## Favourite scholarly papers that I can remember, part 1-6.

A NON-DETERMINISTIC APPROACH TO ANALOGY, INVOLVING THE ISING MODEL OF FERROMAGNETISM

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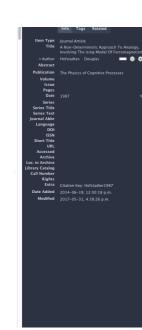
## ABSTRAC

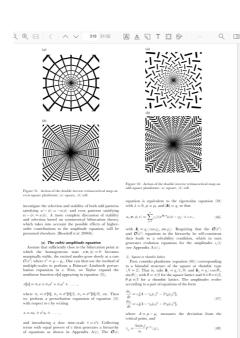
A close analysis of several abstract analogies reveals the critical role played by directed limits created in the act of perceiving the structures in the analogy. These directed limits, having bi-stable orientation properties, are similar to the bi-stable spins of the Ising model of ferromagnetism. The similarity is enhanced by the fact that the limits orientations are not deterministic but stochastic, and the degree of order and disorder in the system is regulated by a formal parameter playing a role analogous to that of temperature in the Ising model. The paper is concerned principally with showing how temperature-controlled flipping of the directed limits allows "perceptual Bloch domains" to emerge, thus facilitating discovery of subtle analogies.

In the making of an analogy between two situations, a critical ngradient is how those situations are framed in terms of known concepts, raming as situation in terms of concepts in smultike visual preception of a kene, in which the goal is to attach numerous labels ("chair", "elephant", Vesurus") to regions of the visual field. The difference is that a situation as generally abstruct rather than risual, and consequently the labels to be titached to perts of it are usually at a higher level of abstraction than those in visual perception.

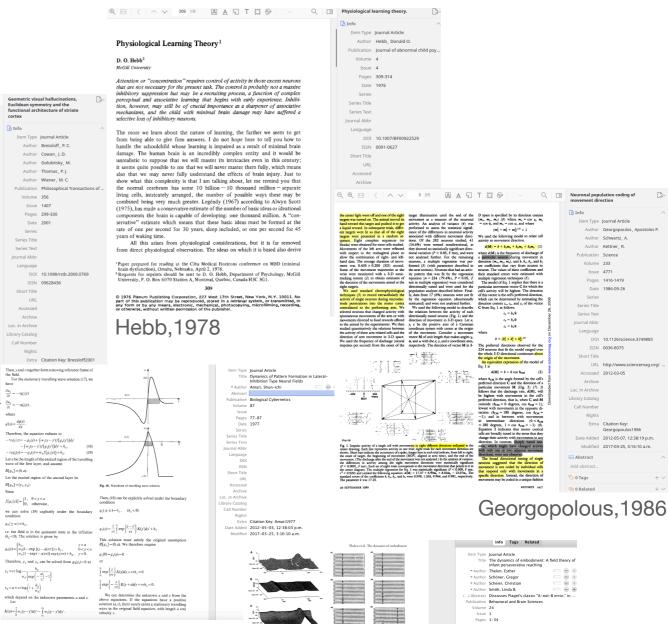
n visual perception. Our work on analogies 1.2.3 involves highly idealized situations epresented by strings of letters of the alphabet. An event in such a

Hofstadter, 1987





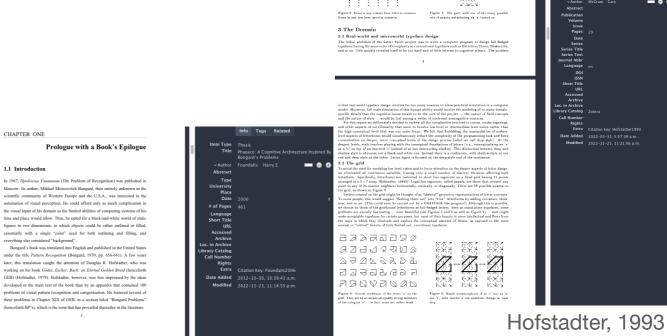
Bresloff,2001



Amari,1977



Thelen,2001



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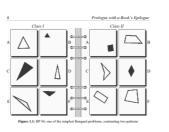
Q ⊕ ⟨ ∧ ∨ 636 1/5 A A T # 8 V

tinguated precedent. Freed (1850) engoised base in the file of the calculation gate entrally applied a factor (1851) engoised that "The nervous states concenitant (cerrelative) with applyincial states are, according to the destrine of evolution, sensori-motor. The highest centres of evolution, sensori-motor. The highest centres of the control of the visual image in the retire and supplied of the control of the visual image in the retire of the control of the visual image in the retire and supplied of the retired displacement. This implies an anticipatory adjustment the visual centre specific for each move-

