

Intuitive explanation of FEP related to cognitive neuroscience and developmental disorders

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“Predictive processing” in a broad sense

- Theory of computational rule of the brain explaining the process of *learning, perception/cognition*, and *behavior*.
- Proposed as independent studies using Bayesian inference models (Rao 1999, Friston 2011, etc.) and neural network models (Tani 2003).

Rao RP, Ballard DH (1999) Predictive coding in the visual cortex: a functional interpretation of some extra-classical receptive-field effects. *Nat Neurosci.* 2: 79-87.

Friston K, Mattout J, Kilner J. Action understanding and active inference. *Biol Cybern.* 2011 Feb;104(1-2):137-60.

Tani J (2003) Learning to generate articulated behavior through the bottom-up and the top-down interaction processes. *Neural Networks* 16: 11-23.

History of key concepts

19th century

“Unconscious Inference”



Hermann von Helmholtz
(1821 – 1894)

Our perceptions are not direct reflections of the world, but rather inferences made by our brain based on sensory inputs.

History of key concepts

19th century

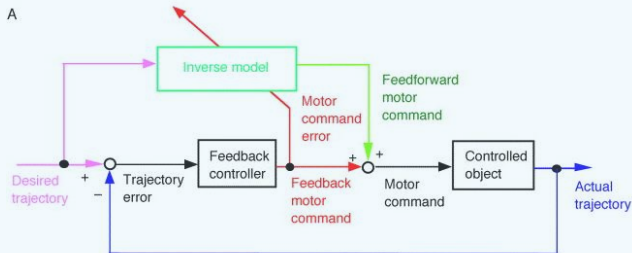
“Unconscious Inference”



Hermann von Helmholtz
(1821 – 1894)

1990s

“Internal model”



(Wolpert1996, Kawato1999...etc)

The brain constructs internal models to predict sensory outcomes of motor commands.

History of key concepts

19th century

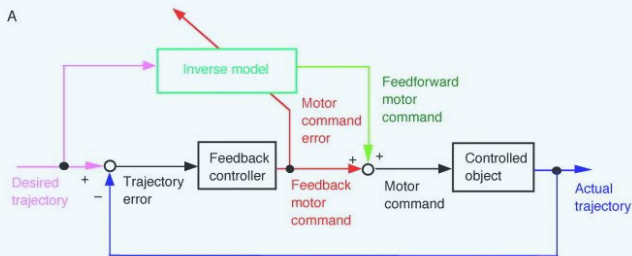
“Unconscious Inference”



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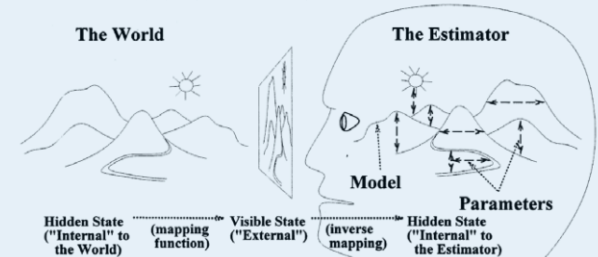
“Internal model”



(Wolpert1996, Kawato1999...etc)

2000s

“Predictive coding”, “Bayesian Brain”,
“Free energy principle”



Rao 1999, Tani 2003, Doya2007,
Friston2006, etc

The brain is constantly making predictions about incoming sensory data and updating these predictions based on the actual sensory input.

History of key concepts

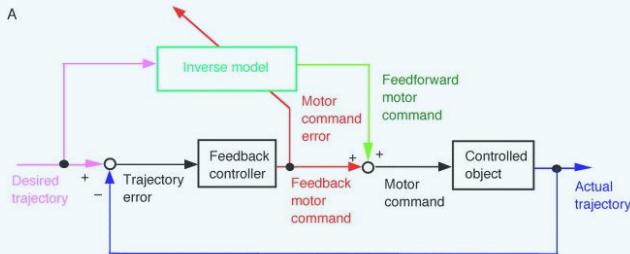
19th century

“Unconscious Inference”



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“Internal model”

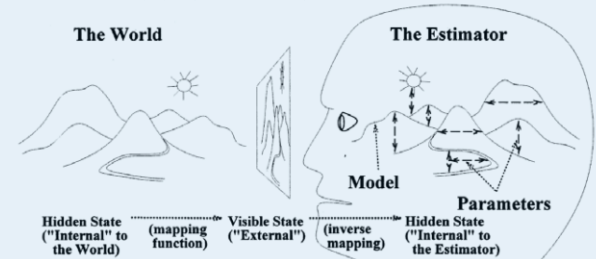


(Wolpert1996, Kawato1999...etc)

1990s

“Predictive coding”, “Bayesian Brain”,
“Free energy principle”

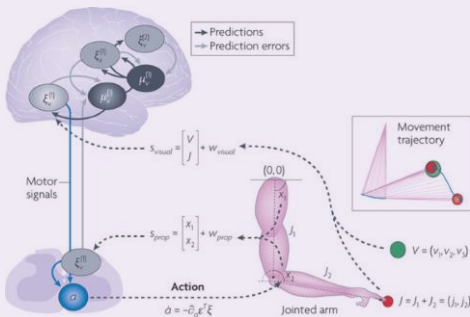
2000s



Rao 1999, Tani 2003, Doya2007,

2010s

“Active inference”

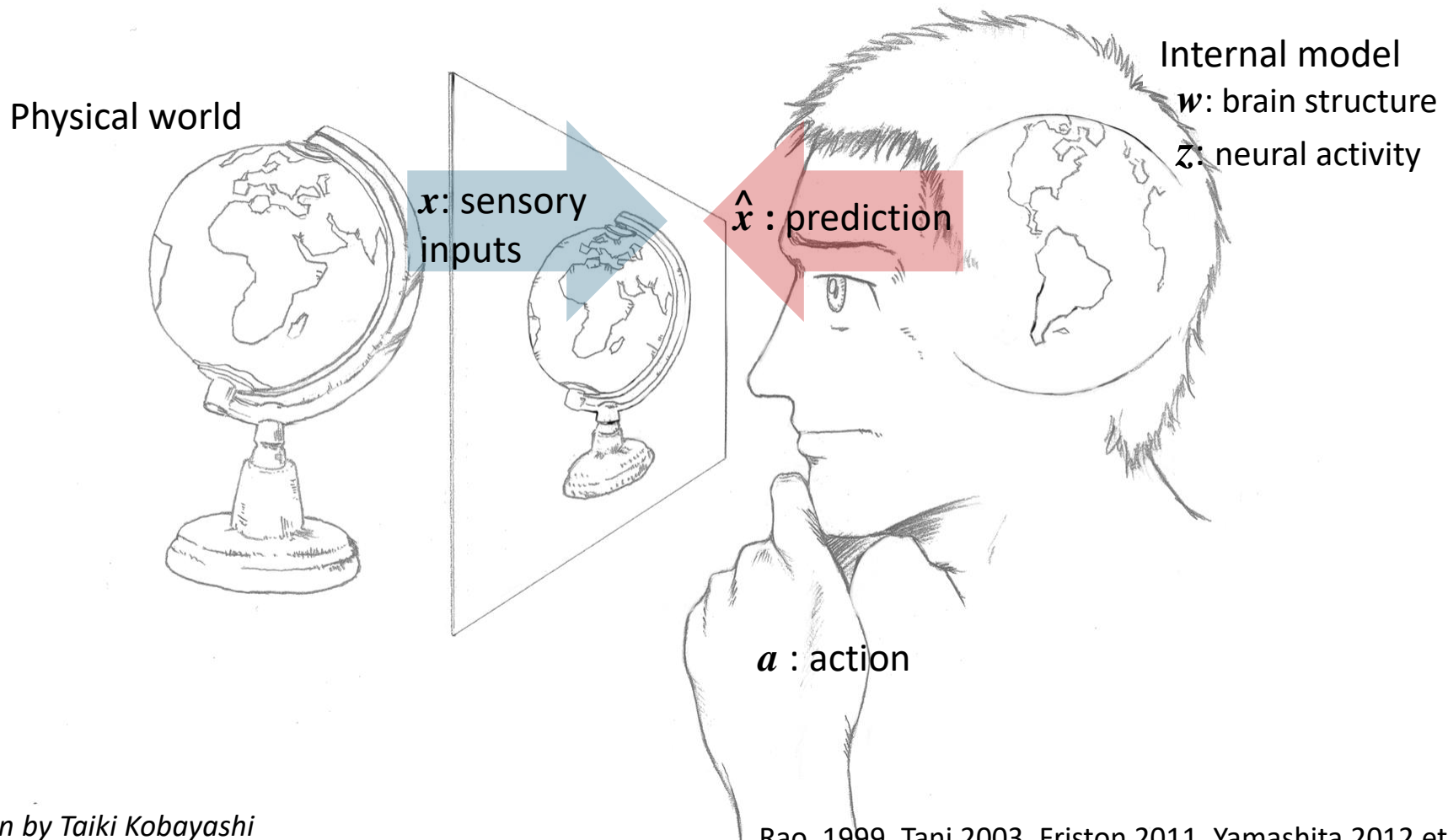


Friston2011, etc Nature Reviews | Neuroscience

Organisms don't just passively predict sensory input but actively engage with the environment to minimize prediction errors.

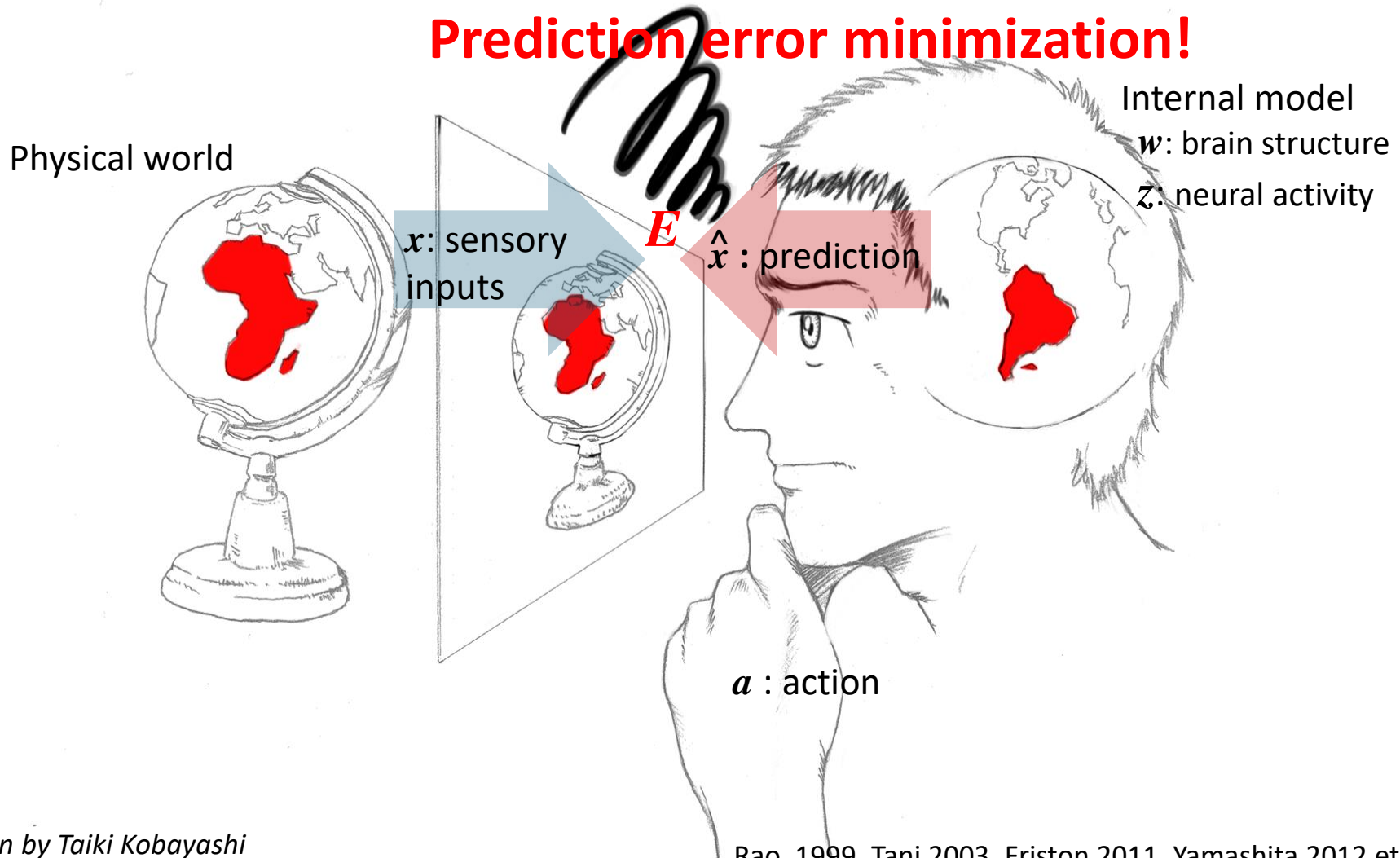
“Predictive processing” in a broad sense

- Brain is a “*predictive machine*” based on internal model of the world.



