

The Fourth Conference on Artificial General Intelligence

AGI Research Progress in Intelligence Science Lab at Chinese Academy of Sciences

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Acknowledgement

- **Key Projects of National Natural Science Foundation of China: No.61035003, 60933004**
- **National Natural Science Foundation of China: No. 61072085, 60903141, 60970088**
- **China National Basic Research Priorities Programme 973: No. 2007CB311004**

Contents Outline



Introduction

Mind Model CAM

Feature Binding

Image Understanding

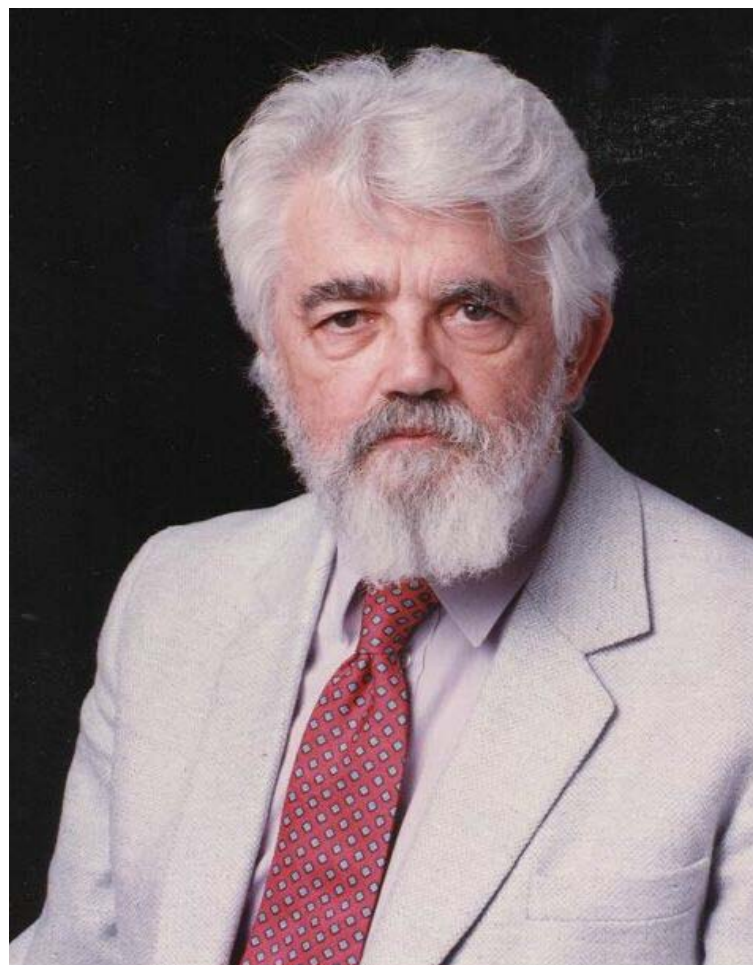
World Wide Brain

Perspectives

Introduction

- McCarthy declared the long-term goal of Artificial Intelligence is human level Artificial Intelligence
- It is not surprising that human-level AI has proved difficult and progress has been slow

John McCarthy. From here to human-level AI. Artificial Intelligence 171 (2007) 1174–



Artificial General Intelligence (AGI)

“The ability to achieve complex goals in complex environments using limited computational resources”

- Autonomy
- Practical understanding of self and others
- Solving problems qualitatively different from those anticipated by the programmers



