

AI : The Standpoint of Developmental Robotics



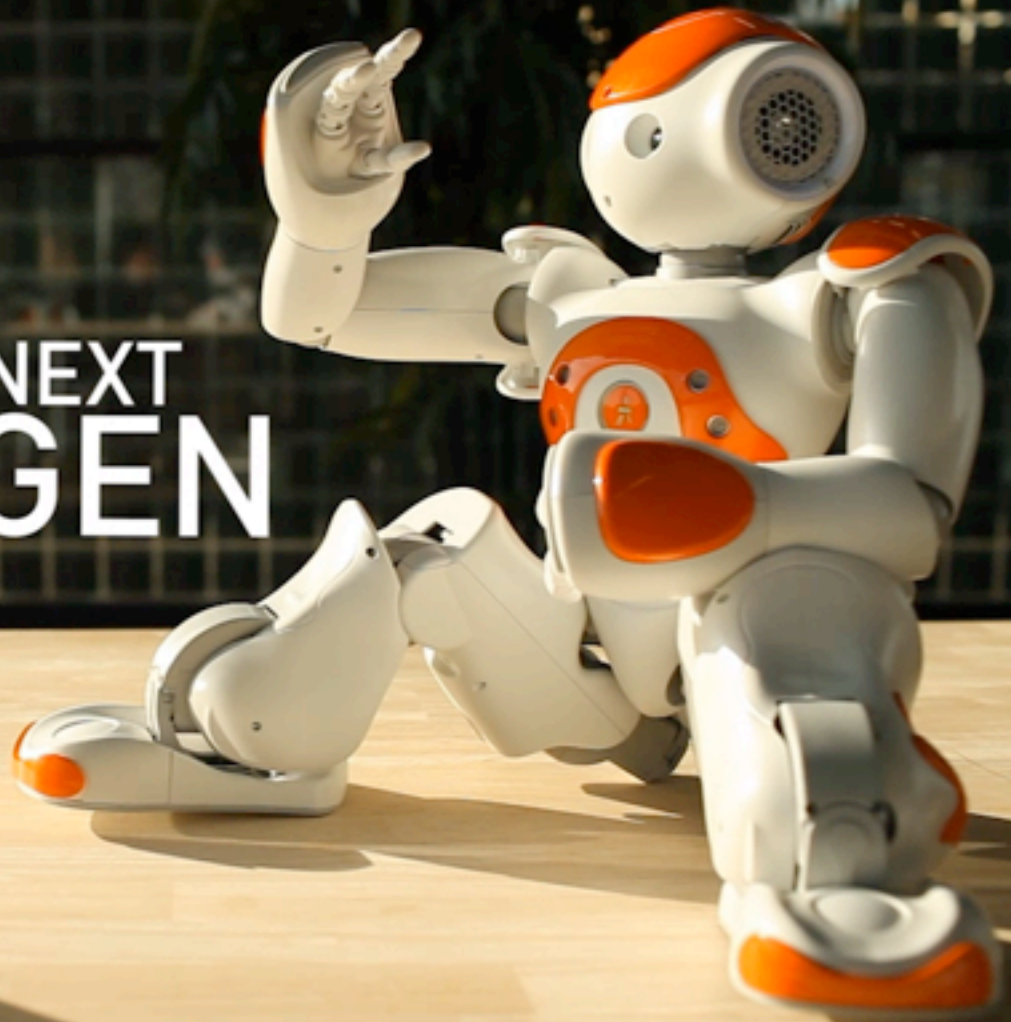
Philosophy & Theory of AI - 2013

Jean-Christophe Baillie
Aldebaran's AI Lab



NMO

NEXT
GEN



About Aldebaran Robotics

California



Boston



Paris



Shanghai



Tokyo



350

Aldebaranians
worldwide

+800 labs,

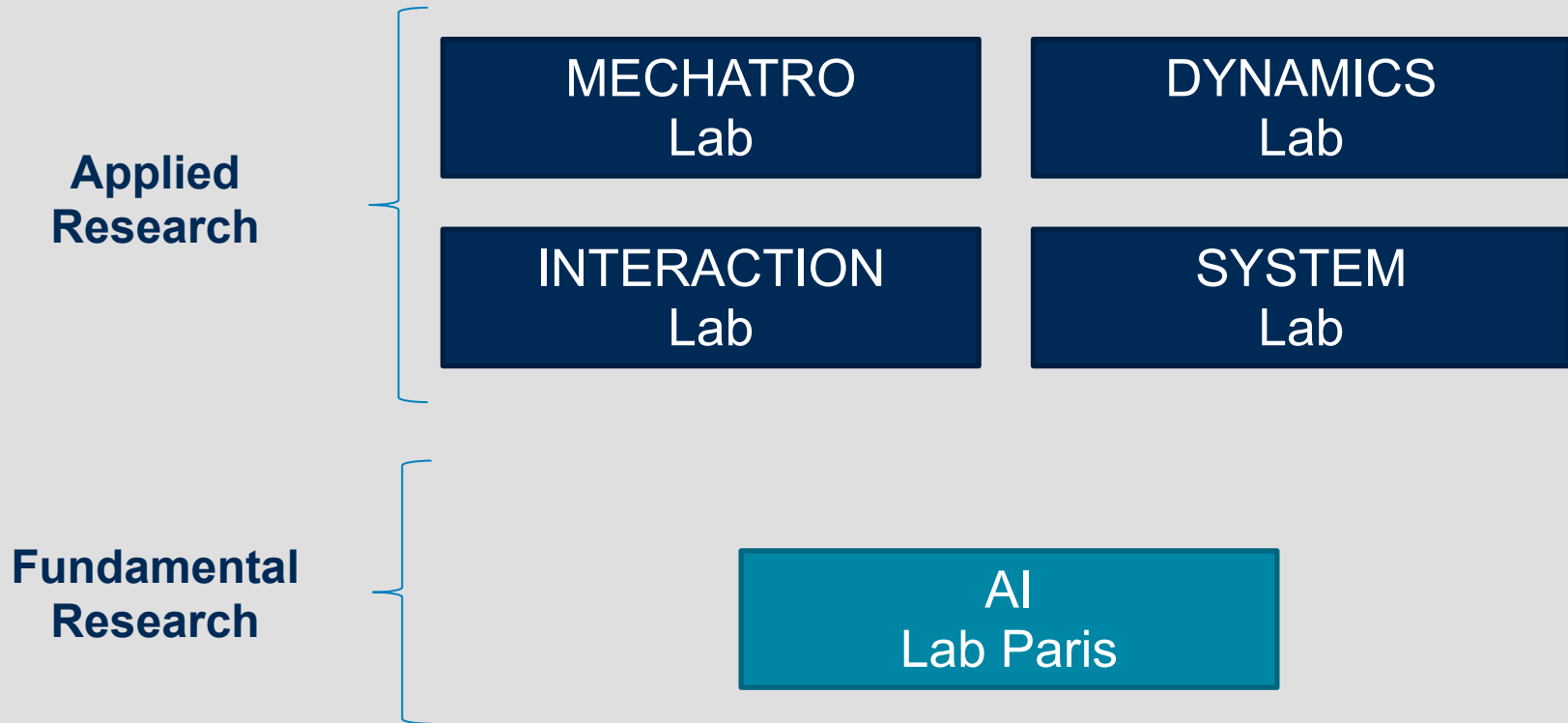
universities & High
Schools as customers

+5000 users

+3500

NAOs in use

Research Groups: the A-Lab



AI Lab: Where do we stand?

Three approaches to AI research:

AI Lab: Where do we stand?

Three approaches to AI research:

- **Applied AI:** solve engineering problems using learning methods, heuristics, etc.

AI Lab: Where do we stand?

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AI Lab: Where do we stand?

Three approaches to AI research:

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- **Theory of Intelligence:** human level intelligence and embodiment. Must *explain* intelligence with *models*.

AI Lab: Where do we stand?

Three approaches to AI research:

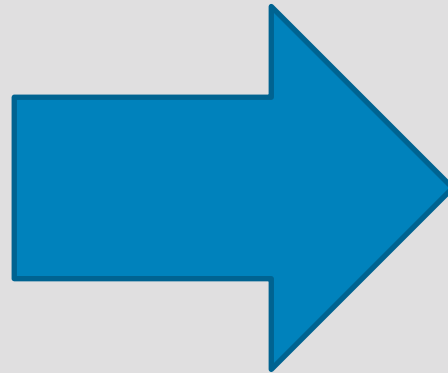
- **Applied AI:** solve engineering problems using learning methods, heuristics, etc.
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AI Lab

Developmental Robotics

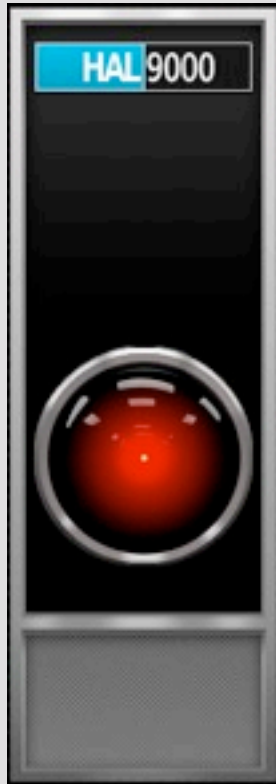


Not the end result

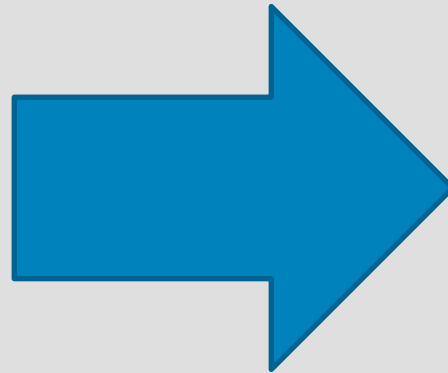


... but the process

Developmental Robotics



Not the end result



inspiration (among others) from
developmental psychology



... but the process

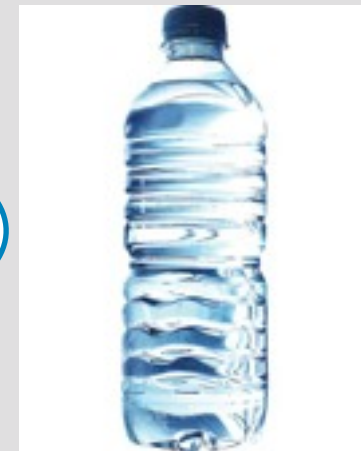
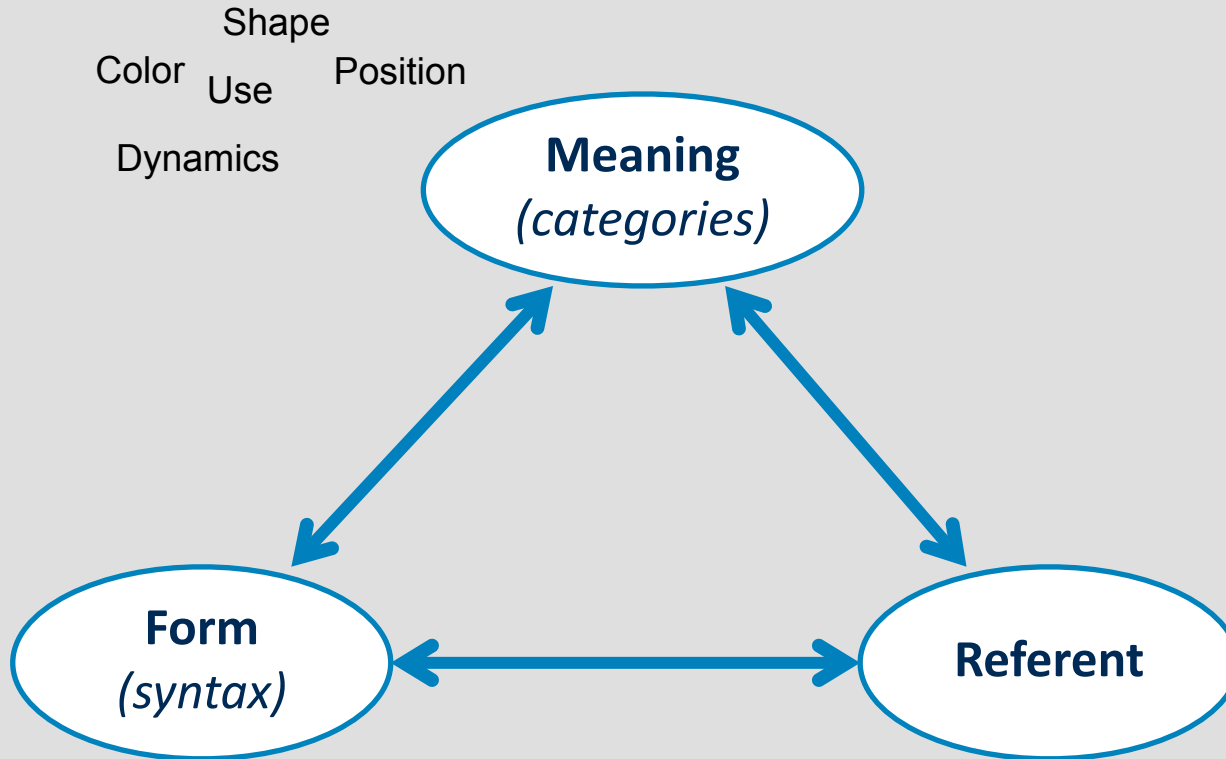
Grounding Symbols