“Predictive processing” in a broad sense

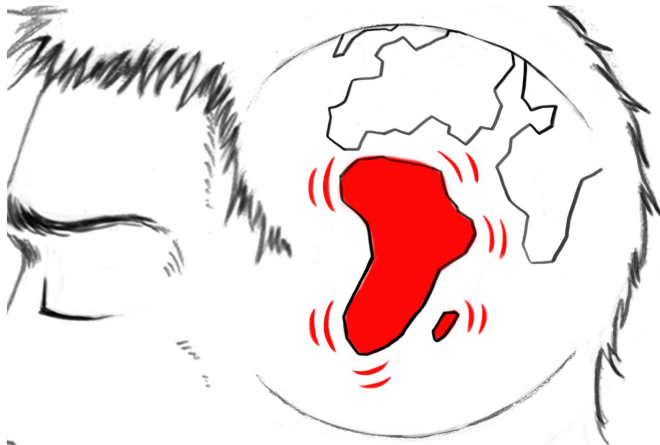
- Interacting with the world via computational rule of “prediction error (PE) minimization”



“Predictive processing” in a broad sense

- Three ways to minimize prediction error

Learning



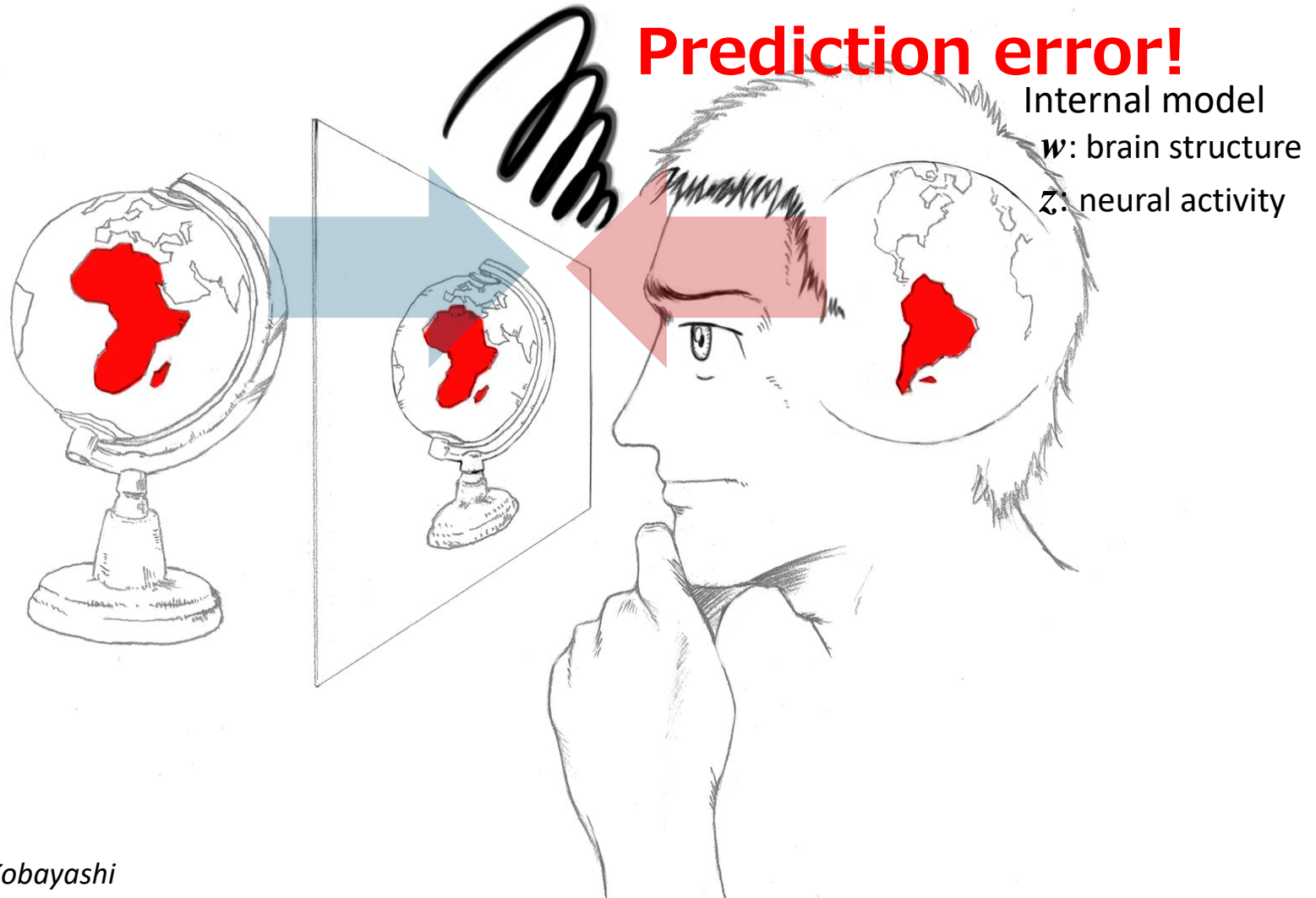
Perception/Cognition



Action



- When there is prediction error...

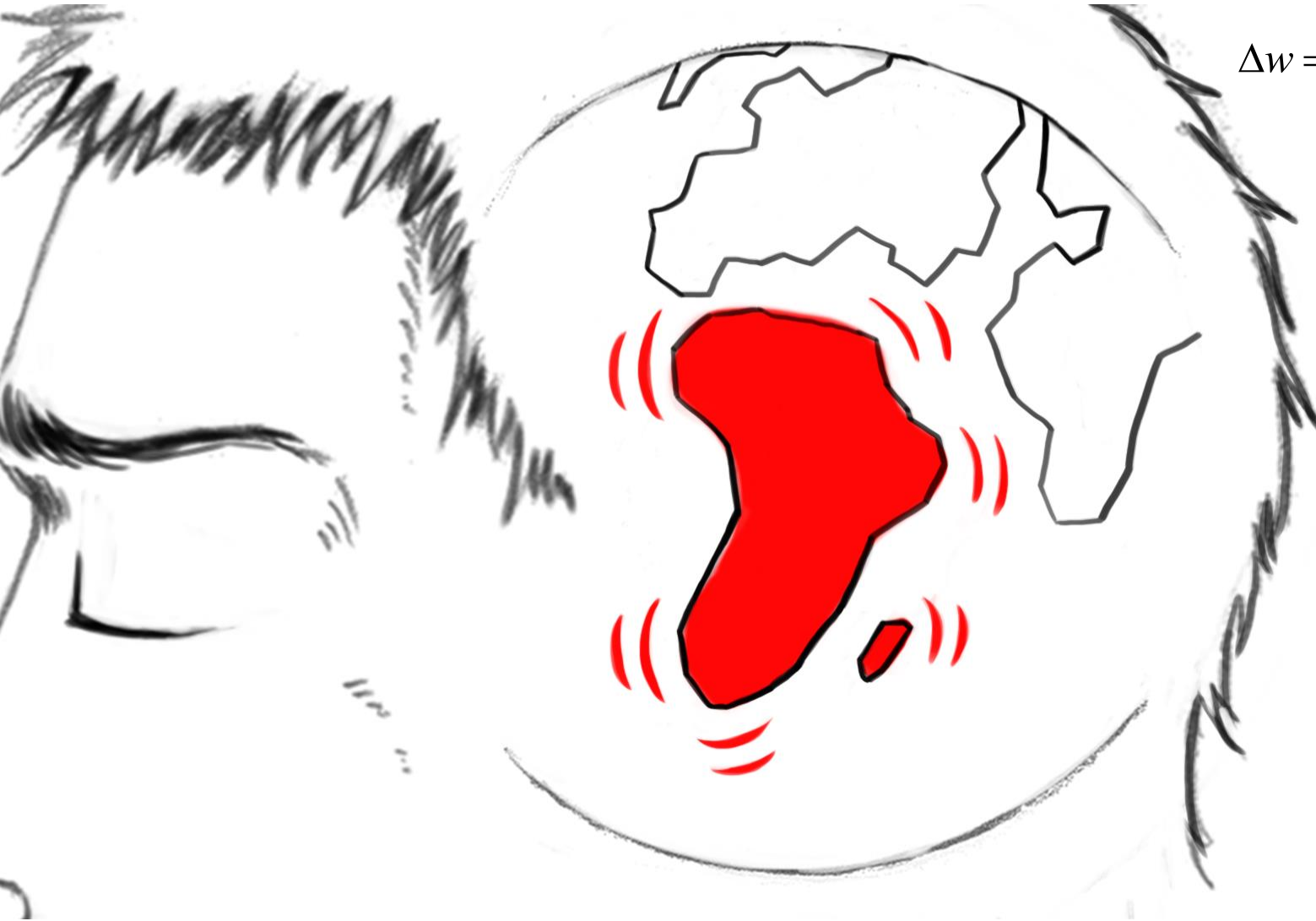


- Internal model might be wrong...



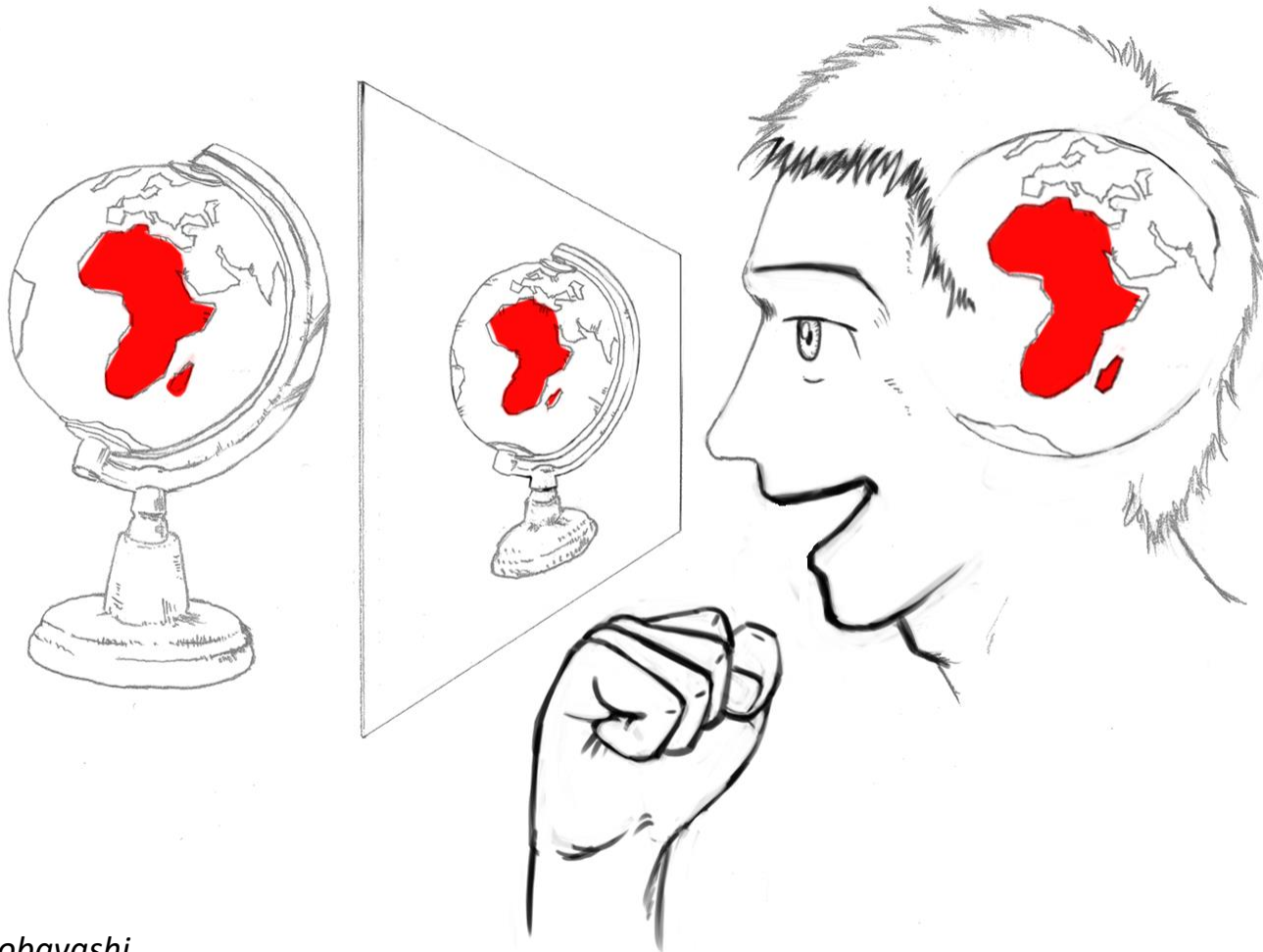
- Update internal model (brain structure)...

$$\Delta w = -\alpha_w \frac{\partial E}{\partial w}$$

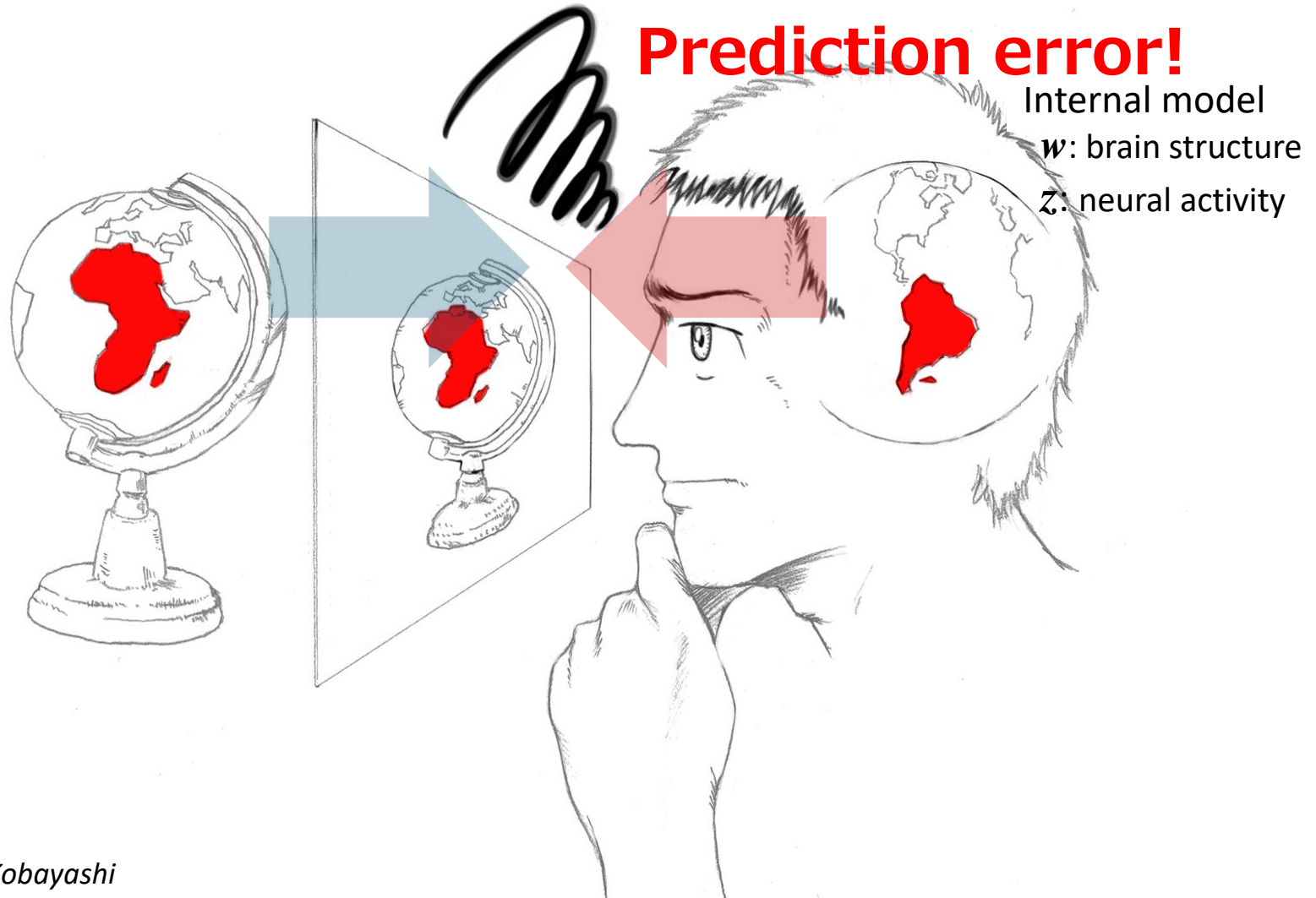


- PE minimization can be achieved through the modifications of internal model (**learning**).

$$\Delta w = -\alpha_w \frac{\partial E}{\partial w}$$



- When there is prediction error...



- Prediction might be wrong...