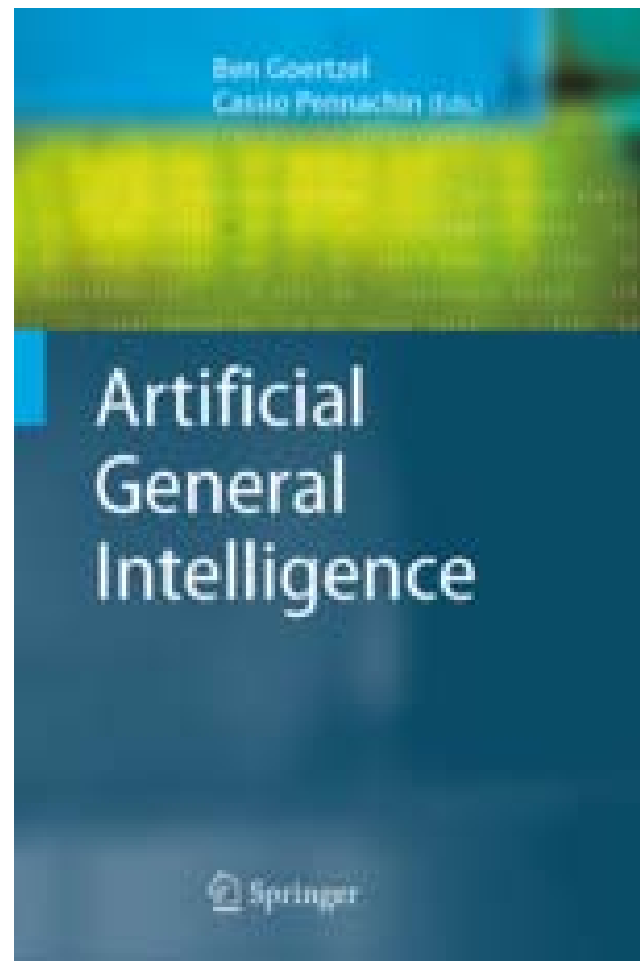
---Ben Goertzel

Artificial General Intelligence (AGI)

Book

Editor» Goertzel, Ben (et al.)
Artificial General Intelligence.
Springer, 2007

This is the first book on current research on artificial general intelligence (AGI), work explicitly focused on engineering general intelligence – autonomous, self-reflective, self-improving, commonsensical intelligence.



Artificial General Intelligence (AGI)

- AGI-2010: Lugano, Switzerland, March 5-8 (In Memoriam Ray Solomonoff)
- AGI-2009: Arlington, Virginia, March 6-9
- AGI-2008: University of Memphis, March 1-3



Artificial General Intelligence (AGI)



A Roadmap Toward Human Level Artificial General Intelligence

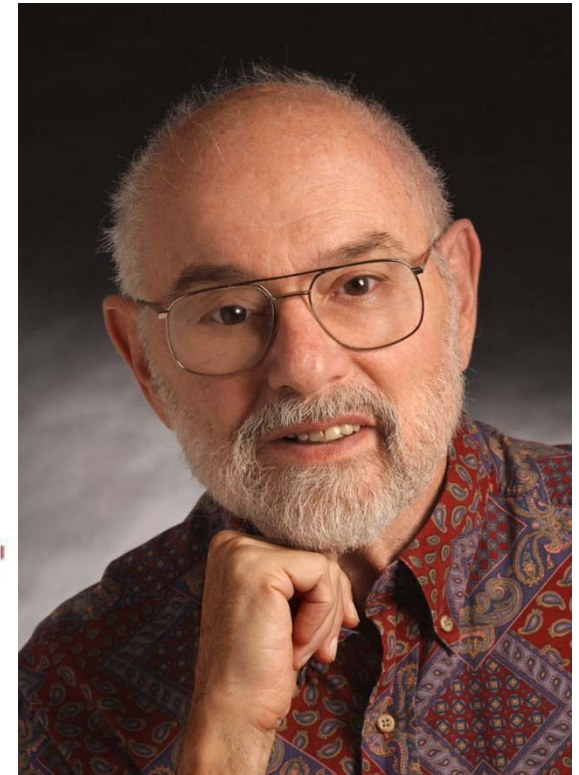
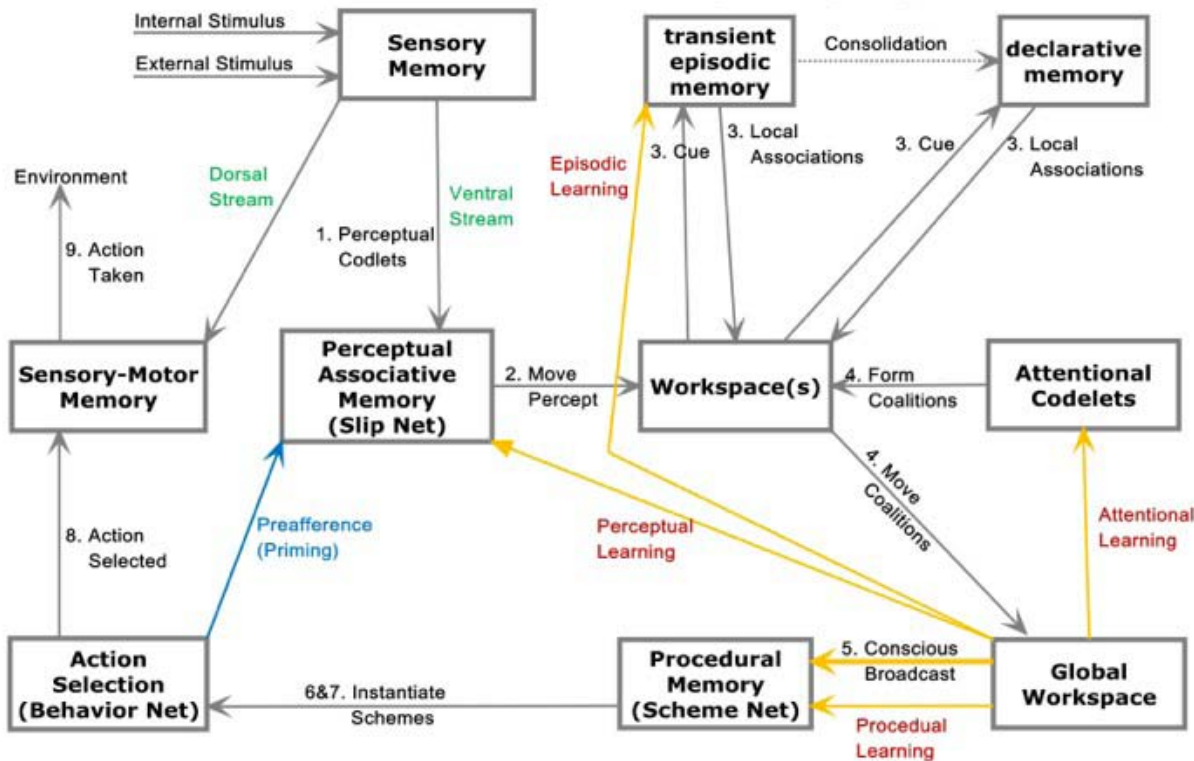
Sam Adams, Itamar Arel, Joscha Bach, Robert Coop, Rod Furlan,
Ben Goertzel, J. Storrs Hall, Alexei Samsonovich, Matthias Scheutz,
Matt Schlesinger, Stuart Shapiro, John Sowa

