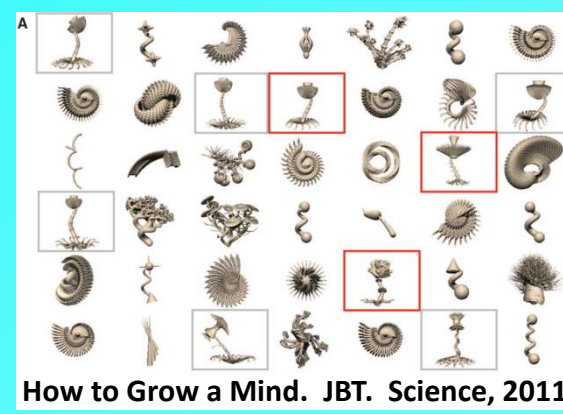


We can **THINK** only
with on-board
knowledge:
on-line does
not cut it.



Learning Categories from a Neuroscience Perspective

Learning Categories: Any parsing of learning methods (aka teaching approaches) into categories **is intrinsically arbitrary**: red-line boundaries do not exist. But there are real differences because different categories engage neocortex/hippocampus in distinct, crucial ways. The 3 specific categories used here differ greatly in terms of style and baud (information transfer rate). **Synaptic Learning Theory** is a network-based formulation of Hebbian Learning provided in 2015 posters: www.zfhindbrain.com—scroll down DMR page

Knowledge
Transfer
Rate

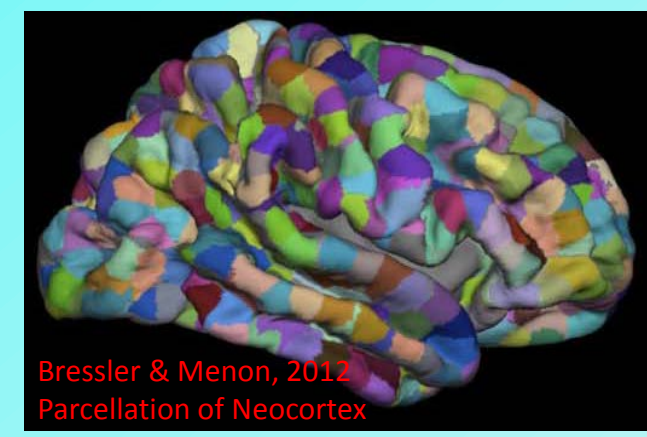
Adaptive Learning

- highest throughput, e.g. Knewton
- automated tidbit-addition, personalized
- but brains are not “flash drives”

Modeled after **Adaptive Testing**, computer algorithms monitor progress and proffer new knowledge tidbits to add to our neocortical database. ≈ Equivalent to rote lecturing, but computer-personalized.

