



OSAKA UNIVERSITY

iSYSLab
Intelligent Systems Laboratory

Robot Learning Group

JST
Japan Science and Technology Agency
国立研究開発法人
科学技術振興機構



Human-Machine
Harmonious
Collaboration

UEC
TOKYO



AIRC
人工知能先端研究センター
Artificial Intelligence eXploration Research Center



EMOTION
Emotion Model For Communication

ChiCaRo



AI × Robotics:

Towards Human-like General Intelligence

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Future of AI/Robots where are we heading?

AI/ロボットのミライ

- tools or partner
- 道具かパートナーか？



単なる道具
意味など理解する必要ない
理解するのは人間
大量のデータと計算パワーがあればOK

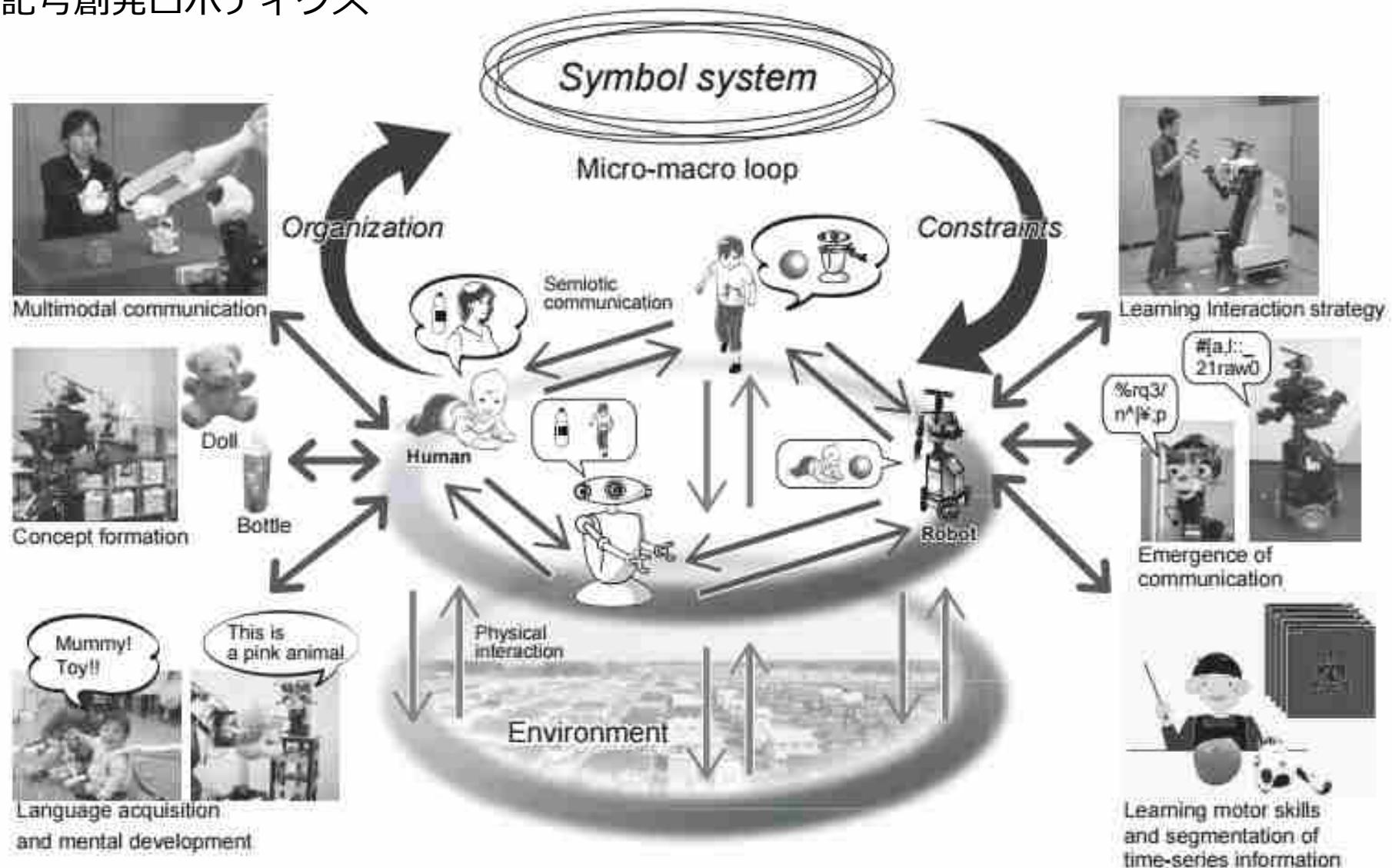


人と関わる相手
意味を人間のように理解する
大量のデータと計算パワーとは異なる問題
⇒意味の理解や相手の理解の本質は？

AI as a partner
Need to understand like humans
Different problem with data and computations
⇒ What is understanding ?

