

# Approaching Artificial Intelligence as an Embodied Developmental Process

LEFT: Chicken neural crest migration, MIDDLE: Axolotl developing nervous system,  
RIGHT: elongation of Zebrafish notochord.

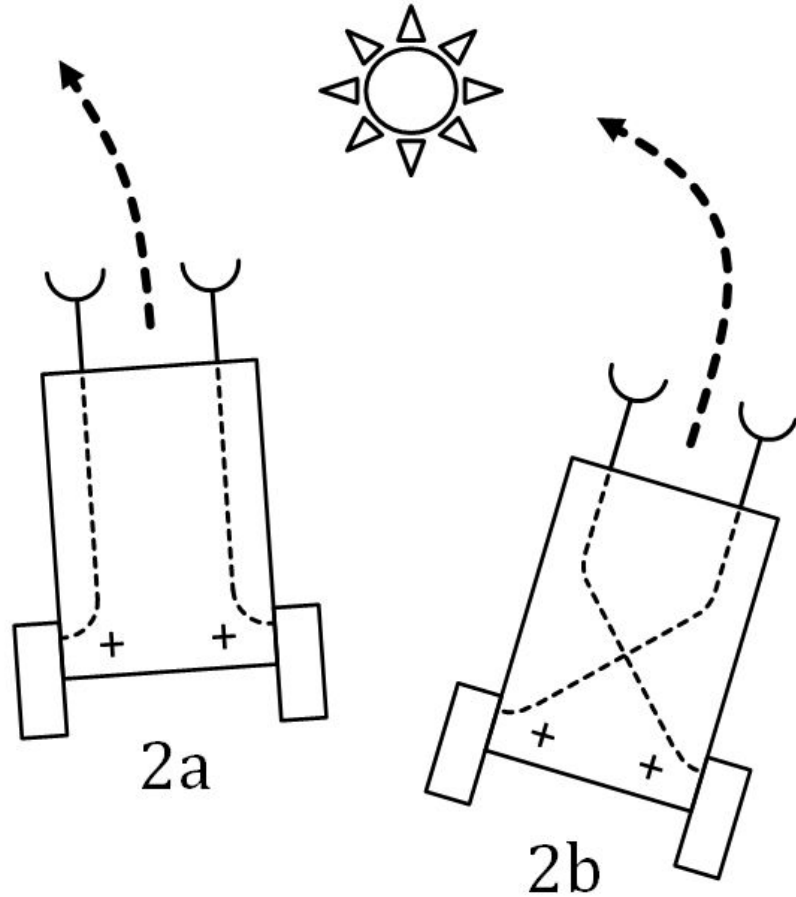
**Bradly Alicea, Rishabh Chakrabarty, Akshara Gopi, Anson Lim,  
Furkan Özçelik, and Jesse Parent**

# Braitenberg Vehicles

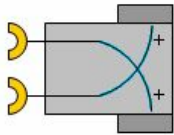
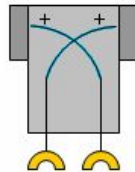
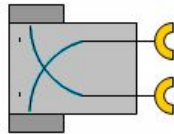
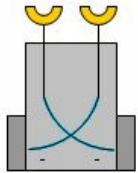
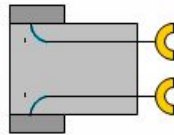
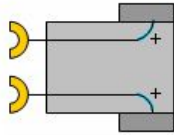
First-order stimulus-response mapping:

- observe empirical world, hard-wired to elicit response.
- vehicle types yield taxis, or reflexive behavior.

Very simple neural networks that are embodied agents (effects of body are explicit).



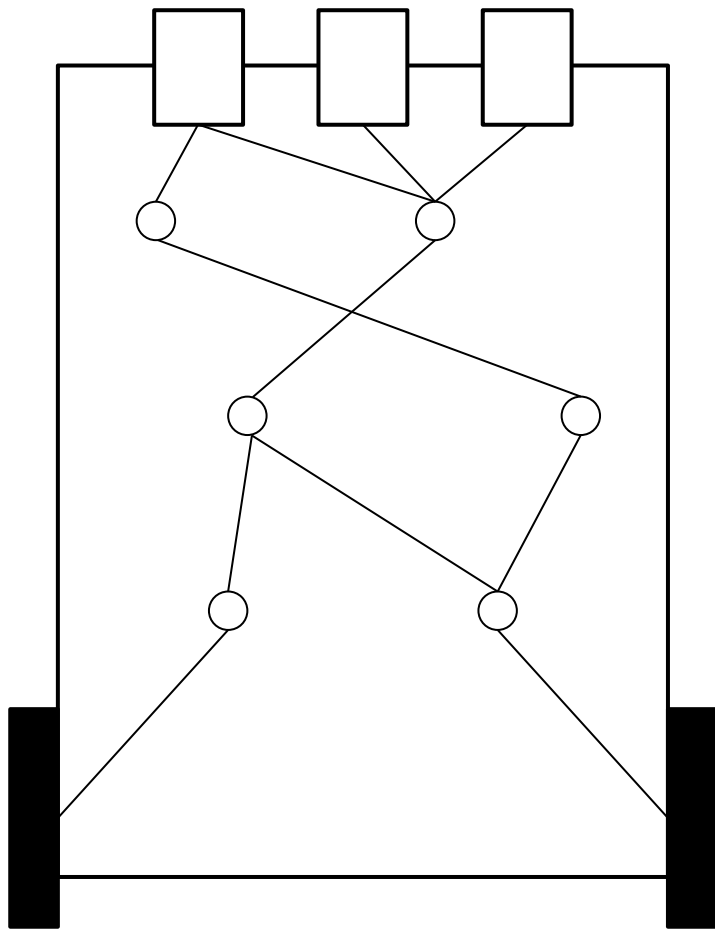
# Developmental Braitenberg Vehicles (dBVs)