Neural Basis of Adaptive Learning

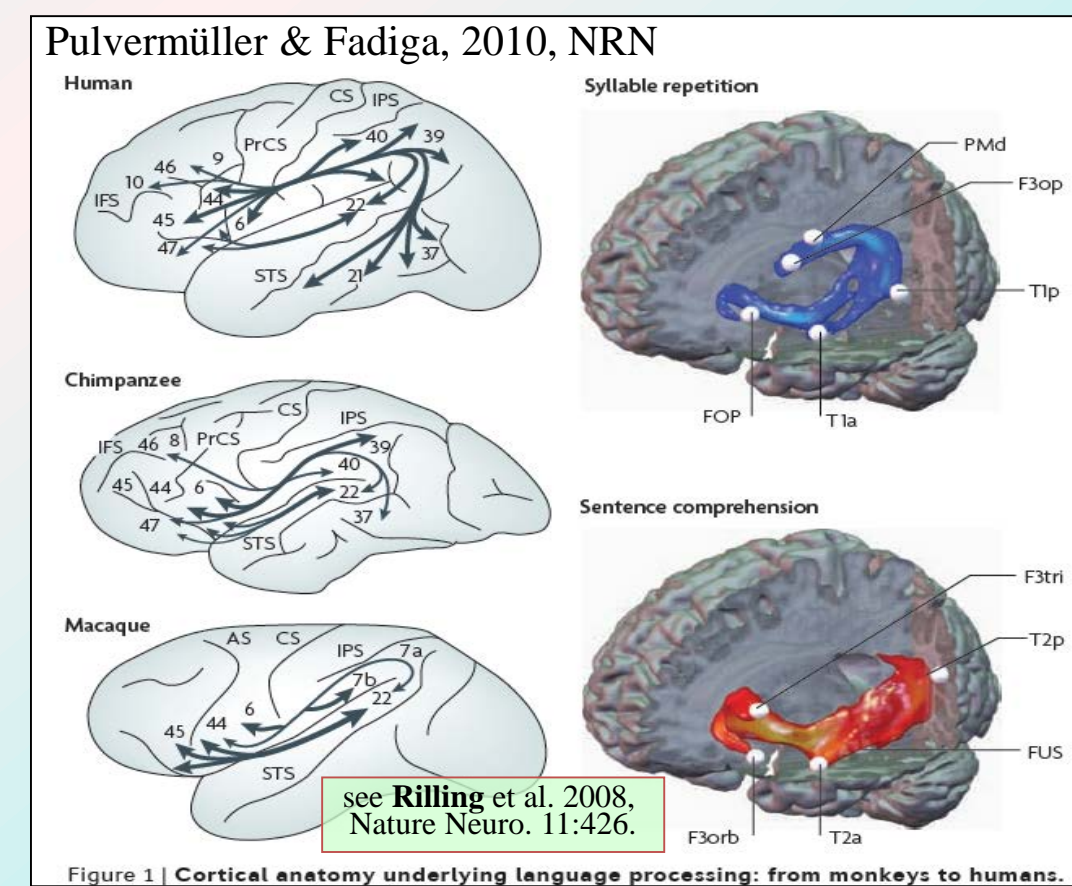
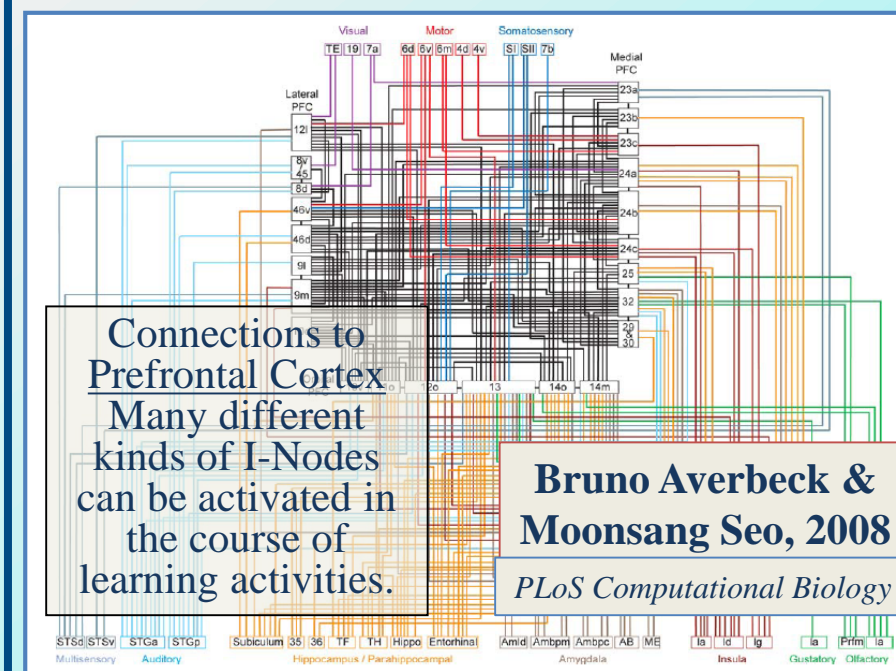
Basic neural mechanisms here are used in all 3 L-categories: “word” nodes are activated and linked, e.g. camels-humps, tigers-stripes. Utilizes neocortical **I-nodes** and “hippocampal context” plus neocortical broadcast systems. see **SNOPs** below



Neocortex has about 200 Parcels and each P. has about 100 million neurons or about **10,000 I-nodes** (2 million I-nodes total, each might store 1000's of patterns). But many I-nodes do other brain tasks, besides learning. Assumes 10k neurons / I-node, which equates to 50 I-nodes per fMRI voxel.