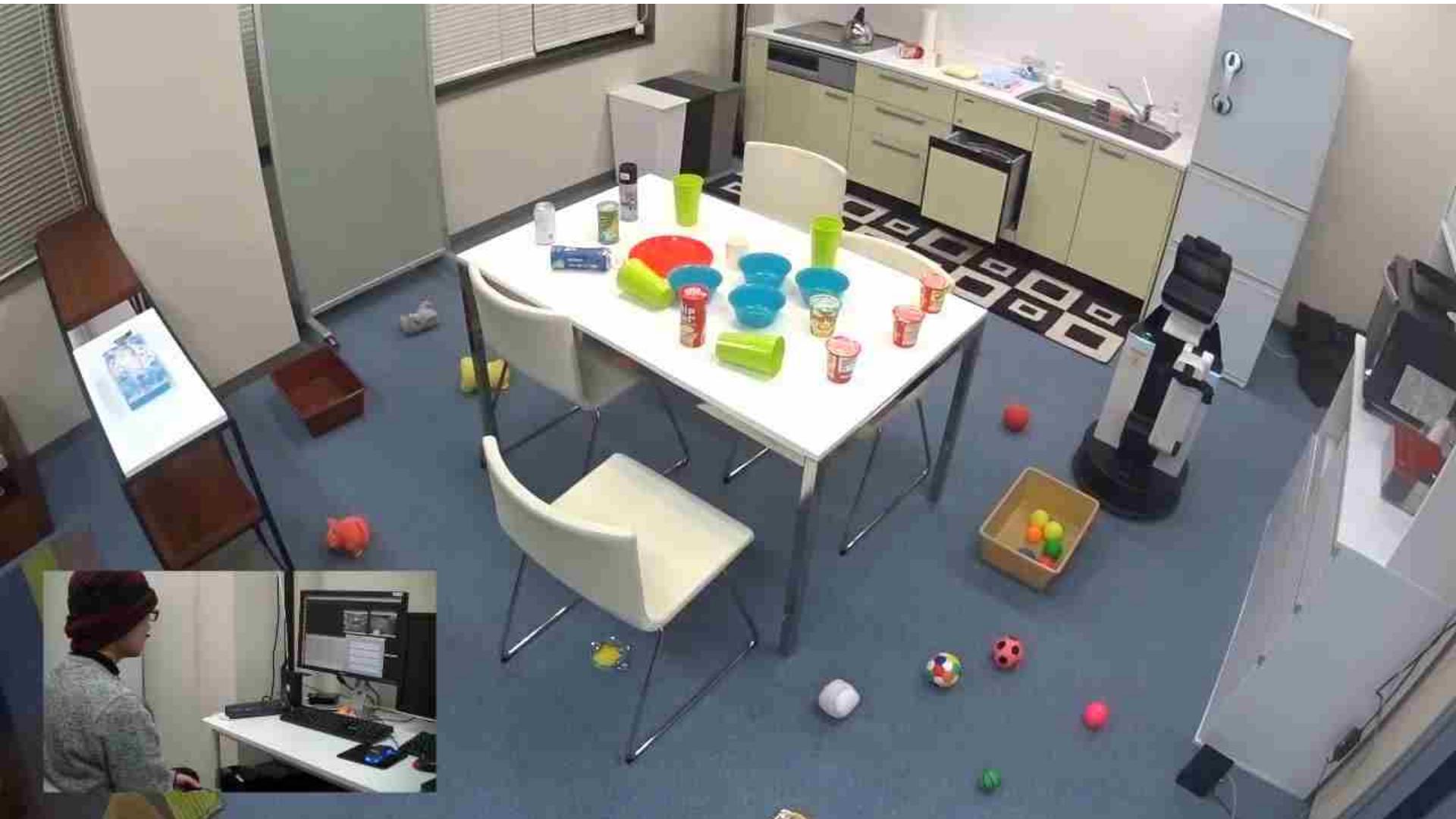
Symbol Emergence in Robotics

記号創発ロボティクス



Example of home service robot

家庭用ロボットの例



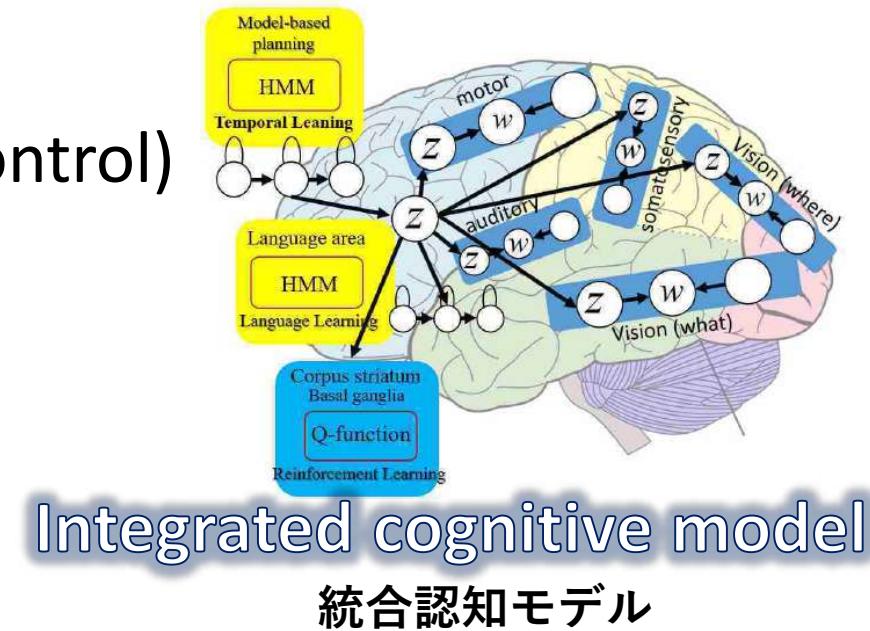
Integrate as a whole!

全体を統合する！

身体	
body	
感情	
affect	
記号	
symbol	
論理推論	
logical inference	
プランニング	
planning	
反実仮想	
counterfactual	
意識	
consciousness	



身体制御(motor control)
言語(language)
推論(inference)
因果(causality)



- How can we integrate everything?

What is “Intelligence”?

知能とは何か？

Efforts

to see what cannot be seen

見えないものを見る努力

Prediction

予測



Language an aspect of intelligence

- ✓ ここにはないものについて語ることができる
- ✓ 言葉の意味とは？
 - 正しい意味などどこにもない
- ✓ 言葉を理解するとは？
 - 予測という視点でとらえることができる
- ✓ その言葉を聞いたときに予測したもの
 - それがその言葉のその人にとってのその時の意味
- ✓ ロボットにできない理由はない！

We can talk about something even doesn't exist

What is the correct meaning?

There must be "NO" correct meaning

What is understanding?

"Prediction" can answer this question

Something that comes to mind

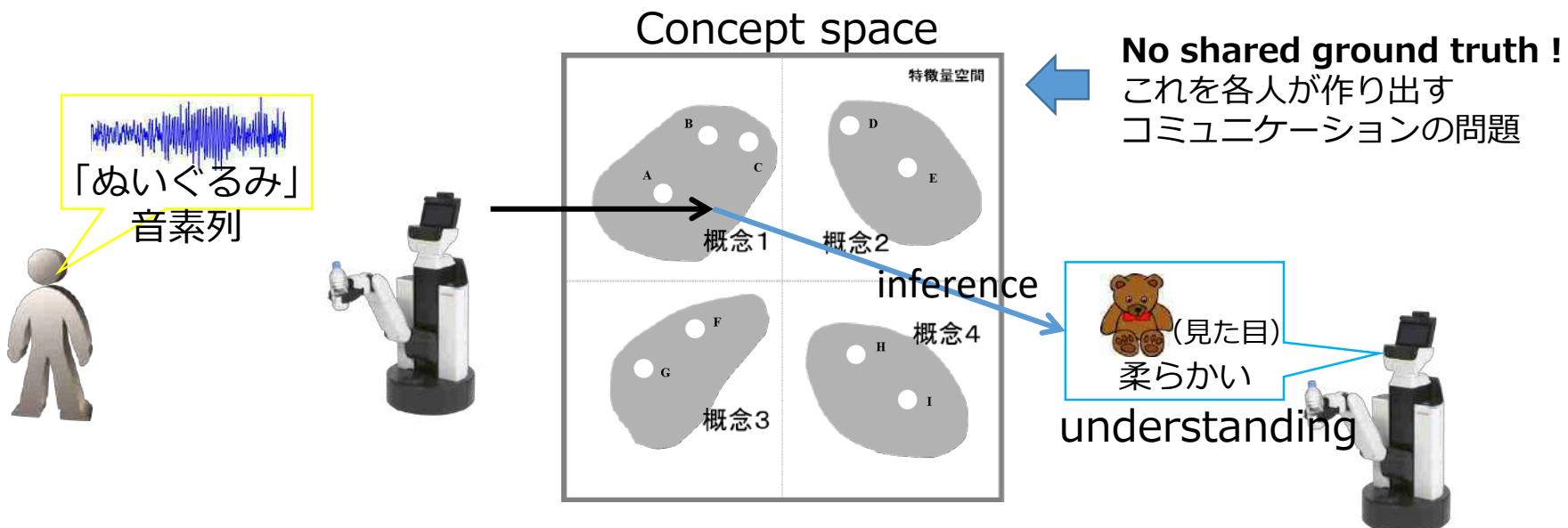
That is what it "means" for the person at the moment

I'm sure robots can do this!