The Symbol Grounding Problem

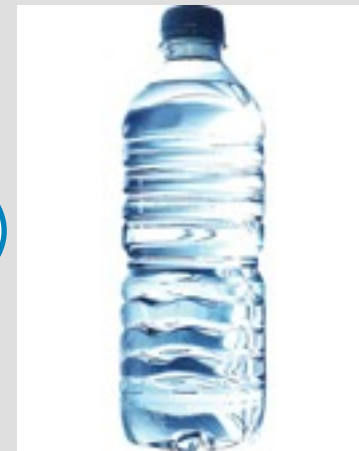
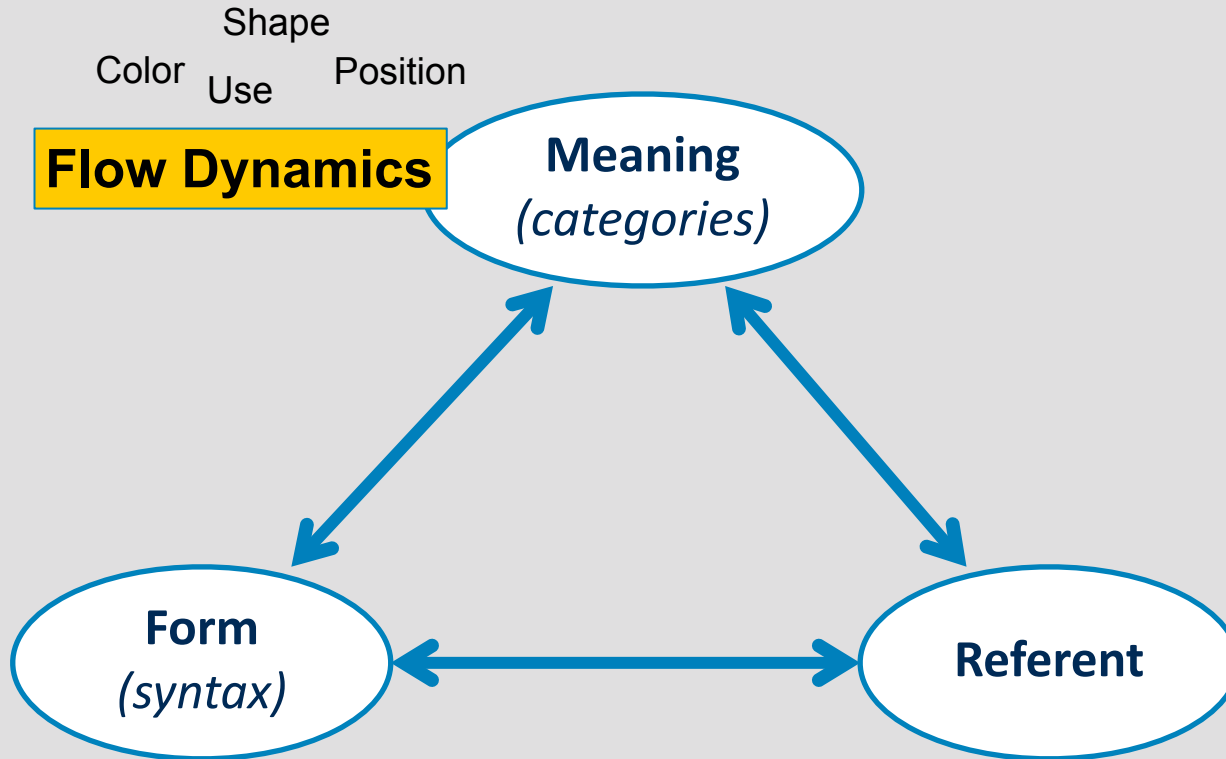
Stevan Harnad, 1990:

*“How can the semantic interpretation of a formal symbol system be made **intrinsic** to the system, rather than just parasitic on the meanings in our heads?”*

Semiotic Triangle



Semiotic Triangle

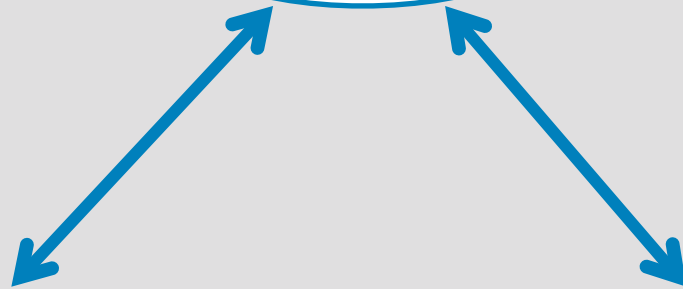


Semiotic Triangle

Shape
Color Use Position

Flow Dynamics

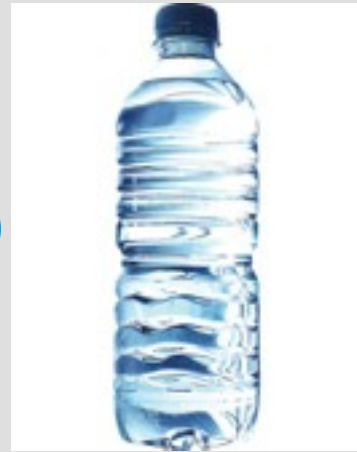
Meaning
(categories)



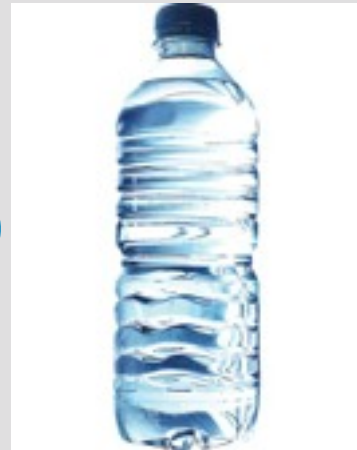
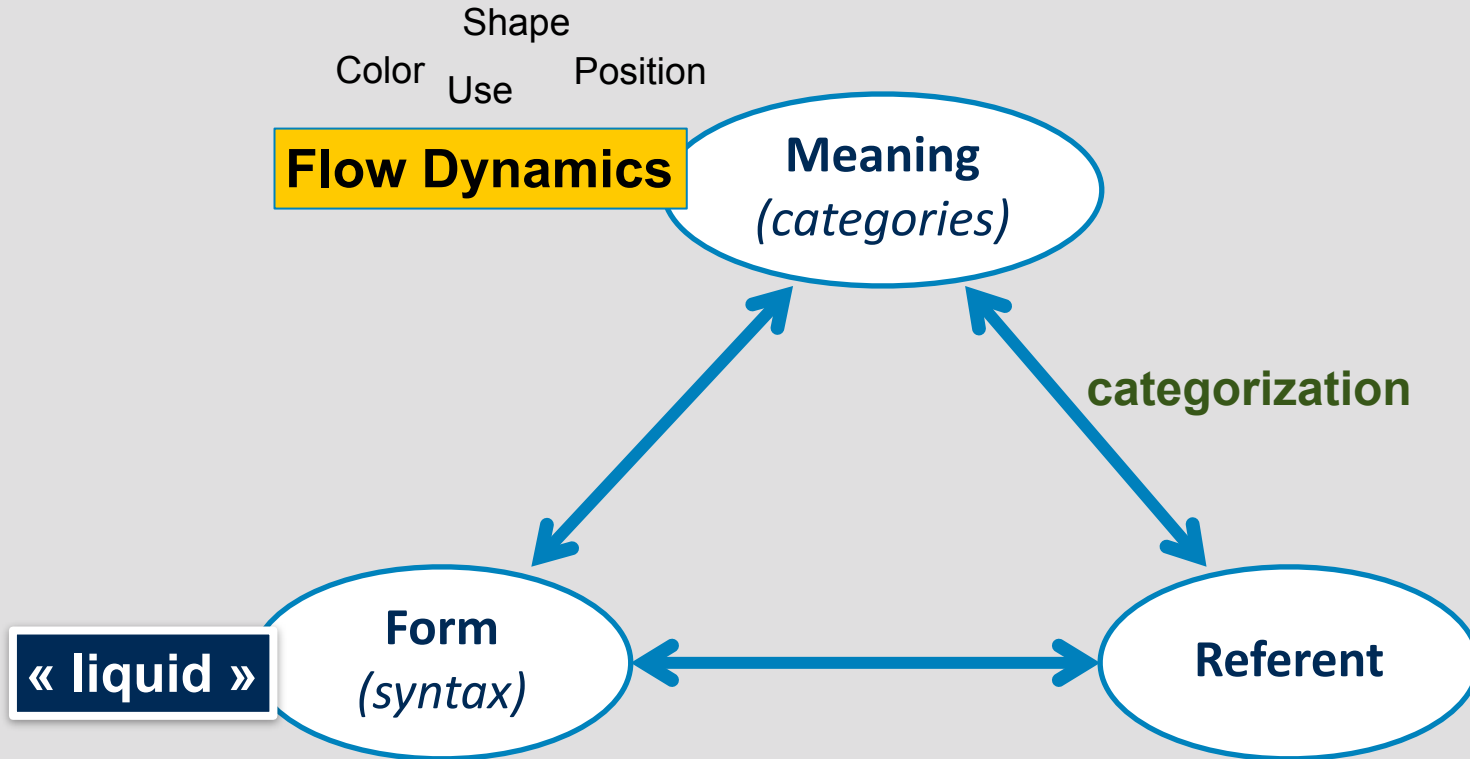
« liquid »

Form
(syntax)

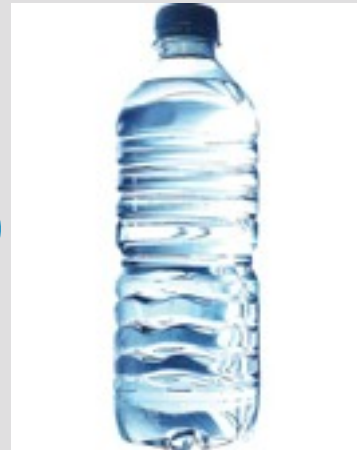
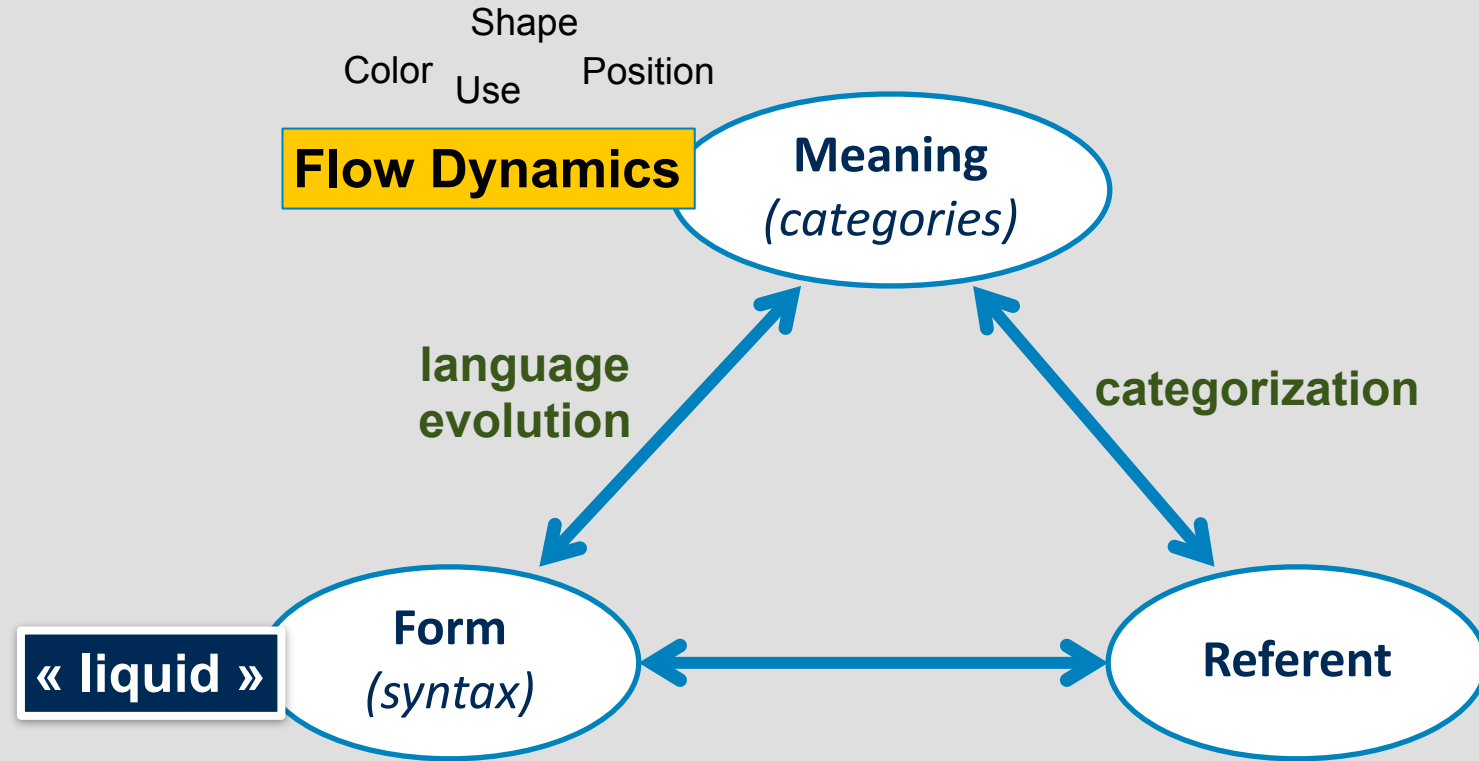
Referent



Semiotic Triangle



Semiotic Triangle

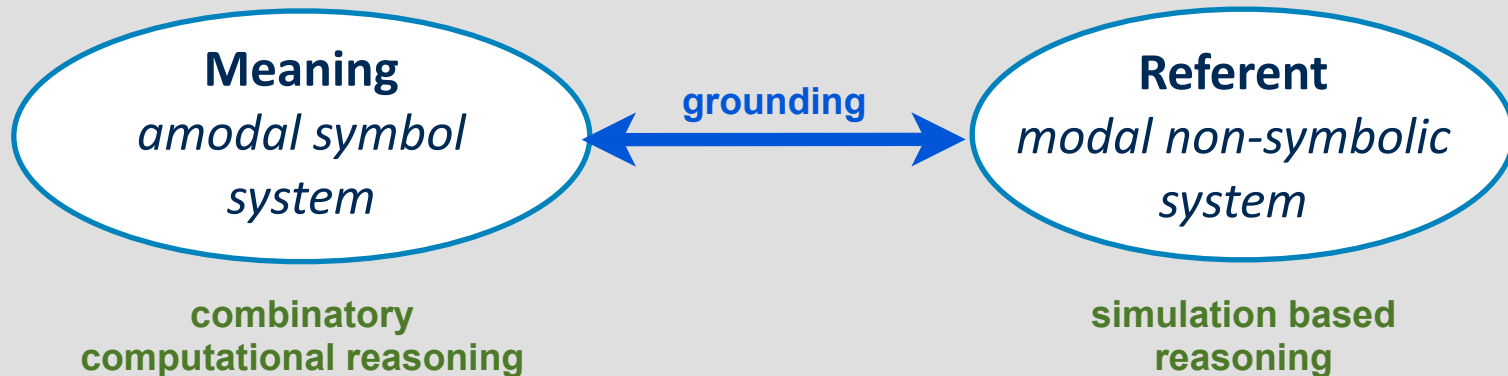


Why is grounding important for AI?

- because grounded cognition is important for robots and embedded systems? (*and they interact with us*)

Why is grounding important for AI?

