Live Long and Prosper

1. Why is there no *Flow Chart* **from Neural Representation to Knowledge Integration?**

Details on Neural Representation are presently sketchy, so it is difficult to chart a path whereby new items become integrated into our still mysterious knowledge architectures. We highlight here some key obstacles along with thoughts on how they might be addressed.

10. Can Active Learning Help?

- extrinsic motivation boosts synaptic activation
- amplified by **intrinsic motivation** of games, case studies
- neuronal summation (multisensory, connections) \rightarrow LTP
- active learning facilitates knowledge integration
- SLT framework might inspire new A.L. approaches?

NeuroMaze Level 1	<u>VISIT</u> : wv	ww.mazefire.com for	HARVARD M
Q. Which animal exhibits SOM processing, e.g. responding producing meaningful sourcessing.	IE aspect of language g to symbols and/or ds?		MEDICAL SCHOOL
A. parrots			
B. border collies			
C. vervet monkeys		C	- A Te Speci
D. bonobos (great apes)		C	
E. ALL of the above		✓	
Digital Maze Games us	e Intrinsic Motivation	to boost synantic learnin	ng MED science cases are <i>Active</i> :



<u>9. How does Language relate to Knowledge?</u>

- words, by themselves, mean nothing
- words and phrases: just "tags" for non-linguistic items?
- universal grammar is built upon universal physics
- aka evolutionarily deep objects, actions, relationships
- language offers fully symbolic operations, BUT
- sub-linguistic SNOPs might entail massive SCIPs



Rilling showed, in humans, enhanced trans-cortical STS connectivity which might facilitate fully symbolic neuronal operations (SNOPs) aka Language.

Suggested Readings			
Adcock RA, Thangavel A, Whitfield-Gabrieli S, Knutson B, Gabrieli JDE (2006). Reward-motivated learning: Mesolimbic activation precedes memory formation. Neuron 50:507-517.			
Baddeley A (1992) Working Memory. Science, 255:556-559.			
Baars BJ, Franklin S (2003) How conscious experience and working memory interact. Trends Cogn. Sci. 7:166-172.			
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Brady TF, Konkle T, Alvarez GA and Oliva A (2008). Visual long-term memory has a massive storage capacity for object details. PNAS 105:14325-14329.			
Eichenbaum H (2014) Time cells in the hippocampus: A new dimension for mapping memories. Nature Reviews Neuroscience 15:732-744.			
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Mashour GA (2006) Integrating the science of consciousness and anesthesia. Anesth. Analg. 103:975-982.			
O'Keefe J, Dostrovsky J (1971) <u>The</u> hippocampus as a spatial map. Preliminary evidence from unit activity in the freely-moving rat. Brain Res. 34:171–175.			
Parker ES, Cahill L, McGaugh JL (2006) A case of unusual autobiographical remembering. Neurocase 12:35-49.			
Quiroga RQ, Reddy L, Kreiman G, Koch C, Fried I (2005) Invariant visual representation by single neurons in the human brain. Nature 435:1102-1107.			
Song S, Miller KD, Abbott LF (2000) Competitive Hebbian learning through spike-timing-dependent synaptic plasticity. Nature Neuroscience, 3, 919-926. Doi:10.1038/78829			
Tononi G (2008) Consciousness as integrated information: a provisional manifesto. Biol. Bull. 215:216-242.			
VanRullen R, Koch C. (2003) Is perception discrete or continuous? Trends Cog. Sci. 7:207-213.			
Ward, LM. (2003) Synchronous neural oscillations and cognitive processes. Trends Cog. Sci. 7:553-559.			

AVAILABLE RESOURCES Origins of Syntax & Semantics (2011) Neural Words (technical report) Synaptic Learning Theory white paper ongoing-work: Motivation & Reward Systems



Knowledge Acquisition, Integration and Reward: a Neuronal / Synaptic Perspective of STEM Learning Annie G. Bryant AND Donald M. O'Malley Behavioral Neuroscience Program & Dept. Biology, NU, Boston MA

In education, Big Data is all about finding multivariate correlations and using that to improve learning and teaching outcomes: this is called *evidence-based* teaching or learning. But this does not explain how we learn. In our extension of Synaptic Learning Theory (SLT), we seek a path to a more principled understanding of the neural processes that acquire and integrate knowledge, as well as the nature of mammalian knowledge architectures. Advancing learning science without these principles might be akin to advancing molecular biology without knowing what codons are or what restriction enzymes do. Linking KA's to synaptic events could take 100 years or more, but could have far-reaching ramifications. Coming soon: Synthetic Neuroscience Institute: d.omalley@neu.edu

8. How is NEW Knowledge Integrated?

- role of **SCIP** 4 Sub-Conscious Information Processing
- valuation of connections in trans-neuronal representations
- expansion and enrichment of Knowledge Architectures
- how are new CONCEPTS created?
- role for Symbolic Neuronal Operations (SNOPs)?
- percolation for Pavlovian-connections & processing?

HOCs Rock! Higher Order Correlations

- within DMRs (correlations between DMR epochs)
- SCIP scanning of knowledge architectures for best fits
- roles for hippocampus, Bayesian inference? J. Tenenbaum - percolation theory and Auto Associative Networks
- cortical proofreading mechanisms (www.syndar.org)

Bound items in DMRs relate to diverse, trans-cortical representations involving many parts of neocortex. Hippocampus might sequence DMR epochs via γ on Φ rhythms.