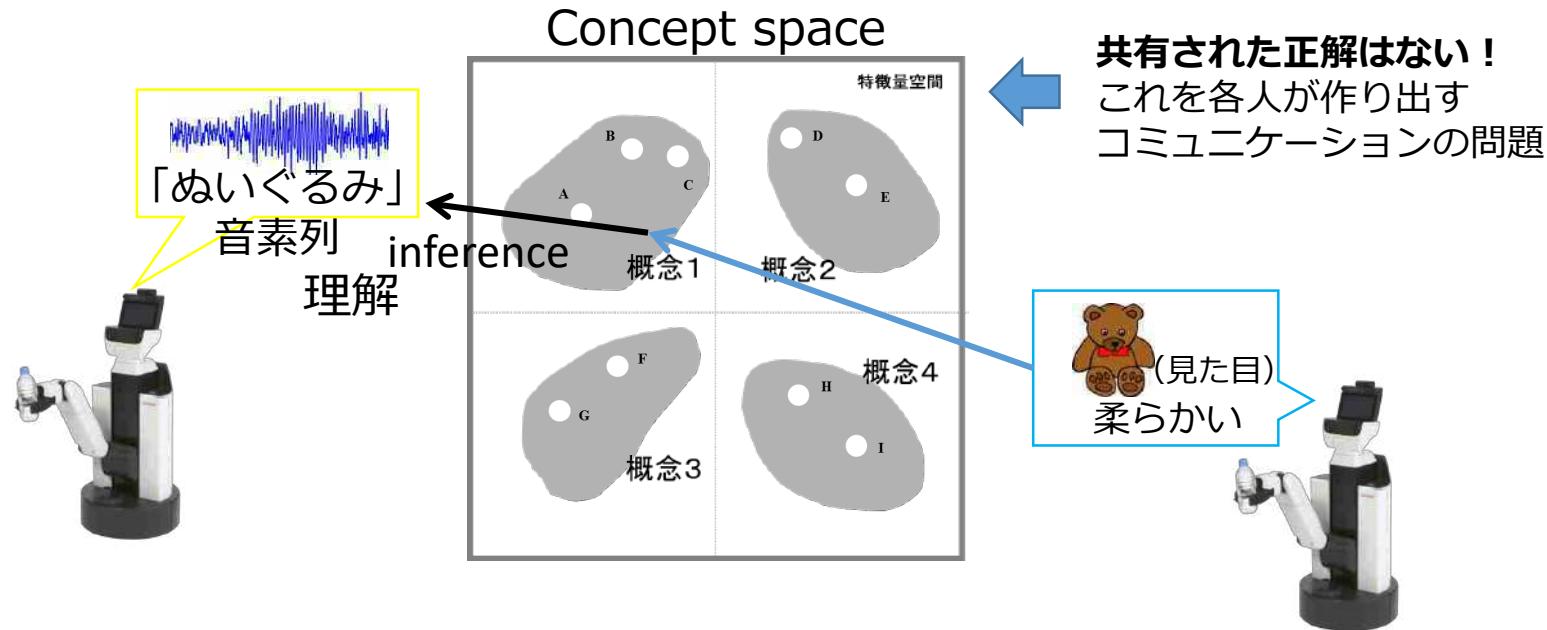


What is “understanding”? Naïve view

- ロボットによる実世界理解
 - “**理解**”：概念を通した未観測情報の予測
 - “**意味**”：予測した内容
 - “**概念**”：経験（マルチモーダルデータ）のカテゴリ分類によって形成される
- Symbol grounding problem



The other way round

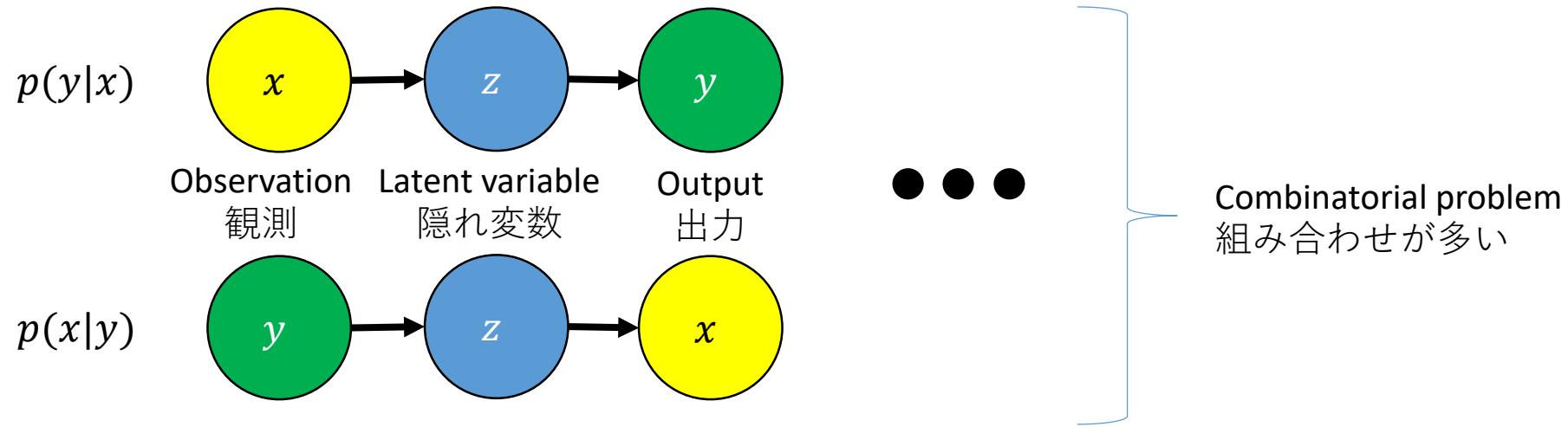


- 単一の識別タスク $p(y|x)$ をモデル化するだけではダメ
- 同時分布をモデル化する

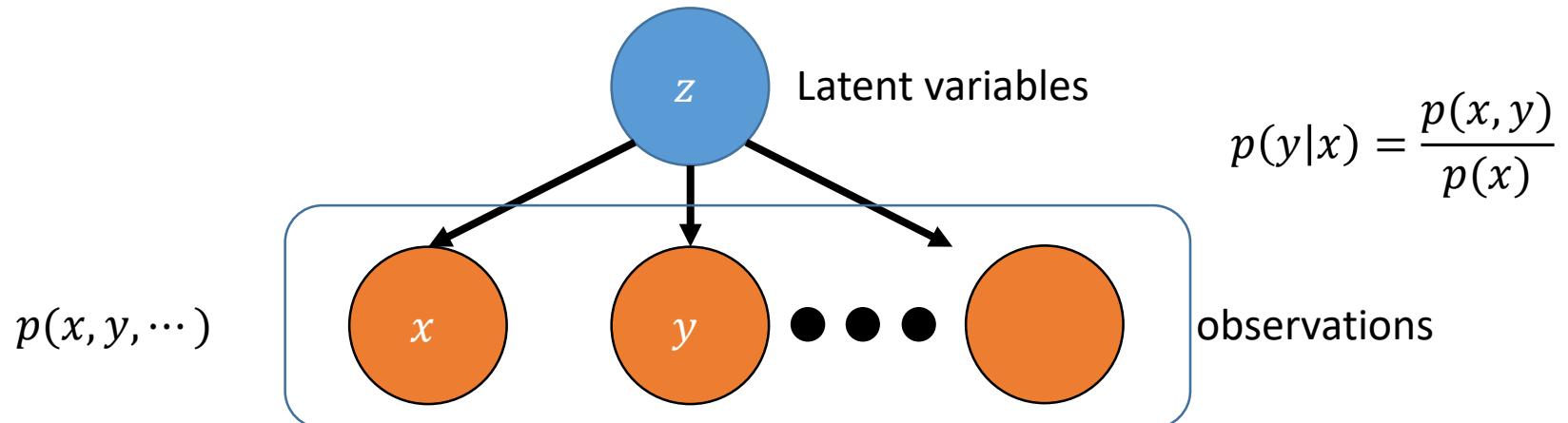
$$p(x, y, \dots) = \sum_z p(x, y, \dots | z)p(z)$$