Embodied nervous system, neural network develops inside of a simple phenotype.

Development: network morphogenesis

- initial condition: a mapping between sensor and effector
- complex network of interneurons that attenuate the initial feedforward signal.



Sensor  
Nodes

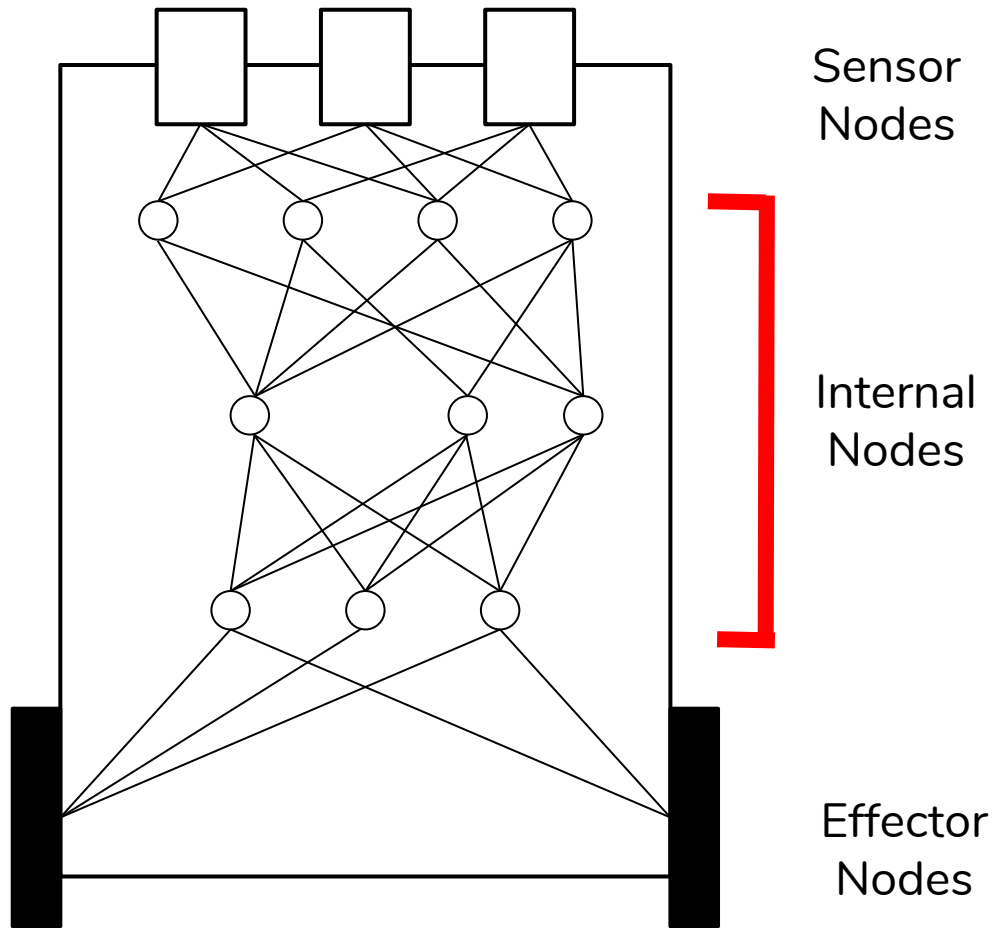
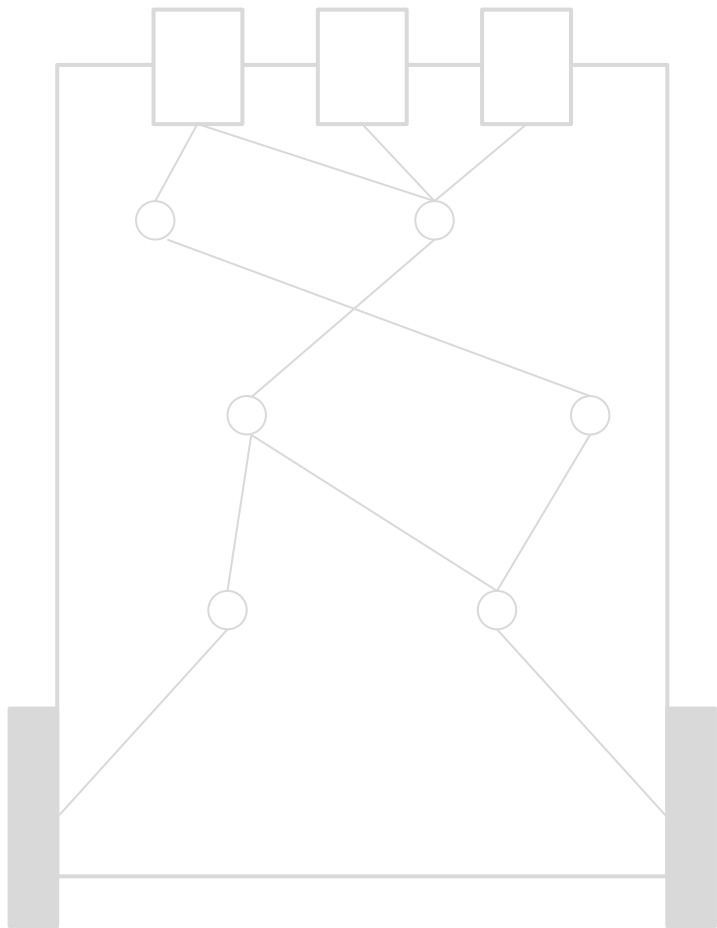
Simple mapping between sensors and effectors modified over time by intermediate (internal) nodes.