Feinberg,1978

4 6 8 LENGTH (CM)

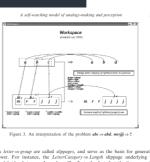
Coren, 1986

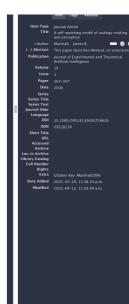


Foundalis, 2006

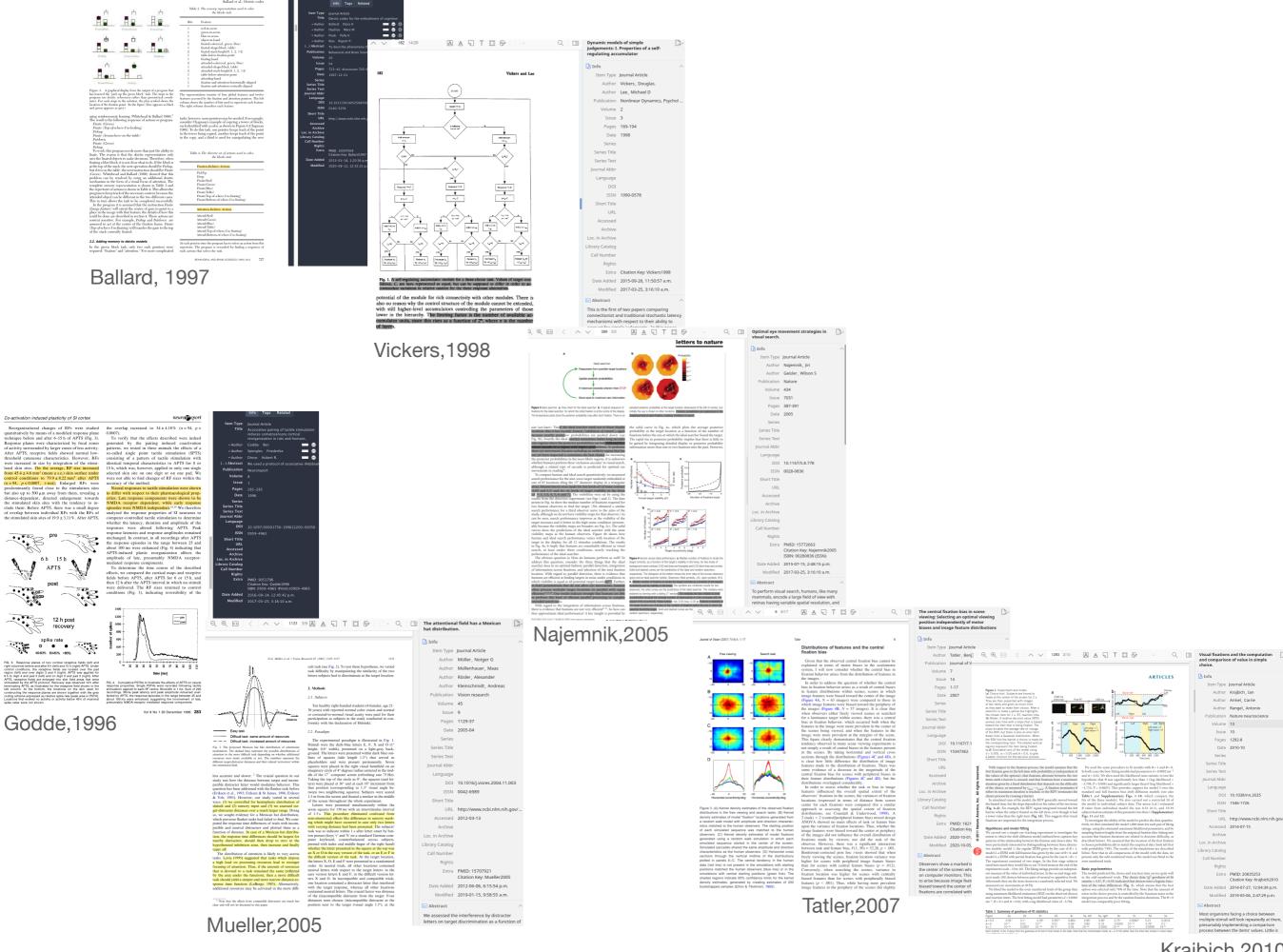


Spering,2013



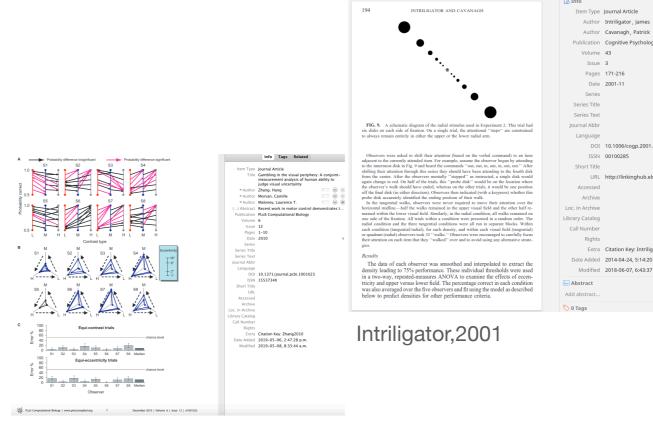


Marshall, 2006



## Using Dynamic Field Theory to Rethink Infant Habituation Eather Thelen Rethink University But Information of Pages Sectioner Reth University Most of the purphosphosphose when taken and experiment and experiment has destroyed to the property and the property