Multimodal Generative Model (unsupervised learning)

Multimodal supervised learning (discriminative model)

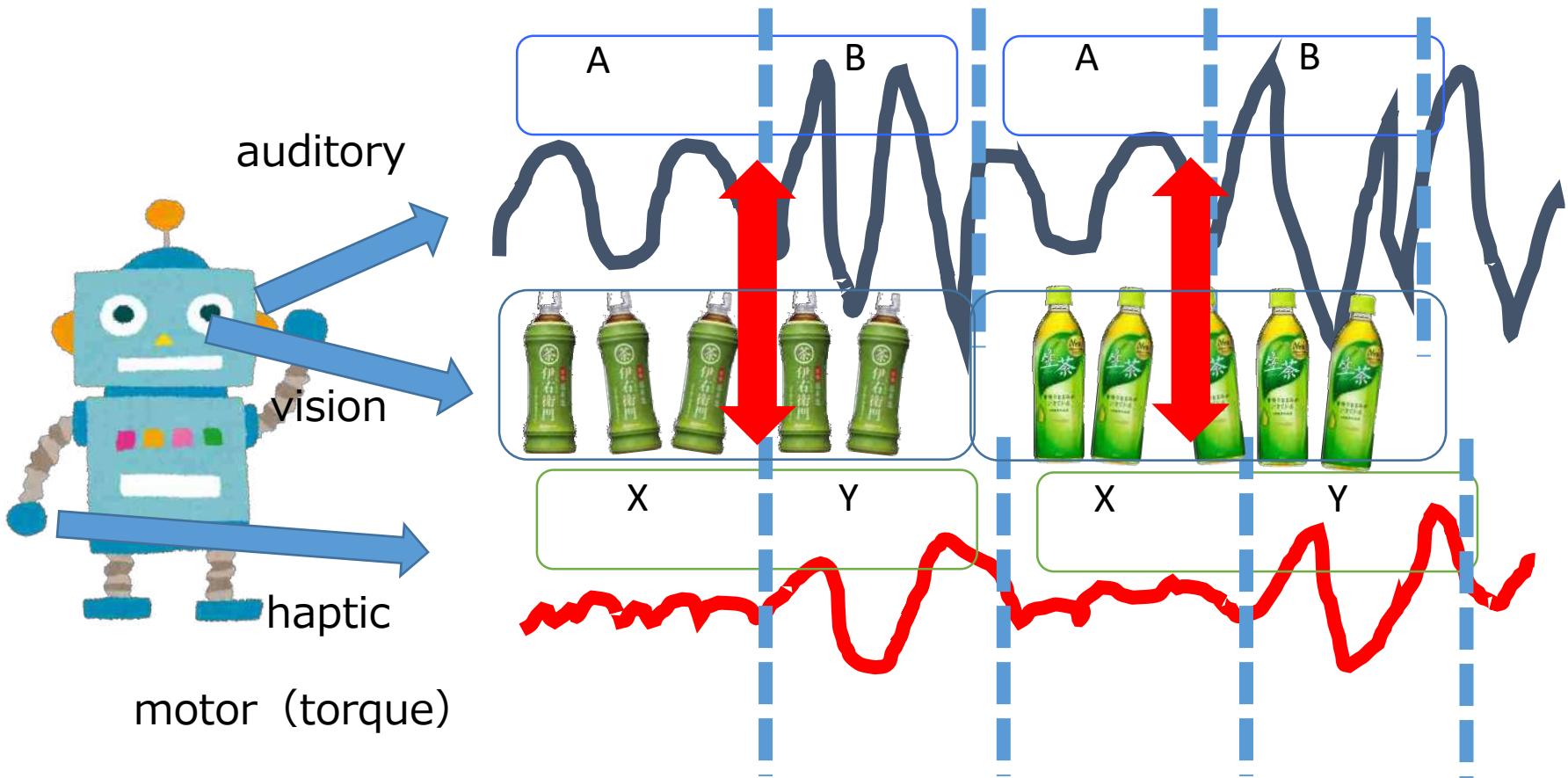


Multimodal unsupervised learning (generative model)



Idea

- Segmentation and categorization
- Gaussian process and hidden Markov models

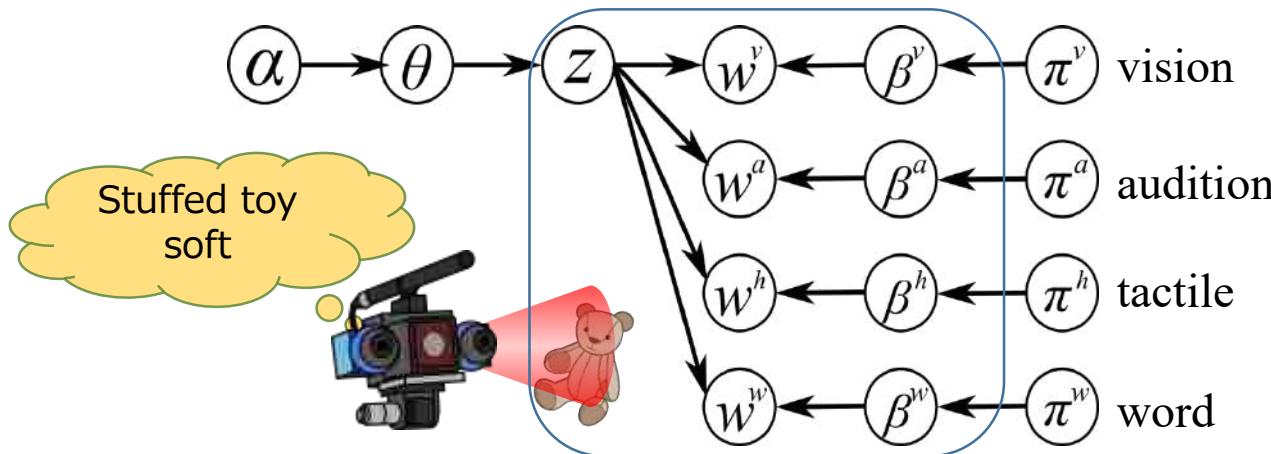


Multimodal categorization

□ Categorization of multimodal data

- ・ マルチモーダルカテゴリゼーション
- ・ **Multimodal Latent Dirichlet Allocation** (MLDA, MHDP, …)

[Nakamura+ 09]