June 25, 2010

Abstract

We present the broad outlines of a roadmap toward Human Level artificial general intelligence (AGI). The roadmap is not targeted at any specific design or architecture, but is rather intended to encourage the development of multiple AGI systems along the same pathways, toward the same end goals and passing through the same set of milestones. A number of specific “scenarios” for the development of Human Level AGI are considered (including virtual-world, robotic and purely text/image-based “school” scenarios at various levels, along with others involving real-world robotics and video-game learning), along with a set of competency areas that may be explored in multiple scenarios and that together constitute a large fraction of Human Level intelligent functionality. It is proposed to develop a series of concrete tasks within each of the described scenarios, implementing the described competency areas. Such task series would then serve as a concrete, detailed roadmap to Human Level AGI for researchers accepting the given scenario as a useful one.

Framework of LIDA

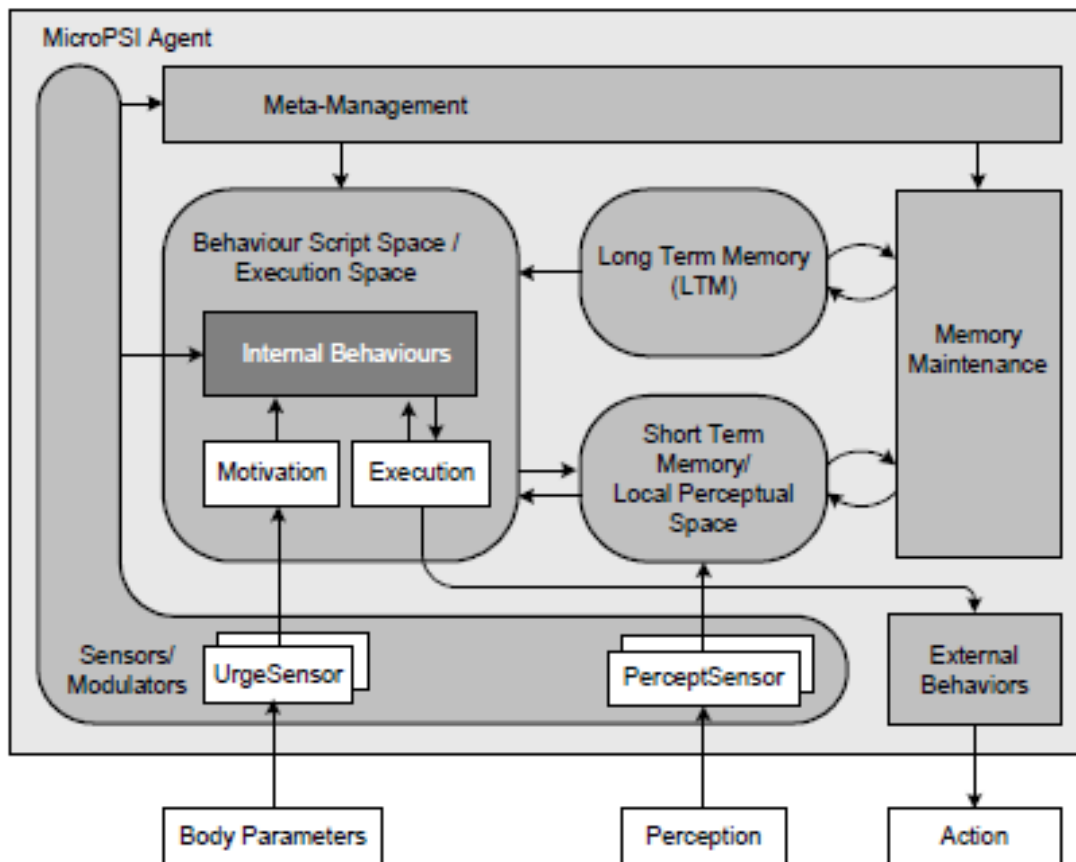


Javier Snider, Ryan Mccall and Stan Franklin. The LIDA Framework as a General Tool for AGI. AGI2011