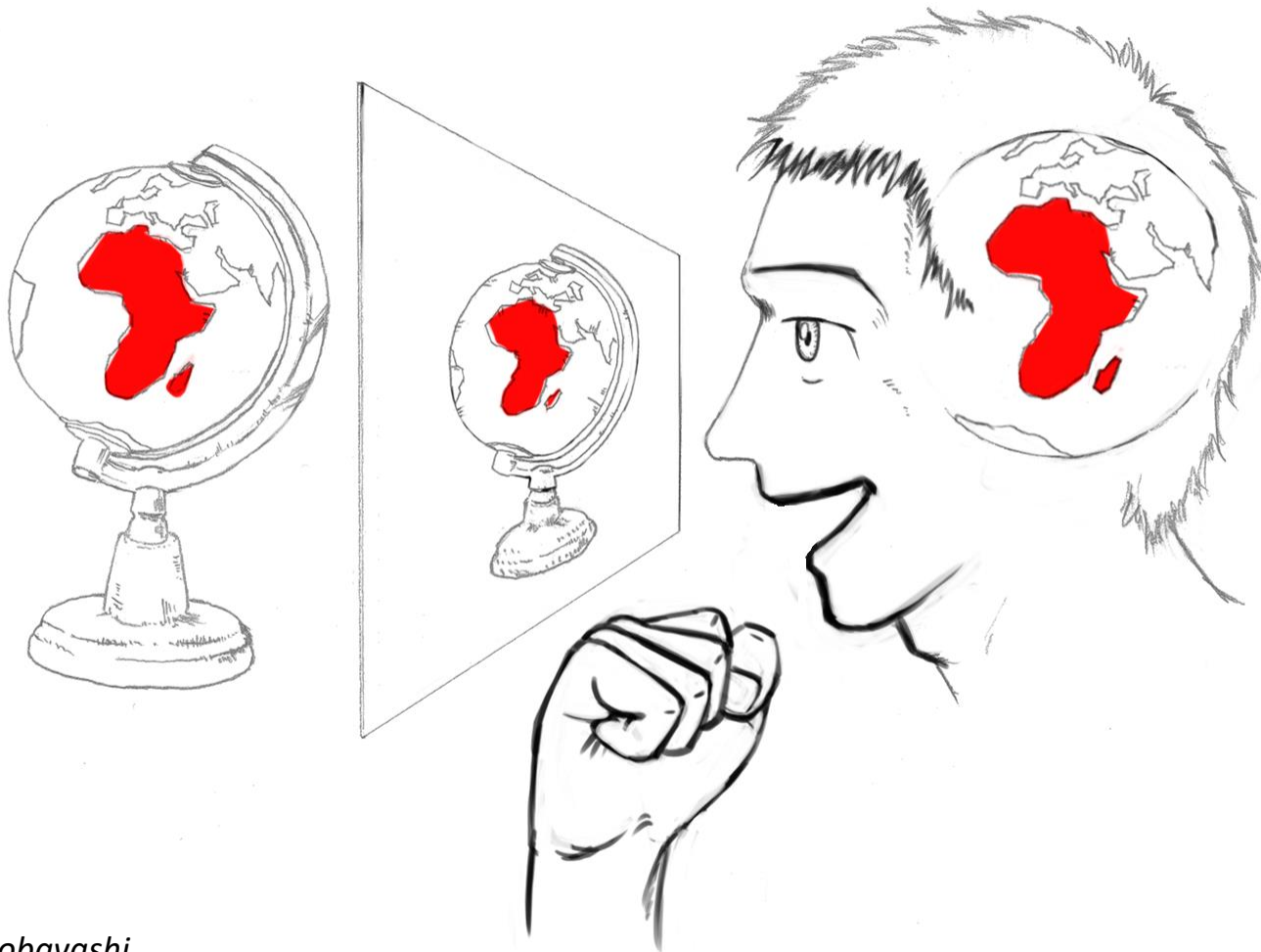


- Change the prediction via updating brain states

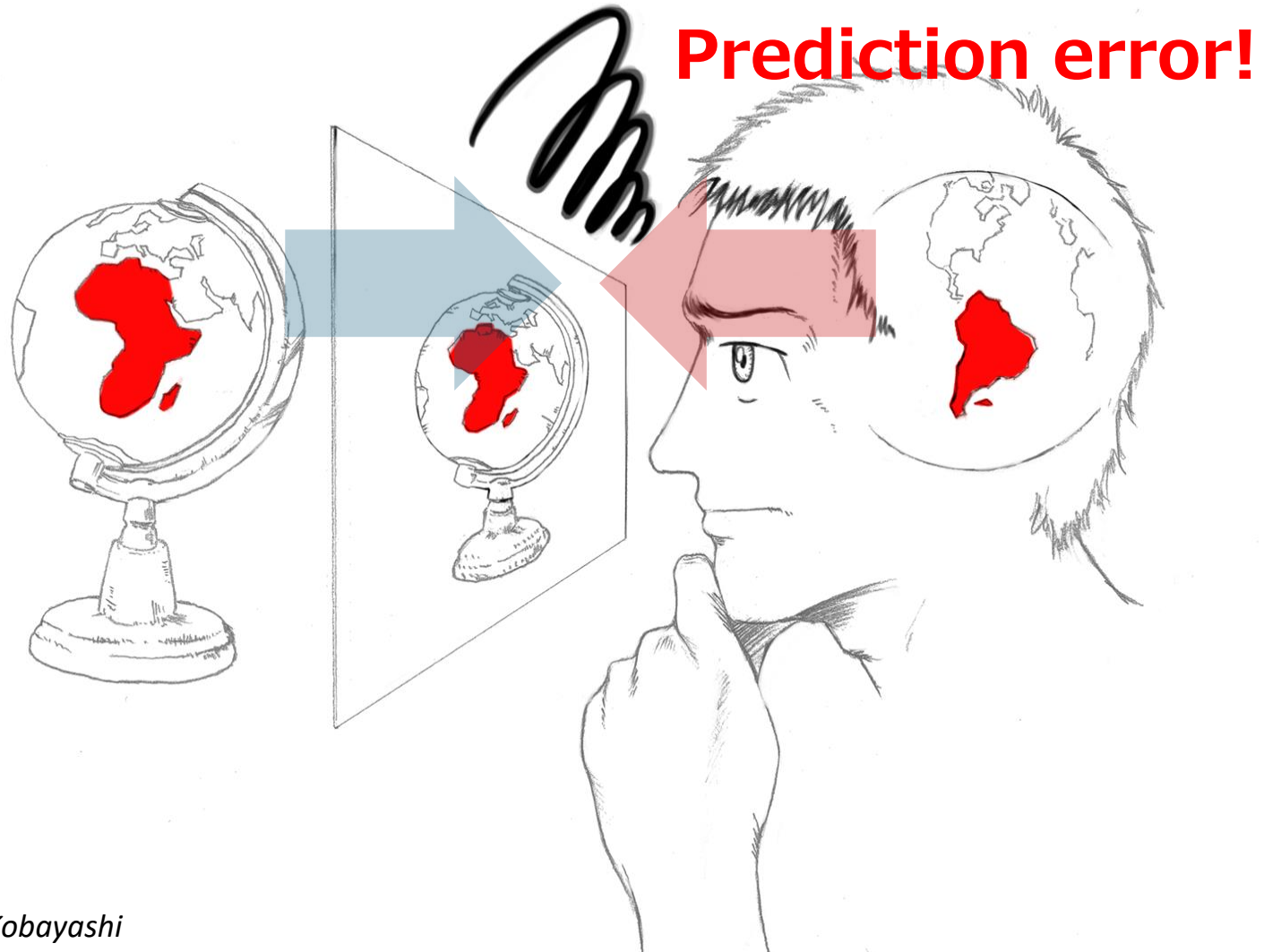


- PE minimization can be achieved through the modifications of internal/brain states
(perception/cognition)

$$\Delta z = -\alpha_z \frac{\partial E}{\partial z}$$



- When there is prediction error...



- External world might be wrong...



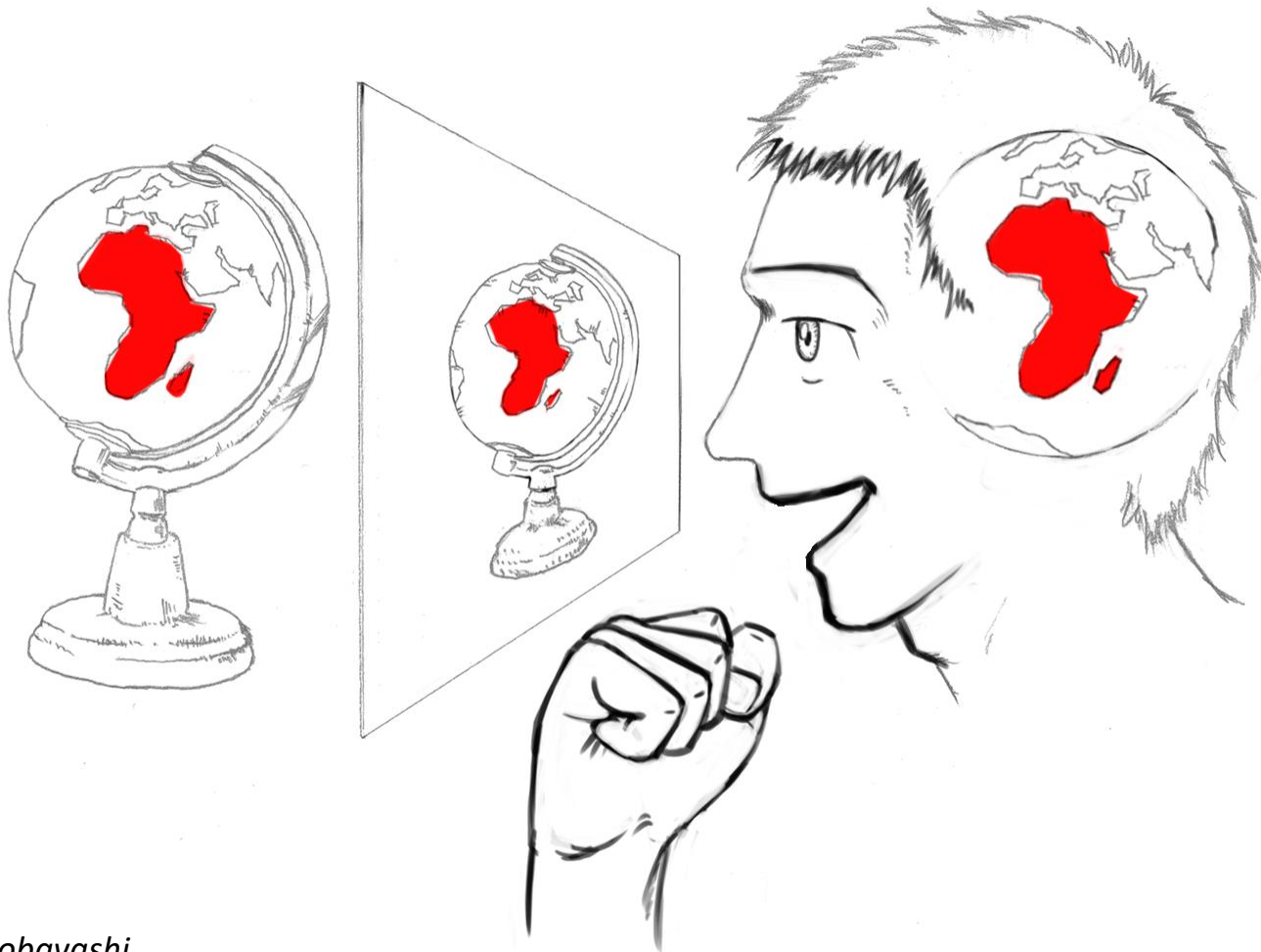
- Change the external world through action...

$$\Delta a = -\alpha_a \frac{\partial E}{\partial a}$$



- PE minimization can be achieved through the modifications of sensor (change the world via **action**).

$$\Delta a = -\alpha_a \frac{\partial E}{\partial a}$$



Predictive processing

- Explaining wide range of brain functions including the learning, perception/cognition and action based on the prediction error minimization.

$$\Delta w = -\alpha_w \frac{\partial E}{\partial w}$$



“Learning”: updates internal models

$$\Delta z = -\alpha_z \frac{\partial E}{\partial z}$$

“Perception/cognition”: updates neural states



$$\Delta a = -\alpha_a \frac{\partial E}{\partial a}$$



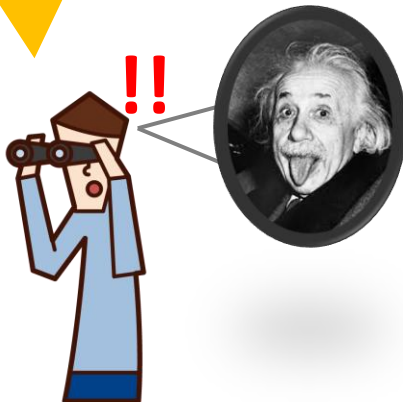
“Action”: changes sensory inputs

“Precision” is important

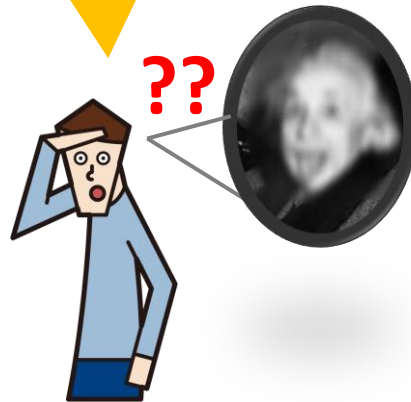
“Precision” in predictive processing

- Estimation of precision works as an important parameter in PE minimization process.
- For example...
 - ✓ Estimate low sensory precision (high variability) => PE should be ignored
 - ✓ Estimate high sensory precision (low variability) => PE should be respected

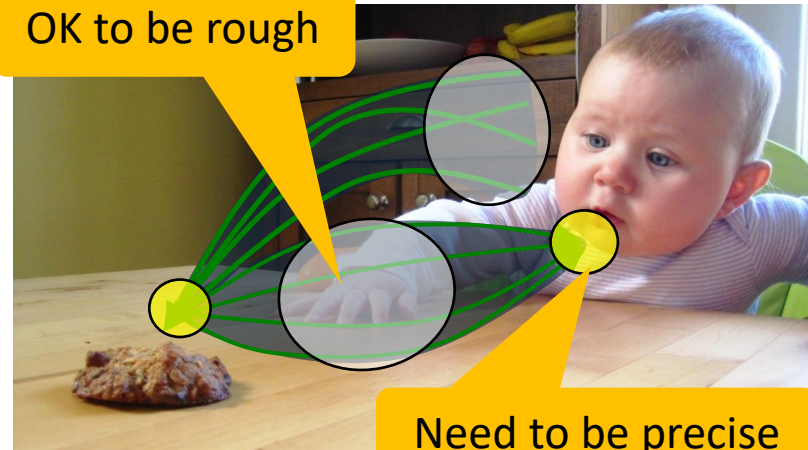
Reliable senses



Unreliable senses



OK to be rough



Need to be precise

**Psychiatric disorders
as altered predictive processing:**

A major theory in “Computational Psychiatry”

