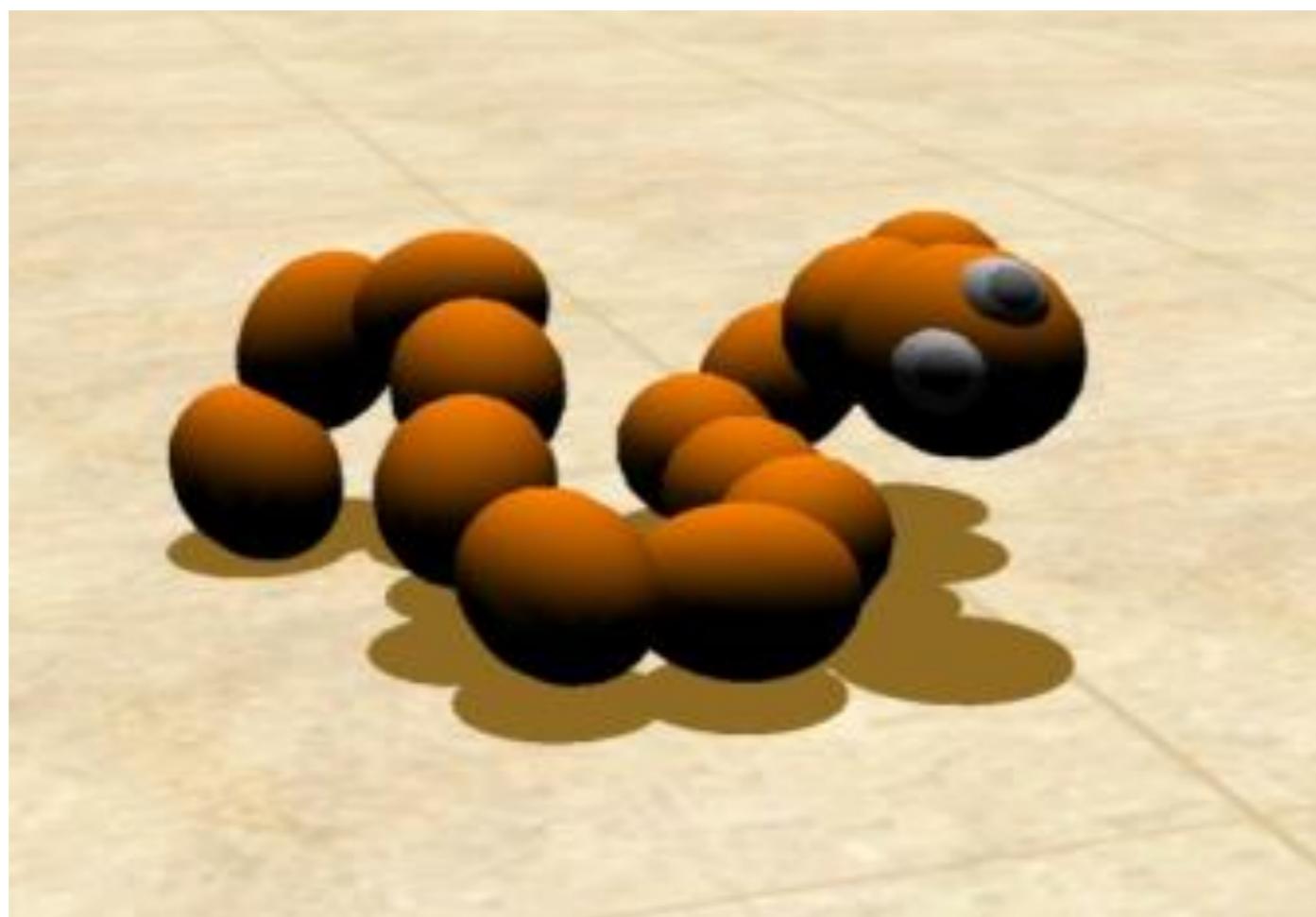
Why not using information metrics to implement an emergent control?