Internal  
Nodes

Internal node mediate between sensor and effector, combine sensory signals.

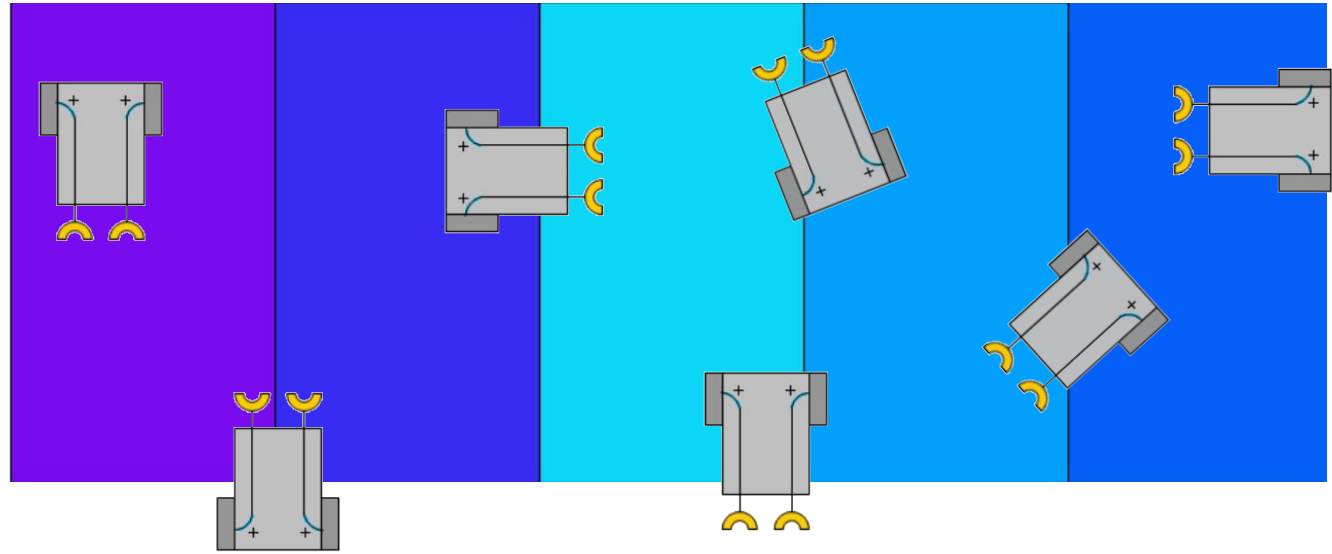
Effector  
Nodes

CGS kernel embedded among the internal nodes, transforms signals into fuzzy combinations (probabilistic signals).



# Chromotaxis

Example of behavior  
generated by Meta-brain  
Model that utilizes a dBV.