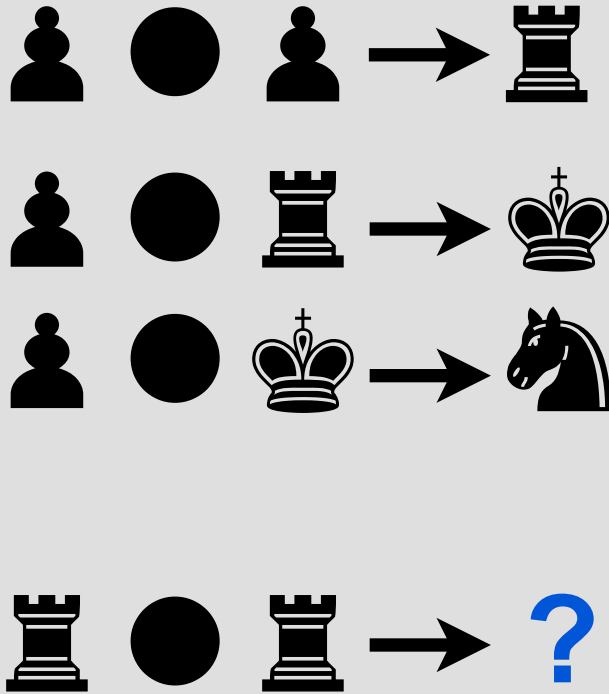
- because grounded cognition is important for robots and embedded systems? (*and they interact with us*)
- because grounded cognition provides access to **grounded forms of reasoning** (Barsalou 2008)



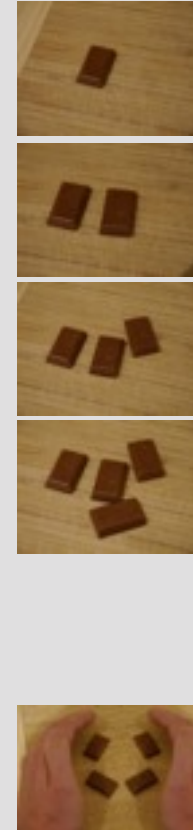
One example of the benefit of grounding



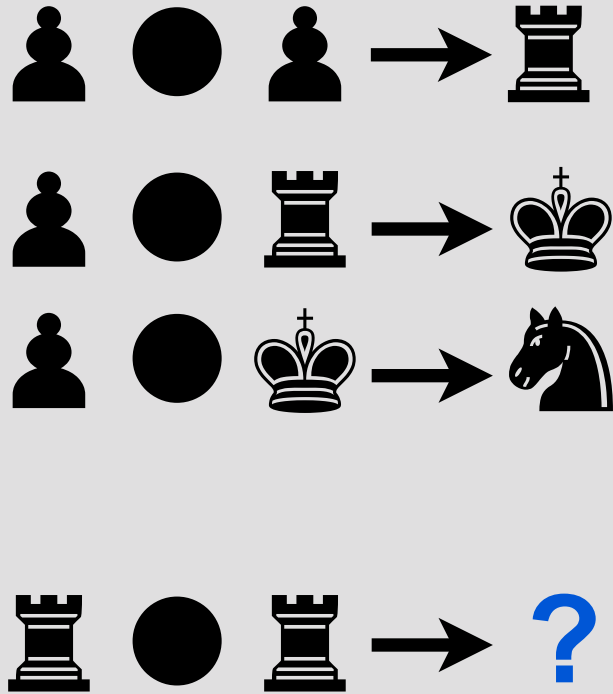
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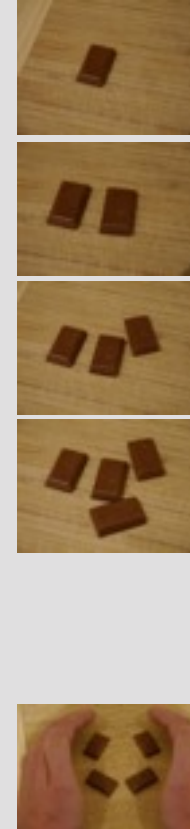
grounding



One example of the benefit of grounding



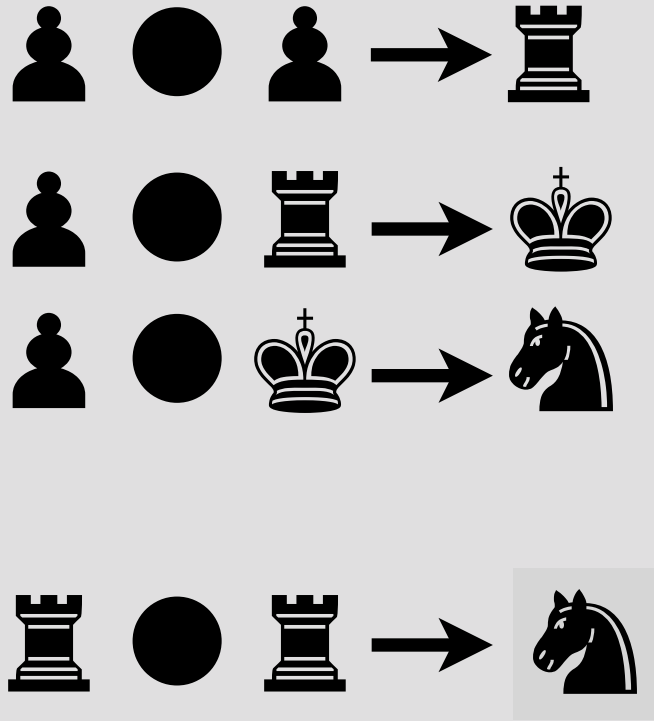
grounding



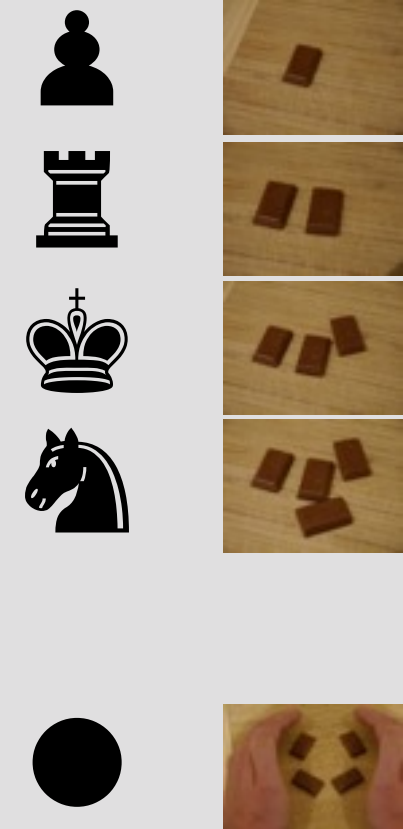
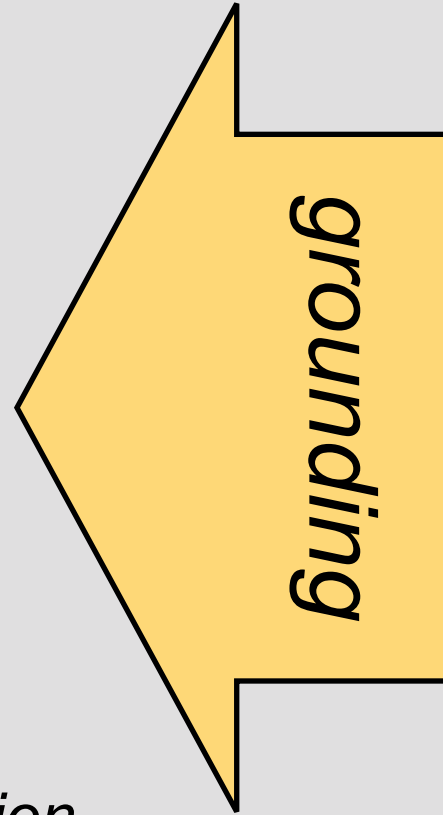
sensor modules

motor module

One example of the benefit of grounding



simulation



sensor modules

motor module

One example of the benefit of grounding

$$1 + 1 \rightarrow 2$$

$$1 + 2 \rightarrow 3$$

$$1 + 3 \rightarrow 4$$

$$2 + 2 \rightarrow 4$$

1



2



3



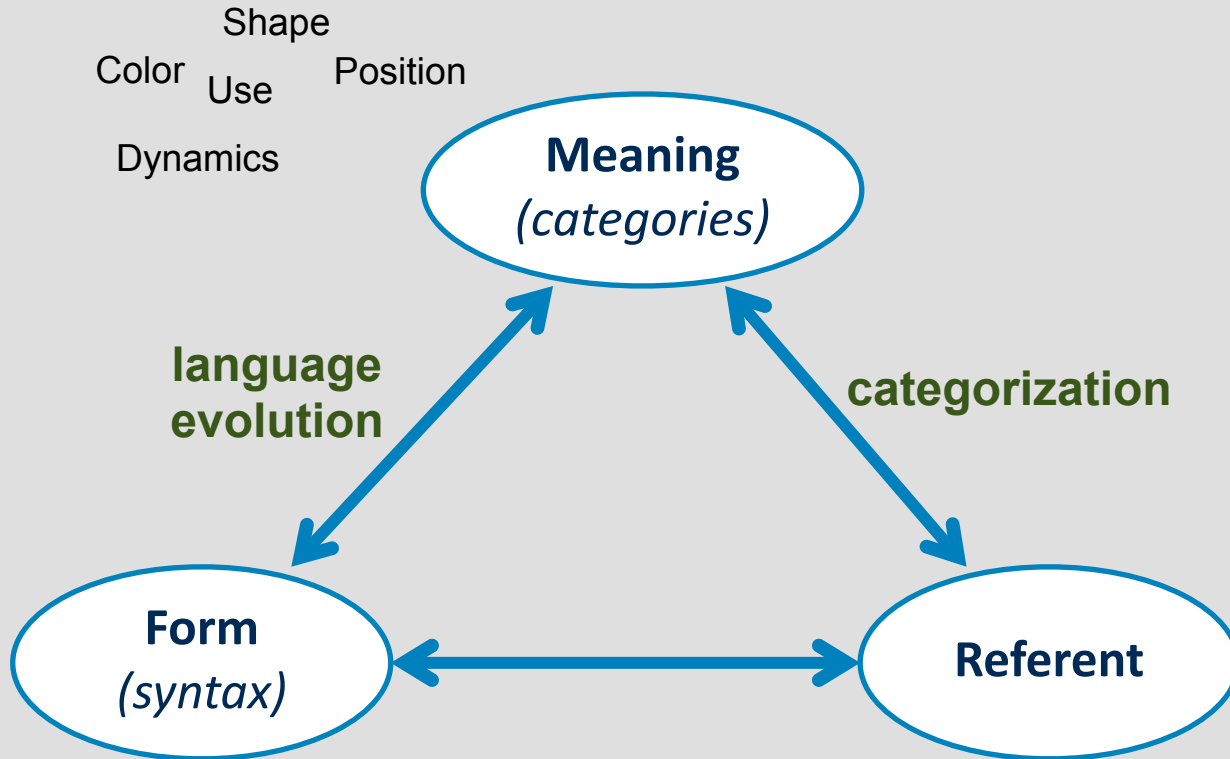
4



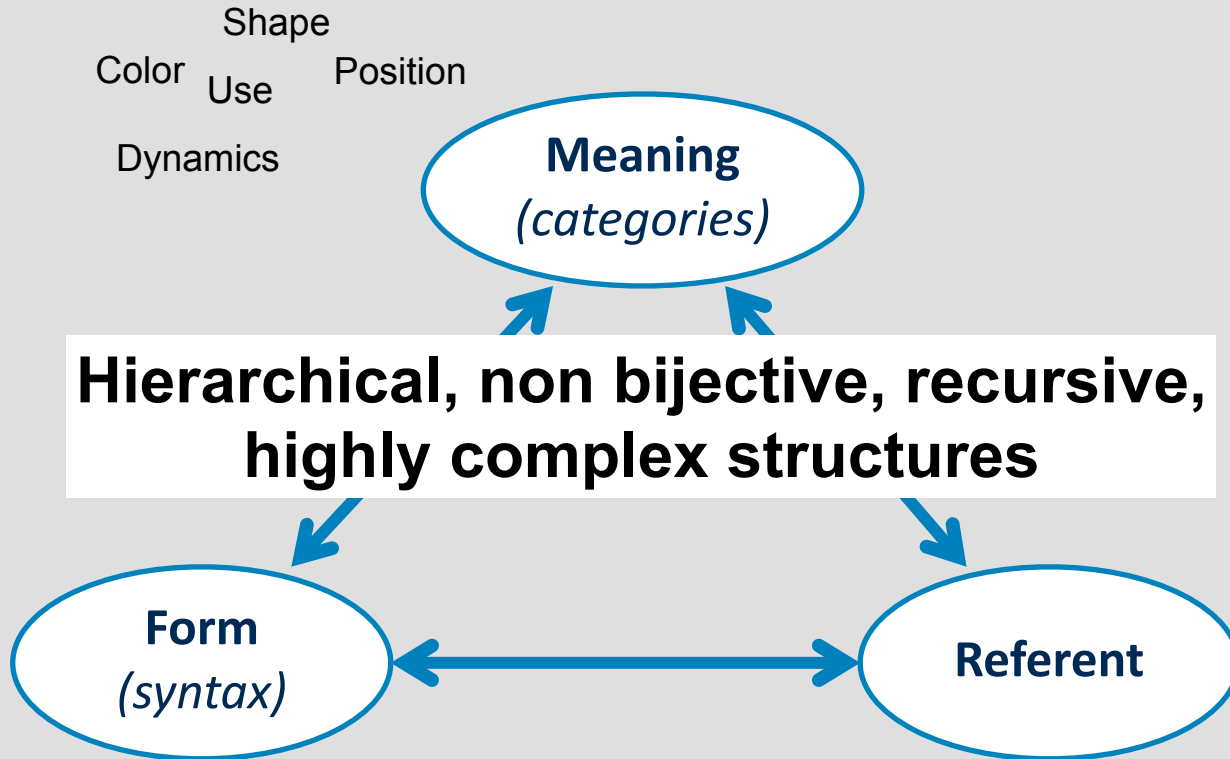
+



Semiotic Triangle



Semiotic Triangle



Evolving complex structures

“[...] evolution by cumulative [...] selection is the only theory we know of that is in principle capable of explaining the existence of organized complexity.”

