



Brains as a Model for Thermodynamic Computing

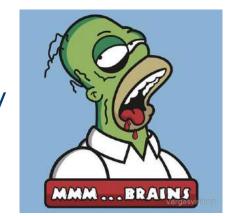
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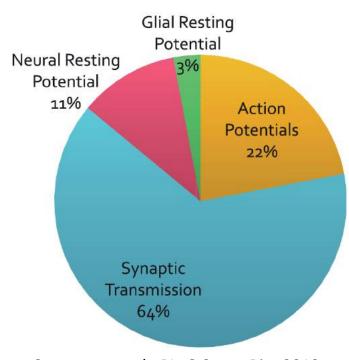
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Brains Reduce Energy Consumption

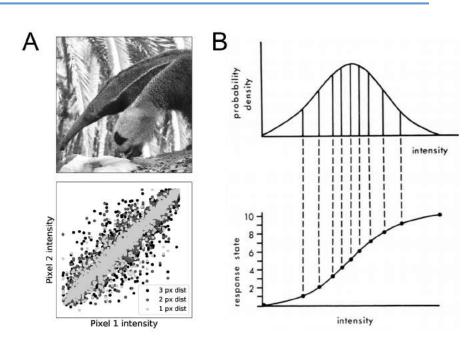
- At every level, the nervous system uses strategies to maintain high performance and information transfer, while minimizing energy expenditure.
- These range from ion channel distributions, to coding methods, to wiring diagrams (connectomes).



Sengupta, et al., PLoS Comp Bio, 2013.

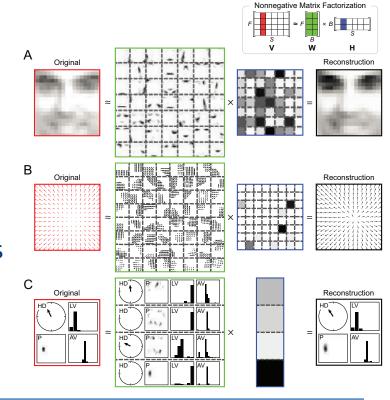
Organisms Must Be Efficient to Survive

- Nervous systems adapt their responses to the regularities of their input to increase the amount of transmitted information.
 - Maximize efficiency (reduce redundancy).
 - Responses should be independent of one another (decorrelation).
 - A stimulus should involve only a small fraction of the available neurons (sparse).
 - Neurons should span the input space.



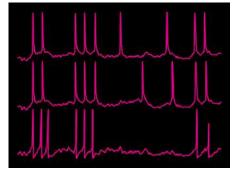
Sparse and Reduced Representations

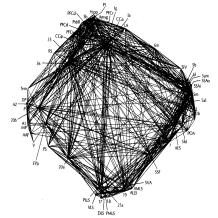
- Sparse coding and dimensionality reduction is a ubiquitous coding strategy across brain regions and modalities,
- Allows neurons to achieve nonnegative sparse coding (NSC) by efficiently encoding high- dimensional stimulus spaces using a sparse and parts-based population code.
- Reducing the dimensionality of complex, multimodal sensory streams is critically important for metabolically constrained brain areas to represent the world.



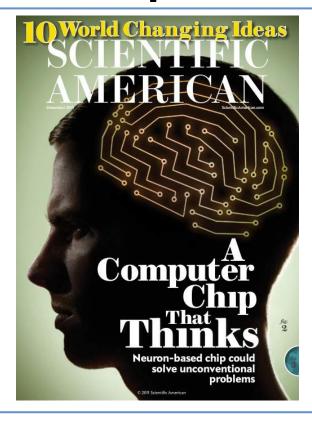
Brains Use Sparse Communication

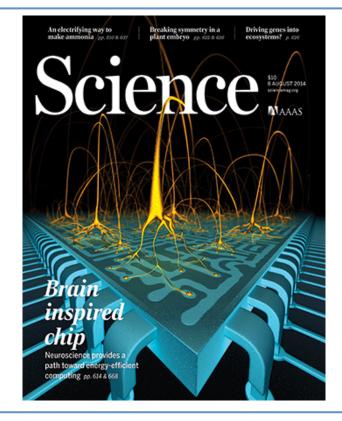
- The brain uses local analog computing and binary-spike communication.
 - Event-driven, asynchronous.
 - Minimal energy expenditure between spikes.
 - Low frequency output.
- The brain is a small world network.
 - Any two neurons are only a few hops away from each other.
 - Reduces wiring, but maximizes information processing.