Meta-brain Model: Developmental  
Embodied Agents as Meta-brain  
Models. *Developmental Neural  
Networks Workshop, Artificial  
Life 2020.*



As a phototaxis, simple attraction/repulsion

As a preference, a separate continuum of emotional valence.

Difference between congruent stimuli (color, shape) enables  
Generalized Hebbian learning (GHA).

Differentiation between incongruous colors enables spatial learning  
(assembly via genetic algorithm -- BraGenBrain).



# $W_{ij}$

## Connectivity Activation Encoding

	SL	SM	SR	IN	WL	WR
SL	0	0	0	1	1	0
SM	0	0	0	1	1	0
SR	0	0	0	0	0	1
IN	1	1	0	0	1	1
WL	1	1	0	1	0	0
WR	0	0	1	1	0	0

$$W_{ij}$$

Each developmental time point produces a connectivity matrix  $W_{ij}$ .

Each matrix has a corresponding determinant  $|A|$

$$W_{ij} = |A_n|$$



$W_{ij}$ 

What is  $W_{ij}$ ? A summary of connections among a network of size  $k$ . In development, the number of neurons ( $k$ ) can grow and synapses (connections  $i,j$ ) can fluctuate in terms of their strength and efficacy. So  $W_{ij}$  is scaled by parameter  $k$ .

$$W_{ij} = \frac{1}{|A_n|}$$

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Universal functions rather than transformation of  $W_{ij}$  actually describe the link between neurobiological complexity and an artificial model.

Universality and individuality in neural dynamics across large populations of recurrent networks. *Advances in Neural Information Processing Systems*, 32, 15629–15641.

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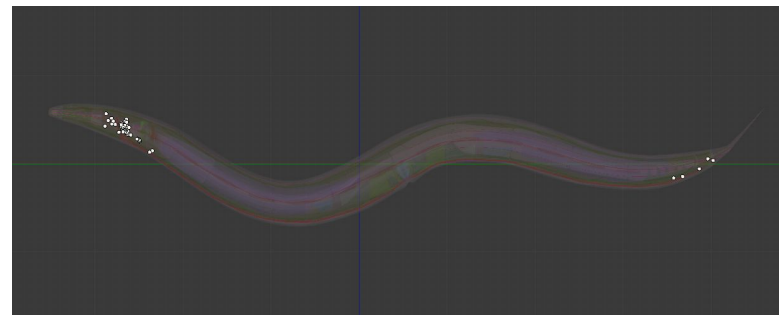
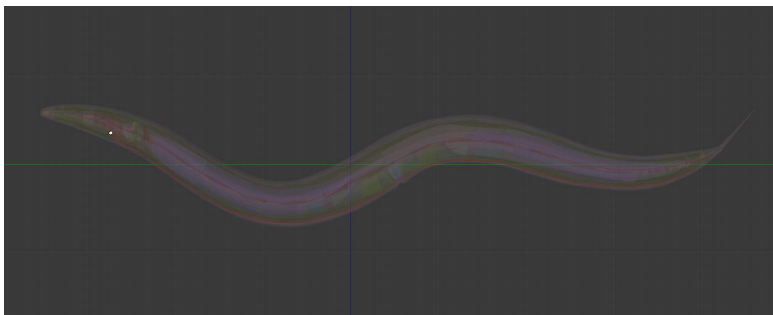
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$$\frac{W_{ij}}{|A_n|} =$$

Reductionist representation: add all components of a biological neural network (synaptic signaling, glia, astrocytes, protein expression, intercellular matrix).

What do we need to add to  $W_{ij}$  to make this model more “biological”? Or can we add biological realism without finding universal features or making model more descriptive?

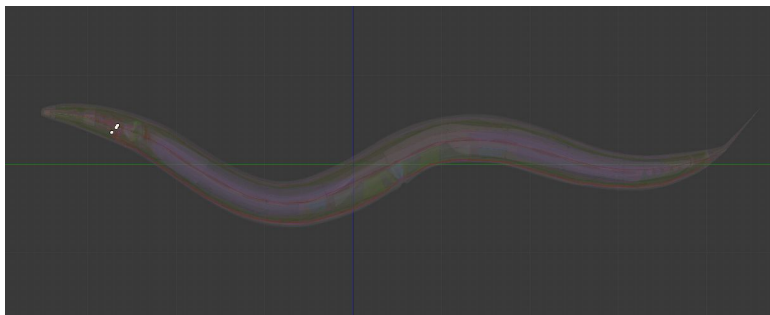


$$W_{ij} = \frac{1}{|A_1|}$$

$$W_{ij} = \frac{1}{|A_2|}$$

$$W_{ij} = \frac{1}{|A_3|}$$

Birth of neuronal cells (white) in *C. elegans* embryo at 265, 280, 300 minutes post-fertilization.