Several researchers have shown the importance of Information Driven Self Organization (IDSO).

In particular Prokopenko, Ralf Der and others have shown simple demonstrators, mainly in simulation, with snake-bots, humanoids and grasping systems.

These approaches seem very promising.

Snakebot



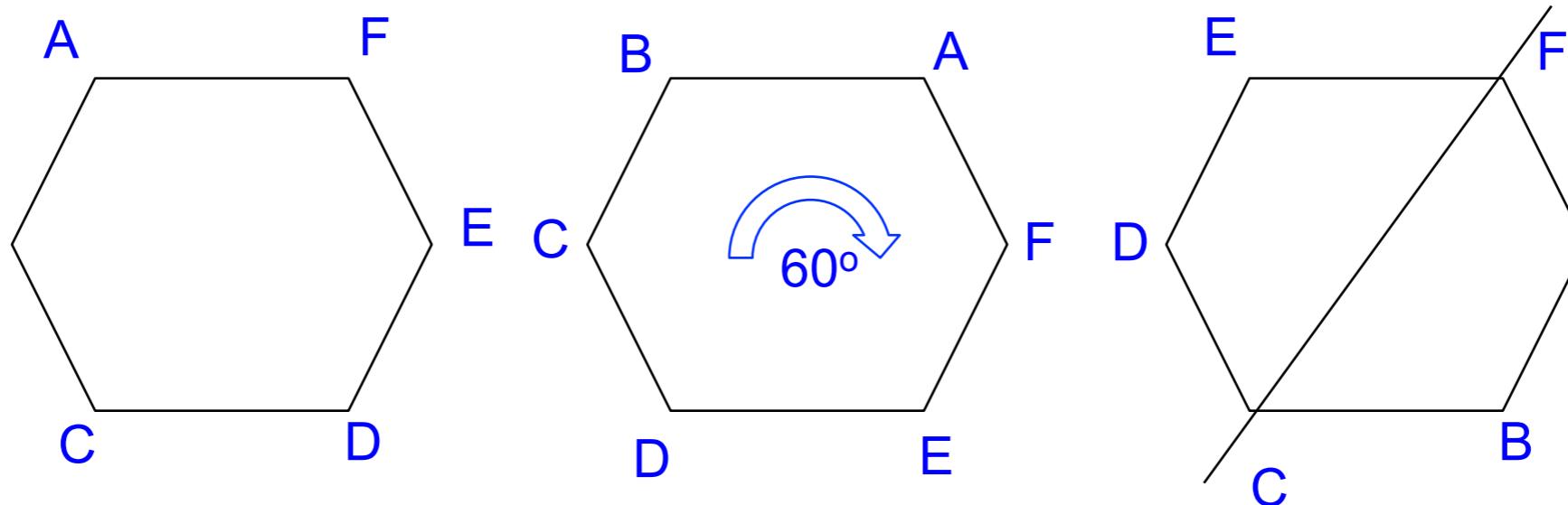
see: Tanev et. al, IEEE TRO, 2005

Maybe not GOF Euclidean space? :-)