Why Complementary Memory Systems?

- Hippocampus is FAST LEARNING system law MMR 1995
- Neocortex slowly updates Knowledge Systems
- OR perhaps fast neocortical learning is key?
- DMRs likely stored in Neocortex (zfhindbrain.com)
- Linking of DMR items: fast & experience-based
- MMR = **McClellan et al.** 1995, Psychological Review



Rilling showed enhanced trans-cortical STS connectivity in humans which might facilitate fully symbolic neuronal operations (SNOPs) aka Language. Requires SCIP = sub-conscious information processing

Universal Physics Universal Grammar

Universal Physics

Neocortical Modules know myriad aspects of the world from visual cliffs to joining visual blips into objects. **This is the neural foundation** that supports all learning and ultimately language.

Innate Knowledge aka: Evolutionary Learning

Darwin's Contribution

Evolution organized the vertebrate brain to create maps, assess stimuli and make decisions. Much world knowledge is encoded in our genome, e.g. fear of snakes.

Analogical Reasoning

Flexible Dancers vs.
The Cognitively Inflexible

Resonant Brain Modules

The modular nature of neocortical columns offers a natural means to form resonant structures in discrete Knowledge Domains. **Analogies can thus build new Knowledge Architectures by ≈ replicating the structure of existing AAN / I-node-clusters.** Abstract analogies are derived from real-world U.P. constructs (which also enable SNOPs)

Nodes and Pathways

- DTI and PFC maps highlight major thoroughfares
- 20 billion neurons means lots of Info. Nodes (I-nodes) aka auto-associative networks (all-to-all wiring is not possible)
- coherent packets must be routed between nodes
- How e.g. *Pink* can be associated w/ *Elephant* is unknown!

Auto-Associative Networks Store Information

- The hippocampus & neocortex have many AANs
- during ENCODING new patterns are stored
- during RETRIEVAL partial input recalls full pattern
- AAN's can tolerate extensive loss of cells, synapses & might be used for long-term neocortical storage

Learning in Neural Space: Nodes & Symbols

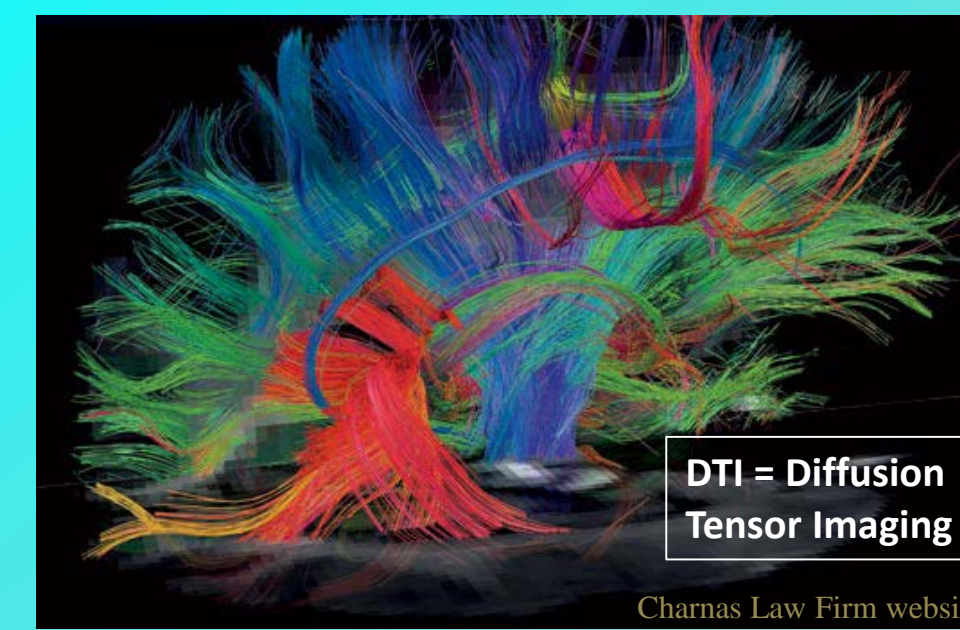
SCIP = Sub-Conscious Information Processing