α : Dirichlet prior

θ : multinomial parameters

z : categories

w^* : multimodal information

β^* : multinomial parameters

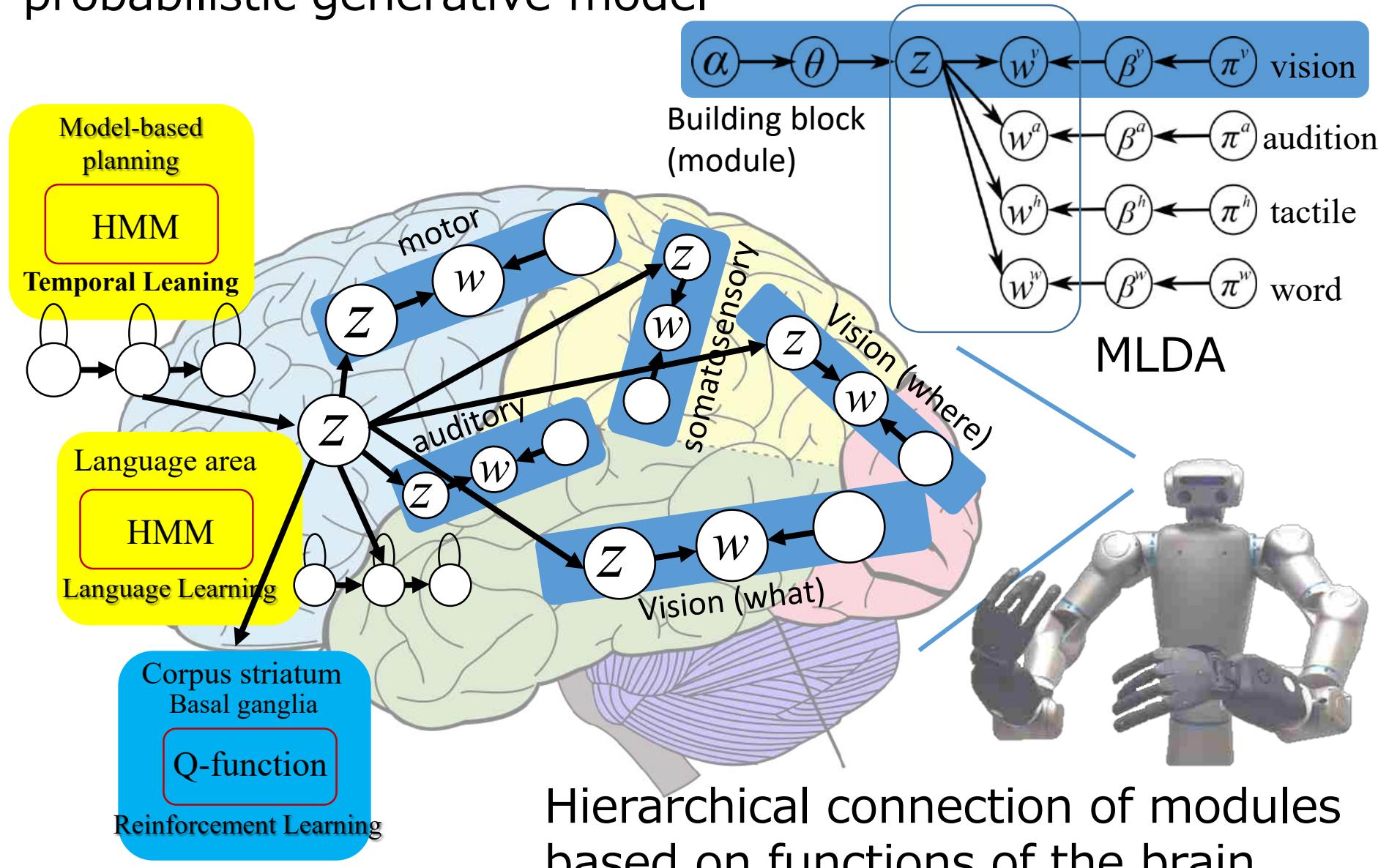
π^* : Dirichlet prior

Inference of the parameters β^* and θ by Gibbs Sampling

[Nakamura + 09] Nakamura,T. et al., Grounding of word meanings in multimodal concepts using LDA, in Proc. IROS2009, pp.3943–3948, 2009

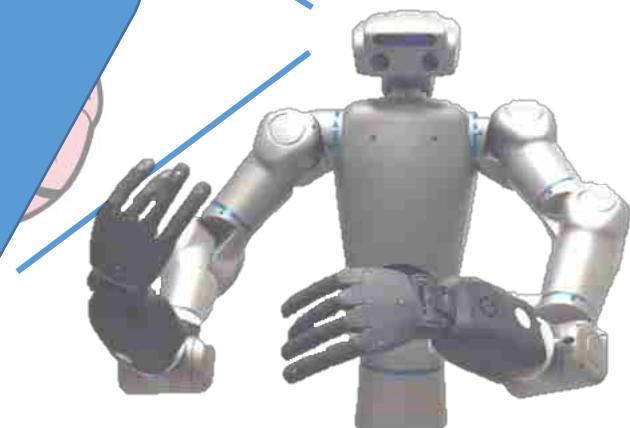
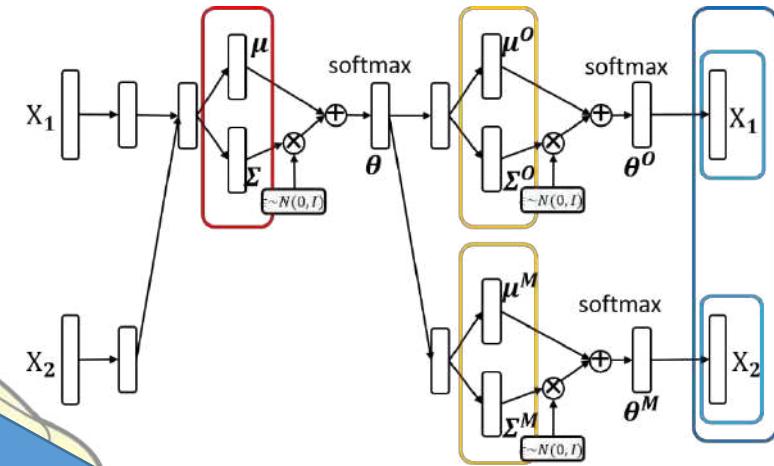
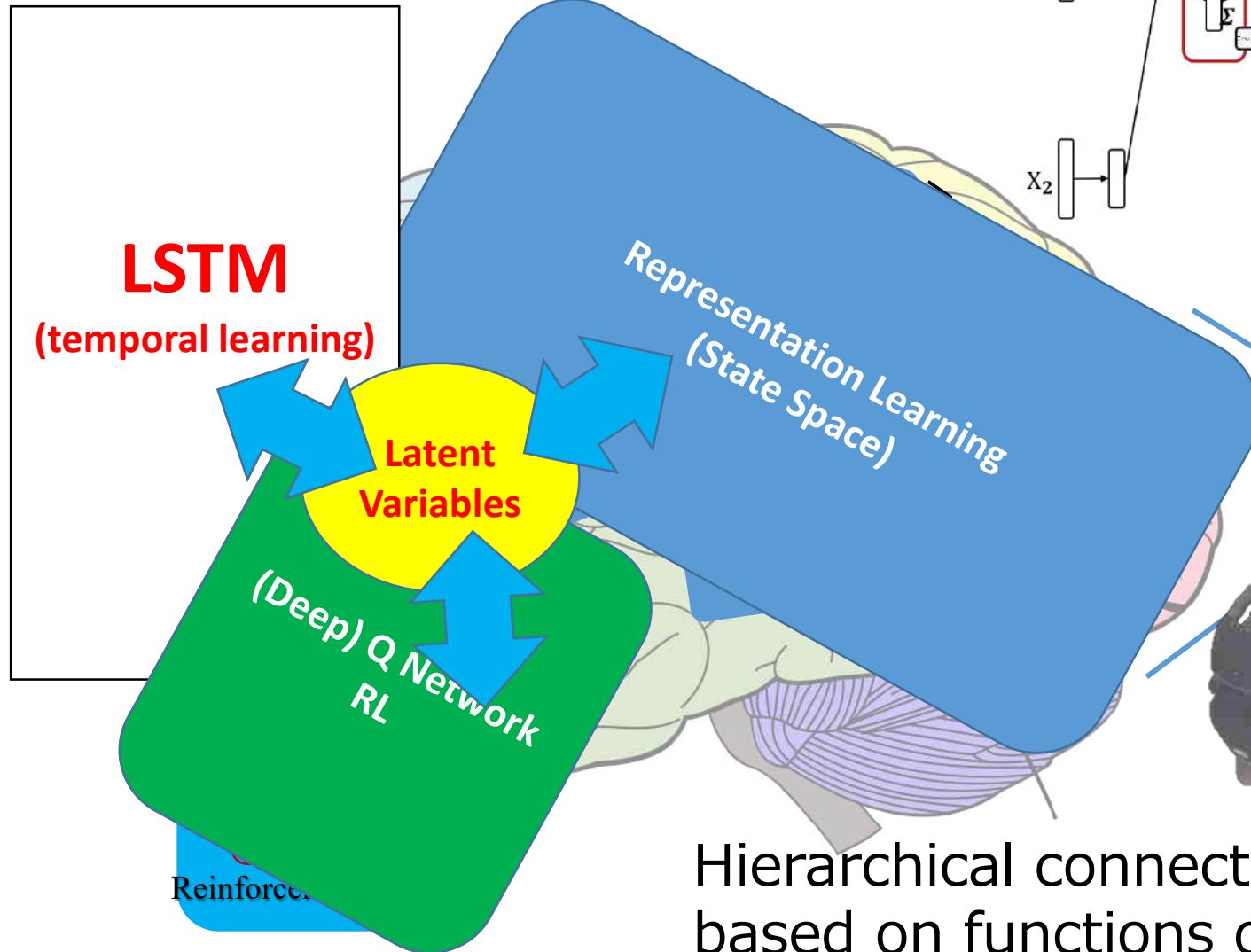
Integrated cognitive model