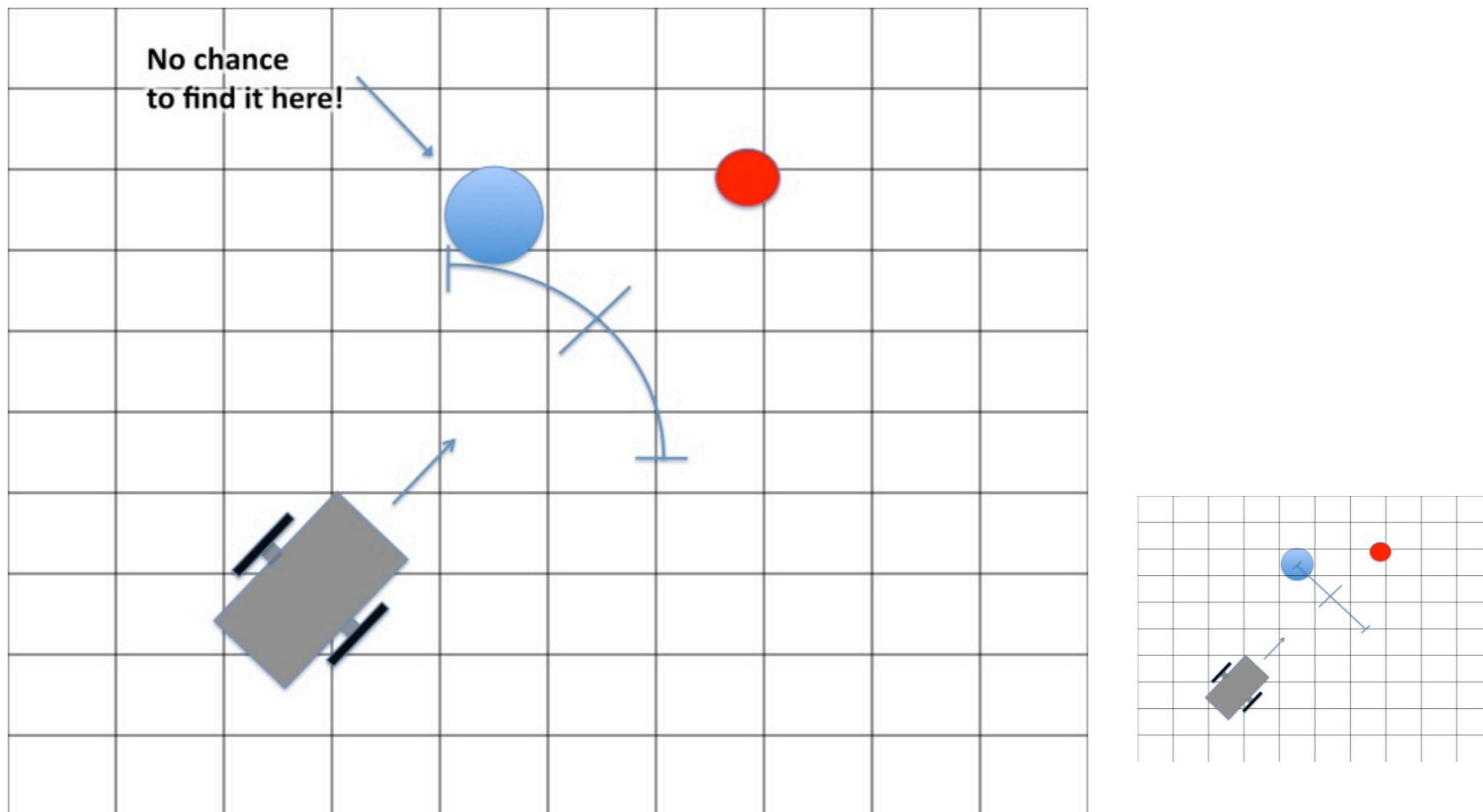
Unfortunately IDSO approaches are computationally heavy.

In [Chirikjian, 2010; Bonsignorio, 2010] it is argued that the recognition of the Lie group structure of the mobility space may help planning methods based on searching in the configurations space.

In Bonsignorio, Artificial Life, 19(2), 2013 the possibilities of the IDSO on Lie groups are shown from a theoretical standpoint.