All our thoughts and sentences emerge from SCIP and appear in Stream of Consciousness likely due to focal γ -band in neocortex. Then DMR “excerpts” are stored and are the basis of all long-term Declarative Memory (episodic + semantic) i.e. of all knowledge. **SCIP runs all of this w/ help of motivation.**

SNOPs are Symbolic Neuronal Operations

Linguistic & Physical Items are Richly Entangled

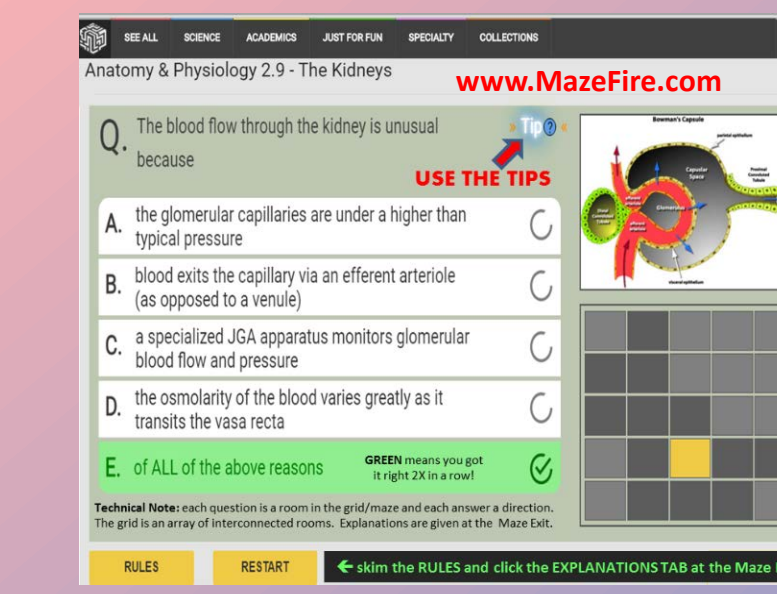
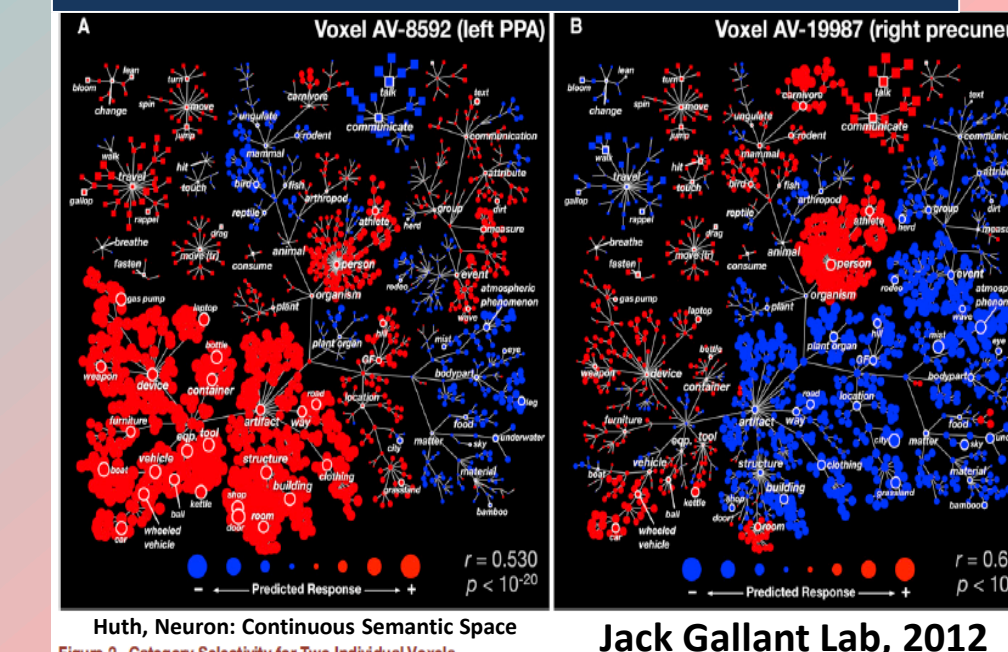
- the Chimp brain represents pre-linguistic encoding
- massive neocortex expansion co-occurred w/ language
- both linguistic tags & real-world items are deeply connected
- but “new conversations” are largely symbolic, fragile
- sub-linguistic SNOPs might entail massive SCIP