probabilistic generative model



Integrated cognitive model

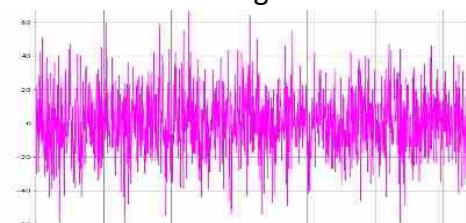
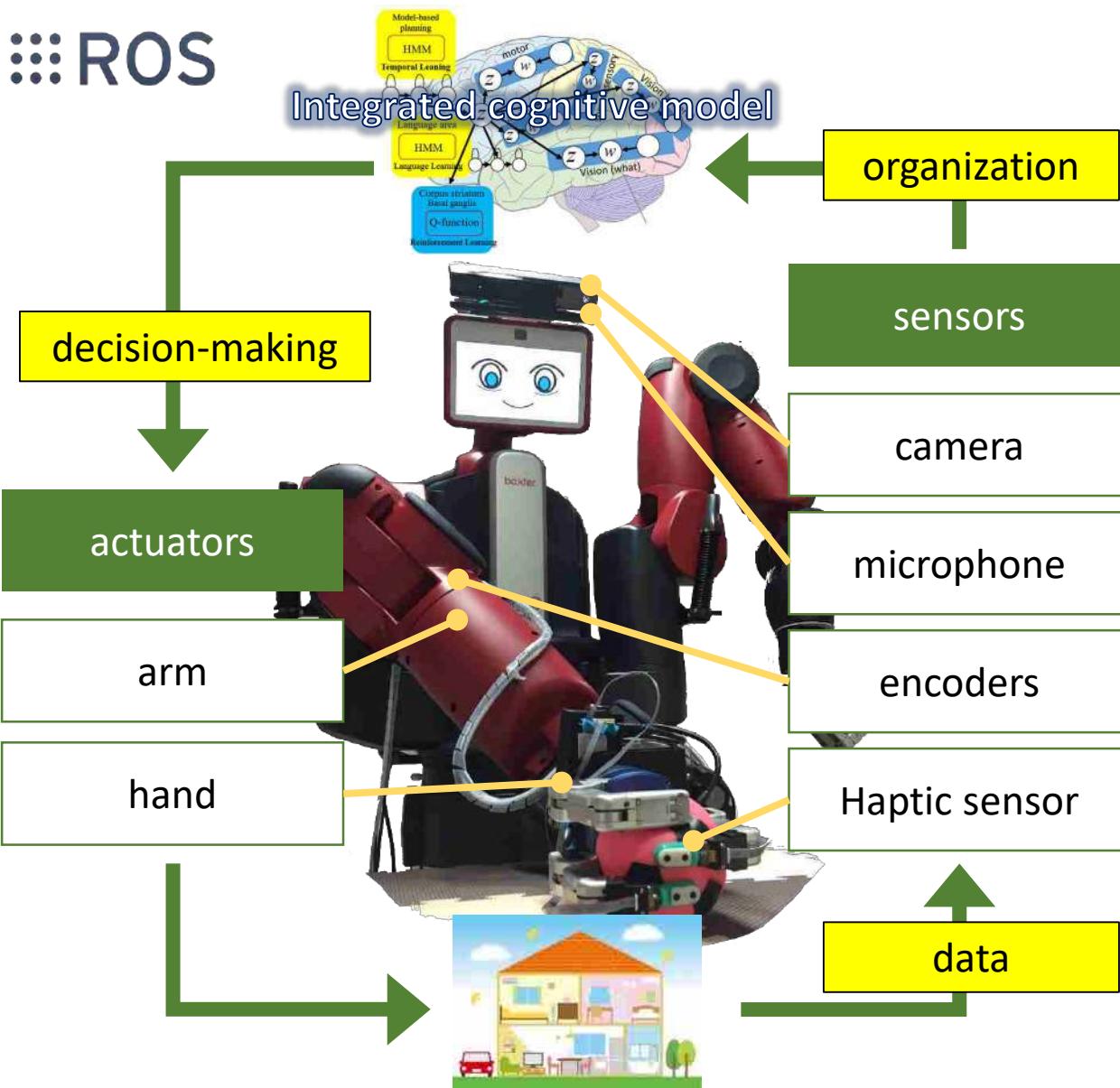
deep generative model