## Schöner, 2006

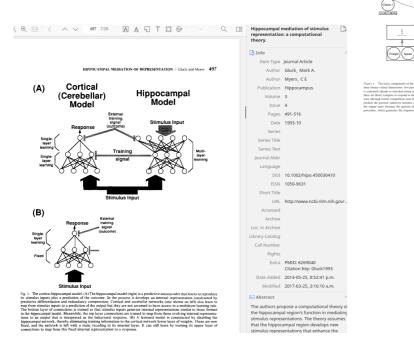


Q ⊕ → C 194 2446 A A T T B Spatial Resolution of Visual Attention

Zhang,2010

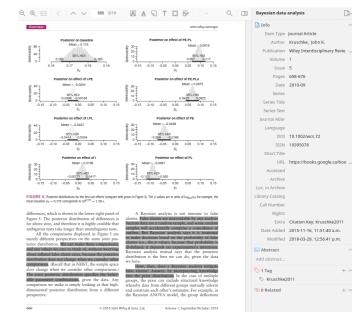


Shepard, Hovland, Jenkins, 1961

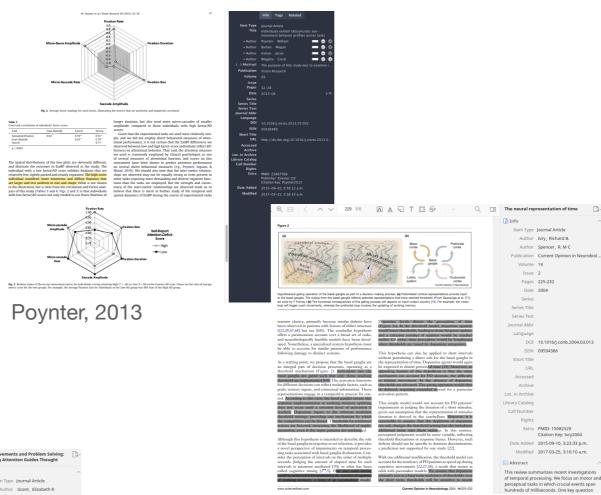


Love,2004

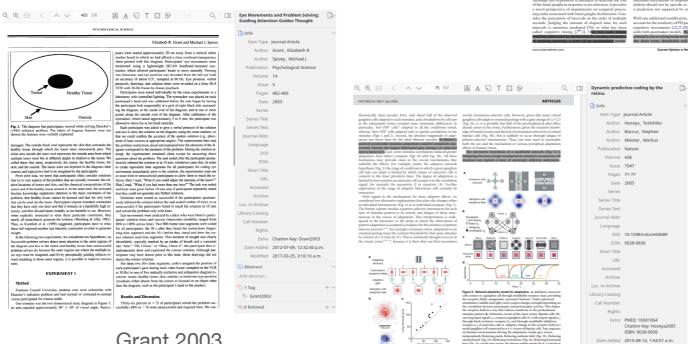
Gluck,1993



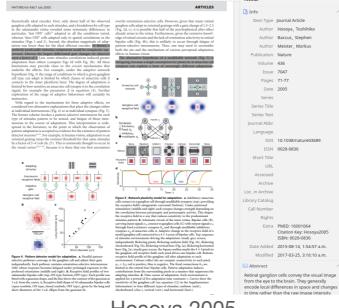
Kruschke,2010



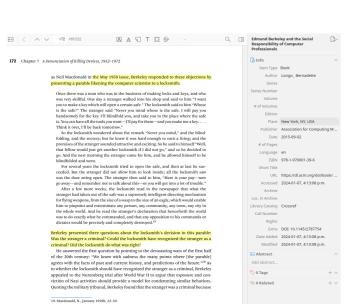
lvry,2004



Grant,2003



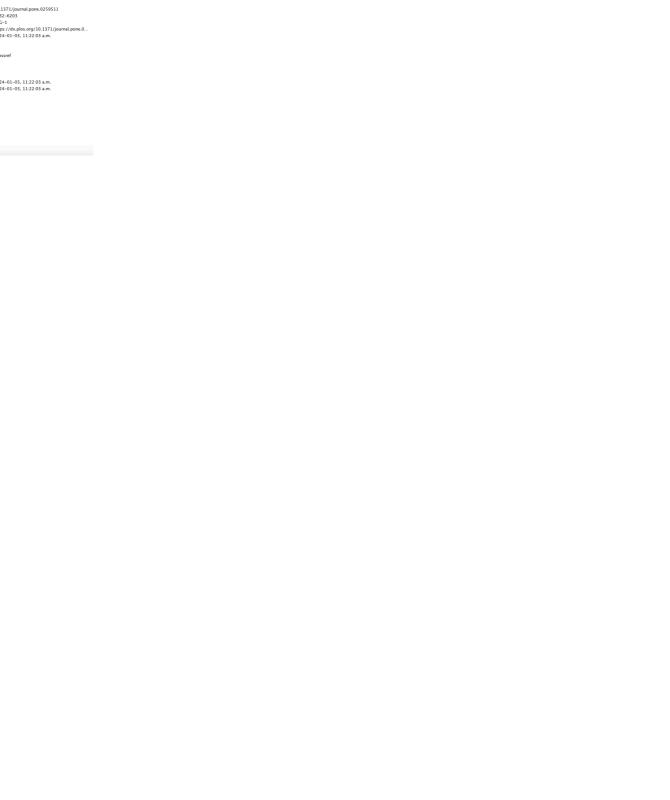
Hosoya,2005



Berkeley,1958

# The Respectability of Medicates in Many County Search County Sea

Chomsky, 1967

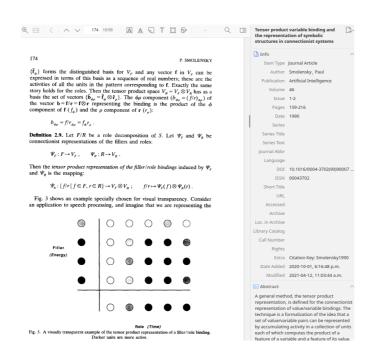


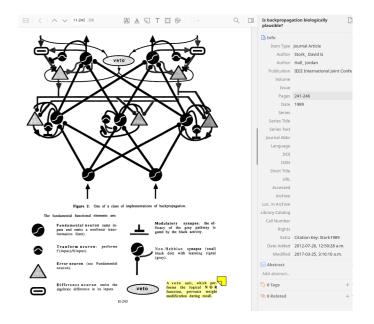
## Wilson & Nisbett, 19??

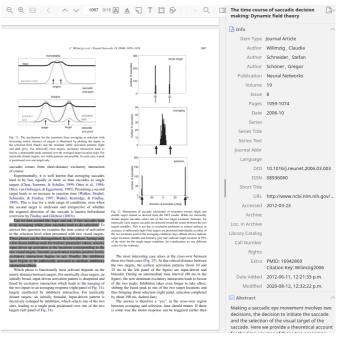
Stork, 1989

McClelland,2006

Craig,2015 Mitkus,2017







Wilimzig,2006



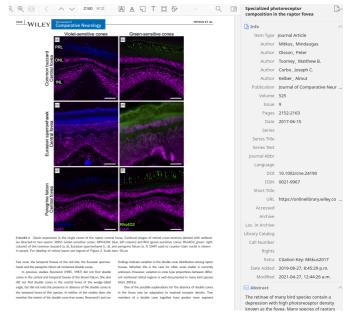
based on O'Reilly's LEABRA learning rule (O'Reilly, 1996; O'Reilly, this volume), is simply to combine error-correcting and Hebbian learning. In LEABRA, the signal that drives the connection weight combines the Hebb, learning rule of Equation 3 with an error-correcting term. If we were to apply this suggestion to capture the role of accuracy feedback in our /V-/r/ learning experiment, we would need to imagine that the this as the source of the correct target information required in standard error correcting learning, which then augments the Hebbian part of the learning when accuracy feedback is available.