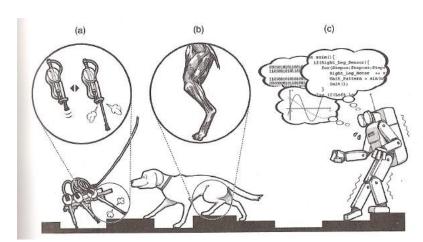
Neuromorphic Chips



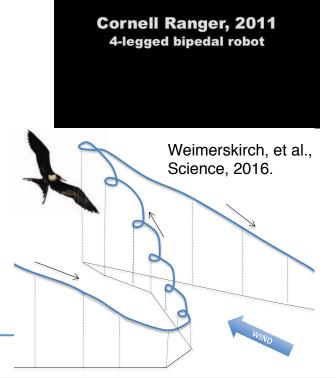


Embodiment: Non-Neural Theromodynamic Computing

- Morphological Computation
 - Certain processes are performed by the body that free up brain processing.

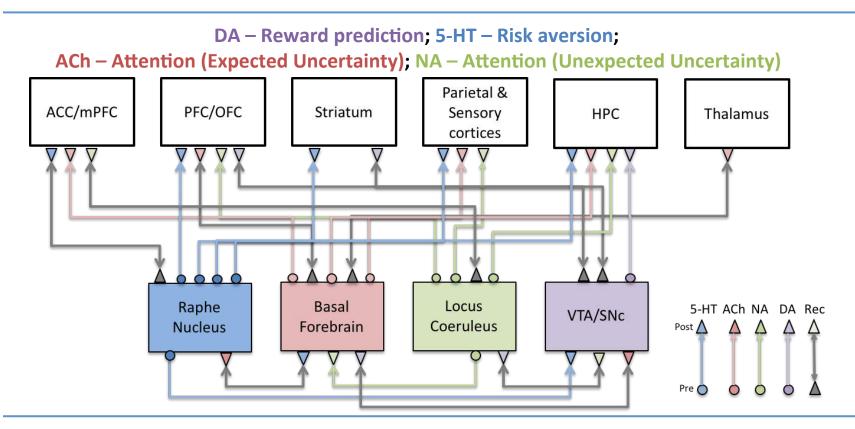


Pfeifer & Bongard, How the Body Shapes the Way We Think.



Value and Prediction

- Intelligent agents are equipped with a value system which constitutes a basic set of assumptions about what is good or bad.
- Organisms adapt their behavior through value systems.
- Minimizing surprise, by predicting future outcomes, minimizes the expenditures required to deal with unanticipated events.



Reducing Surprise Through Prediction

- Prediction is crucial for fitness in a complex world and a fundamental computation in cortical systems.
- This requires the construction and maintenance of an internal model.
- The brain maintains internal models for a wide range of behaviors; from motor control to language processing.



Krichmar, Hwu, Zou & Hylton, Cognitive Computation and Systems, In Press.

Biological Organisms Minimize Free Energy

- Natural systems must adapt or evolve to resist a tendency toward disorder in an everchanging environment (Ashby, 1947).
- Any self-organizing system that is at equilibrium with its environment must minimize its free energy (Friston, 2010).

Free Energy Principle - Unifying Brain Theory

Attention and biased competition

 $\mu_v = \arg \min \int dt F$

Optimization of synaptic gain representing the precision (salience) of predictions

Associative plasticity

$$\ddot{\mu}_{\theta ...} = -\partial_{\theta ...} \varepsilon^T \xi$$

Optimization of synaptic efficacy

Perceptual learning and memory

 μ_{θ} = arg min $\int dt F$

Optimization of synaptic efficacy to represent causal structure in the sensorium

Probabilistic neuronal coding

 $q(\theta) = N(\mu, \Sigma)$

Encoding a recognition density in terms of conditional expectations and uncertainty

Predictive coding and hierarchical inference

 $\dot{\mu}_{v}^{(l)} = D\mu_{v}^{(l)} - \partial_{v} \varepsilon^{(l)T} \xi^{(l)} - \xi_{v}^{(l+1)}$

Minimization of prediction error with recurrent message passing

The Bayesian brain hypothesis

 $\mu = \arg \min D_{\kappa_l}(q(\theta)||(p(\theta|\tilde{s})))$

Minimizing the difference between a recognition density and the conditional density on sensory causes

The free-energy principle

 $a, \mu, m = \arg \min F(\tilde{s}, \mu | m)$

Minimization of the free energy of sensations and the representation of their causes