Computational psychiatry is rising to prominence

- Novel area of psychiatric research drawing researchers' attentions.
- New specialist journal "Computational Psychiatry" has been opened (2017).
- Director of NIMH listed computational psychiatry as one of the "three particular areas of interest" in psychiatry

TRENDS in Cognitive Sciences

Review

Special Issue: Cognition in Neuropsychiatric Disorders

Computational psychiatry

P. Read Montague^{1,2}, Raymond J. Dolan², Karl J. Friston² and Peter Dayan³

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²Wellcome Trust Centre for Neuroimaging, University College London, 12 Queen Square, London, WC1N 3BG, UK
³Gatsby Computational Neuroscience Unit, Alexandra House, 17 Queen Square, London, WC1N 3AR, UK

Review

THE LANCET Psychiatry

Computational psychiatry: the brain as a phantastic organ

Karl J. Friston, Xiao-Bo Stephan, Read Montague, Raymond J. Dolan

Computational Psychiatry

Xiao-Jing Wang^{1,2,3*} and John H. Krystal^{4,5,6}

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<http://dx.doi.org/10.1016/j.neuron.2014.10.018>

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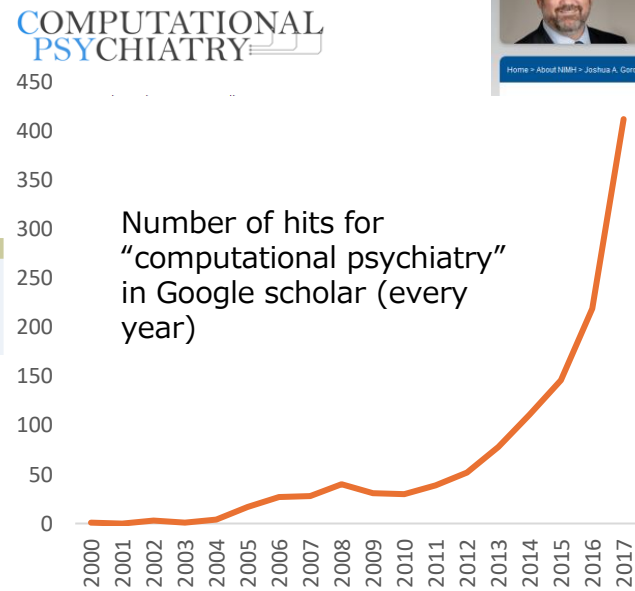
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Transforming the understanding and treatment of mental illnesses.

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DIRECTOR'S MESSAGE

Joshua A. Gordon, M.D., Ph.D.
 Director of NIMH

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Computational Neuroscience: Deciphering the Complex Brain

By Joshua Gordon on February 7, 2017

I wrote in my welcome message about my priorities. First, we need to fund excellent science. Second, we should support studies that will yield benefits on short, medium, and long-term timescales. I also have three particular areas of interest: neural circuits, computational and theoretical psychiatry, and suicide prevention. Here I will discuss computational and theoretical approaches to mental health research. These approaches can be applied across the entire NIMH portfolio, and have the potential to yield benefits in the short, medium, and long-term.

On an unusually hot day in Frankfurt in the summer of 2015, 44 scientists sat in a sweltering room as fans hummed in the background. Along with my co-organizer David Redish, a neuroscientist at the University of Minnesota, we had brought together a group comprised of theoretical neuroscientists, schooled in the challenging art of creating detailed computational models of the brain, and psychiatrists, experienced in the equally challenging art of diagnosing and treating individuals suffering from the most misunderstood of brain

**Autism spectrum disorder (ASD)
as altered predictive processing:**

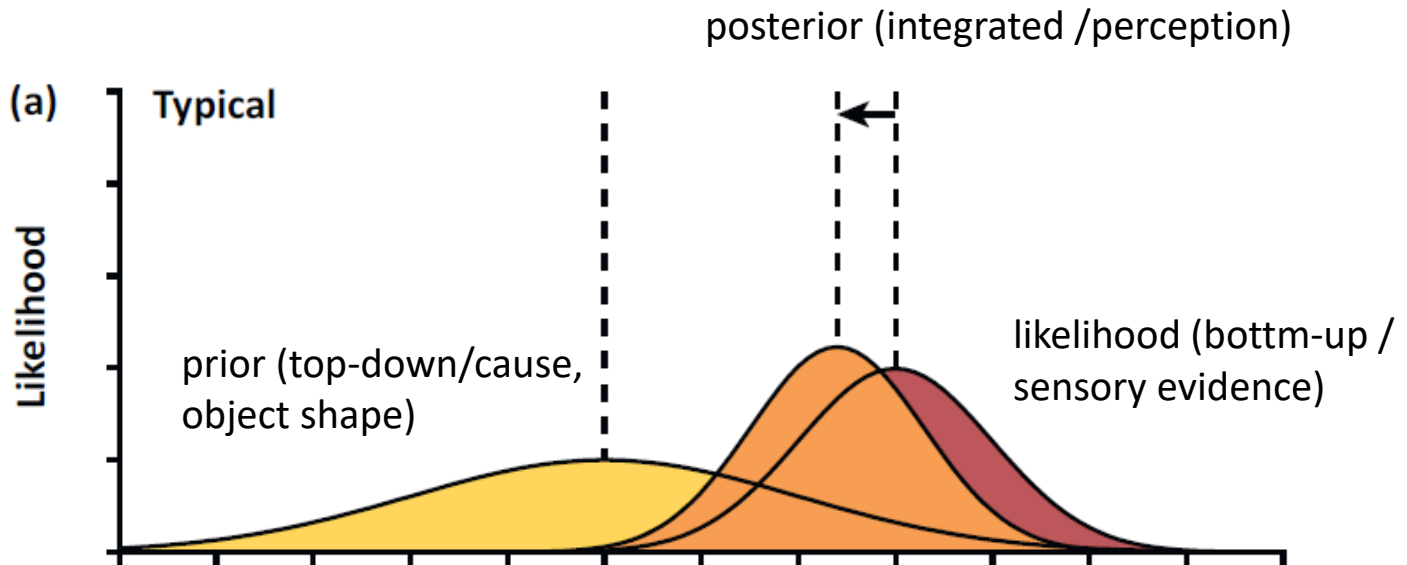
Autism spectrum disorder (ASD): DSM-5 criteria

Many recent studies of ASD focus on the non-social symptoms.

A. Persistent difficulties in the social use of verbal and nonverbal communication

- B. Restricted, repetitive patterns of behavior, interests, or activities,
- Stereotyped or repetitive motor movements, use of objects, or speech
 - Insistence on sameness, inflexible adherence to routines, or ritualized patterns of verbal or nonverbal behavior
 - Highly restricted, fixated interests that are abnormal in intensity or focus
 - Hyper- or hyporeactivity to sensory input or unusual interests in sensory aspects of the environment

Perception as Bayesian inference



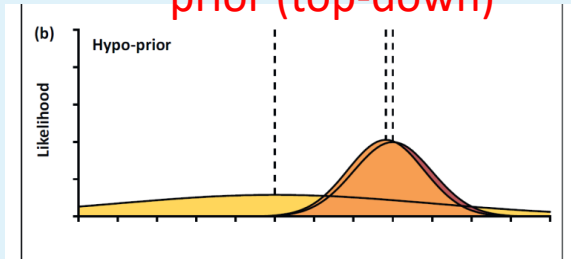
$$p(S|I) \propto p(I|S) p(S)$$

Probability that object shape is S given sensory input I ($p(S|I)$) is determined by the integration of the likelihood of sensory inputs given object shape S ($p(I|S)$) and prior expectation of object shape ($p(S)$).

Aberrant precision accounts for ASD

- Reduced precision of prior leads to high reliability of sensor inputs.
- Similar phenomenon can be expected by increased sensory precision.

Primary impairment in prior (top-down)

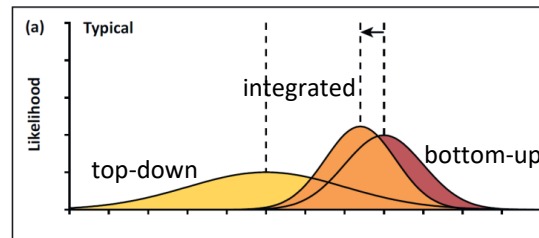


- “Hypo-prior (Increased variance of prior)”: *Pellicano & Burr, 2012*

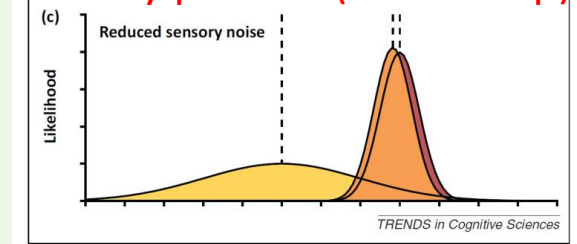
$$\Delta \text{belief} \propto \frac{\text{precision}_{\text{input}}}{\text{precision}_{\text{prior belief}}} \times \text{prediction error}$$

Balance??

- “Imbalance of the precisions of prior belief vs. sensory evidence”: *Lawson et al., 2014*



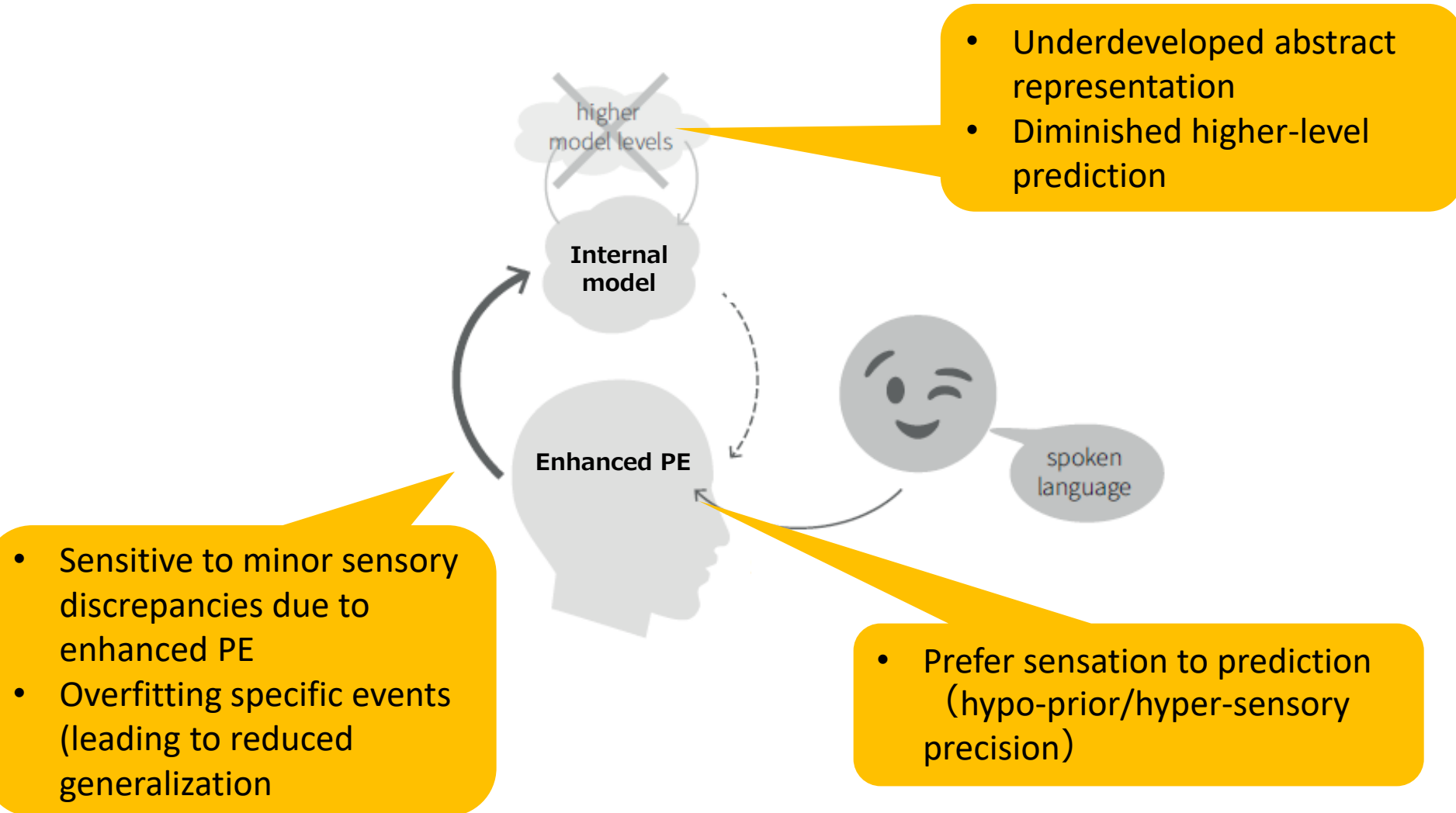
Primary impairment in sensory process (bottom-up)



- “Reduced variance of sensory noise”: *Brock, 2012*
- “High and inflexible precision of PE”: *Van de Cruys et al., 2014*
- “Reduced endogenous neural noise”: *Davis, 2015*