The LEABRA approach is certainly worth exploring, but does introduce some processing complexity that has led me to consider other alternatives. To compute the error-correction component of the weight update, LEABRA uses a second pass through the activation settling process with the teaching input provided, after the first pass of activation in the absence of the teaching input. O'Reilly and I are currently at work on a successor to the LEABRA algorithm that attempts to eliminate the separate second pass. In the meantime, the two proposals considered below are perhaps mechanistically simpler than the existing version of LEABRA, and have thus been the focus of the modeling effort by Vallabha and McClelland (2004)

The first of the two ideas is to use the feedback signal to produce a reward signal I will call R(F). and then use this to modulate Hebbian learning:

To apply this idea to the results of the experiments reviewed above, in which we see evidence of learning without any feedback, we would require that R(F) have some positive value in the absence of any R(F) above its baseline value, and feedback indicating that the response is incorrect could reduce it below baseline, or potentially (as in many applications of reward-driven or reinforcement learning, c.f. Barto, 1992) reverse its sign. The second idea is to use the accuracy feedback signal to derive the identity of the correct response, and use this to adjust the activation of the output unit before applying Hebb, rule of





Long-Term Effects of Stimulant Treatment for ADHD: What Can We Tell Our Patients?

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## My scholarly papers, p1

PLOS ONE LAG-1: A model of learning, attention, and gaze

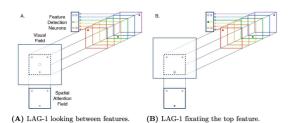


Fig 2. Information processing schema. Schematic of the relationships between experiment input, the Feature Detection Neurons. Visual Field, and the Spatial Altention Field. A) The forea is indicated by the dashed grey circle in the center of the Visual Field. B) After an eye movement to the bottom middle feature, the green sensitive Feature Detector is activated and the Spatial Altention Field is boosted at its associated location. The Visual Field is coding spatial information in retinotopic coordinates while the Spatial Attention Field is coding information in apaintopic coordinates.

https://doi.org/10.1371/journal.pone.0259511.g002

The Spatial Attention Field in LAG-1 represents changes in attentional priority to locations in a spatiotopic reference frame. The different frames of reference of the Spatial Attention Field and the Visual Field are depicted in Fig 2. Notice that there is no change in the location of features on the Spatial Attention Field between Fig 2A and 2B, despite the shifting of the bottom feature onto the fovea of the Visual Field.

Changes in activation on the Spatial Attention Field are described by similar equations as those used for the Visual Field, in Equation 13.

Saccade Motor Field. Candidate saccade target locations compete on the Saccade Motor Field [48, 49]. Unlike the Spatial Attention Field which binds a retinue of competing attentional priorities with spatial locations, the Saccade Motor Field resolves competition for attention at locations other than the current locus of fixation. The dynamics of this field resolves the competition between locations to be the target of the next eye movement according to Equation 17.

Saccade Timing System overview. Under normal viewing conditions, humans make about three saccades per second [77]. The period of relative spatial stability between saccades is the fixation duration. The parameterizations used in the present simulations were chosen for their rough correspondence with the normal range of saccade and fixation durations. In the model, as in the brain, the timing of an eye movement is affected by previous experience. In people, this includes factors like expected processing difficulty, but category learning experiments are designed to minimize such effects, so changes in fixation duration exhibited by LAG-1 are primarily the result of learning the relevance of different features. We have also observed fixation duration differences in LAG-1 when using it for visual search where disorganized inputs yield spatial interactions that speed or slow its saccade onset latency [40, 46, 58, 67, 78].

As the name suggests, the Saccade Timing System is the primary arbiter of decisions to release fixation and foveate a new location. The trio of neurons controlling this system are: the Gaze Change Neuron, the Fixation Neuron, and the Saccade Initiation Neuron. These model neurons have a functional correspondence with brain stem neurons referred to as: \*build-up or the state of the state o



Barnes, 2022