Model selection and evolution

 $m = \arg \min \int dt F$

Optimizing the agent's model and priors through neurodevelopment and natural selection

Computational motor control

 $\dot{a} = -\partial_{\alpha} \varepsilon^{T} \xi$

Minimization of sensory prediction errors

Optimal control and value learning

 $a, \mu = \arg \max V(\tilde{s}|m)$

Optimization of a free-energy bound on surprise or value

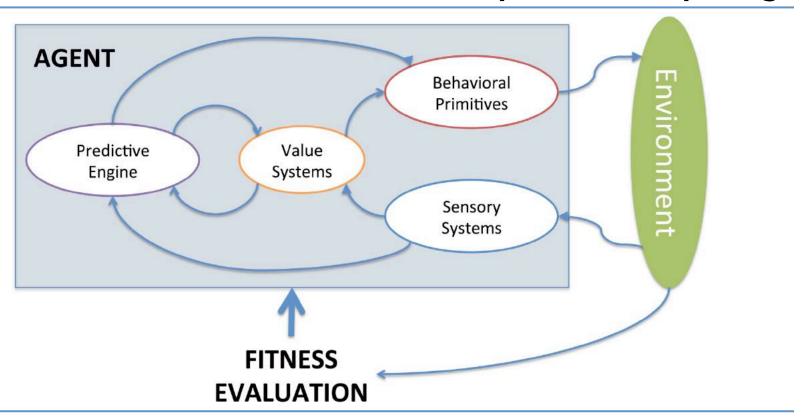
Infomax and the redundancy minimization principle

 $\mu = \text{arg max} \{ I(\tilde{s}, \mu) - H(\mu) \}$

Maximization of the mutual information between sensations and representations

Friston, Nat Rev Neurosci, 2010

Biomimetic Architecture for Thermodynamic Computing

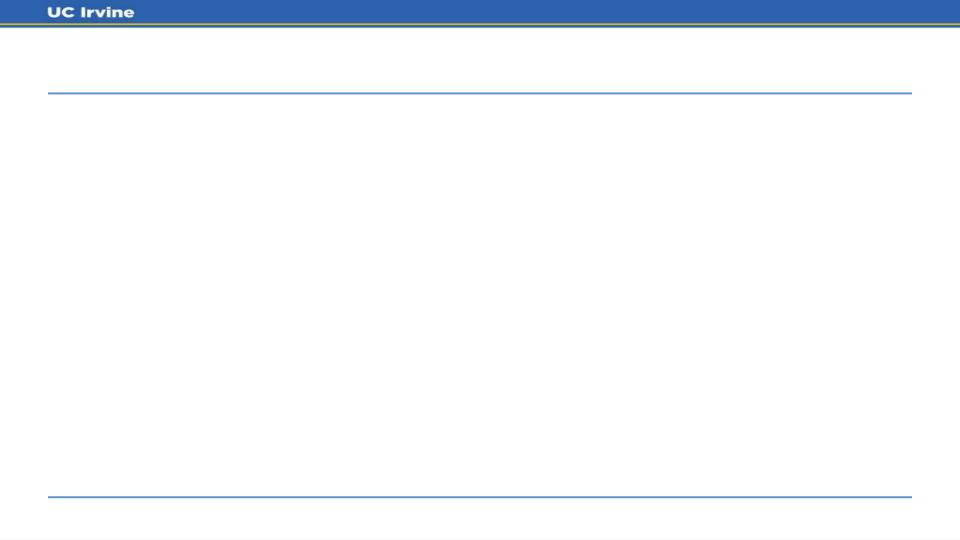


Biomimetic Architecture for Thermodynamic Computing

- A closed loop system where the control (artificial brain) is closely coupled with the body (robot) and the world (environment).
 - The agent has innate values, derived from the environment and value systems, which send signals to the brain to adapt behavior.
 - Since value is inversely proportional to surprise, predicting value is key to the agent's fitness.
- The world is dynamic, the agent must adapt its behavior to survive.
 - Fitness Evaluation is the metric for evolving these algorithms
 - How long the system can perform without intervention.
 - How successful the system is in the task environment.
 - How energy efficient the system is in performing a task.

More to Explore

- Beyeler, M., Rounds, E.L., Carlson, K.D., Dutt, N., and Krichmar, J.L. (2017). Sparse coding and dimensionality reduction in cortex. bioRxiv.
 - https://www.biorxiv.org/content/early/2017/06/14/149880
- Krichmar, J., Severa, W., Khan, S., Olds, J. (2018).
 Making BREAD: Biomimetic strategies for Artificial Intelligence Now and in the Future. arXiv.
 - https://arxiv.org/abs/1812.01184



Special Thanks!!!