G FUTURE SCIENTIST



3. WHERE are Neural Representations Located?

• working memory strongly involves prefrontal cortex (PFC) but PFC-hippocampal connections are also prominent - one trial learning in hippocampus-amygdalar complex shown for "novel" and "familiar" cells (Rutishauser et al. 2006) - locations are best guesses based upon sparse information - both locus & nature of most neural codes are largely unknown







7. How do α , θ and γ Rhythms Contribute? - γ (gamma) rhythms underlie **conscious experience** - γ on θ rhythms enable working memory (WM) - synchrony amplifies signals, LTP & STDP (plasticity) - coupled oscillators might link DMR items - memory recall depends more on θ than on α -rhythm - rhythms intensify upon learning, decision making see e.g. Benchenane et al. 2010, Kendrick et al. 2011, Watrous et al. 2015. • From *Stream of Consciousness* (SoC) we **RECORD** into DMRs

• Excerpts of DMRs are **SAVED** into Long Term Memory (LTMs) • Most learning/memory is **Hebbian**, which requires coincident firing • Multiplexing/concurrent rhythms is a computational multiplier, BUT .

• ARE rhythms computations? [as opposed to mere carrier bands]



SENSORY INPUTS

Bottom-Up





see On Intelligence (Hawkins) and Baars & Franklin (2003, 2005, 2007)

5. How do WM contributions differ from DMRs?

SIGNALS

Top-Down

- WM holds 7 +/- 2 items (see Lisman & Idiart γ on θ model) - DMRs are largely non-linguistic in nature: "knowledge tokens" WM is repetitive and effortful vs. the 1-trial, no effort DMR writing mechanism that might store thousands of items. Both entail conscious states, but they seem rather different!

Consc. Epochs see VanRullen & Koch (2003); Alpha/Gamma see Ward (2003) in Trends in Cogn. Sciences

6. Atoms to Organisms: Learning by Analogy

- Exemplar STEM Knowledge Architecture (KA): *Biology* atoms, molecules, DNA, cells, organisms, clades ... LIFE - KAs are often analogous to other KAs (network duplication?) - Universal Physics (objects, actions, properties) shaped the evolution of CNS, neocortical architectures - Theoretical model of Knowledge Integration (e.g. Hinton,

1986) has no clear correlate in neocortical systems



The number of possible sentences vastly exceeds the number of words. By analogy, the number of knowledge architecture constructs vastly exceeds the number of "stored" working memory tokens.



10. HOCs Rock! [Higher Order Correlations]

- within DMRs (correlations between DMR epochs)
- SCIP scanning of knowledge architectures for best fits
- roles for hippocampus, Bayesian inference? J. Tenenbaum
- percolation theory and Auto Associative Networks
- cortical proofreading mechanisms (www.syndar.org)

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Distributed Representation Across Neocorte			
DMR COMPONENTS	←→	Possible Anatomica	
 Self actions Thoughts Phrases Observed Actions Spatial content Time References Numbers Object recognition Emotions Names Object Location Person Description Direction Quotes Sound Recognition Reflective Thoughts Planning Ahead Smell 		 Frontal, primary mot Frontal, hypothalam Broca's Area, left fro Parietal, premotor m Hippocampus, occip Basal ganglia, occipital Occipital, Temporal Amygdala, hypothala Hippocampal ventral Fornix, parietal, temp Face neurons inferio Mammillary neurons parietal Broca's Area, left fro Auditory cortex, tem Medial prefrontal, left Anterior prefrontal context 	

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Ganz & O'Malley, 2012

al Locations

tor cortex, premotor cortex

ontal, Wernicke's Area, left temporal nirror neurons, occipital, inferior frontal gyrus , parietal, mammillary nucleus, anterior cingulate l, parietal, cerebellum, premotor cortex,

amus

I tegmental area loop poral, anterior cingulate cortex or temporal visual cortex, occipital , head direction cells, hippocampus, occipital,

ontal, Wernicke's Area, left temporal, occipital poral, insular cortex ft anterior middle temporal gyrus, right cerebellum ortex gdala, entorhinal cortex, anterior



Rilling showed, in humans, enhanced trans-cortical ST SconNectivity which might facilitate fully symbolic neuronal operations (SNOPs) aka Language.



- trans-cortical representations (see items #6, #8)
- binding depends on alpha & gamma-band oscillations?
- largely non-linguistic in nature: "knowledge tokens"



Consc. Epochs see VanRullen & Koch (2003); Alpha/Gamma see Ward (2003) in Trends in Cogn. Sciences

5. What exactly DO we RECORD and SAVE? - "salience / novelty" = <u>first pass "filter</u>" into SoC

- from SoC, 2^{nd} filter evaluates and stores epochs
- of ~ 15,000 epochs per day (= 1 DMR) only a small subset are consolidated into long-term memory
- epochs might be assigned value by comparison to
- stored, related items $= 3^{rd}$ filter - subconscious IP, sleep & hippocampus all contribute



- From *Stream of Consciousness* (SoC) we **RECORD** into DMRs
- Excerpts of DMRs are **SAVED** into Long Term Memory (LTMs)
- Rats store contextual LTMs that require an intact hippocampus • Do rats have DMRs? Just ask one! or see Eichenbaum and Fortin, 2003
- Most learning/memory is **Hebbian**, but this requires coincident firing