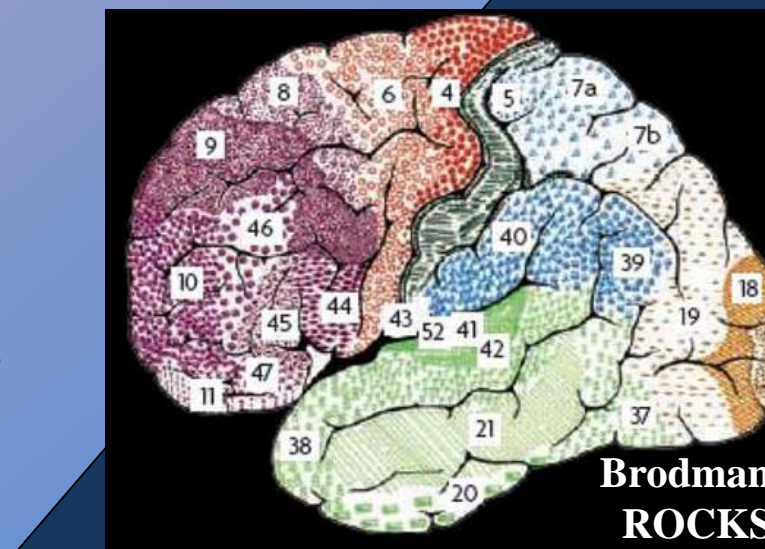
Linguistic SNOPS = vastly expressive

symbol manipulation system, but rides upon SNOPS-nl [non-linguistic symbol system] which is derived from vast evolutionary learning/calculations which stored vast innate knowledge aka U.P.

Semantic Networks: Voxel Eye's View

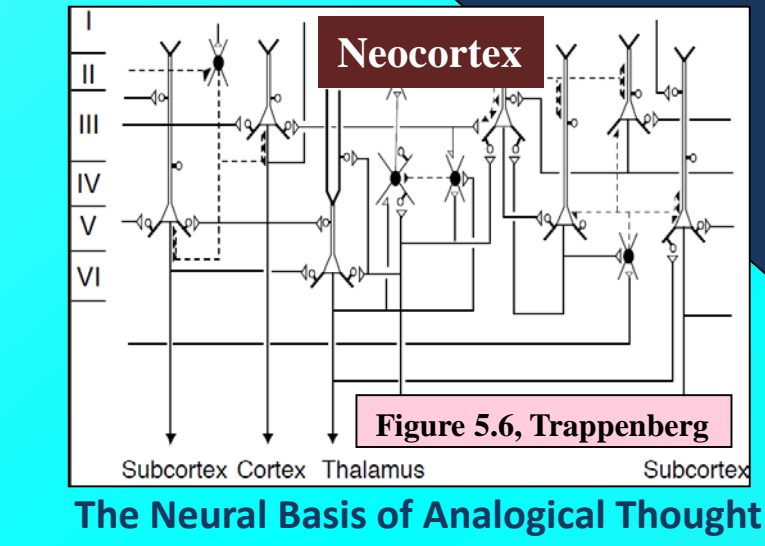


www.MazeFire.com has many **Reflective Learning** games (click-n-play) in e.g. Bio1, Physiology, Neurobiology, Micro, Biochem and Pharmacology