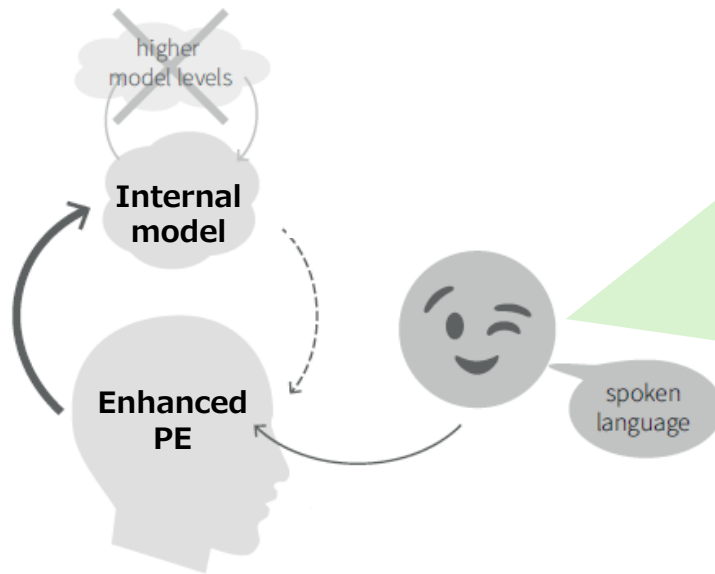
Aberrant predictive processing for ASD

Hypo-prior (hyper-sensory precision) hypothesis



Aberrant predictive processing for ASD

Hypo-prior (hyper-sensory precision) hypothesis



Explaining characteristic symptoms of ASD

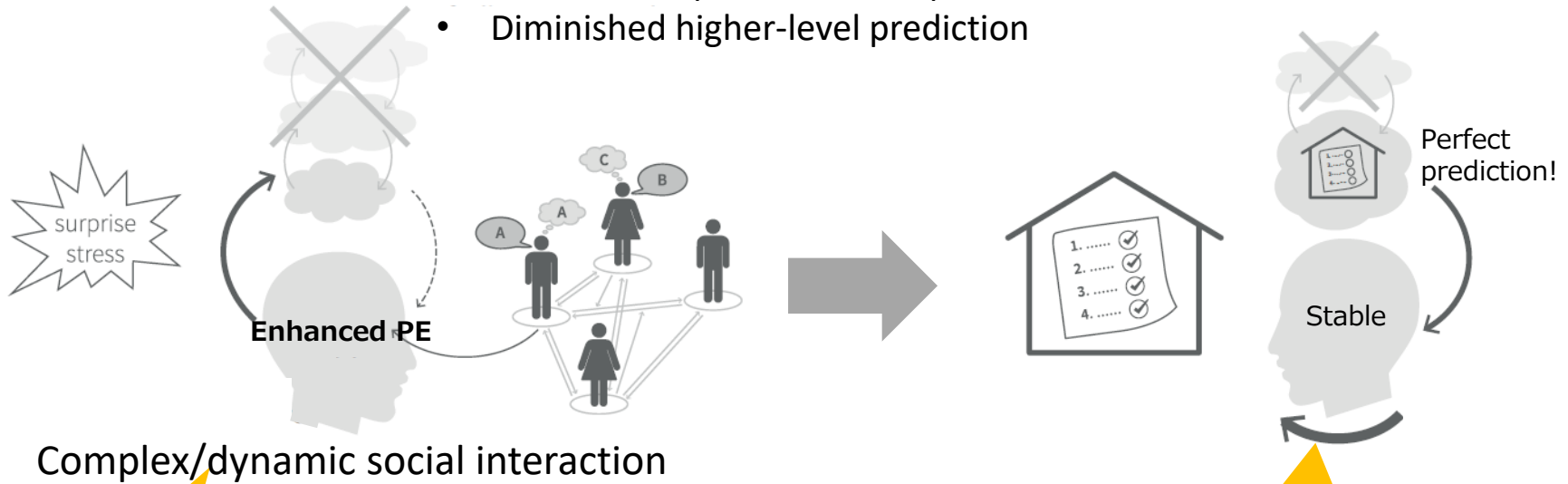
For example...

- More accurate perception
- Lack of advantage using prior
- Hyper sensitivity (sensory overload)
- Reduced spontaneous perceptual shift
- Reduced cognitive flexibility
- Reduced capacity for generalization

Aberrant predictive processing for ASD

Hypo-prior (hyper-sensory precision) hypothesis

- Underdeveloped abstract representation
- Diminished higher-level prediction

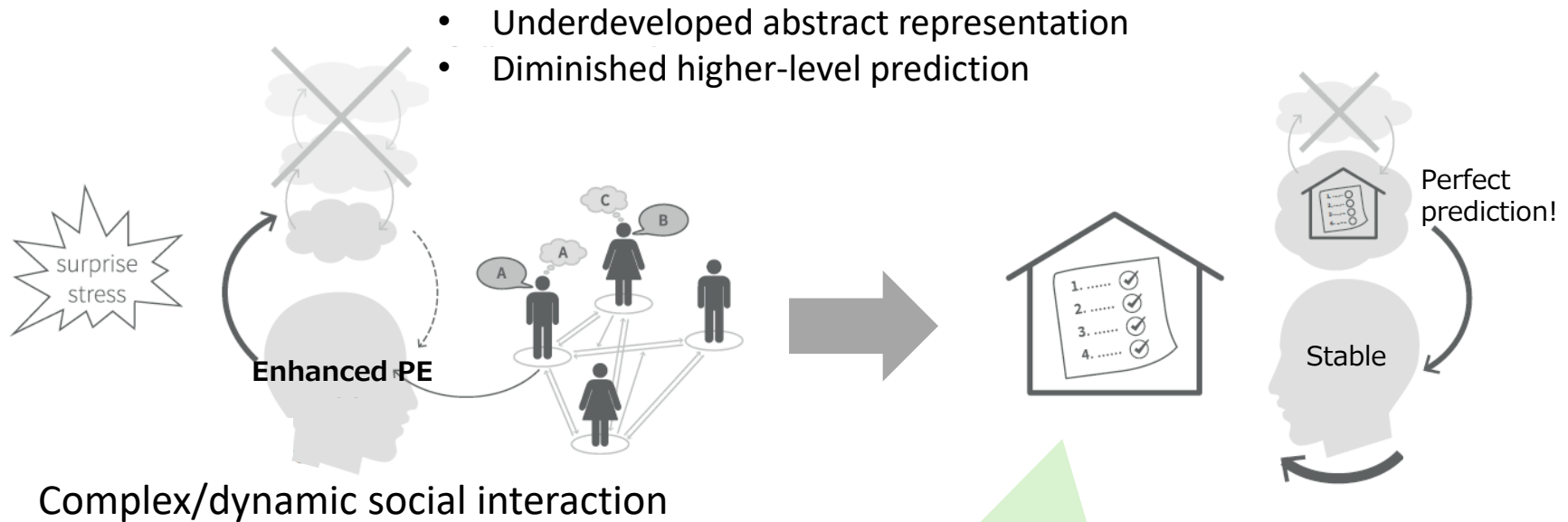


- Larger prediction errors (surprise, stress)

- Prefers objects that allow for accurate prediction
- Urge for repetitive actions in familiar environments

Aberrant predictive processing for ASD

Hypo-prior (hyper-sensory precision) hypothesis

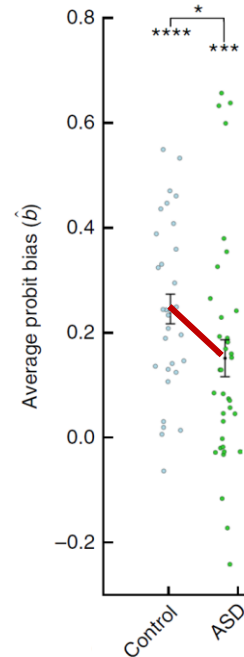
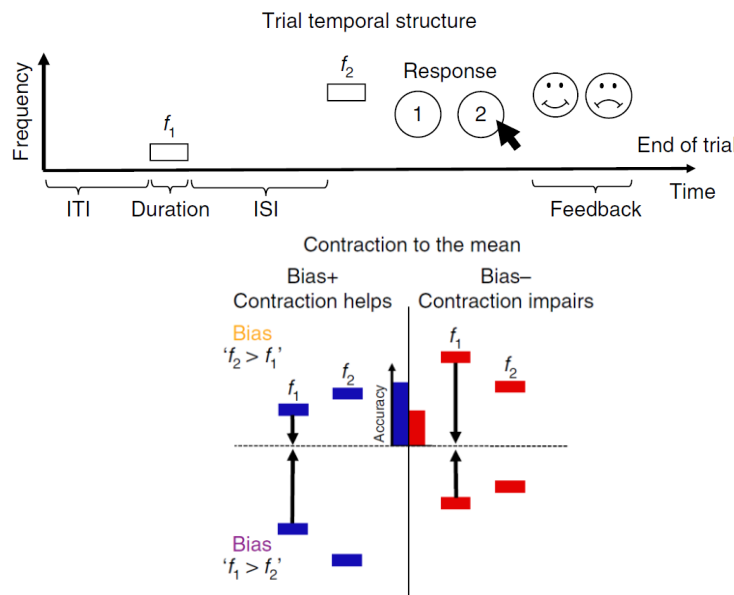


- Desire to sameness
- Repetitive/restricted behavior
- Autistic social interaction

There's also a completely opposite explanation...

Hyper-prior hypothesis for ASD

- Inflexibility or slow-updating of internal state observed in ASD can be considered as “hyper-prior” (strong top-down/resistance to PE).



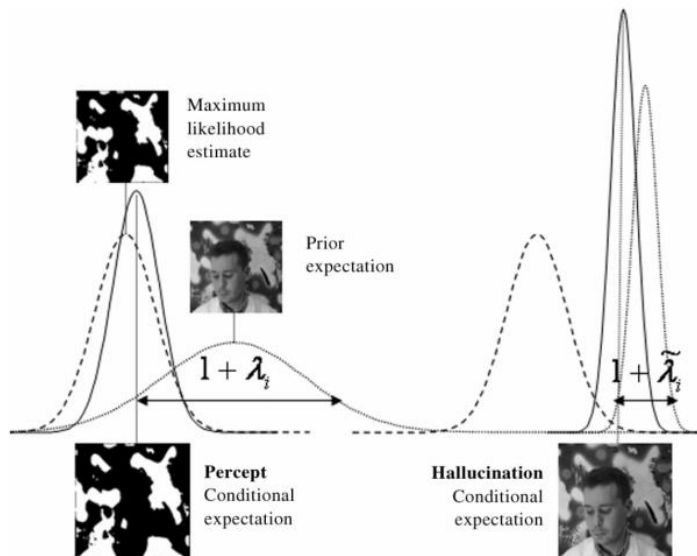
Controversial!?

- Lieder I, Adam V, Frenkel O, Jaffe-Dax S, Sahani M, & Ahissar M (2019) Perceptual bias reveals slow-updating in autism and fast-forgetting in dyslexia. *Nature Neuroscience* 22: 256-264.
- Vishne G, Jacoby N, Malinovitch T, Epstein T, Frenkel O, & Ahissar M (2021) Slow update of internal representations impedes synchronization in autism. *Nature Communications* 12: 5439.

Aberrant precision account for schizophrenia (SZ)

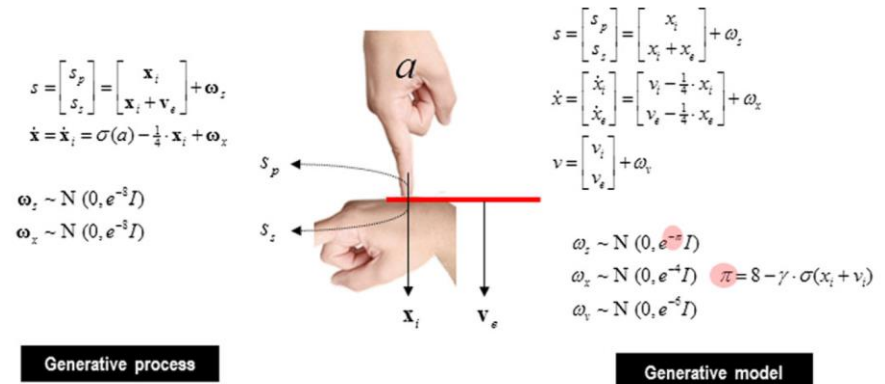
Hallucinations result from the overestimation of the precision of prior beliefs

Hyper-prior precision hypothesis for hallucination



Hyper-sensory precision leads to impaired sensory attenuation and mis attribution of agency

Hyper-sensory precision (hypo-prior) hypothesis for altered sense of agency



Karl J. Friston. Hallucinations and perceptual inference Behavioral and Brain Sciences 28 (6):764-766 (2005)

Brown H, Adams RA, Pares I, Edwards M, Friston K. Active inference, sensory attenuation and illusions. Cogn Process. 2013 Nov;14(4):411-27.

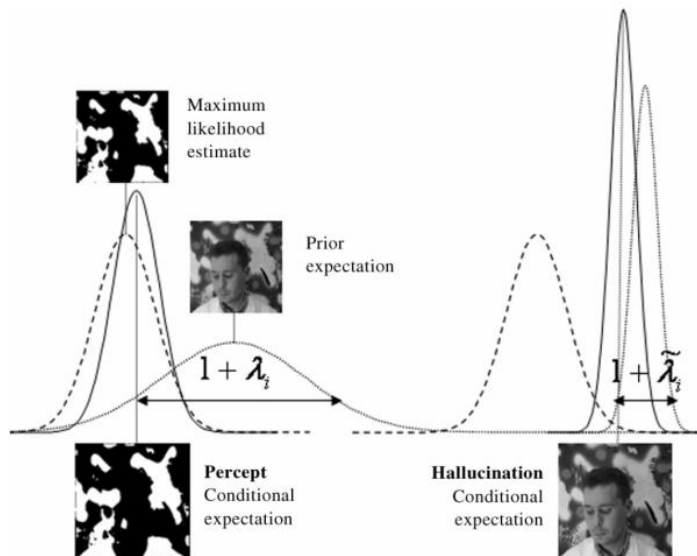
Adams RA, Stephan KE, Brown HR, Frith CD and Friston KJ (2013) The computational anatomy of psychosis. Front. Psychiatry 4:47.

Aberrant precision account for schizophrenia (SZ)

Controversial!?

Hyper-prior precision hypothesis for hallucination

Hyper-sensory precision (hypo-prior) hypothesis for altered sense of agency



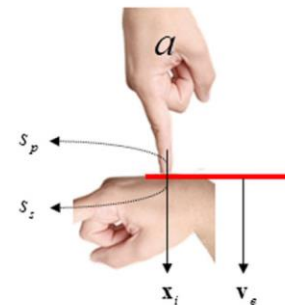
$$s = \begin{bmatrix} s_p \\ s_z \end{bmatrix} = \begin{bmatrix} x_i \\ x_i + v_e \end{bmatrix} + \omega_z$$

$$\dot{x} = \dot{x}_i = \sigma(a) - \frac{1}{4} \cdot x_i + \omega_x$$

$$\omega_z \sim N(0, e^{-3}I)$$

$$\omega_x \sim N(0, e^{-3}I)$$

Generative process



$$s = \begin{bmatrix} s_p \\ s_z \end{bmatrix} = \begin{bmatrix} x_i \\ x_i + x_v \end{bmatrix} + \omega_z$$

$$\dot{x} = \begin{bmatrix} \dot{x}_i \\ \dot{x}_v \end{bmatrix} = \begin{bmatrix} v_i - \frac{1}{4} \cdot x_i \\ v_v - \frac{1}{4} \cdot x_v \end{bmatrix} + \omega_x$$

$$v = \begin{bmatrix} v_i \\ v_v \end{bmatrix} + \omega_v$$

$$\omega_z \sim N(0, e^{-3}I)$$

$$\omega_x \sim N(0, e^{-4}I)$$

$$\omega_v \sim N(0, e^{-5}I)$$

$$\pi = 8 - \gamma \cdot \sigma(x_i + v_i)$$

Generative model

Karl J. Friston. Hallucinations and perceptual inference Behavioral and Brain Sciences 28 (6):764-766 (2005)

Brown H, Adams RA, Pares I, Edwards M, Friston K. Active inference, sensory attenuation and illusions. Cogn Process. 2013 Nov;14(4):411-27.

Adams RA, Stephan KE, Brown HR, Frith CD and Friston KJ (2013) The computational anatomy of psychosis. Front. Psychiatry 4:47.