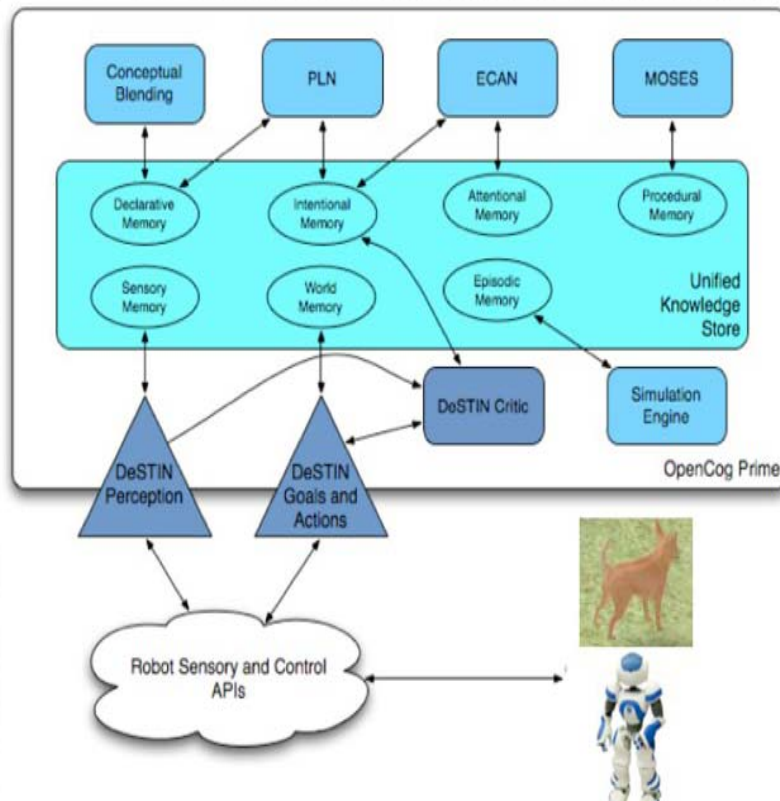
MicroPsi Architecture



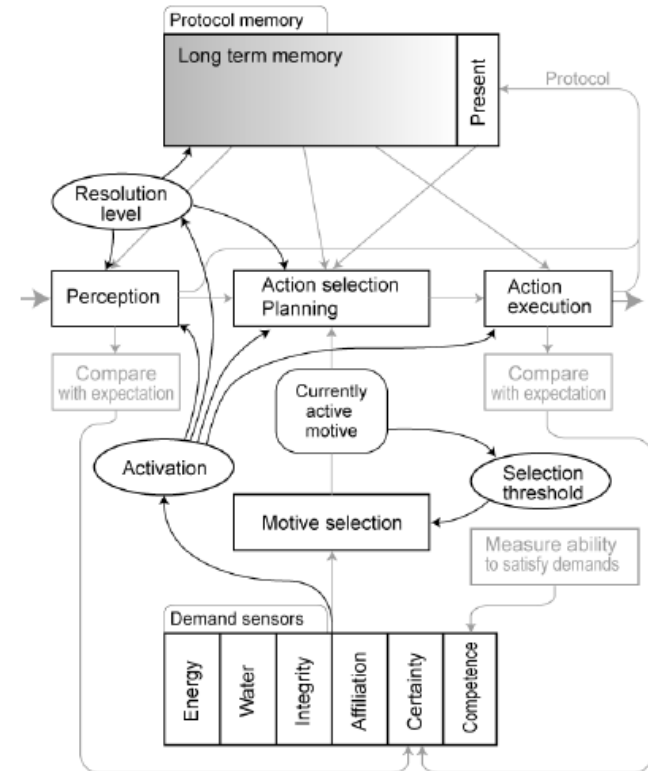
Dr. Joscha Bach