Cognitive Anteater Robotics Lab (CARL)



Back row: Hirak Kashyap, Alexandro Ladron, Celeste Crowder, Emre Neftci, John Shepanski, Georgios Detorakis

Front row: Nikil Dutt, Ting-Shou Chou, Jeff Krichmar, Xinyun Zou, Tiffany Hwu



Todd Hylton UCSD



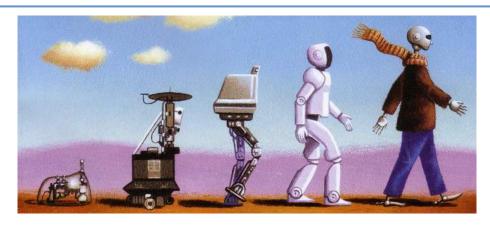
Tom Conte Georgia Tech



Yiorgos Makris UT Dallas

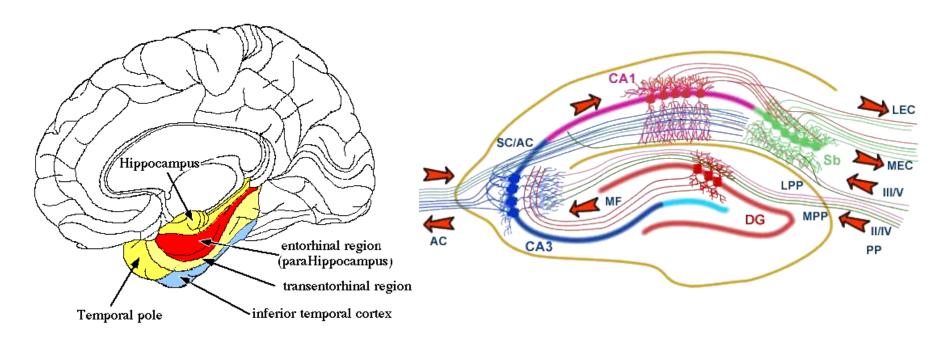
http://www.socsci.uci.edu/~jkrichma/

Future Outlook...



- To get a truly cognitive system one must study and be inspired by the brain and body of natural systems.
- Biological intelligence is an existence proof and currently our only working model.
- Following this will ultimately lead to intelligent cognitive robots and assistants.

Hippocampus – Learning and Memory



Hippocampus – Learning and Memory



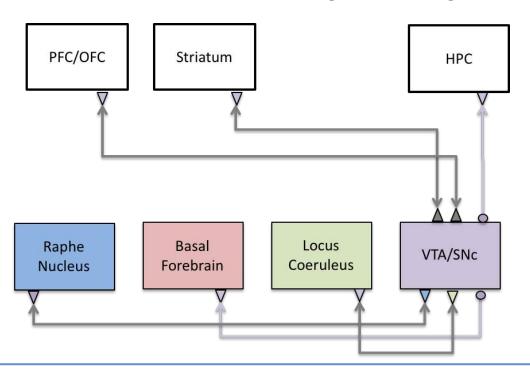


Self-Driving using CNNs on Neuromorphic Hardware

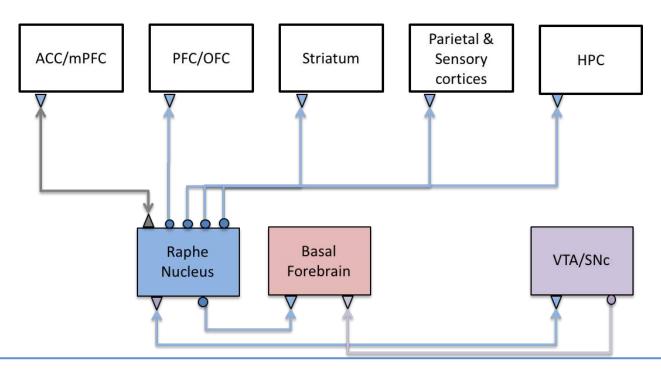




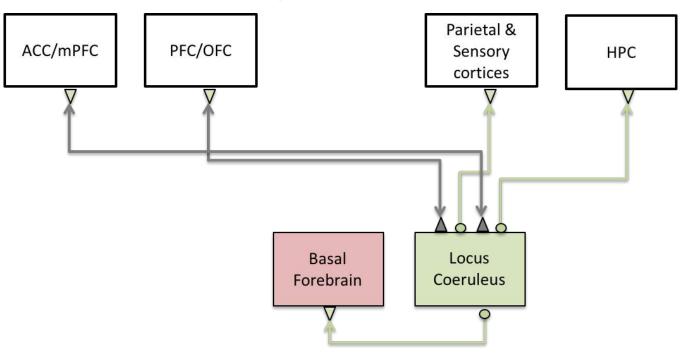
Dopamine - Reward, Saliency, Novelty, Invigoration