Altered AiF process of neurorobots

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ICDL2024 Tutrial@May20, 2024

Predictive processing: A major theory in “Computational Psychiatry”

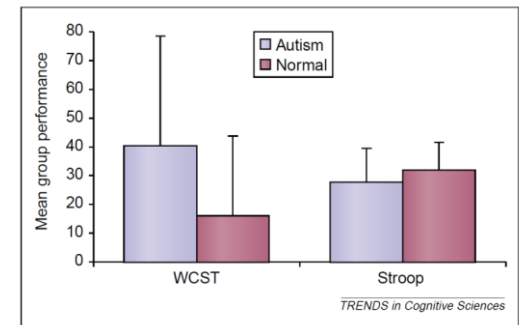
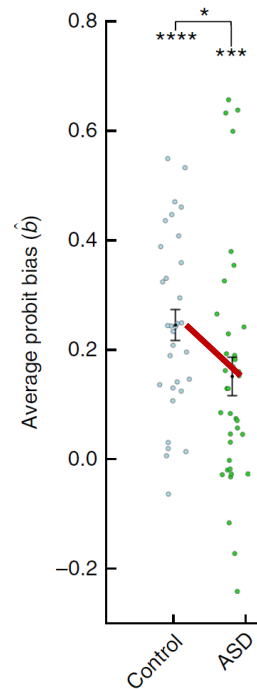
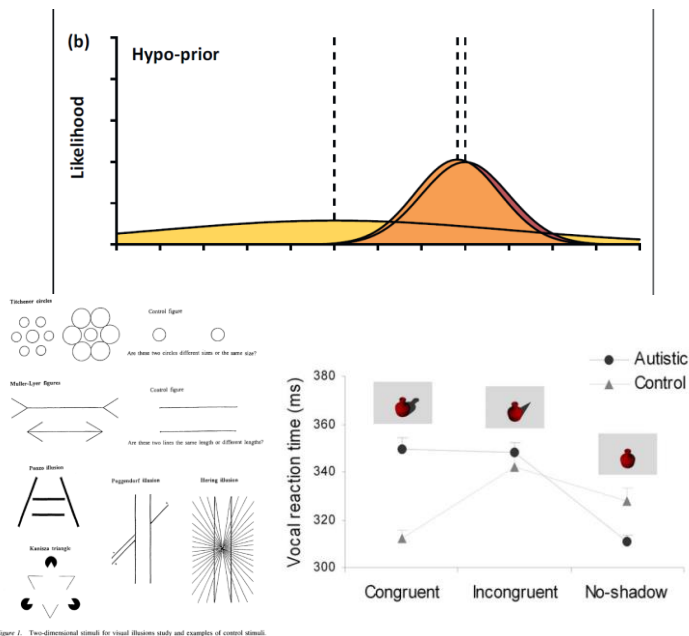
There is hope that the pathology of ASD and SZ can be explained as altered predictive processing.

However....

- Similarity of the pathologies postulated in SZ and ASD (altered prior)
- Opposing mechanistic hypotheses for SZ and ASD (hypo-prior vs hyper-prior hypothesis)

Controversy in aberrant predictive processing in ASD

- Reduced precision of prediction (hypo-prior hypothesis)
 - ✓ More accurate perception, hyper-sensitivity, reduced generalization
- Slow-updating of prediction (hyper-prior hypothesis)
 - ✓ Reduced cognitive flexibility



Pellicano E, Burr D. When the world becomes 'too real': a Bayesian explanation of autistic perception. Trends Cogn Sci. 2012 Oct;16(10):504-10.

Vishne G, Jacoby N, Malinovitch T, Epstein T, Frenkel O, & Ahissar M (2021) Slow update of internal representations impedes synchronization in autism. Nature Communications 12: 5439.

Controversy in aberrant predictive processing in SZ

- Excessive precision at the sensory level = reduced precision of prediction (hypo-prior hypothesis)
 - ✓ Abnormal response to PE (misattribution of beliefs, aberrant salience)
- Excessive precision of prediction (hyper-prior hypothesis)
 - ✓ Reduced response to PE (perceptions/beliefs not based on external stimuli (hallucinations and delusions))

Table 1. Predictive Coding and Positive Symptoms: Theory and Controversy

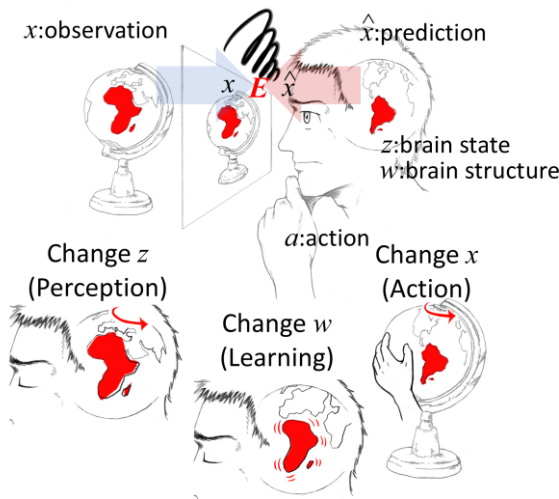
Symptom	Feature	Theory	Literature	Controversy
Hallucinations	Percepts without external stimulus	Strong perceptual priors	Powers <i>et al.</i> (120)	Entails weak and strong prior beliefs—for perception and action—in the same brain at the same time
	Speech from external agents	Weak corollary discharge	Thakkar <i>et al.</i> (86)	
Delusions	Delusional mood/aberrant salience	Weak perceptual priors	Corlett <i>et al.</i> (121)	Necessitates a transition from weak to strong priors as delusions form, foment, and become ingrained
	Fixed in the face of contradictory evidence	Strong memory reconsolidation/ strong conceptual priors	Corlett <i>et al.</i> (103); Schmack <i>et al.</i> (72)	

Our perspective

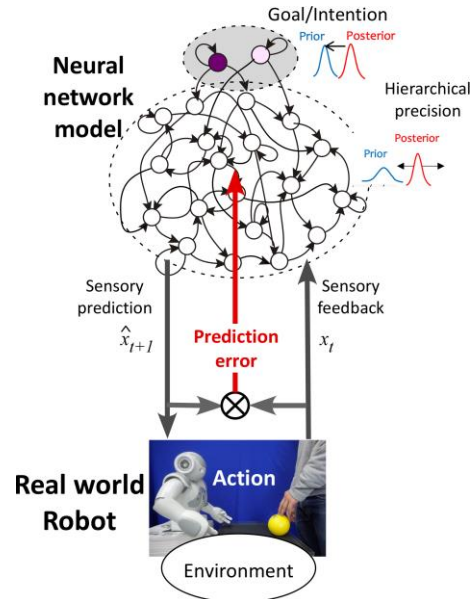
Need for a unified framework that can deal with

- Developmental *persistent trait* and impact of *episodic change*
 - Detailed consideration of hierarchy of precision
- => **Developmental neurorobotics approach!!**

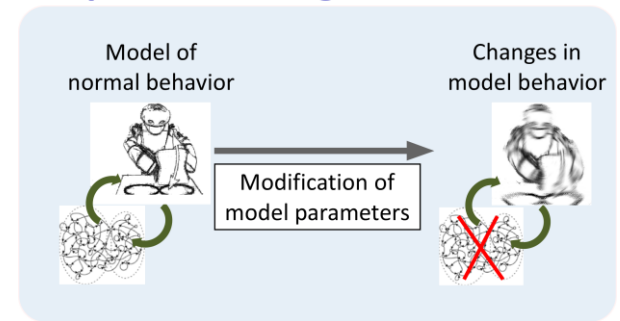
Predictive processing theory



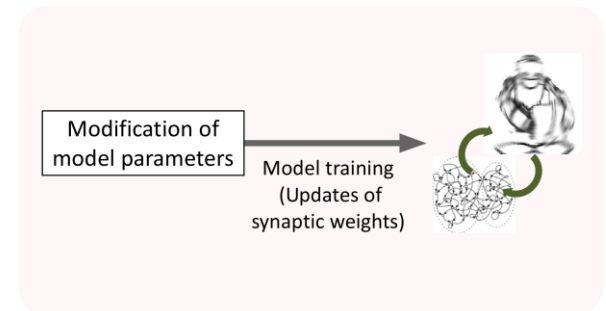
Neurorobotics system with hierarchical neural network



Episodic change simulation

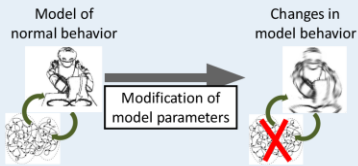


Developmental learning simulation

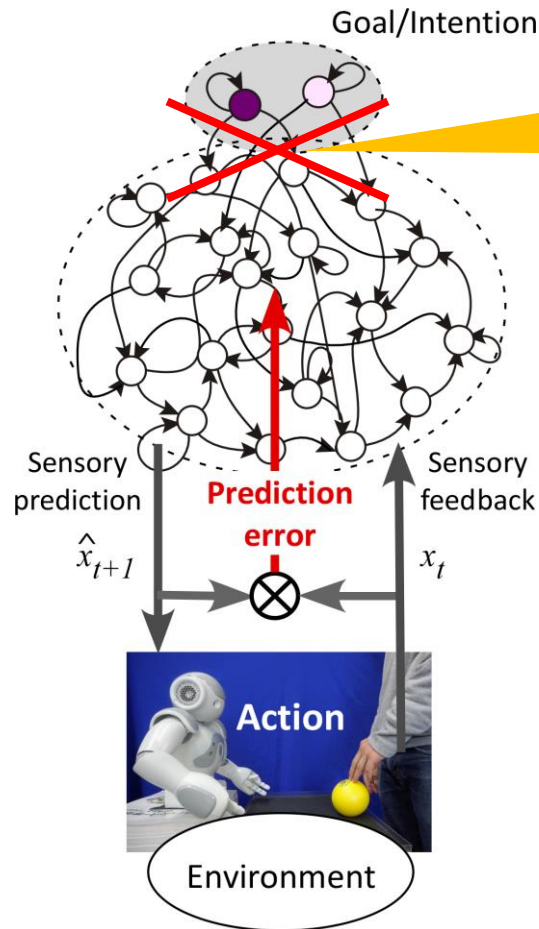


Part1:

Functional disconnection as
acute episode (SZ)/inherent characteristics (ASD)



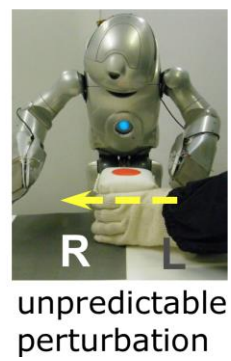
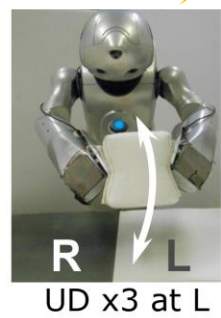
Episodic change simulation (Normal learning => Functional disconnection simulation) [Yamashita2012]

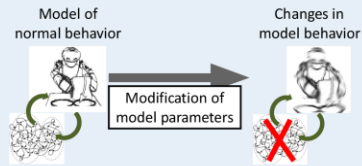


Functional disconnection was applied after normal learning...

$$w^{\text{dis}} = w + U(|Kw|)$$

Task that requires switching actions depending on the situation



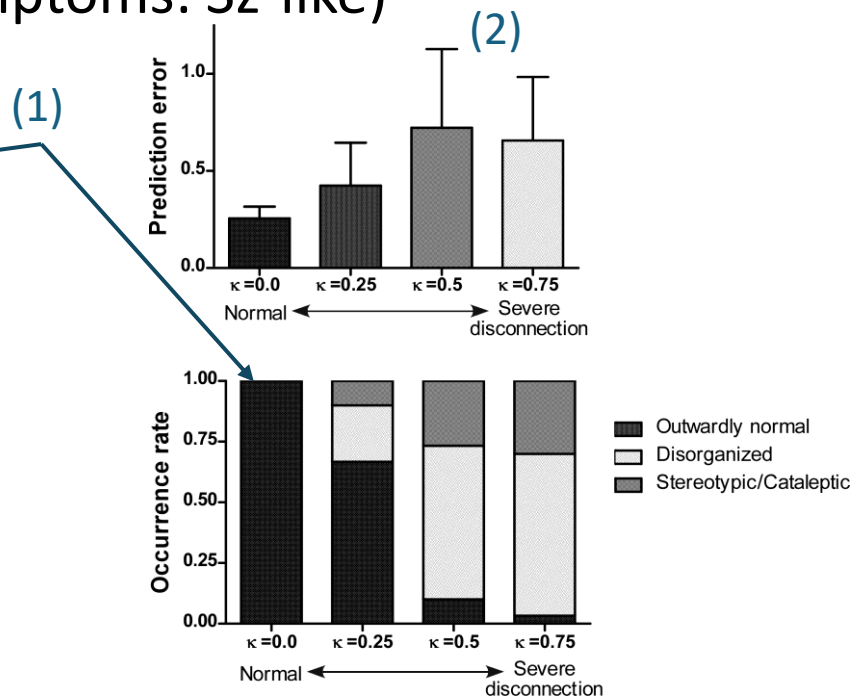
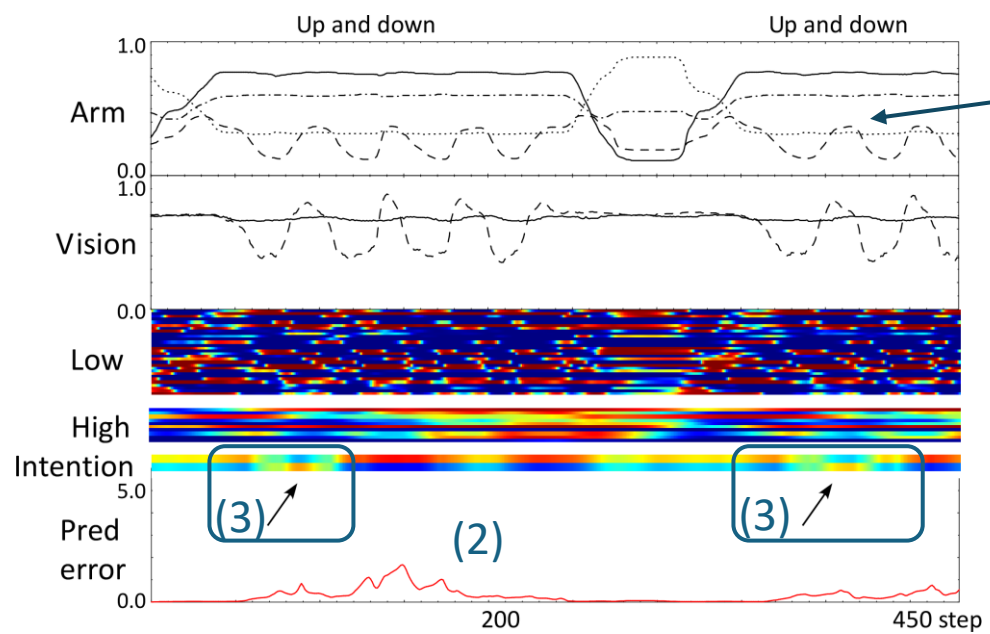