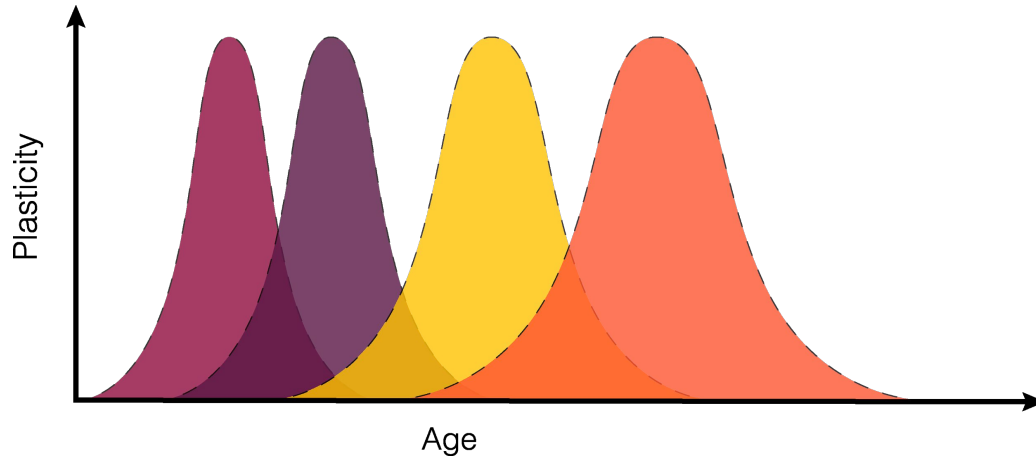


*C. elegans* embryonic connectome: *Biosystems*, 173, 247-255 (2018).

# Critical Periods

Critical period: window of developmental time that features an enhanced ability to learn.

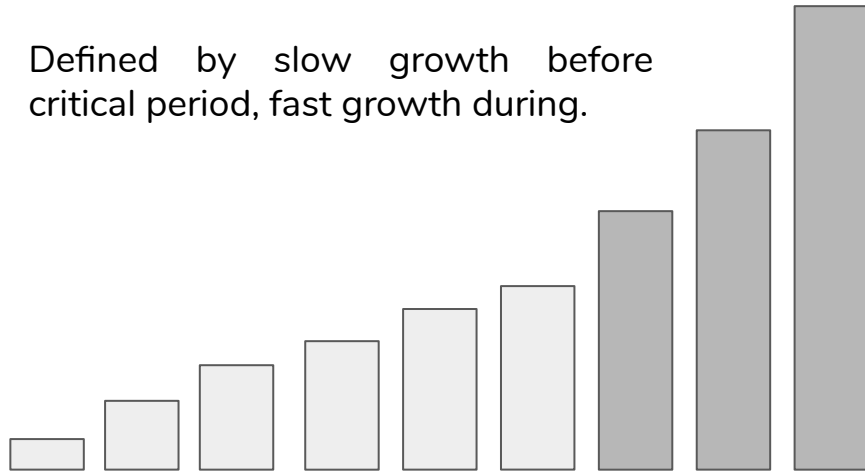
- associated with physiological reorganization.
- critical period window can have variable width, shifted earlier or later in development.



# Critical Periods

## Morphogenesis

Defined by slow growth before critical period, fast growth during.