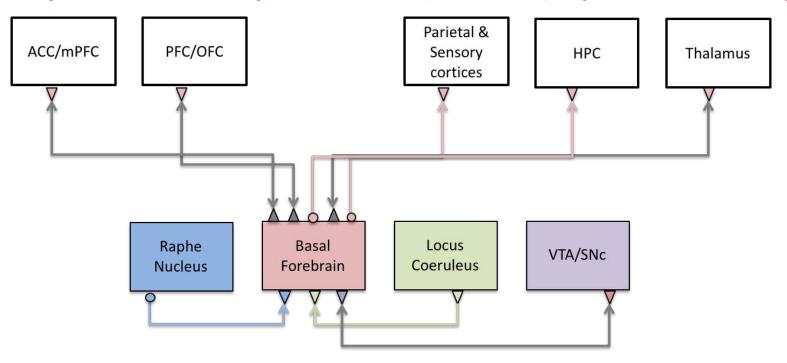
Serotonin - Harm Aversion, Anxious States, Withdrawal



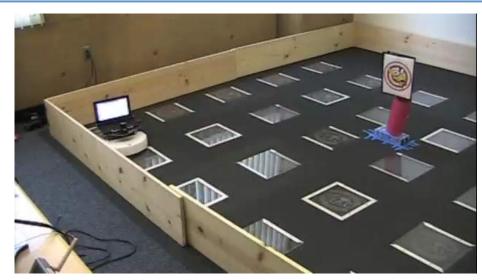
Norepinephrine - Vigilance, Unexpected Uncertainty



Acetylcholine – Memory Consolidation, Attention, Expected Uncertainty



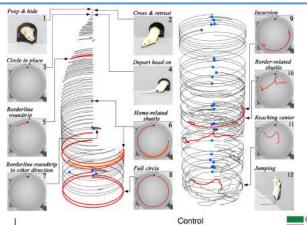
Value-Based Neurorobot

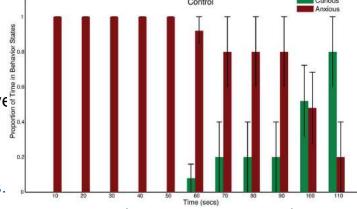


- Open Field Test for Rodents and Robots
 - Gold standard test for anxiety and obsessive compulsive disorders in animal models.
 - Typical behavior

Anxious – stay close to the walls until convinced it is safe.

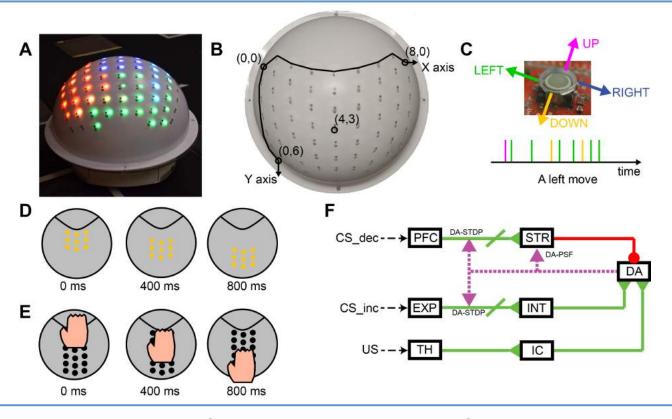
Curious - make crossings in the open and explore novel objects.





Krichmar, Front. Neurorobot, 2013

Predicting User Preferences with a Tactile Robot



Chou, T.-S., Bucci, L. D., & Krichmar, J. L. Front. Neurorobot, 2015.