Episodic change simulation (Normal learning => Functional disconnection simulation) [Yamashita2012]

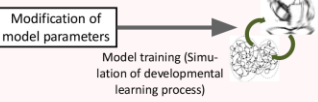
Functional disconnection after normal learning can induce

- Seemingly normal behavior (1)
- Spontaneous prediction error generation (2)
- Aberrant higher-level modulation (3)

=>Miss-attribution of behavioral intention?
(delusion of control/passivity symptoms: Sz-like)

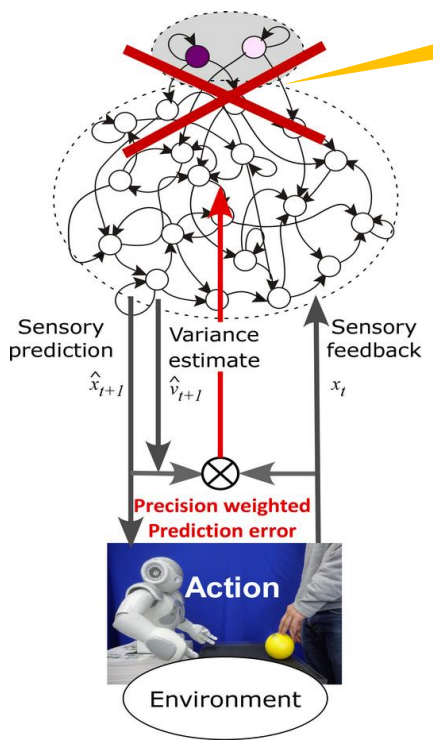


Developmental learning simulation (Functional disconnection simulation => Network Learning) [Idei2021]

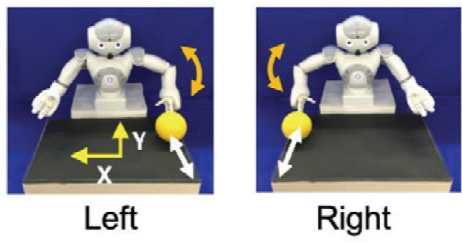


Functional disconnection was applied during developmental learning...

$$w^{dis} = w + U(|Kw|)$$



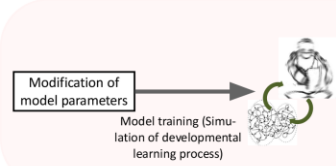
Learning phase



Adaptation phase



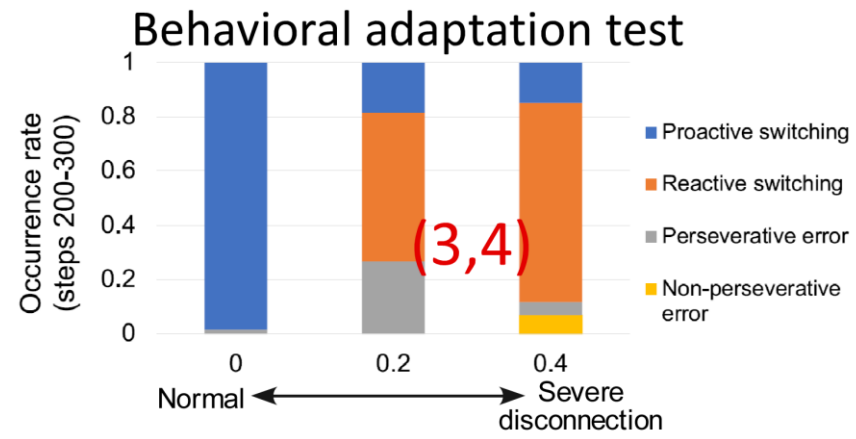
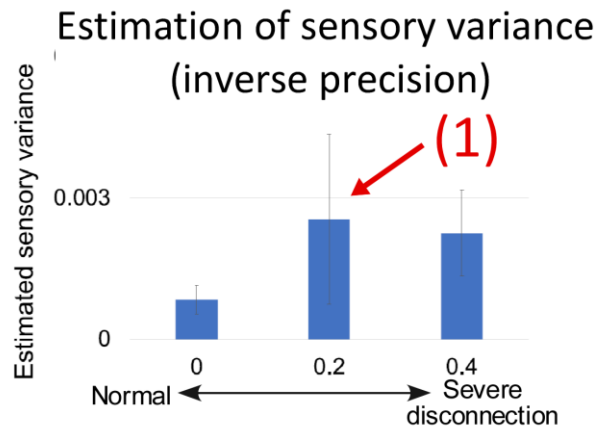
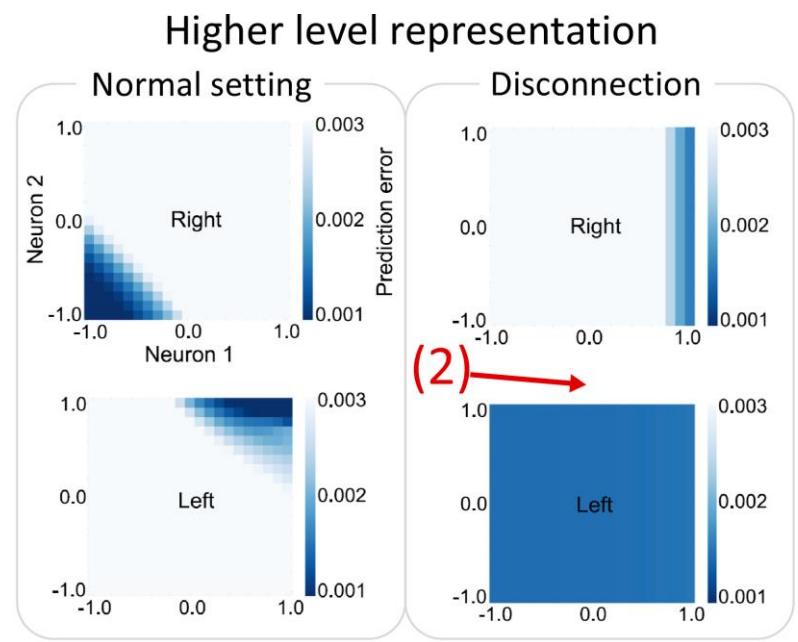
$$L_{out} = \prod_{s \in I_S} \prod_{t=1}^{T^{(s)}} \prod_{i \in I_O} \frac{1}{\sqrt{2\pi v_{t,i}^{(s)}}} \exp\left(-\frac{(y_{t,i}^{(s)} - \hat{y}_{t,i}^{(s)})^2}{2v_{t,i}^{(s)}}\right)$$



Developmental learning simulation (Functional disconnection simulation => Network Learning) [Idei2021]

Development under functional disconnection leads to ASD-like features including...

- Reduced estimation of sensory precision (1)
- Reduced precision (vague behavioral representation) at higher levels (2)
- Coexistence of hyper- and hypo-reactivities to sensations (3)
- Reduced flexibility of behavior (4)



Part2

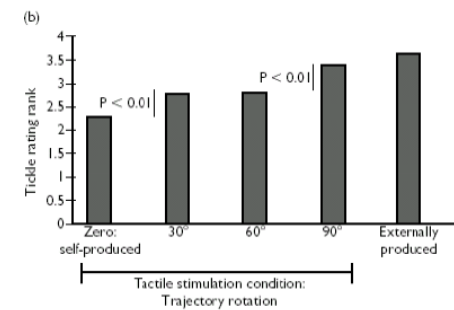
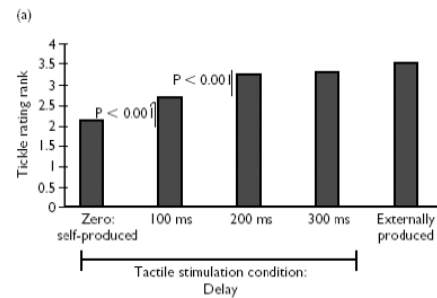
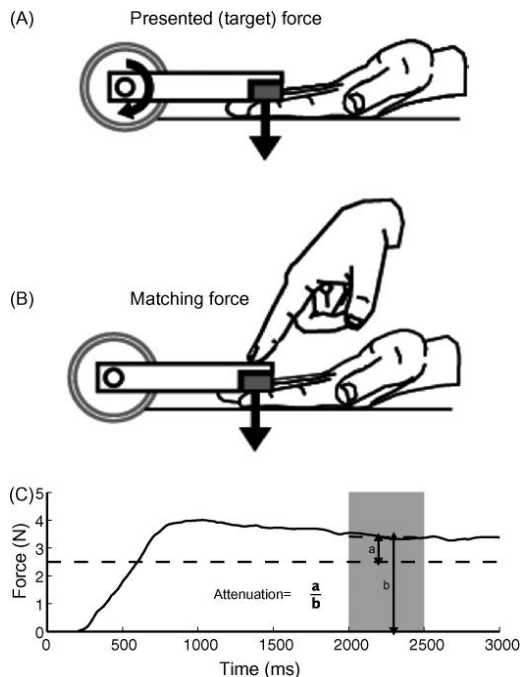
Hierarchy of precision and traits for SZ and ASD

Sensory attenuation (SA)

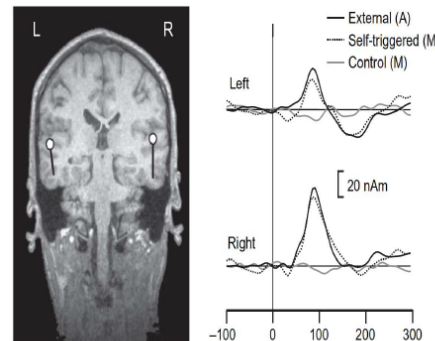
- Phenomenon in which the level of perception/neural activity is diminished when generated by oneself, compared to that generated externally.

Ex) Force matching, tickling, self-generated voice

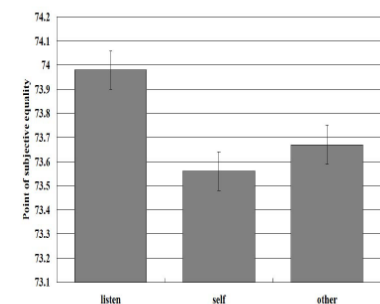
- Reduced SA in SZ
- Intact SA in ASD



Blakemore et al., 1999, J Cogn Neurosci.



Martikainen et al., 2004, Cereb Cortex



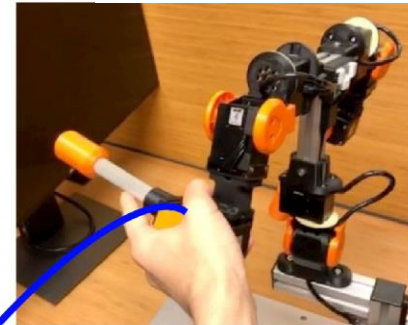
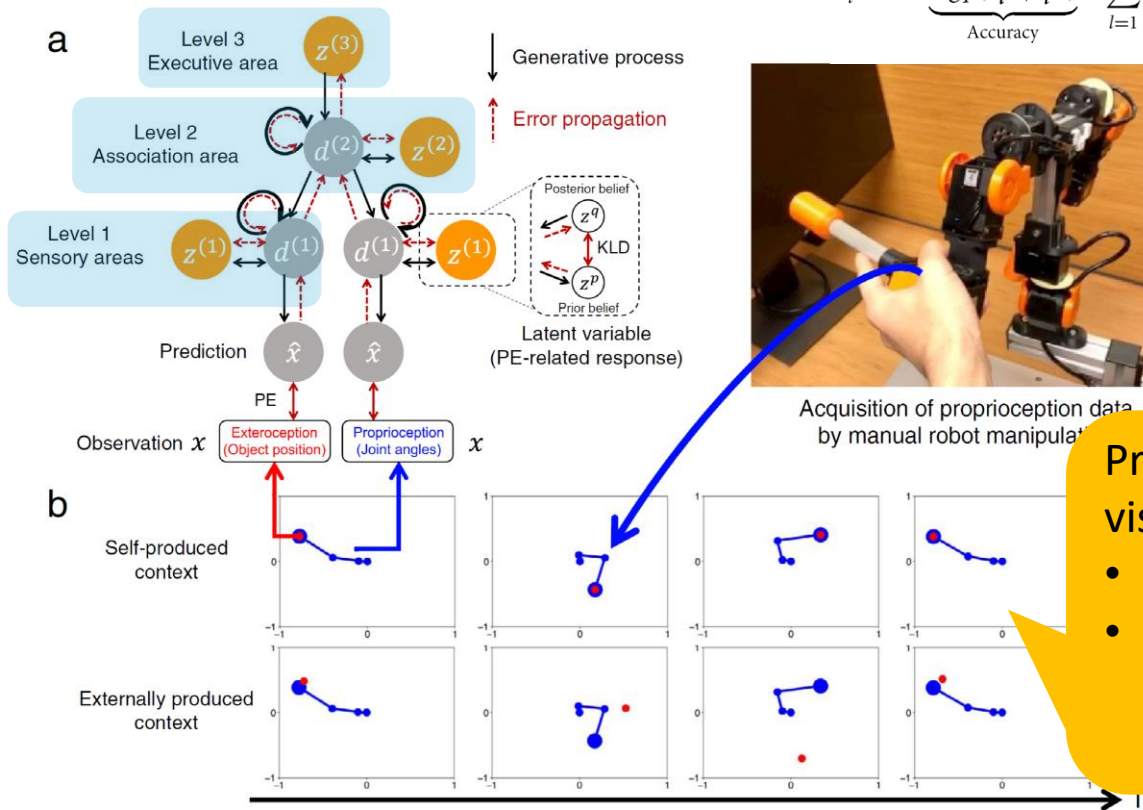
Sato, 2008, Conscious Cogn.