Maybe not GOF Euclidean space? :-)



see: Bonsignorio, Artificial Life, 2013

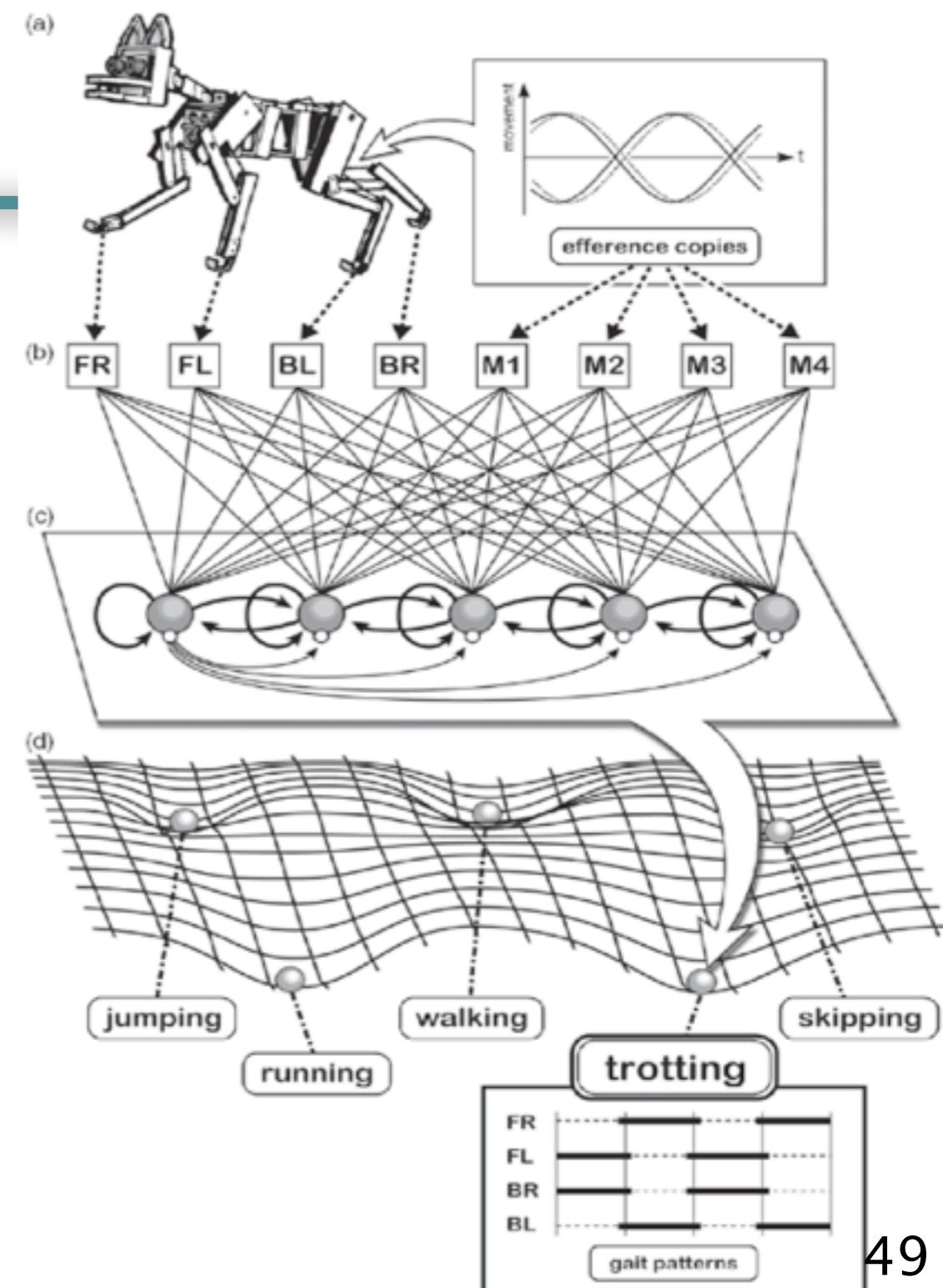
Building grounded symbols (labeling!)

Human: grasping object — patterns of sensory stimulation “match” morphology of agent

Puppy: patterns from pressure sensors or joint angle trajectories: match morphology of agent



grounding for “high-level” cognition



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End of lecture 8

Thank you for your attention!

stay tuned for lecture 9

“Towards a theory of intelligence”



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rob