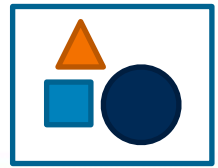
The Blind Watchmaker, R.Dawkins

Talking heads experiment



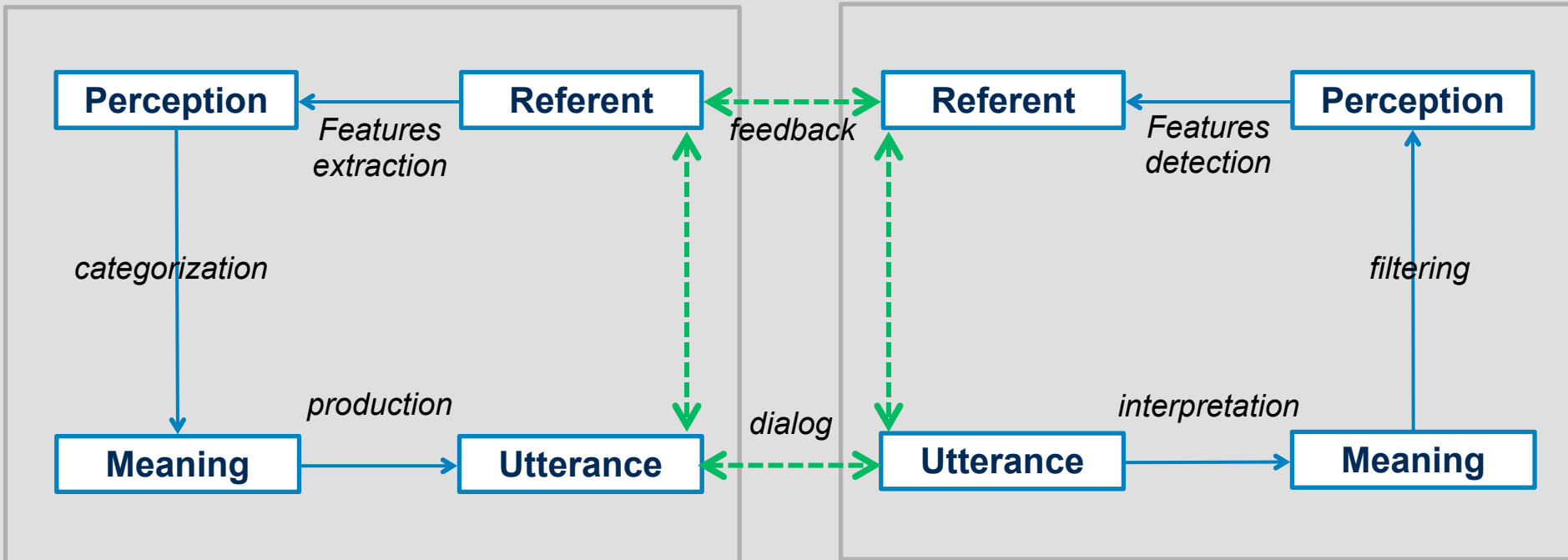
Luc Steel, Frédéric Kaplan, "*Bootstrapping grounded word semantics*", Linguistic evolution through language acquisition: formal and computational models, 1999 Cambridge University Press

Guessing game



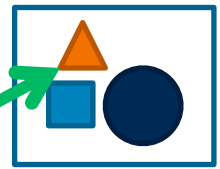
Speaker

Hearer



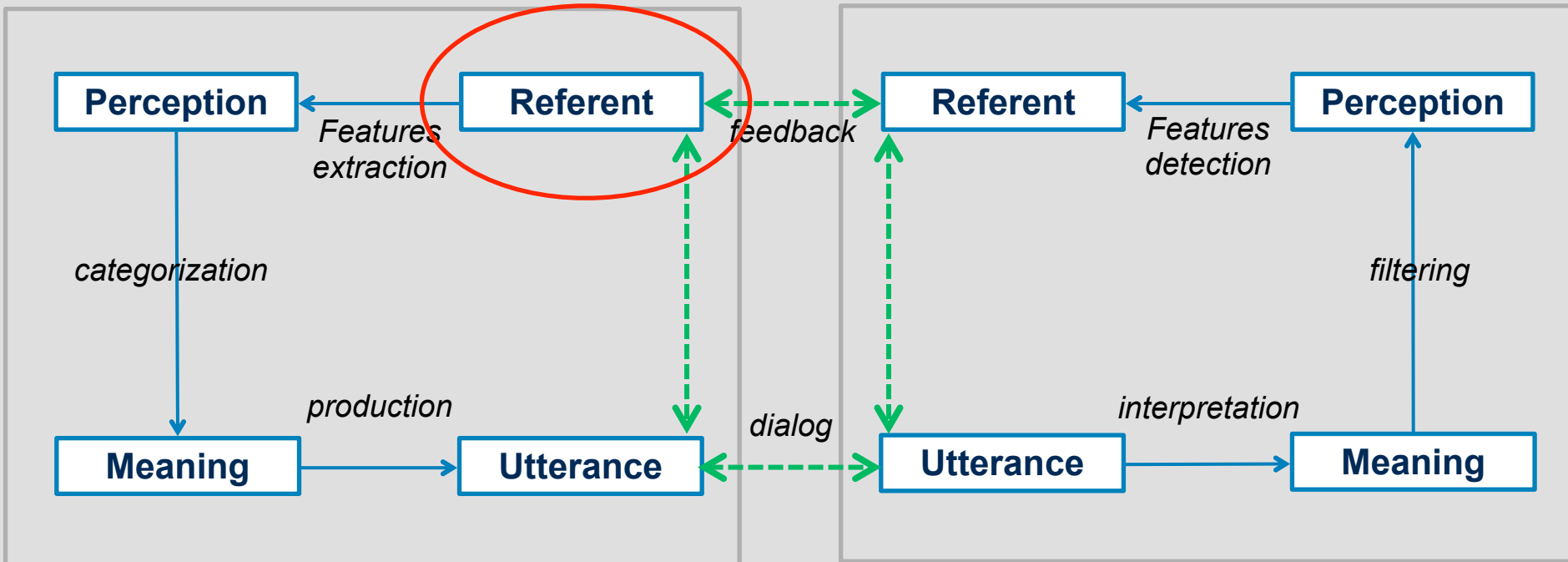
from Luc Steel, "Language games for autonomous robots", IEEE INTELLIGENT SYSTEMS 2001 vol 16

Guessing game



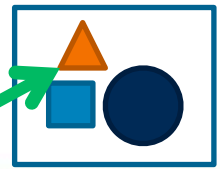
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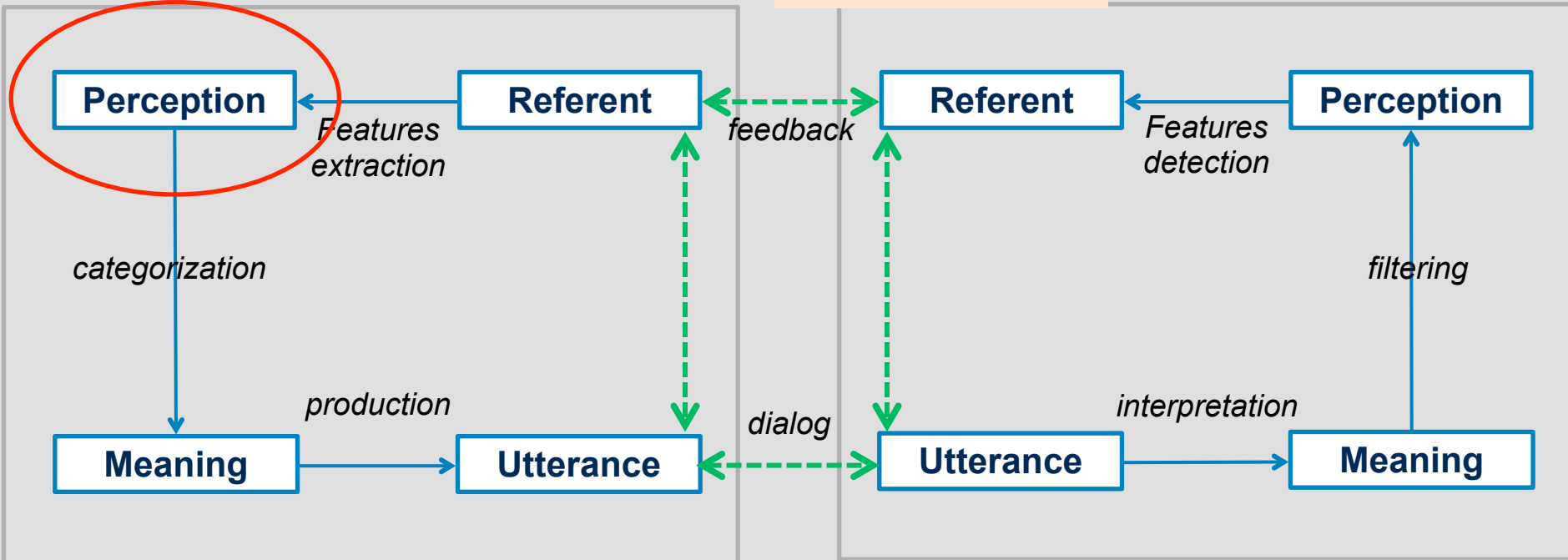
Guessing game



Color: (142;12;1)
Size: 124
Position: x=10
Shape: 3

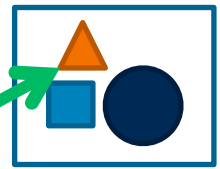
Speaker

Hearer



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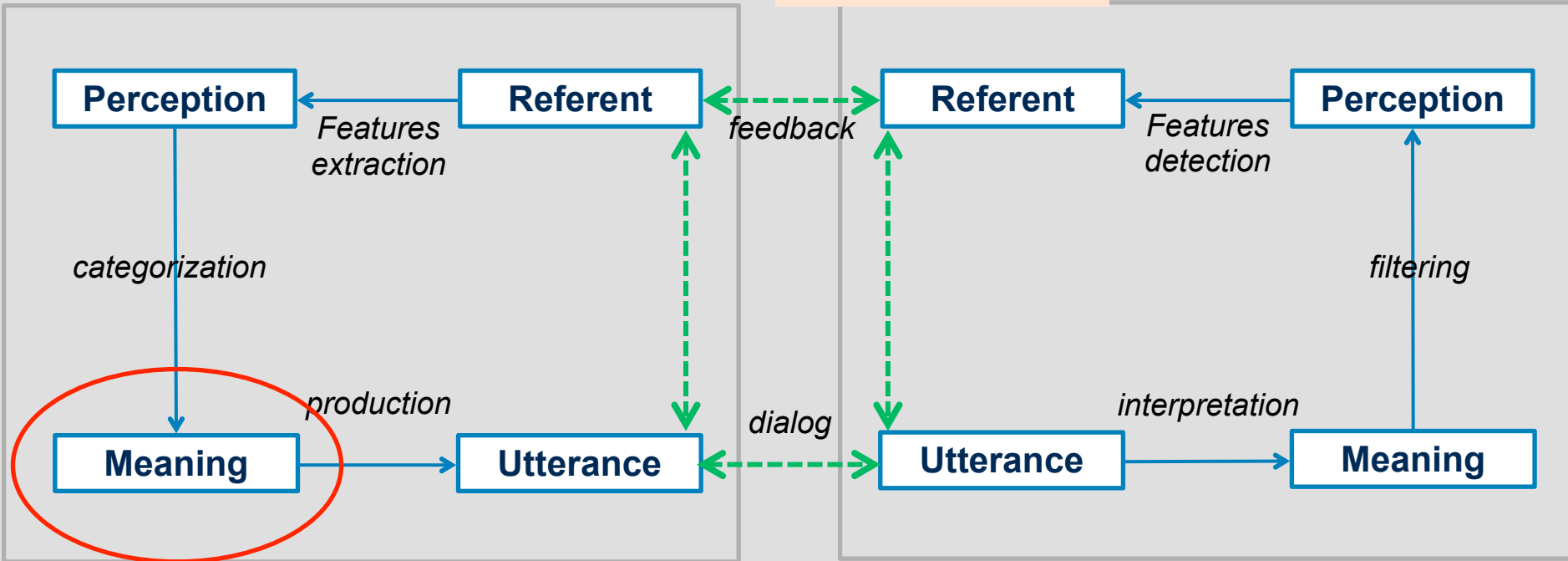


Color: R > 100
G, B < 20

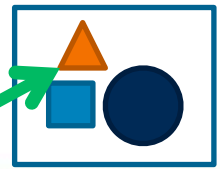
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Speaker

Hearer



Guessing game



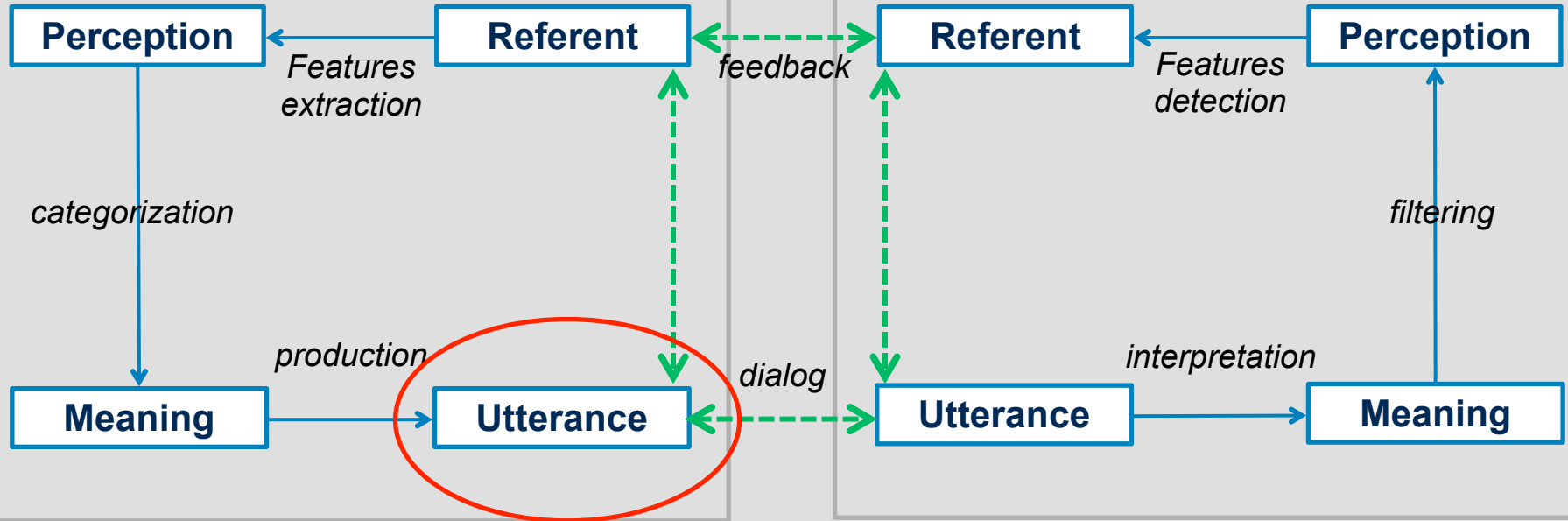
"mikari"

Color: R > 100
G, B < 20

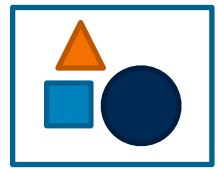
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Size: 124
Position: x=10
Shape: 3