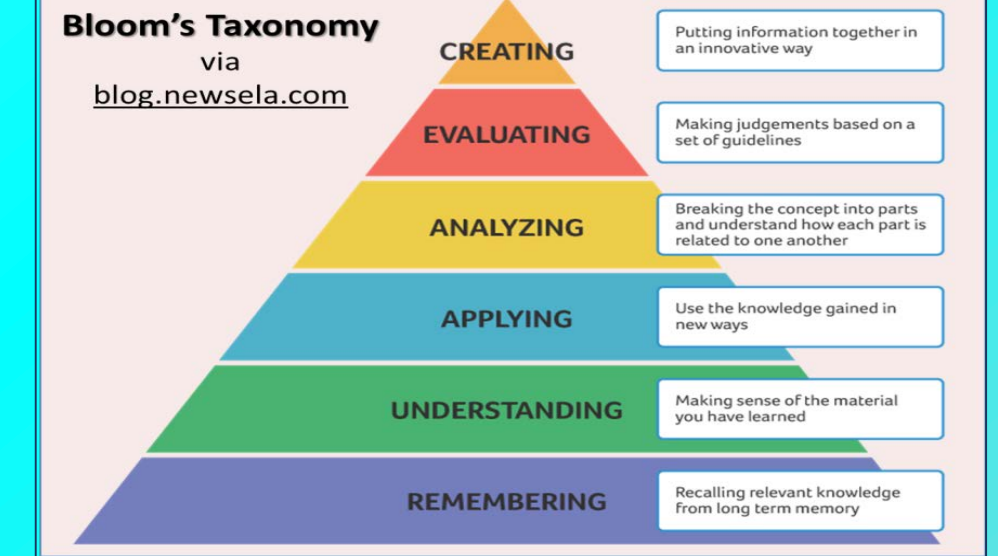


Connecting Learning Science to Neocortical Networks & Synapses

NODAL THEORY: The linking of ANY two items REQUIRES that those items become connected in **Neocortical Space**. Neuroscientists do not know how this happens but activation of **information nodes** (aka Auto-Associative Networks) via compact neural messages (packets, neural words) is central to both rote learning and the making of new, higher-level connections aka **Cognitive Advancement**. *Packet Routing and Nodal Theory* are discussed in Belloch-O'Malley, 2017 on DMR page at www.zfhindbrain.com (at bottom). For deeper dive see *Neural Words* essay, mid-page.



Not an Endorsement of Taxonomy



While **Adaptive Learning** typically entails Levels 1 and 2, **Reflective Learning** moves higher and **Active/Project Learning** can invoke the highest levels. But this is NOT EVEN CLOSE to being a hard and fast rule.

Active Learning: e.g. Projects, Cases

- slowest throughput, but ≈ deepest learning
- acquire & apply knowledge, build on experience
- concept maps, peer review, collaborations
- *example*: build DNA model, figure out base-pairing

Neural Basis of Active Learning

Likely uses the greatest variety of neocortex resources / regions. Conversations, tactile inputs, motor acts and other elements all engage more I-nodes and lead to greater convergence onto new concept nodes: e.g. build double-helix with H-bonded base pairs → easier retrieval, “knowing”.

**** While Active / Project Learning** uses a greater variety of resources, **Adaptive Learning** uses specific resources much more intensively/repetitively, leading to fatigue, tedium and ↓ motivation: think MCAT prep!

Self-Regulated Learning: entails cognition, metacognition and motivation; including forethought, planning (Schunk, 2005; Schraw et al., 2006). **Metacognition:** “Do I understand this topic?” or “How do I best study this topic”. SRL entails monitoring self-progress and figuring out how to improve studying. SRL is unequivocally “reflective” but can be part of active L. as well

Learning Categories: Audience Participation Module (comments welcome; your name or proponent optional)

1. Active
2. Reflective
3. Adaptive (Knewton Co.)
4. Self-Regulated
5. Rote
6. Socratic
7. Incidental
- 8.
- 9.
- 10.
- 11.
- 12.
- 13.
- 14.
- 15.

note: **Evolutionary Learning** is outside the scope of this meeting even though this learning mechanism shapes/constrains the scope of all **Learning Neurodynamics**.

Melissa B. McElligott & Donald M. O'Malley: How Reflective and Adaptive Learning Strategies relate to a Nodal Theory of Neocortical Computation, Northeastern University