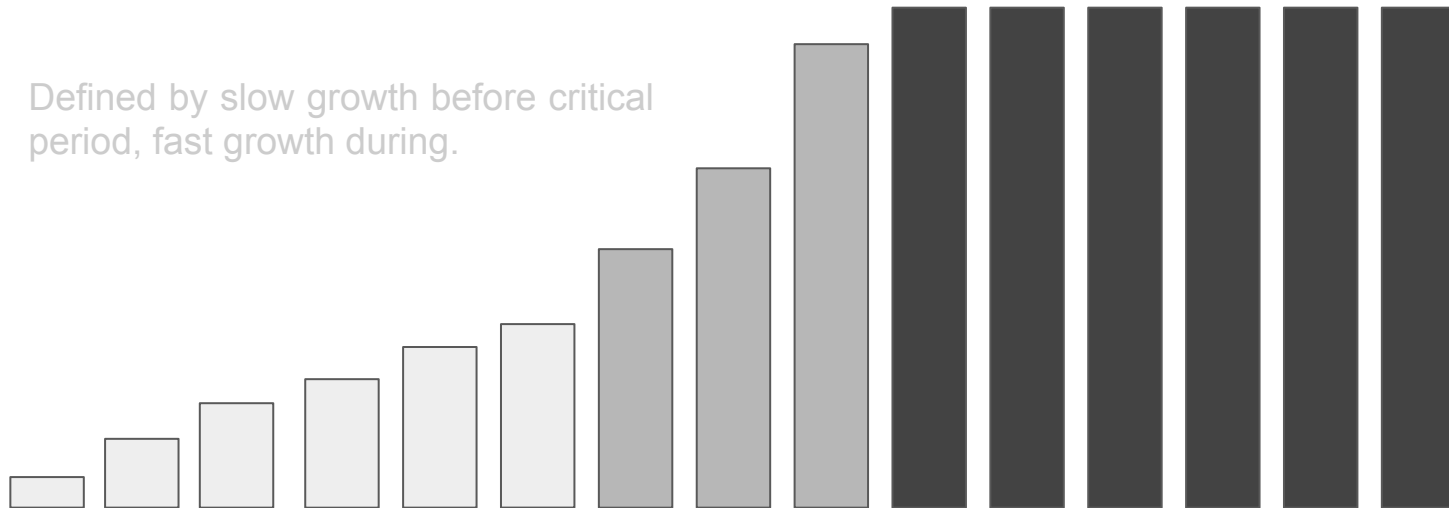


**Critical Period**  
(dark gray)

# Critical Periods

## Morphogenesis

Defined by slow growth before critical period, fast growth during.



Defined by no growth in size of matrix and relative stability among connections. Enables **developmental freedom**, or ability to learn on a stable network.

**Developmental Learning**

## CRITICAL PERIOD

Adult neural network: tends to be more homogeneous,

Less time for preliminary or auxiliary nervous system or behavioral traits to develop

Fewer varieties of experience prior to maturity.

## EARLY

## LATE

## DEVELOPMENT FREEDOM



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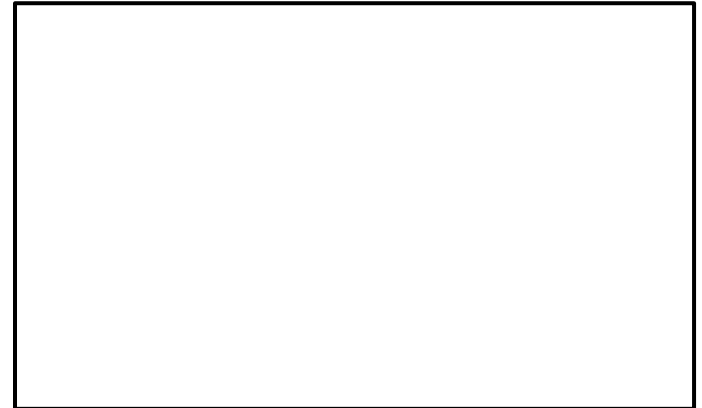
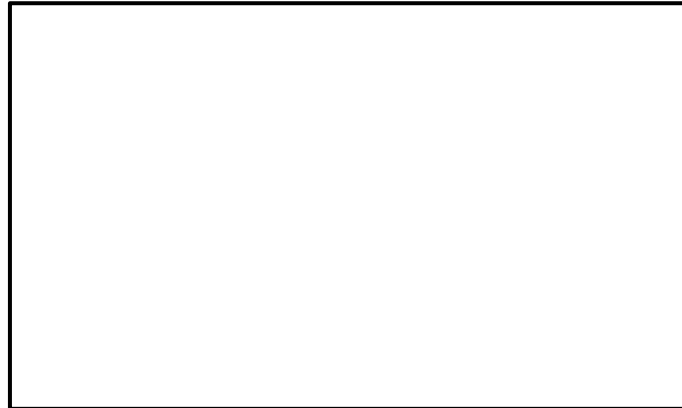
Adult neural network: can exhibit a larger number of possible configurations across a population of dBVs.

Slower development periods afford more time for exploration and 'wandering'.

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Low degree of overall developmental freedom.

Early critical period allows for faster ascendance to a fully mature suite of behaviors suite and capabilities.