Speaker

Hearer

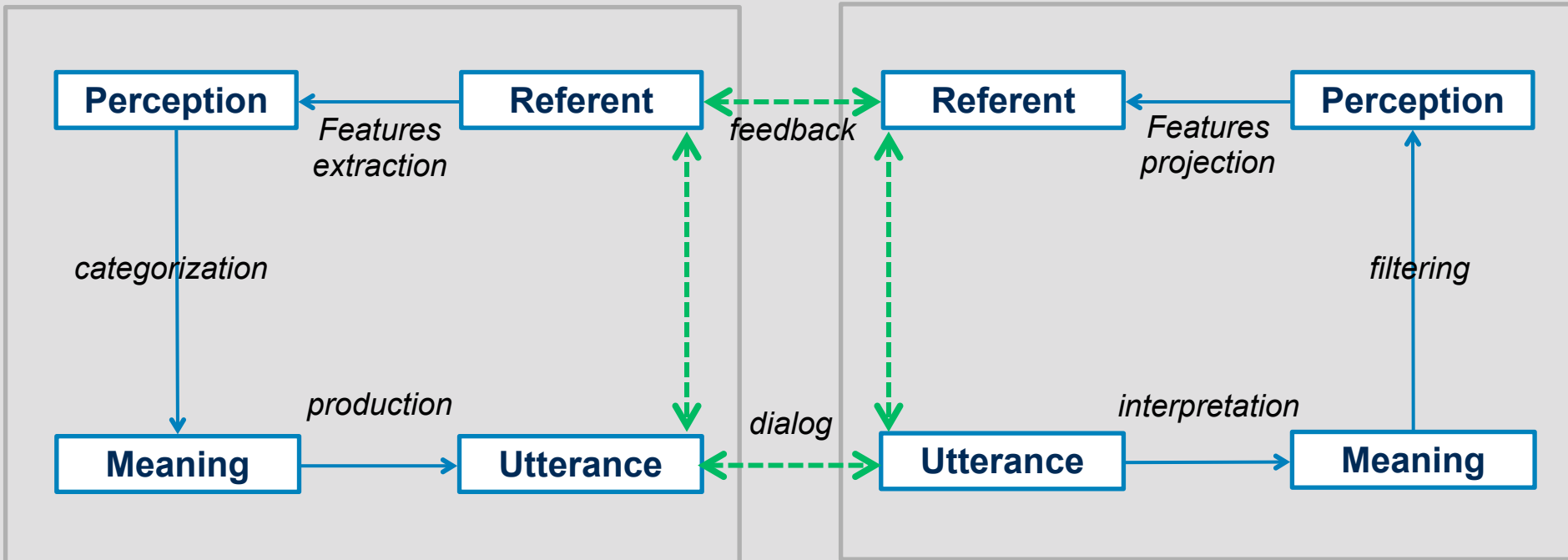


Guessing game



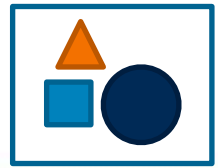
Speaker

Hearer



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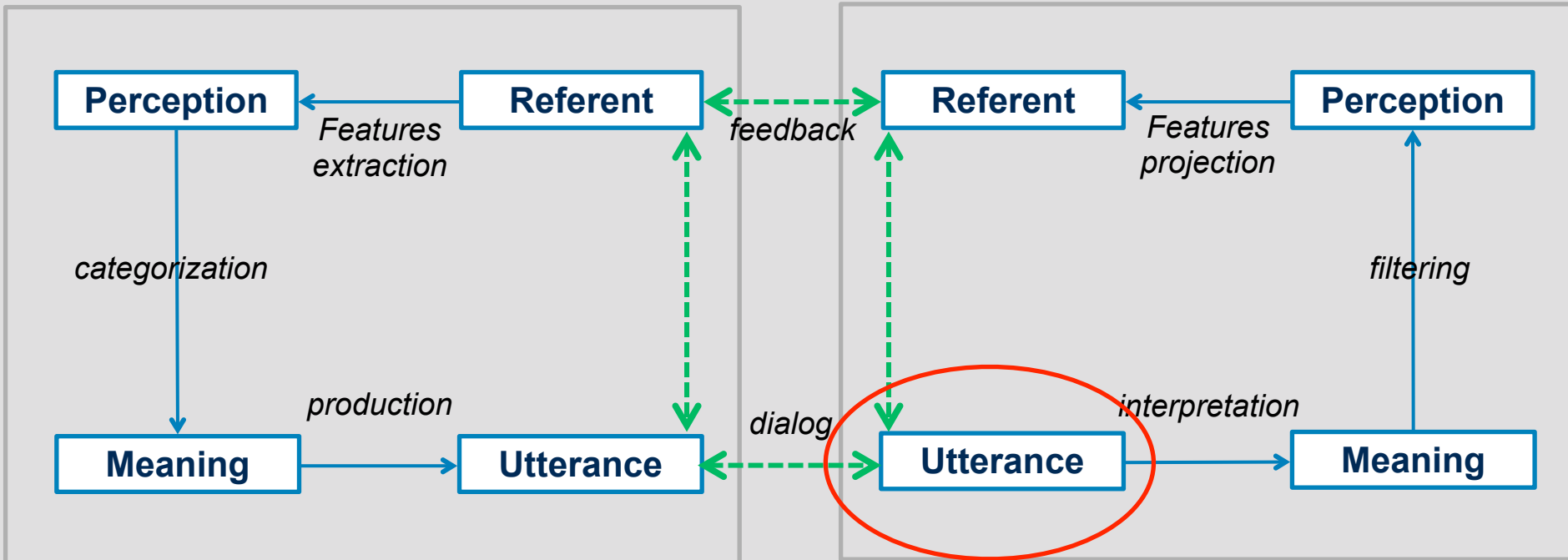
Guessing game



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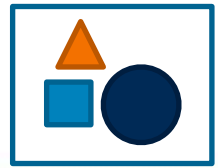
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Guessing game

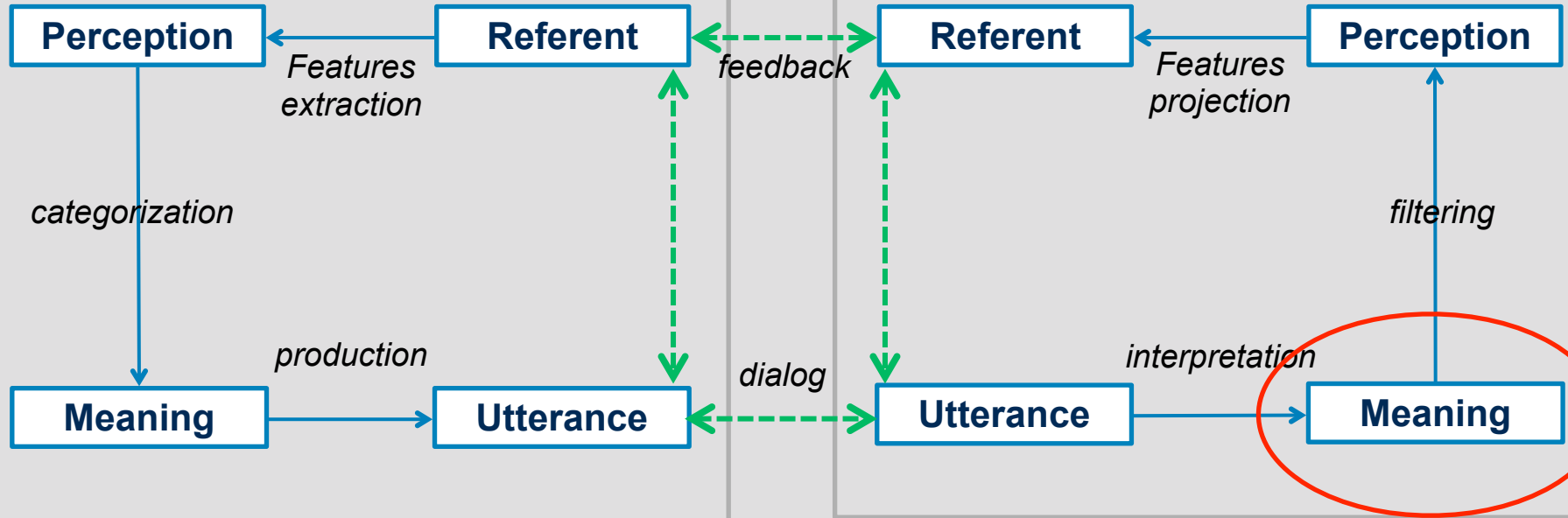


"mikari"

Position: $x > 5$

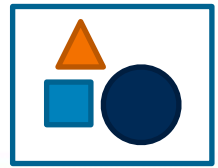
Speaker

Hearer



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Guessing game



"mikari"

Position: $x > 5$

Color: (142;12;1)

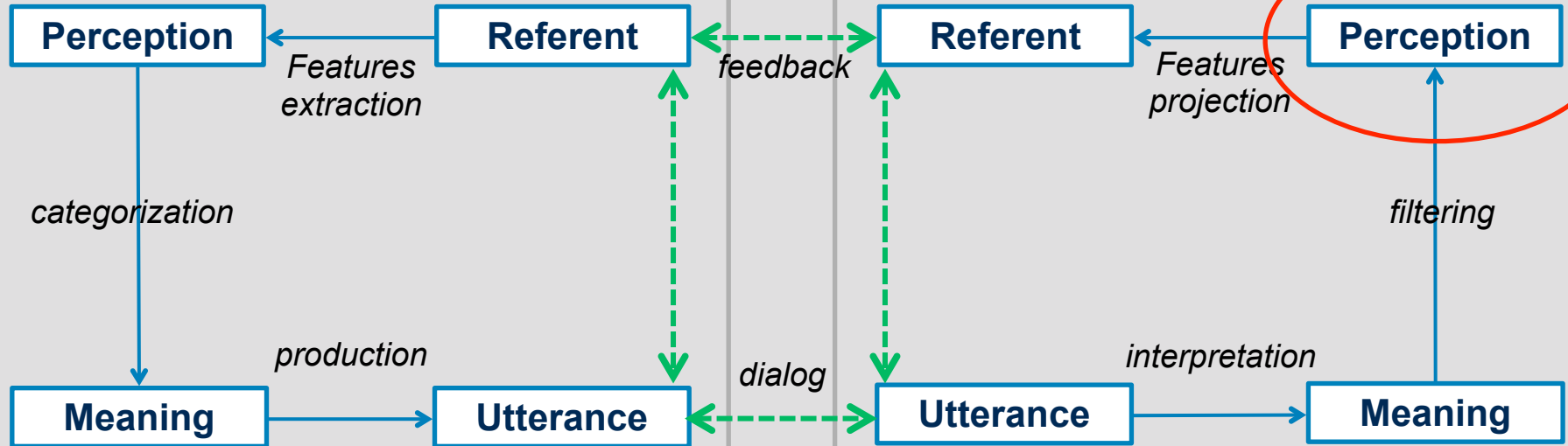
Size: 124

Position: $x=10$

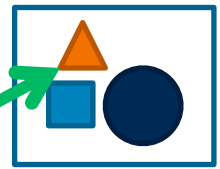
Shape: 3

Speaker

Hearer



Guessing game



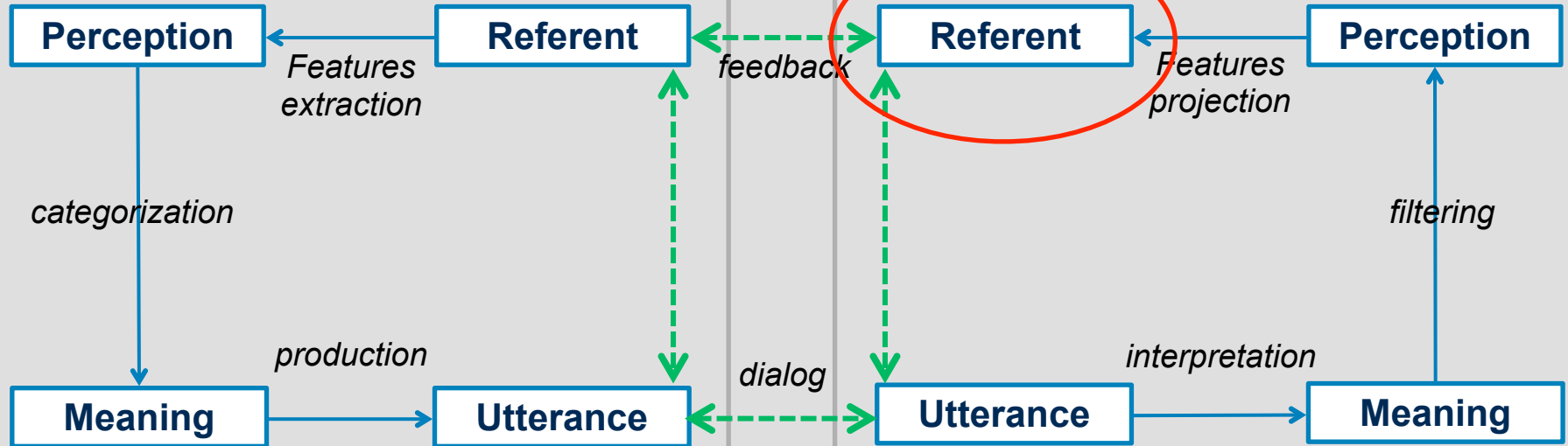
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Position: $x > 5$

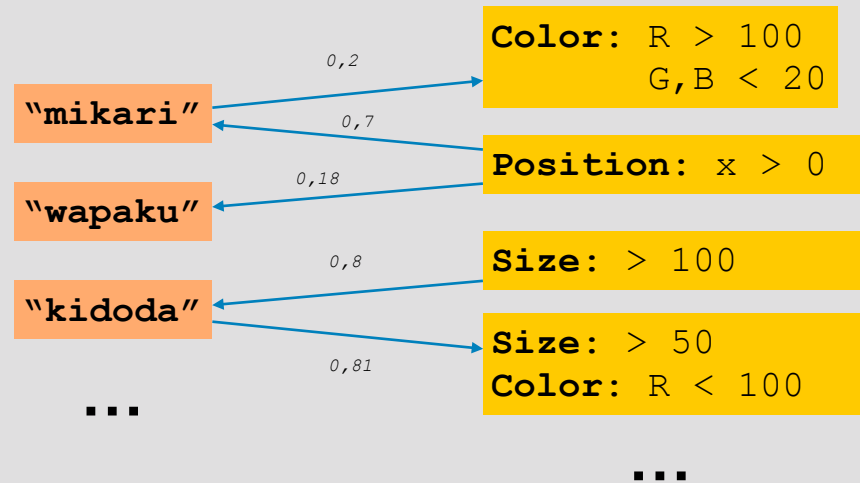
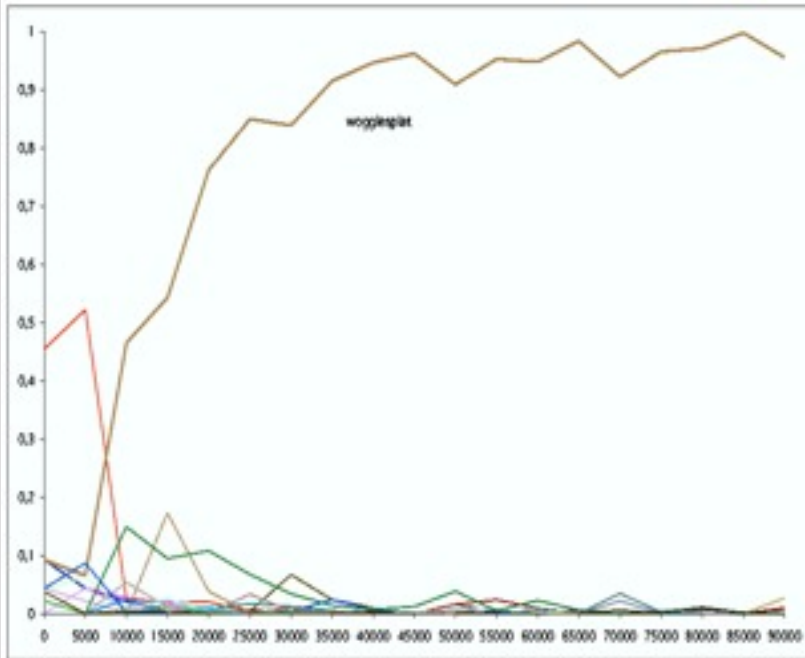
Color: (142;12;1)
Size: 124
Position: $x=10$
Shape: 3

Speaker

Hearer



Grounded dynamics



After a few thousand iterations, agents **synchronize a shared grounded lexicon** to describe their world.

Why is it an important experiment?