Predicting User Preferences with a Tactile Robot

CARL-SJR - FeedMe Game
Tactile Sensing and Human-Robot
Interactive Reinforcement Learning

Ting-Shou Chou, Liam D. Bucci, and Jeffrey L. Krichmar Cognitive Anteater Robotics Laboratory University of California, Irvine

Mental Imagery

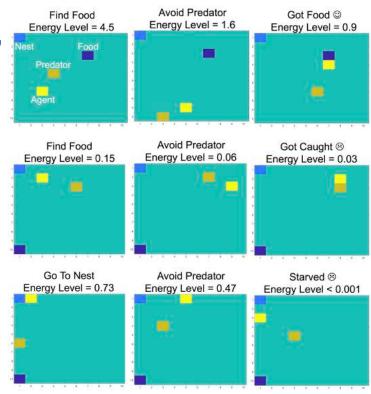
- Mental simulation can predict outcomes and overcome uncertainty.
- Mentalizing
 - Ability to understand another's state and plan accordingly
- Mental imagery is thought to be an important component for developing artificial systems that are cognitive or conscious.





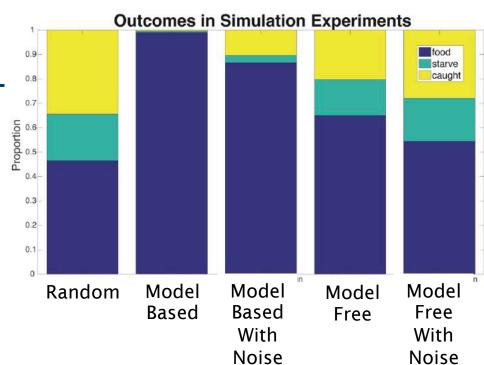
Predicting Value through Mental Imagery

- Developed a simple model based algorithm to recursively "imagine" how the effects of its own actions would affect another's actions.
- Designed a predator-prey scenario
 - Agent maximizes positive value
 - · Acquiring food.
 - Minimizes negative value
 - · Eaten by a predator.
 - · Starving.



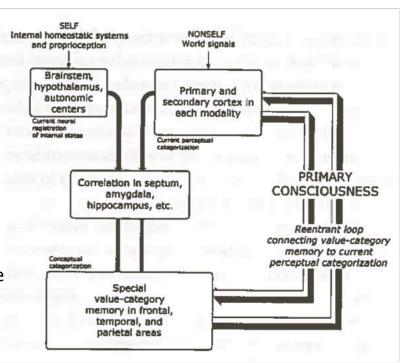
Predicting Value Through Mental Imagery

- Model-based system that could mentalize outperformed a modelfree reinforcement learning algorithm.
- Highlights advantage of planning ahead before taking actions in dynamic, noisy environments.



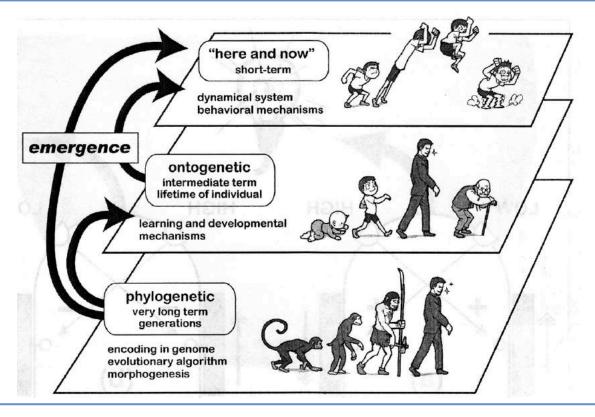
Theory of Neuronal Group Selection

- Plasticity is modulated by value
 - Signaled by neuromodulatory systems and hormonal systems.
 - Predicted by frontal cortex and limbic system.
- Value systems control which neuronal groups are selected and which actions lead to evolutionary fitness
 - Predicting outcomes that lead to positive value and avoid negative value. In this sense, predicting value is inversely proportional to surprise.
- From a nonequilibrium thermodynamic perspective
 - Value is associated with those actions that minimize the increase of entropy (e.g., feeding, predicting outcomes, gathering information).
 - Efficient movement, efficient alert sensory scanning of the environment, and foresight.



Edelman, Neural Darwinism, 1993.

Emergence of Cognition Occurs at Multiple Timescales



Brain Computing Facts

- That 1.3 kilogram lump of neural tissue you carry around in your head accounts for about 20 percent of your body's metabolism.
 - An average basal metabolism of 100 watts, each of us is equipped with the biological equivalent of a 20-Watts supercomputer.
 - Today's most powerful computers, running at 20 million Watts, can't come close to matching the brain.
- The brain has relatively shallow but massively parallel networks.
 - At every level, from deep inside cells to large brain regions, there are feedback loops that keep the system in balance and change it in response to activity from neighboring units.
 - The ultimate feedback loop is through the muscles to the outside world and back through the senses.