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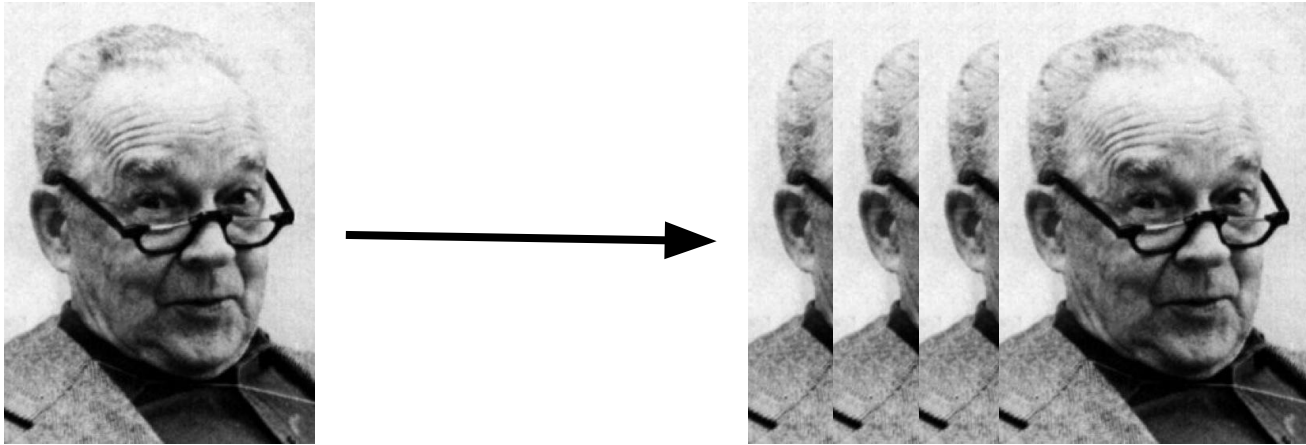
Low degree of overall developmental freedom.

Early critical period allows for faster ascendance to a fully mature suite of behaviors suite and capabilities.

High degree of overall developmental freedom.

Later critical period allows for a slower ascendance to adult behaviors and capabilities.

# Gibsonian Information



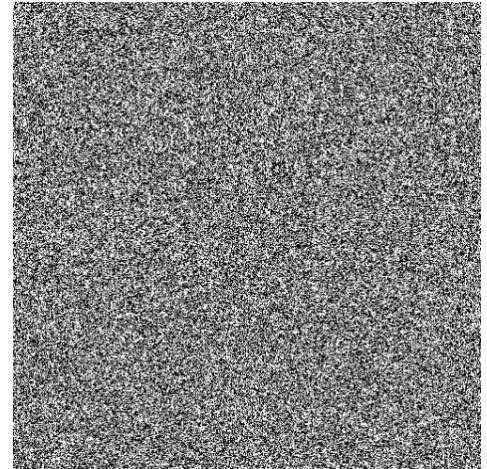
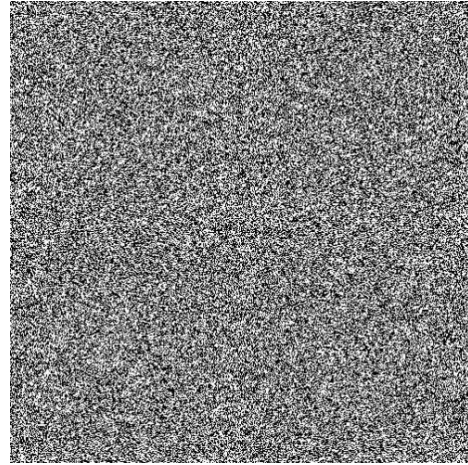
Motion provides information!

Optical flow, relative motion between object and observer, structure from motion

- 1) Flow, motion, and contrast.
- 2) Disjoint distribution (moving vs. stationary pixels).
- 3) Time-series structure (temporal-dependence).
- 4) Maximum vs. minimum coherent movement.