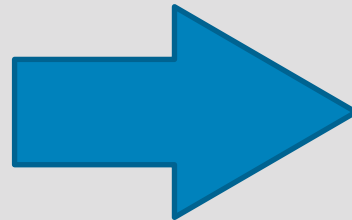
- Symbols are **socially** grounded: the lexicon is shared
- Symbols are grounded in the **environment**
- Symbols are grounded in the **task** (here, the game)

Next step: grounding grammar



Grammar: set of conventionalized strategies to disambiguate

“John gives a ball to Mary”



John *give* *???* *ball*
 Mary

from Generative to Construction Grammar

Generative Grammars (Chomsky)

$S \Rightarrow NP + VP$

$NP \Rightarrow N$

$NP \Rightarrow Det + N$

$VP \Rightarrow V$

$VP \Rightarrow V + NP$

...

from Generative to Construction Grammar

Generative Grammars (Chomsky)

**Combinatorial explosion
+ Syntax only**

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...

from Generative to Construction Grammar

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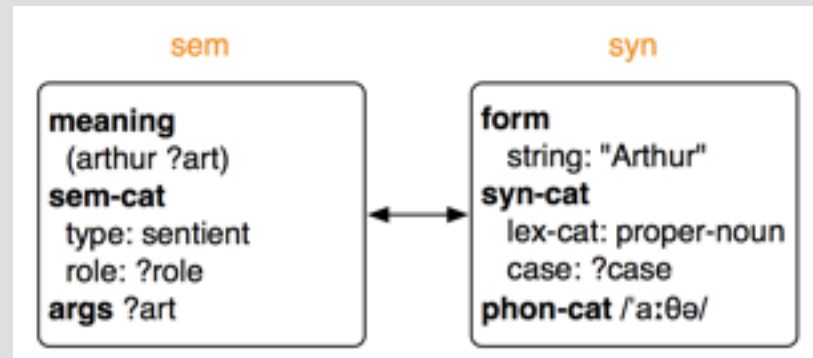
NP \Rightarrow Det + N

VP \Rightarrow V

VP \Rightarrow V + NP

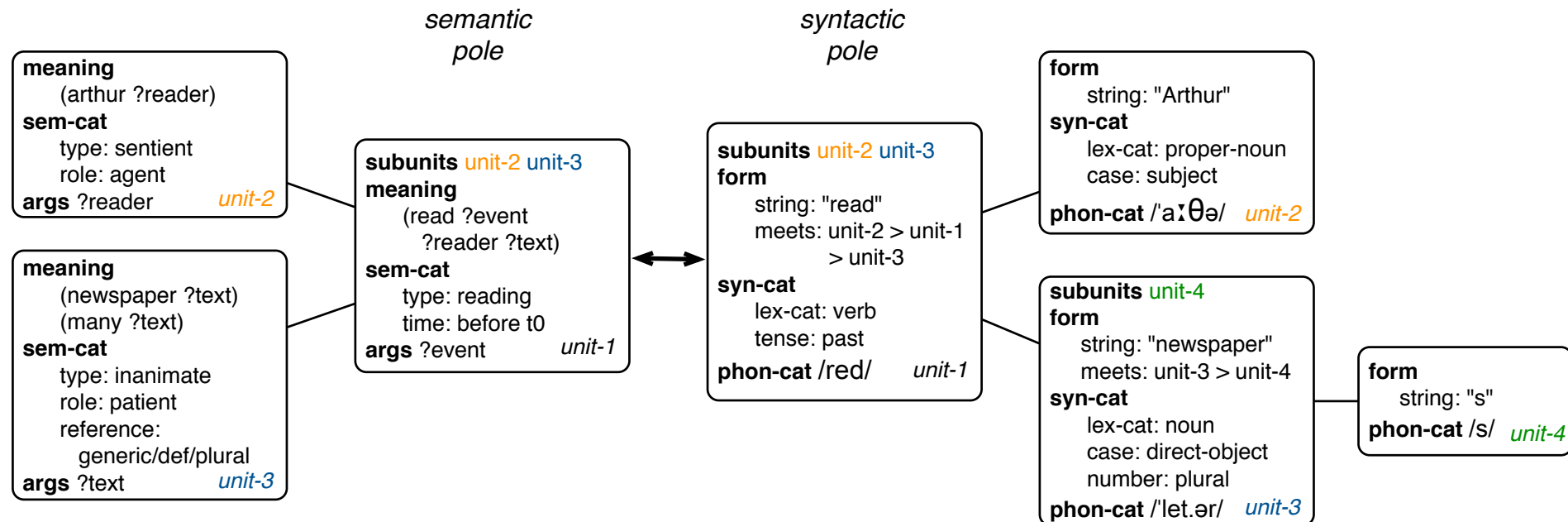
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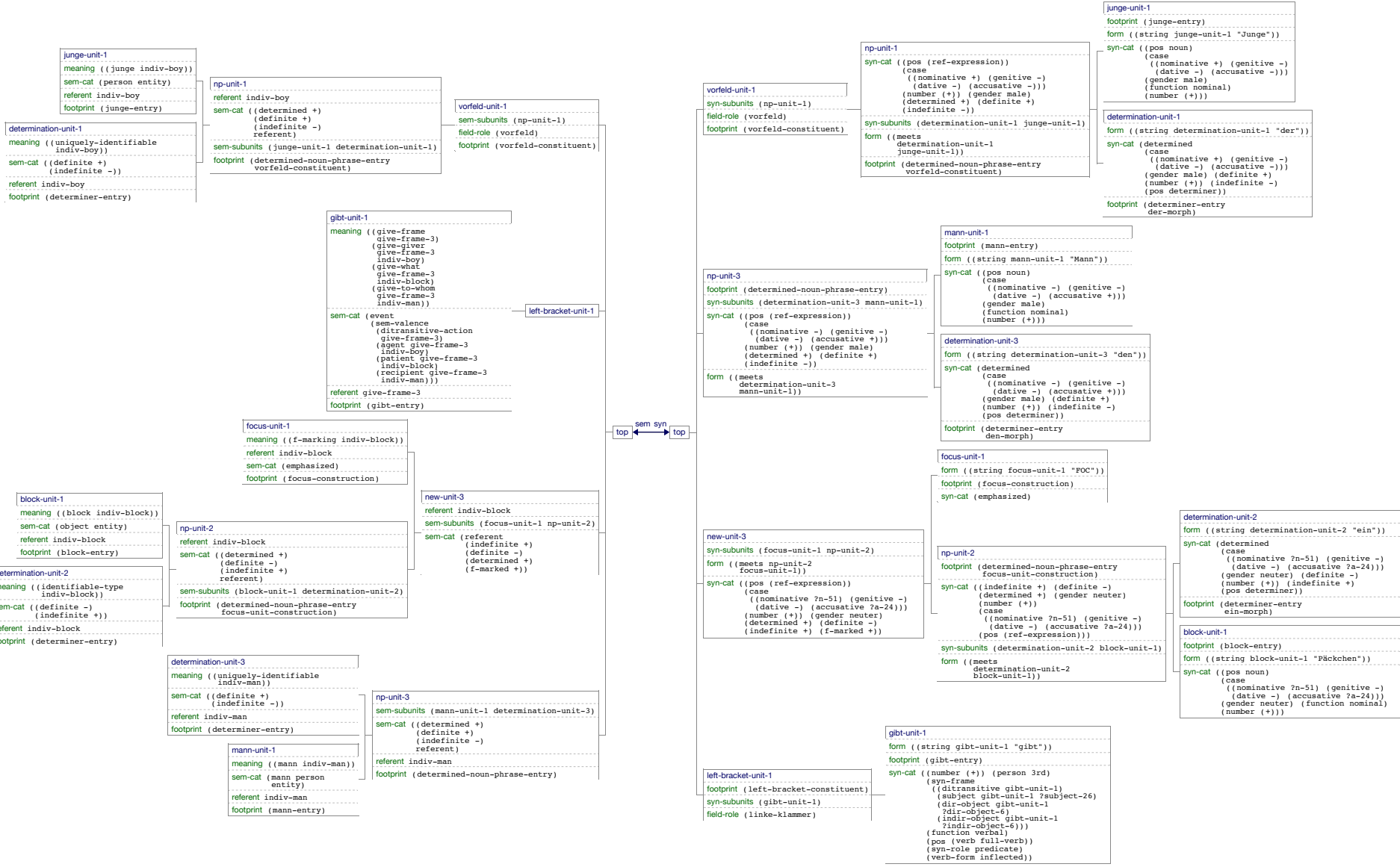
N



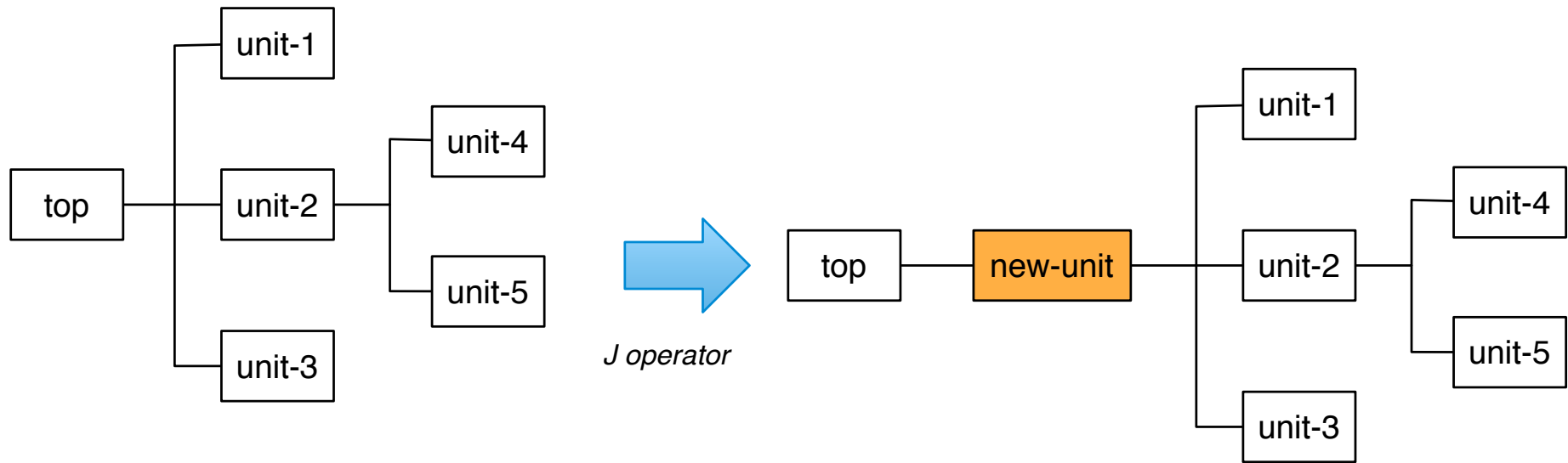
Fluid Construction Grammar

Arthur read newspapers





Operators to perform unit transformations



FCG: Analogy with chemistry

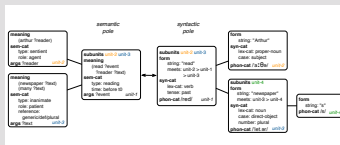
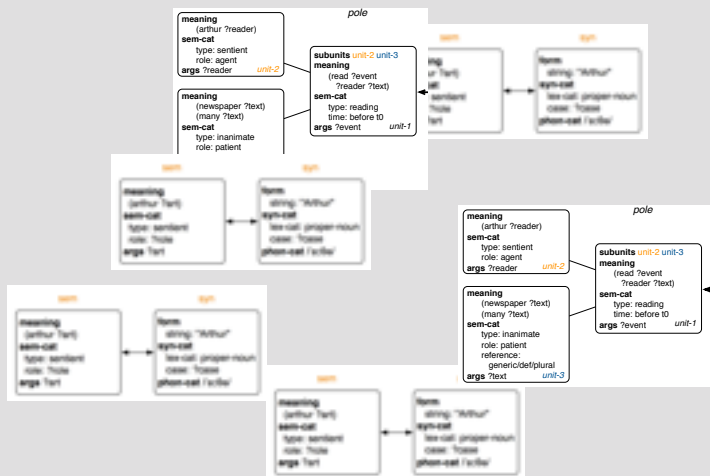
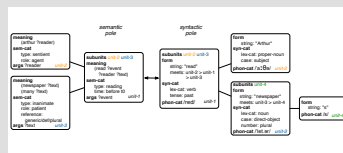
complex “soup” of potentially relevant constructions

Meaning

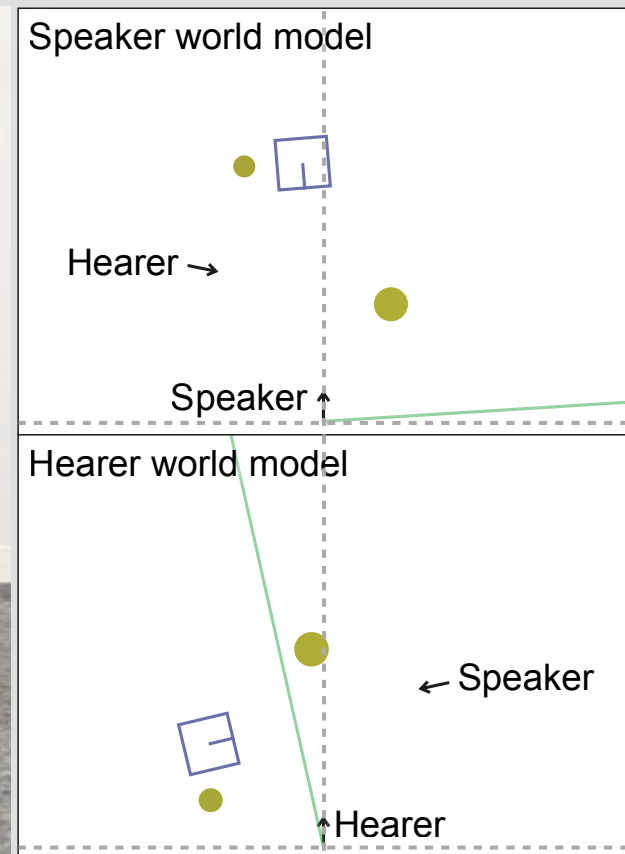
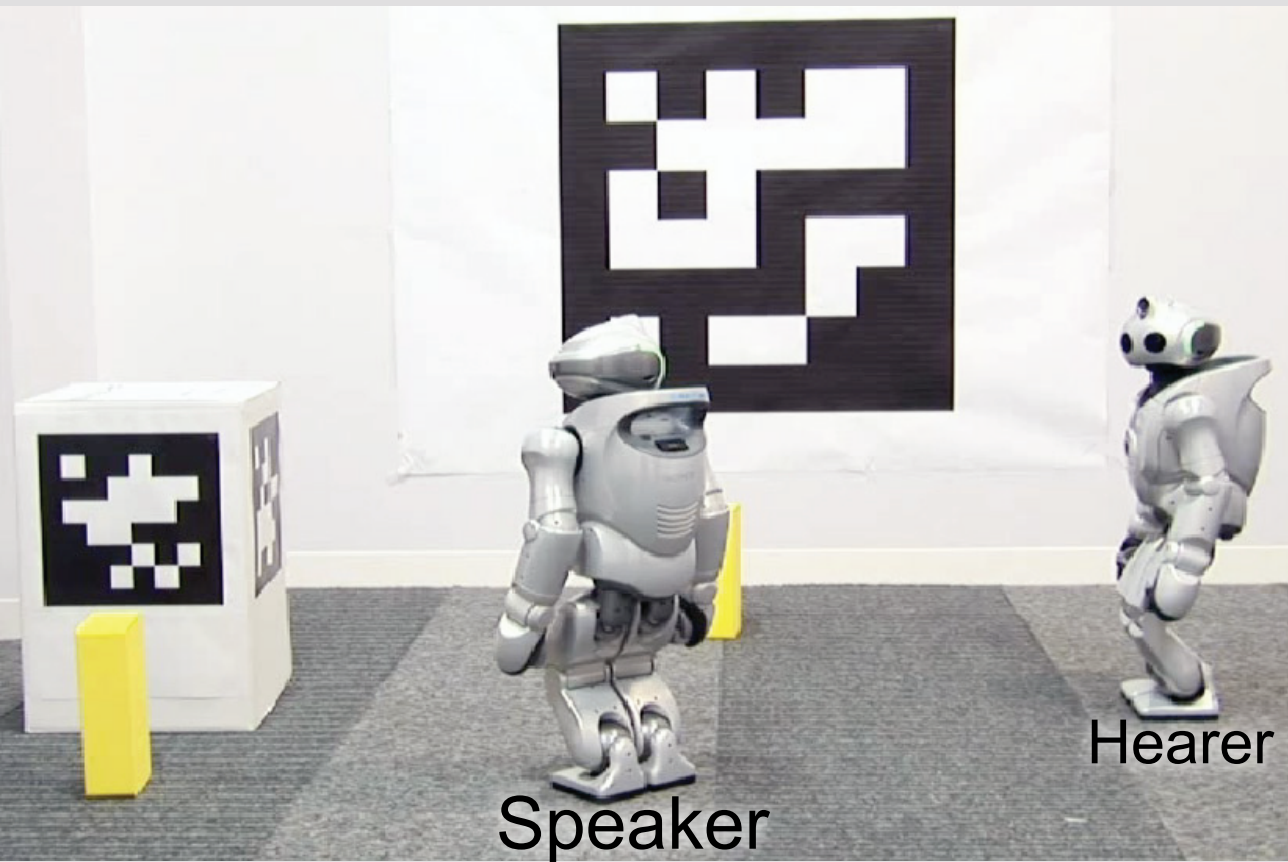
production