Developmental learning simulation

(Modulate intrinsic relative impact of prior between hierarchy => Traits in cognition and sensory attenuation) [Idei2022, Idei2023]

- Free energy (FE) minimization model with hierarchical latent state precisions
- Self/non-self action generation conditions

$$F_t^{(s)} = - \underbrace{\log p(x_t^{(s)} | d_t^{(s)})}_{\text{Accuracy}} + \sum_{l=1}^3 \underbrace{W^{(l)} \left(D_{KL}[q(z_t^{(l),(s)} | e_{t:T}^{(s)}) || p(z_t^{(l),(s)} | d_{t-1}^{(l),(s)})] \right)}_{\text{Complexity}}$$



Acquisition of proprioception data by manual robot manipulation

Proprioception & vision are

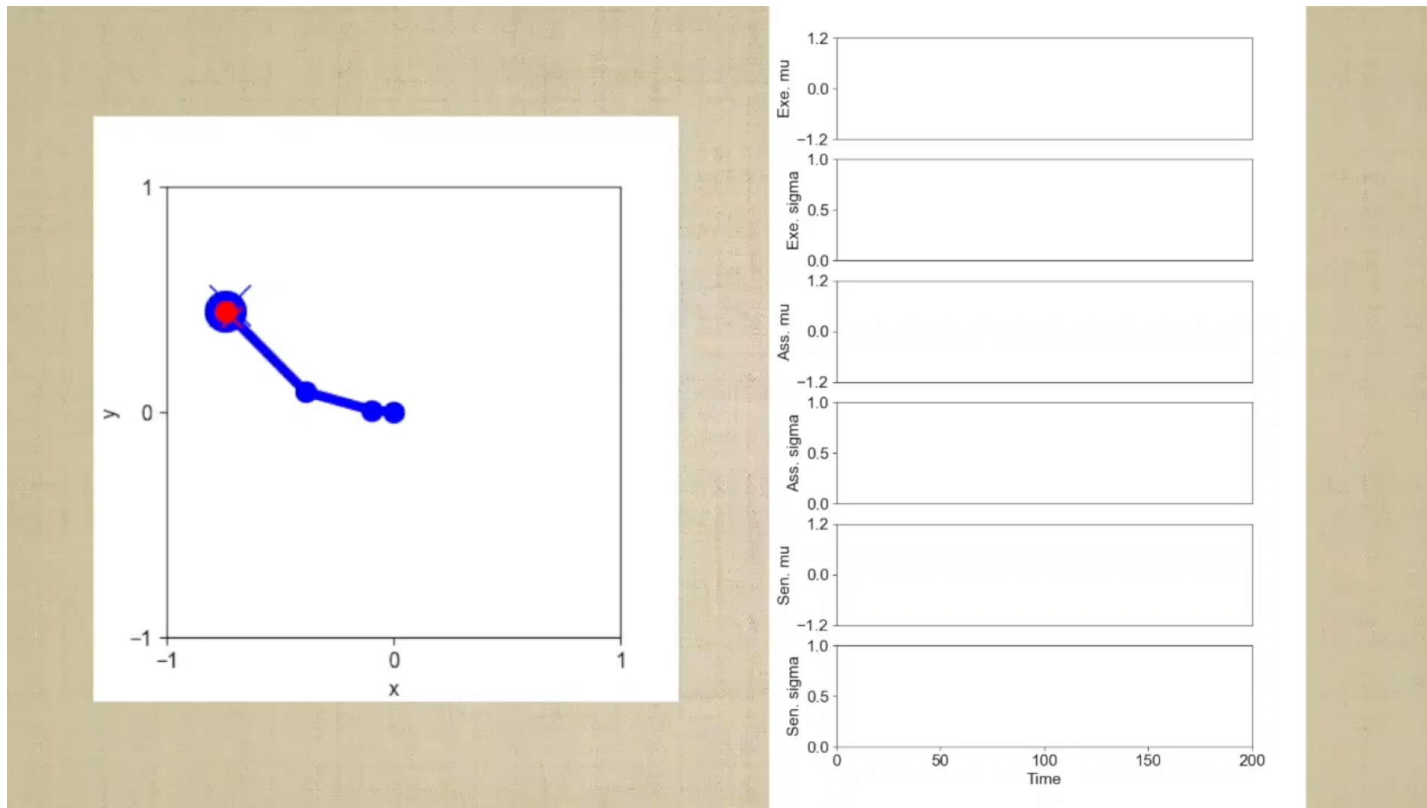
- consistent => self
- Inconsistent => non-self (externally generated)



Developmental learning simulation

(Modulate intrinsic relative impact of prior between hierarchy => Traits in cognition and sensory attenuation) [Idei2022, Idei2023]

- Free energy (FE) minimization model with hierarchical latent state precisions
- Self/non-self action generation conditions

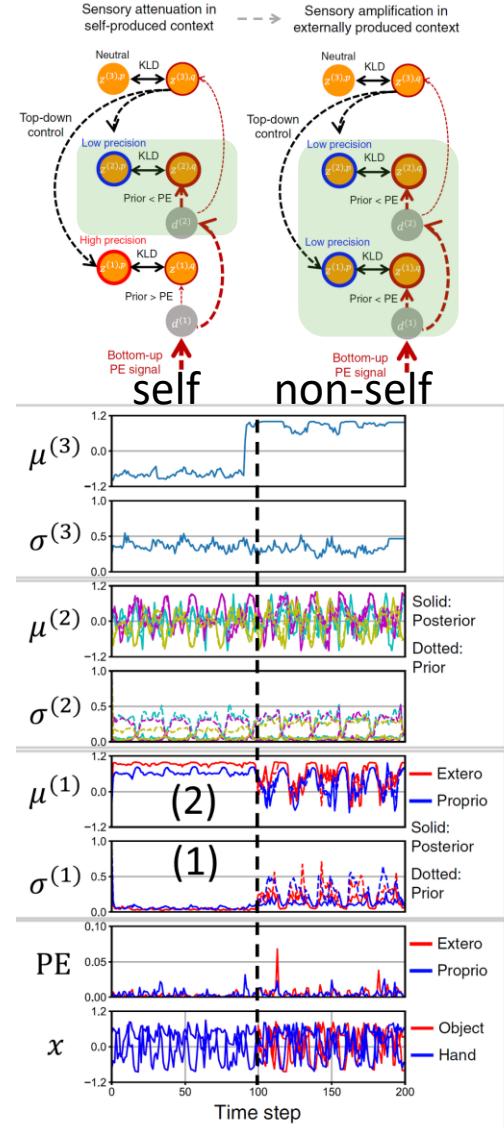
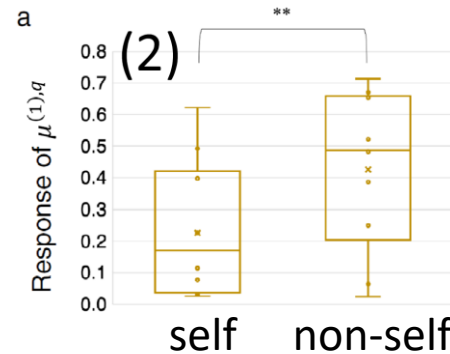
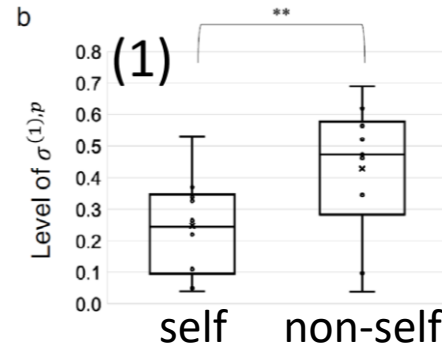
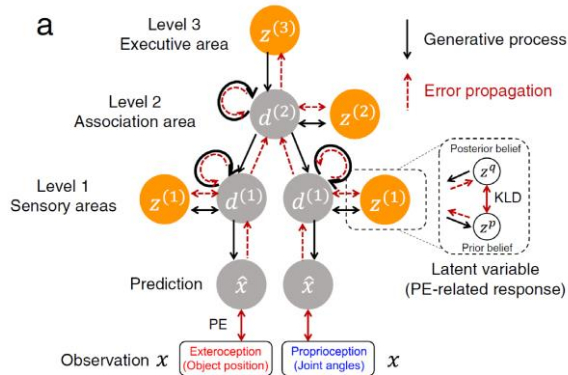


Developmental learning simulation

(Modulate intrinsic relative impact of prior between hierarchy
 => Traits in cognition and sensory attenuation) [Idei2022, Idei2023]

Through the experience of self/non-self conditions

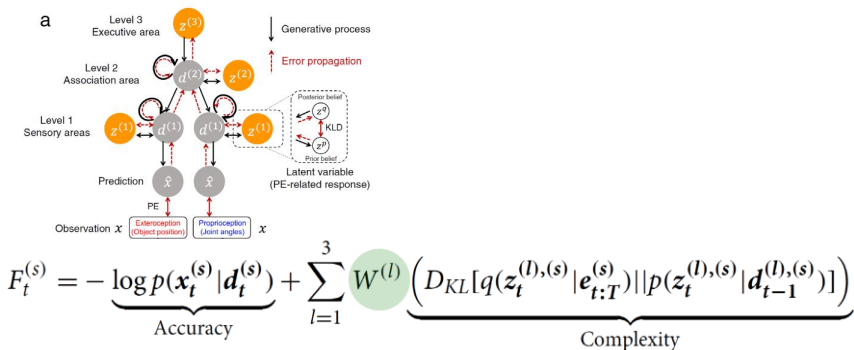
- Self-organized temporally dynamic changes in precisions of latent states (1)
- Emergence of “sensory attenuation (SA)” phenomenon (2)





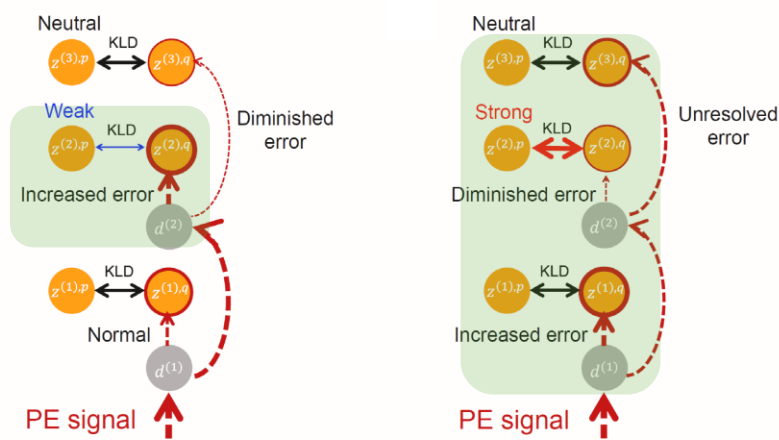
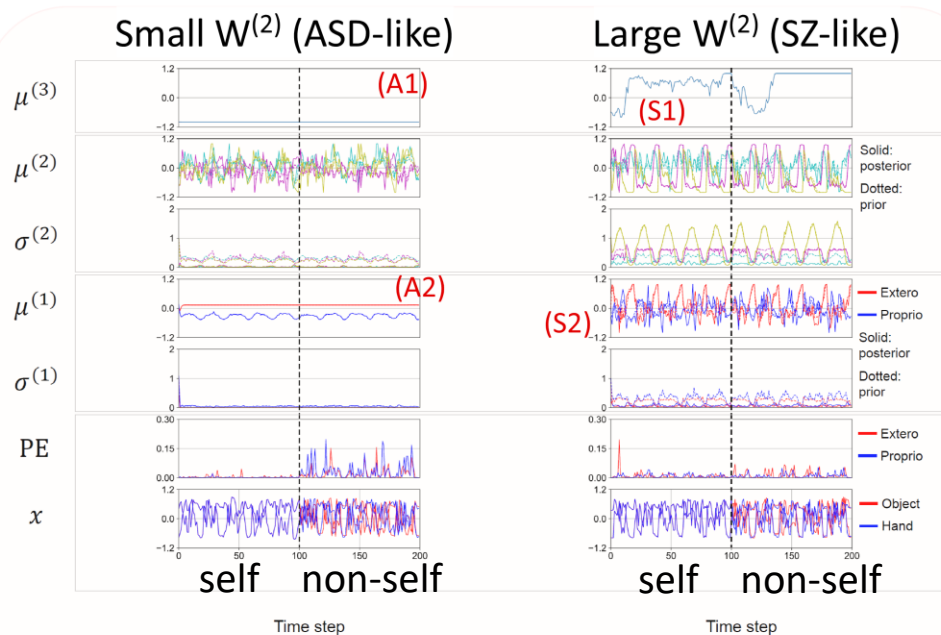
Developmental learning simulation

(Modulate intrinsic relative impact of prior between hierarchy
=> Traits in cognition and sensory attenuation) [Idei2022, Idei2023]



Difference in the “intrinsic” relative influences of prior between hierarchies (W -value) results in SZ-like and ASD-like phenotype including...

- Weak associative level prior => Over-fitting, reduced flexibility(A1), intact SA(A2) (ASD-like)
- Strong associative level prior => Spontaneous recognition switch(S1), reduced SA(S2) (SZ-like)



Take-home messages

- Functional disconnection in hierarchical predictive processing system can result in distinct symptom formations of SZ and ASD.
- By considering detailed hierarchical and developmental learning aspects, distinct sets of primary and secondary alterations of hierarchical precision estimation might capture differences between SZ and ASD.
- Developmental neurorobotics approaches may serve as a complementary research framework for computational psychiatry with the predictive processing theory.



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Elucidating multifinal and equifinal pathways to developmental disorders by constructing real-world neurorobotic models

[Hayato Idei](#), [Yuichi Yamashita](#) [👤](#) [✉](#)

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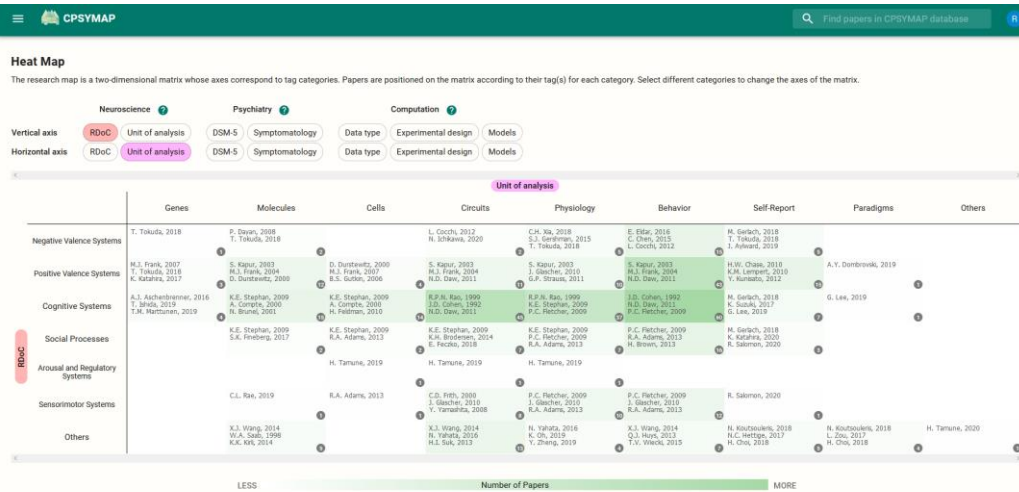
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Idei, H., & Yamashita, Y. (2024). Elucidating multifinal and equifinal pathways to developmental disorders by constructing real-world neurorobotic models. *Neural Networks*, 169, 57–74.

Thank you for your attention.



Computational Psychiatry Research Map (CPSYMAP)



<https://ncnp-cpsy-rmap.web.app/>



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- Shingo Murata (Keio Univ.)
- Hayato Idei (NCNP)

We are looking for researchers, research associates, and students who are interested in computational psychiatry!

Contact: yamay@ncnp.go.jp

Grants: JSPS KAKENHI (JP20H00625,JP24H00076, JP24K00499)

JST CREST (JPMJCR21P4), JST Moonshot R&D (JPMJMS2031)

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