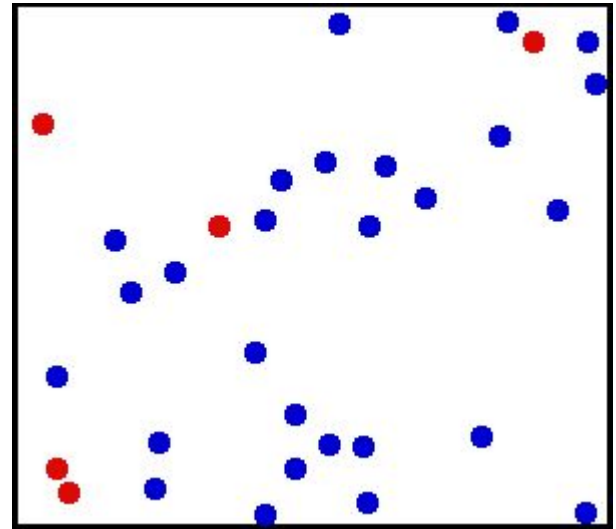
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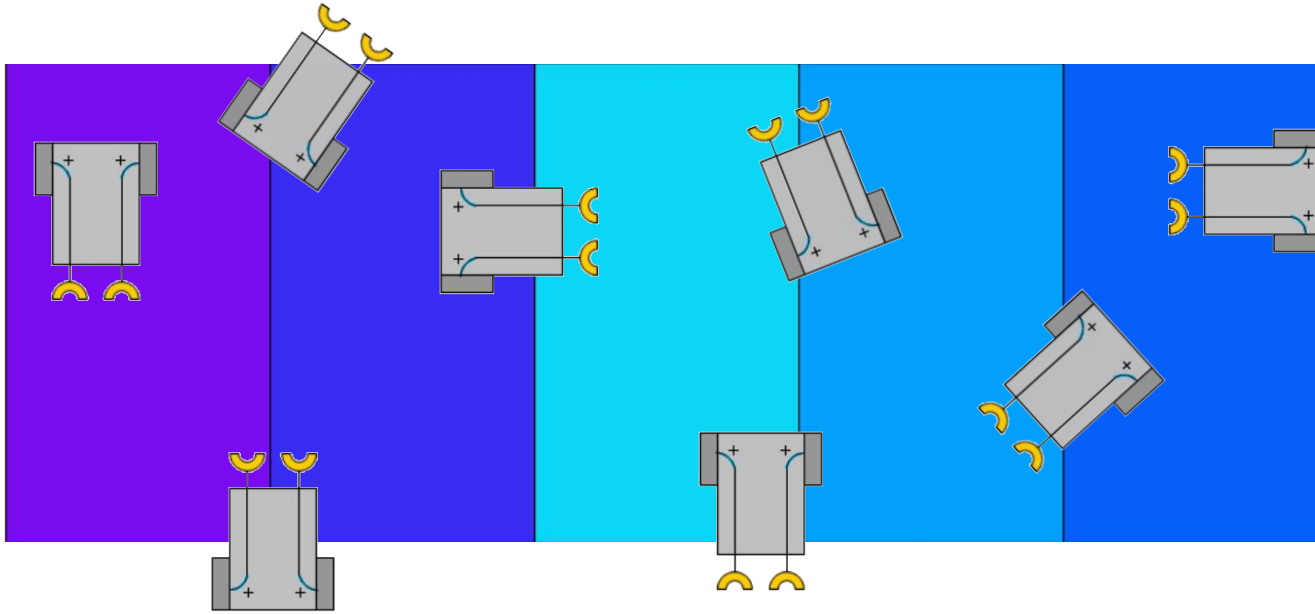
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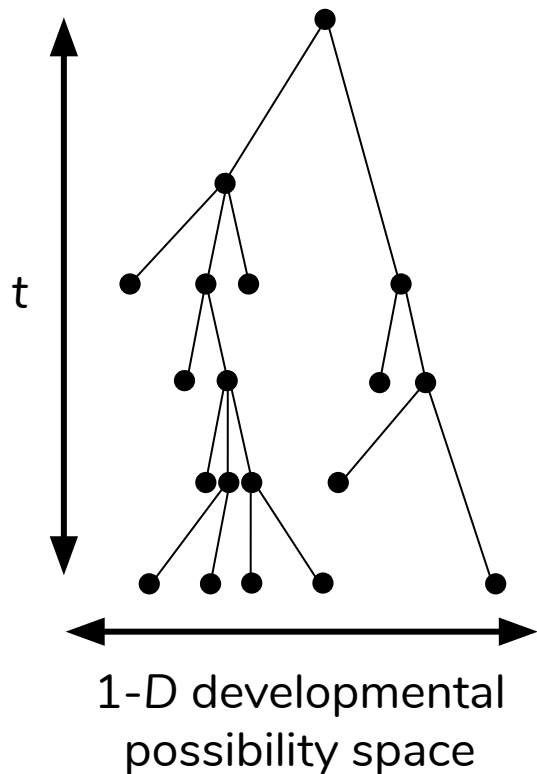


### **Towards a Developmental Reinforcement Learning (dRL) approach:**

- development of discrete layers will help agents learn better policies.
- growth of nervous system counters potentially large, continuous action space.
- curriculum learning (easy examples, then harder examples over time).

# Contingency in Development

Interpreting scenarios A-D



Each node is a developmental event, branches are possible developmental pathways.

Morphogenetic events (growth of matrix  $W_{ij}$ ) actually restricts: earliest branches (higher in the tree) restrict developmental system in possibility space.

More changes occur during critical period, but results in a network restricted to an increasingly smaller number of possible outcomes over time.

Developmental freedom may allow for limited convergent evolution of development (evo-devo). Brain vs. mind?

# Thanks for Your Attention!

## Acknowledgements

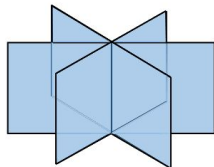
Developmental Braitenberg  
Vehicles

Stefan Dvoretzki  
Ziyi Gong  
Ankit Gupta

Read our preprint on  
ResearchGate!



<http://tiny.cc/q651tz>



Visit our lab website and Github!

<https://orthogonal-research.weebly.com/>

<https://github.com/Orthogonal-Research-Lab>

## REPRESENTATIONAL BRAINS AND PHENOTYPES



Thanks to the  
Saturday Morning  
NeuroSim group