Form

parsing



Grounding spatial grammatical conventions



from Spranger, M., Pauw, S., Loetzsch, M., and Steels, L. "Open ended procedural semantics", Language grounding in robots. Springer, New-York

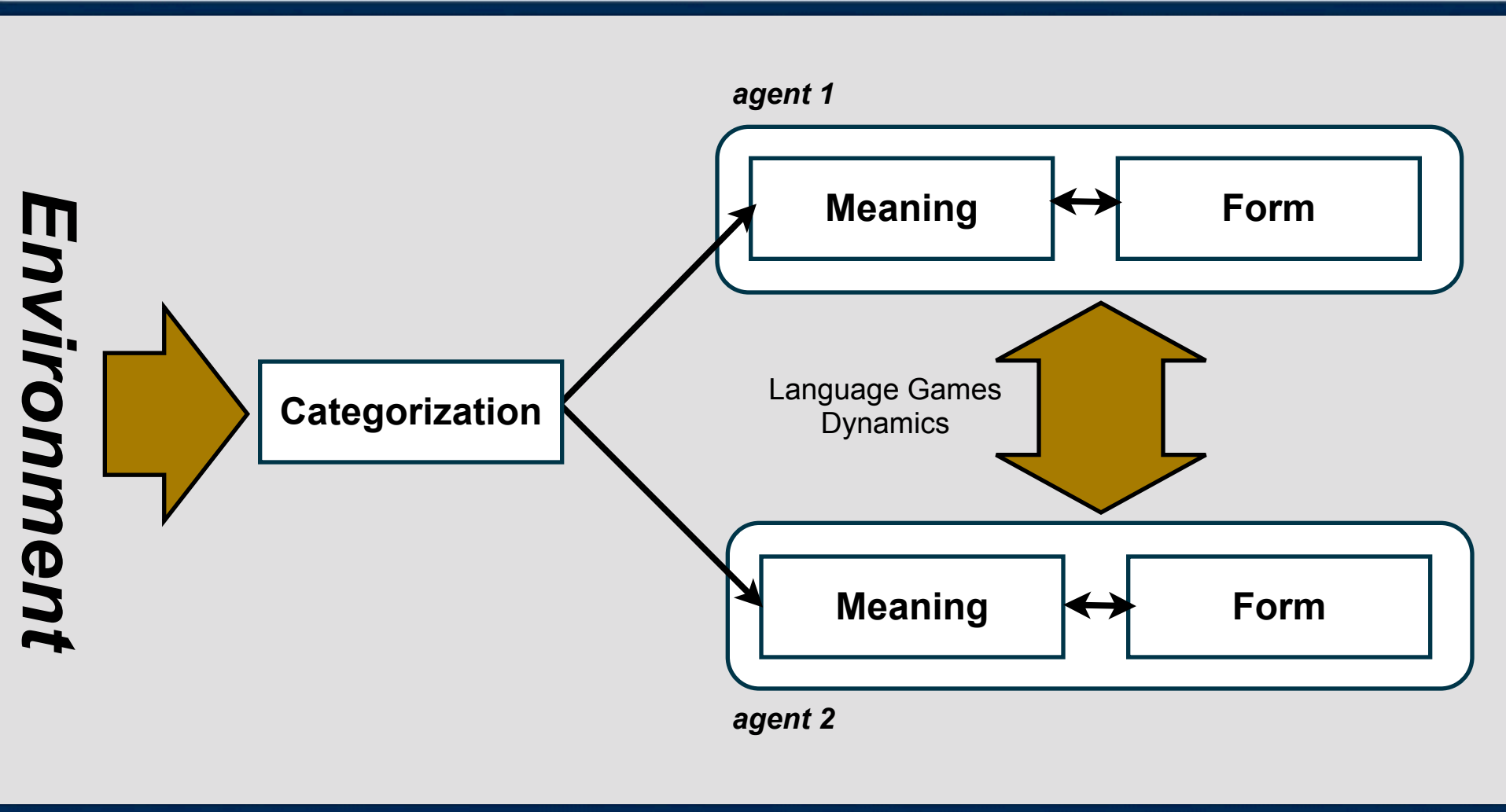
Language development stages

FCG and similar computational models of construction grammars are good candidates to account for **stages of linguistic competences**, in a **usage-based approach** of language learning:

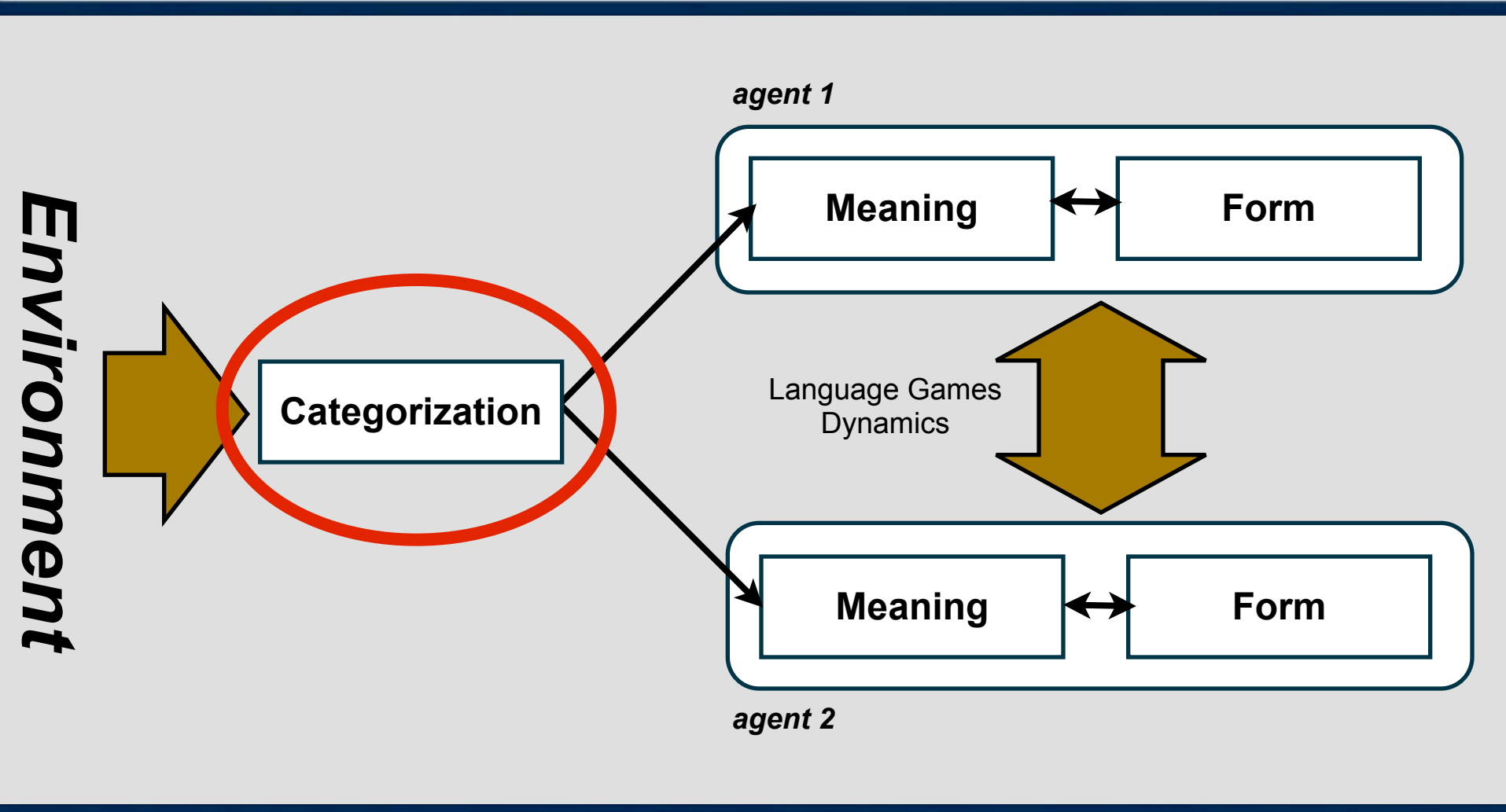
- Holophrases
- Word combinations
- Pivot Schemas
- Item-based constructions
- Abstract constructions

from Tomasello, 2005

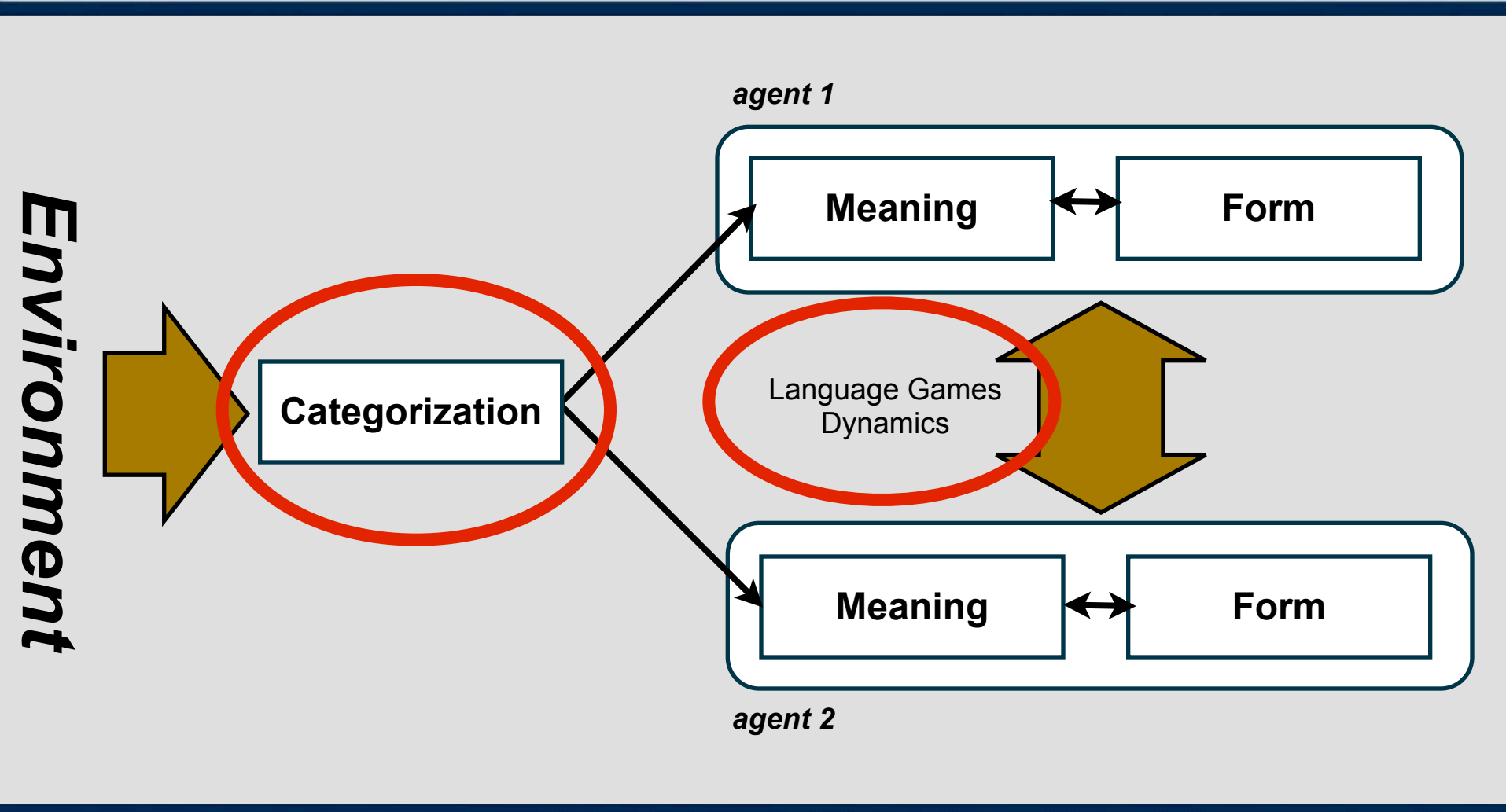
What comes next?



What comes next?

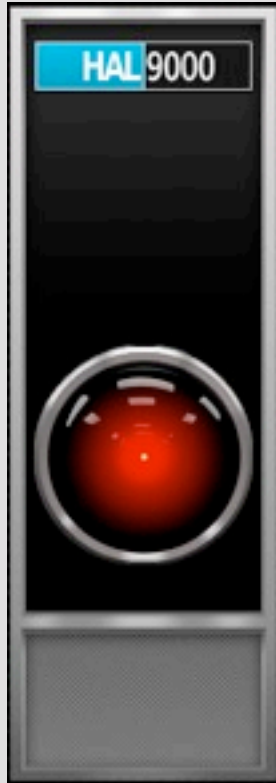


What comes next?