Hierarchical connection of modules based on functions of the brain

Implementation on a real robot

ROS



Data from various sensors

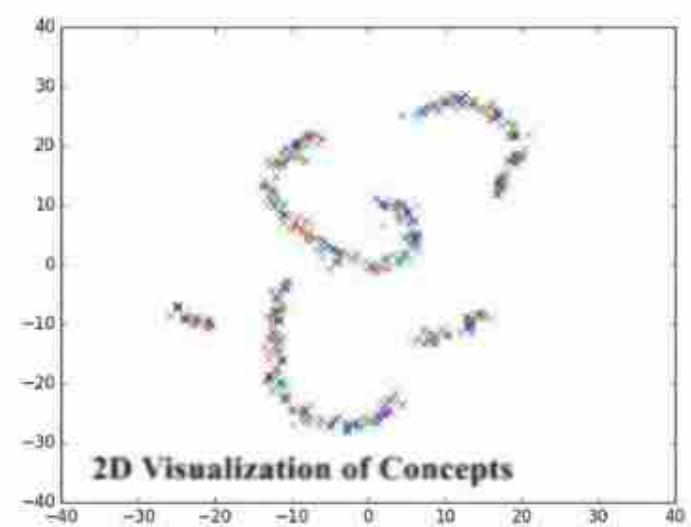
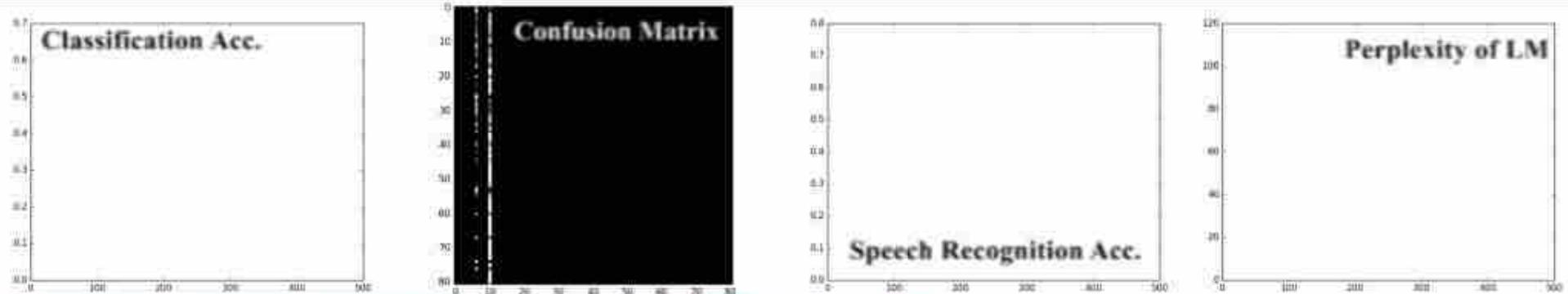
Long-term learning experiment [Nishihara+ 16]

- 3 ~ 5 時間/日 × 1か月 (100 時間以上)
- 500 物体 (81 カテゴリ) を使用
- Over 100 hours of interaction
- Using 500 objects (81 categories)



Long-term learning experiment [Nishihara+ 16]

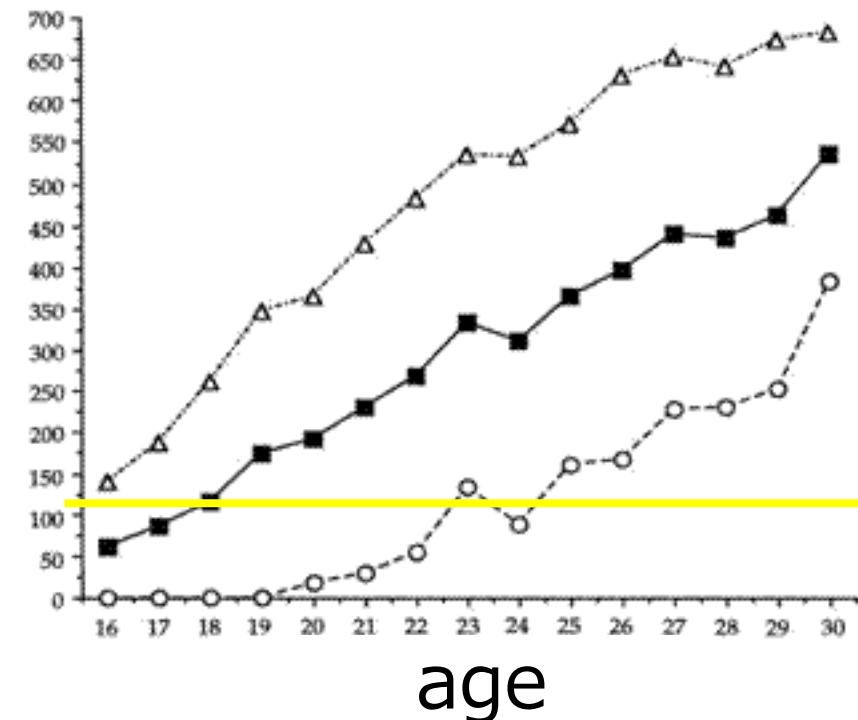
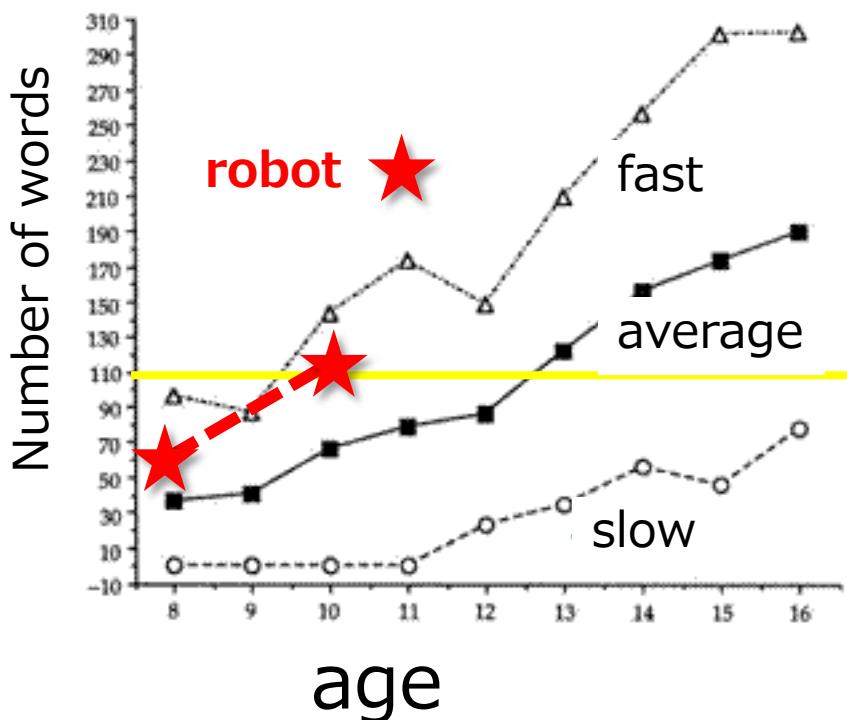
- 3000以上の発話（300ユニーク単語）100単語程度獲得
- 約7割程度の正解率



Vocabulary development

understanding

production



About 180 words in 16-month

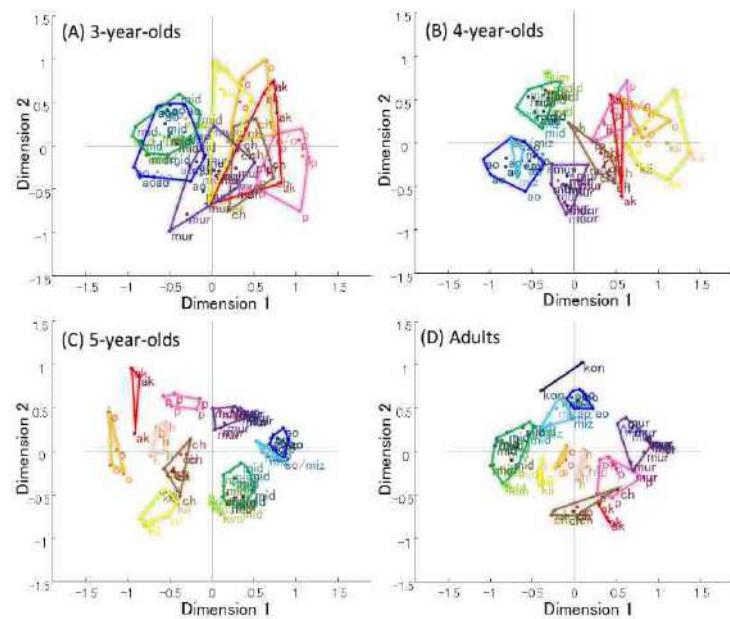
About 60 words in 16-month
About 1000 words in 3 years old