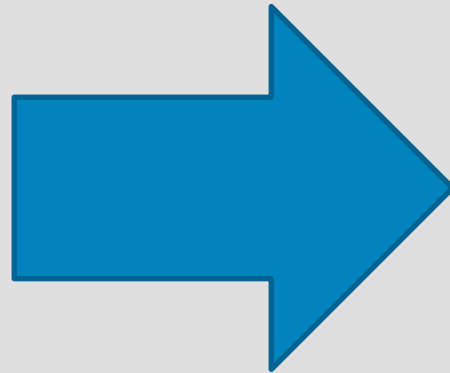


Developmental Robotics

Focus on the learning mechanisms



Not the end result



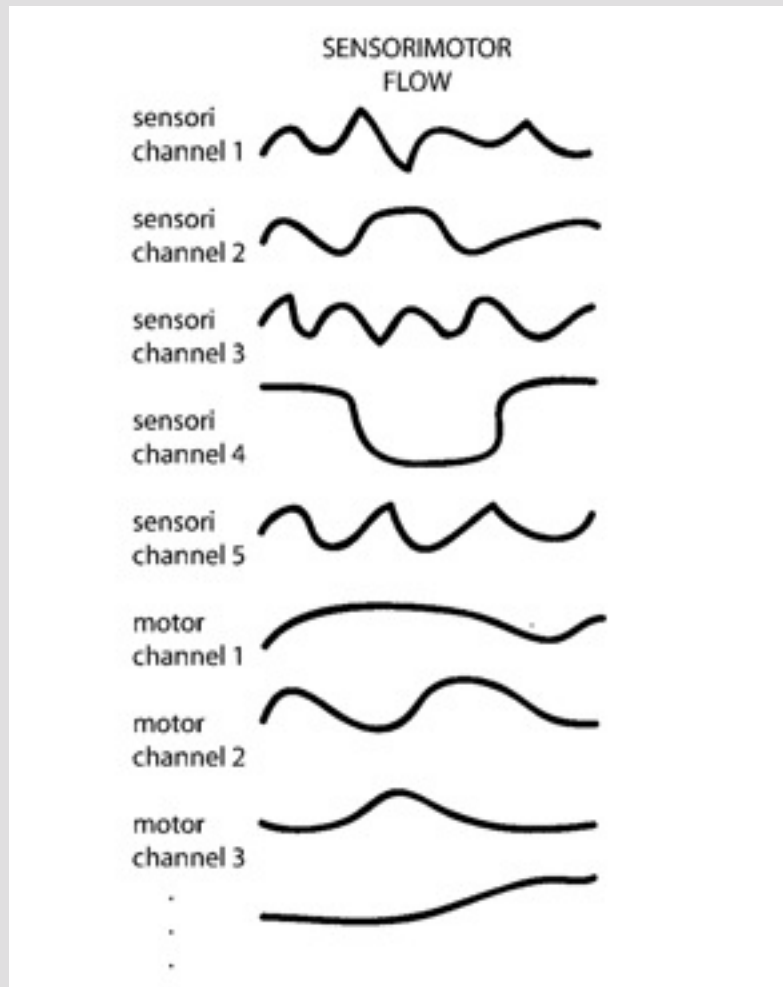
... but the process

Key aspects of the developmental approach

- **Grounded** semantics
- **Staged** development => previous stages help to build the next.
- **Dynamic** systems, Life-long learning

Categorization

What is it like to be a robot?



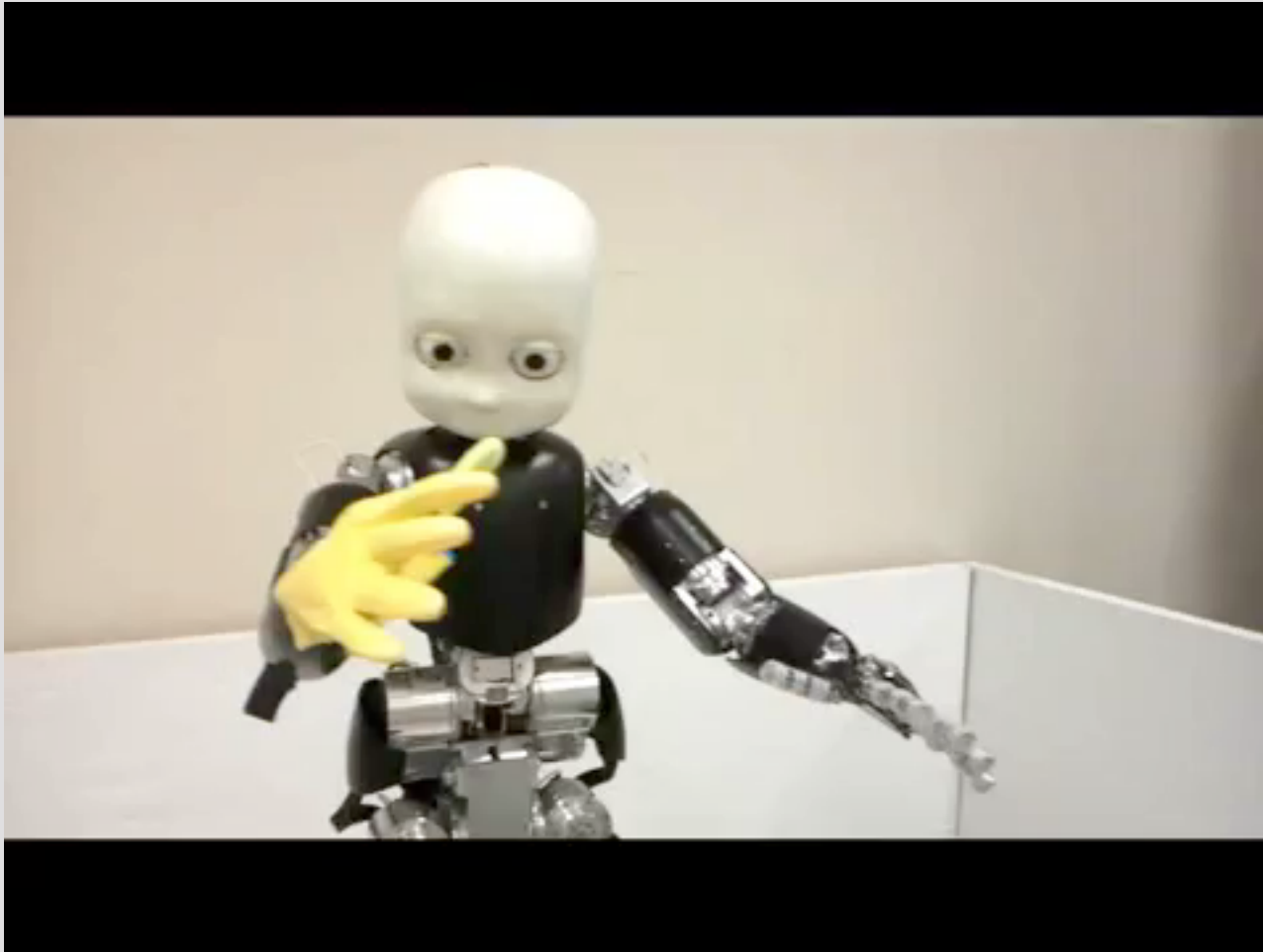
*“The Blooming,
buzzing
confusion”*

William James

Scaffolding knowledge through interaction



Learning its own body



Ref: University of Illinois at Urbana-Champaign

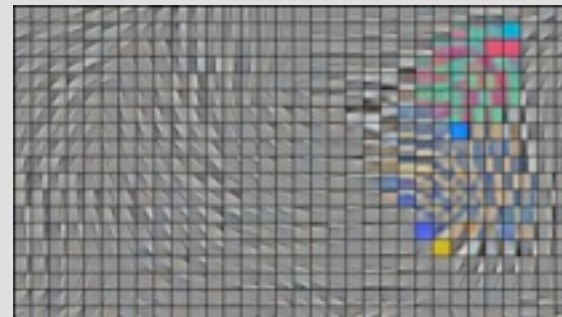
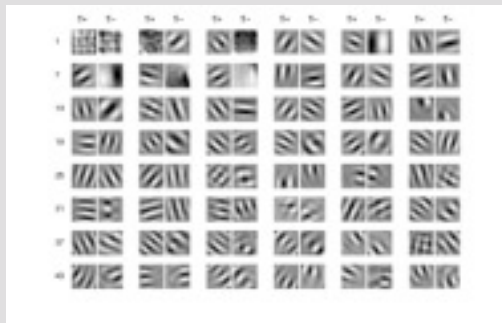
Learning its own body



Ref: Sony CSL

Low level visual and audio pattern extraction

- **Avoid biases as much as possible in the nature and structure of relevant perceptual data for the robot.**
- **Deep Learning:** hierarchical neural approaches in general
- **Slow Feature Analysis:** layered approaches to build up structure
- **Multi-modal cross-correlation** extraction



- Quoc V Le, Rajat Monga, Matthieu Devin, Greg Corrado, Kai Chen, Marc'Aurelio Ranzato, Jeff Dean & Andrew Y Ng. **Building highlevel features using large scale unsupervised learning.** arXiv preprint arXiv:1112.6209, 2011
- Yoshua Bengio. **Learning deep architectures for AI.** Foundations and Trends in Machine Learning, vol. 2, no. 1, pages 1–127, 2009
- Sven Behnke. **Hierarchical neural networks for image interpretation**, volume 2766. Springer, 2003
- Laurenz Wiskott & Terrence J Sejnowski. **Slow feature analysis: Unsupervised learning of invariances.** Neural computation, vol. 14, no. 4, pages 715–770, 2002

Sensorimotor conceptualization

- Gather spatiotemporal correlations into conceptual first level categories.
- Proto-object extraction
- Simple motor skills schemata
- Simple causality, naïve physics

=> Active Learning



- D.H. Rakison & L.M. Oakes. **Early category and concept development: Making sense of the blooming, buzzing confusion**. Oxford University Press, USA, 2003
- Francesco Orabona, Giorgio Metta & Giulio Sandini. **A proto-object based visual attention model**. Attention in cognitive systems. Theories and systems from an interdisciplinary viewpoint, pages 198–215, 2007
- Ryunosuke Nishimoto, Jun Namikawa & Jun Tani. **Learning multiple goal-directed actions through self-organization of a dynamic neural network model: A humanoid robot experiment**. Adaptive Behavior, vol. 16, no. 2-3, pages 166–181, 2008
- Paul Fitzpatrick, Giorgio Metta, Paul Fitzpatrick & Giorgio Metta. **Grounding vision through experimental manipulation**. Philosophical Transactions of the Royal Society of London. Series A: Mathematical, Physical and Engineering Sciences, vol. 361, no. 1811, pages 2165–2185, 2003

Joint Action

towards language games

Scaffolding knowledge through interaction



Some fundamental & open questions

- What is the role of **pointing**?
- What **simulators** do we run in our heads?
- How do we ground a **theory of mind**?
- What is the role of **emotions**?
- How does the system represent a **goal**?
- How do developmental **stages** unfold?
- How do we represent **fluid knowledge**?
- ...

Curiosity & Intrinsic Motivation

What drives an autonomous system?

- *Physiological needs* => food, safety, ...
- *Curiosity* => intrinsic motivation

Curiosity = drive to decrease the rate of prediction error¹
= drive to increase the amount of learning

1: Pierre-Yves Oudeyer; Frédéric Kaplan, "What is intrinsic motivation? a typology of computational approaches", Frontiers in Neurorobotics 2007

The playground experiment



Shared verbal and motor lexicon acquisition

- Evolve shared conventionalized symbols to ground visual categories, motor skills, gestures (lexical level), social interaction patterns
- First stages of joint action
- Role of pointing
- Turn taking & imitation
- Role of emotions
- Artificial Curiosity in social context



- Luc Steels. **Experiments in cultural language evolution**, volume 3. John Benjamins Publishing Company, 2012
- J. Grizou, M. Lopes & P.Y. Oudeyer. **Robot learning simultaneously a task and how to interpret teaching signals**. In IEEE-RAS International Conference on Humanoid Robots, 2012
- L.W. Barsalou. **Grounding symbolic operations in the brain's modal systems**. In Embodied grounding: Social, cognitive, affective, and neuroscientific approaches. Cambridge University Press, 2008
- L. Steels. **Language games for autonomous robots**. Intelligent Systems, IEEE, vol. 16, no. 5, pages 16–22, 2001
- Luc Steels & Manfred. Hild. **Language grounding in robots**. Springer, 2012

Extending Construction Grammars

- Beyond lexicon, ground compositional structures like grammar or composite action plans.
- **Construction Grammars**
- Relationship between **language** and **action**
- Role of **simulation**, in particular in the development of a theory of mind
- Scaffolding **higher-level cognitive** functions.



- Luc Steels. **Design patterns in fluid construction grammar**, volume 11. John Benjamins Publishing Company, 2011
- Benjamin Bergen & Nancy Chang. **Embodied construction grammar in simulation-based language understanding**. Construction grammars: Cognitive grounding and theoretical extensions, pages 147–190, 2005
- E. Bates & J.C. Goodman. **On the emergence of grammar from the lexicon**. The emergence of language, pages 29–79, 1999
- A. Cangelosi, G. Metta, G. Sagerer, S. Nolfi, C. Nehaniv, K. Fischer, J. Tani, T. Belpaeme, G. Sandini, F. Noriet al. **Integration of action and language knowledge: A roadmap for developmental robotics**. Autonomous Mental Development, IEEE Transactions on, vol. 2, no. 3, pages 167–195, 2010.
- L.W. Barsalou. **Grounded cognition**. Annu. Rev. Psychol., vol. 59, pages 617–645, 2008

AI Lab Research Program

Five axis of research



Structuring
Perception
Layers

-
*Patterns
extraction*

- Deep Learning
- Slow Feature Analysis
- Multi-modal cross-correlation

Categorization
For Visual and
Motor Schemes

-
*Environment
modeling, IM*

- Proto-object
- Simple motor skills
- Simple causality
- Active Learning
- Intrinsic Motivation

Theory of Mind, Emotions

-
Social modeling

Grounding of language games and
social interactions

-
Social Motivation Principle

- Language games => how do they evolve?
- Role of pointing, Theory of Mind
- Turn taking & imitation
- Selectionist approach

Grounding of
lexicon and
grammar

-
*Construction of
Meaning*

- Construction Grammars
- Relation language & action
- Role of simulation
- Scaffolding higher-level cognitive functions.

Human Cognitive Developmental Map project

- Gather peer-reviewed known facts about the cognitive and language development of children in a centralized open wiki.
- Examples:
 - Age 12-15 months: coordinated joint engagement: active triadic interaction, directing other's actions, calling for reengagement if interrupted, taking the other's turn (Bakeman, 1984)*
- Allow for wikipedia style **discussion** and **debates** on “Discussion” page, revision history, list of published **references**, etc.
- **Timescale** representation of selected fact categories based on tags
- **Ranking** of contributors, ranking of fact reliability

Conclusions

- **Developmental Robotics as a way to evolve complex representational structures in grounded systems**
- **Grounding is key to AI, to enable simulation-based inferences, creativity and insight**
- **Key challenges: evolve categorization and joint interactions between agents**
- **AI Lab: possibility for joint labs, collaborative projects, partnerships, open positions**