Findings

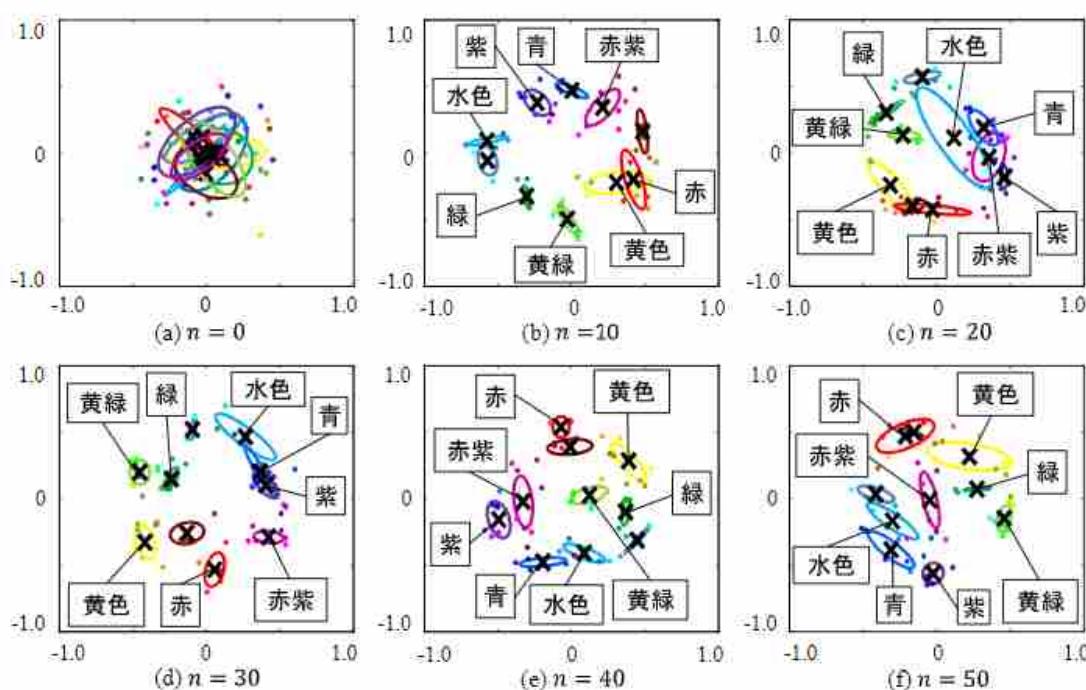
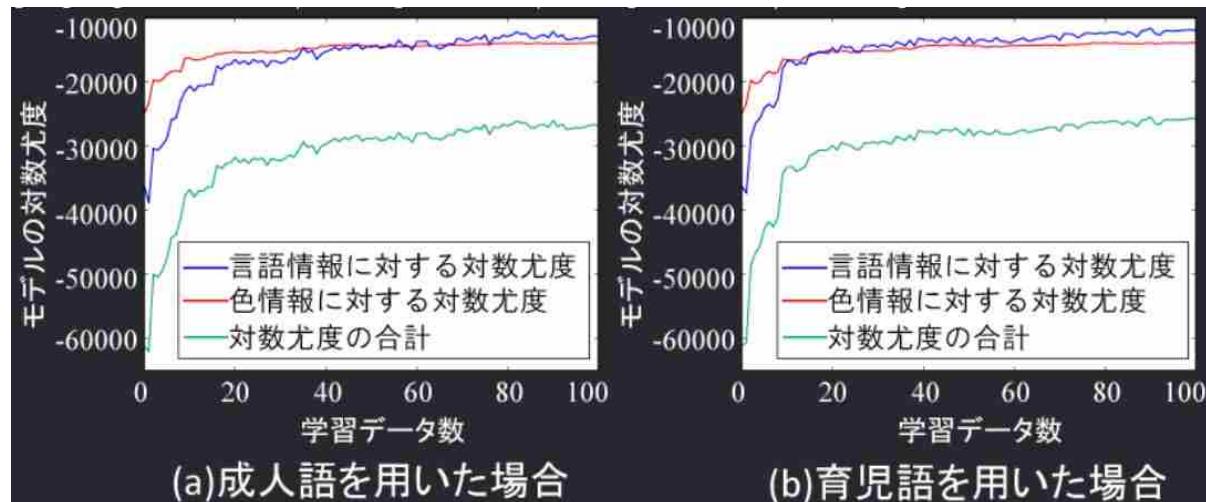
Suggestions obtained by analyzing robot learning

- ✓ The role of child directed speech [Funada+ 17]
- ✓ Learning of color [Funada+ 16]

M.Imai, et.al "Meanings of symbols emerge within a system" JSAI 2015



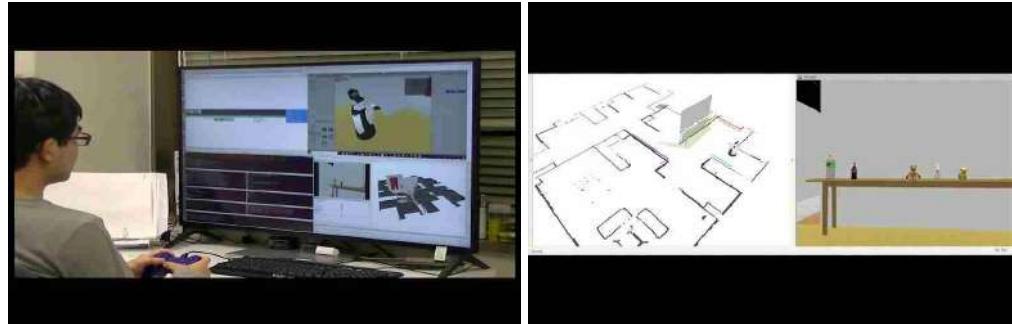
ak = 赤, ao = 青, ch = 茶色, h = 肌色, kii = 黄色, kim = 黄緑, mid = 紫, miz = 水色,
 mur = 桃, o = オレンジ色, p = ピンク色



Learning from teleoperation for domestic service robot



Iwata et al. "Learning and generation of actions from teleoperation for domestic service robots," IROS 2018



Miyazawa et al. "Integrated cognitive architecture for robot learning of action and language," Frontiers in Robotics and AI, 2019

Currently working on experiments @ three different real home environments



Three different real home environments



After learning (autonomous task execution)



Challenges

- Communication (sociality)
 - Self/other discrimination
 - Mirror neuron system
- Emotion/affection
 - Information on body (**interoception**)
 - Rewards
- Higher level cognitive functions
 - Induction, deduction, abduction
 - Causality
 - Counterfactual
 - Explainability
- Consciousness