OpenPsi



OpenCog



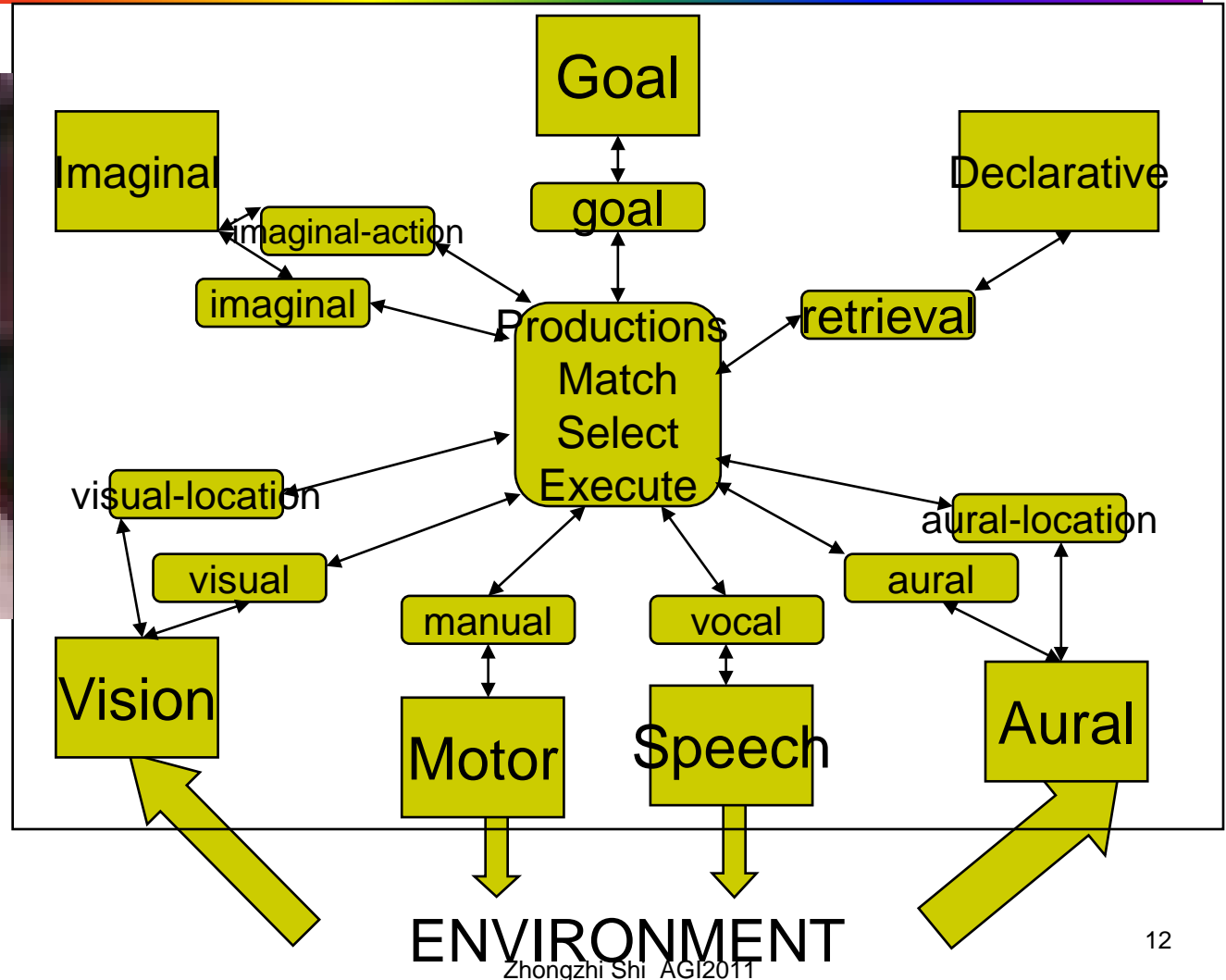
OpenPsi

Dr. Ben Goertzel et al.

ACT-R 6.0



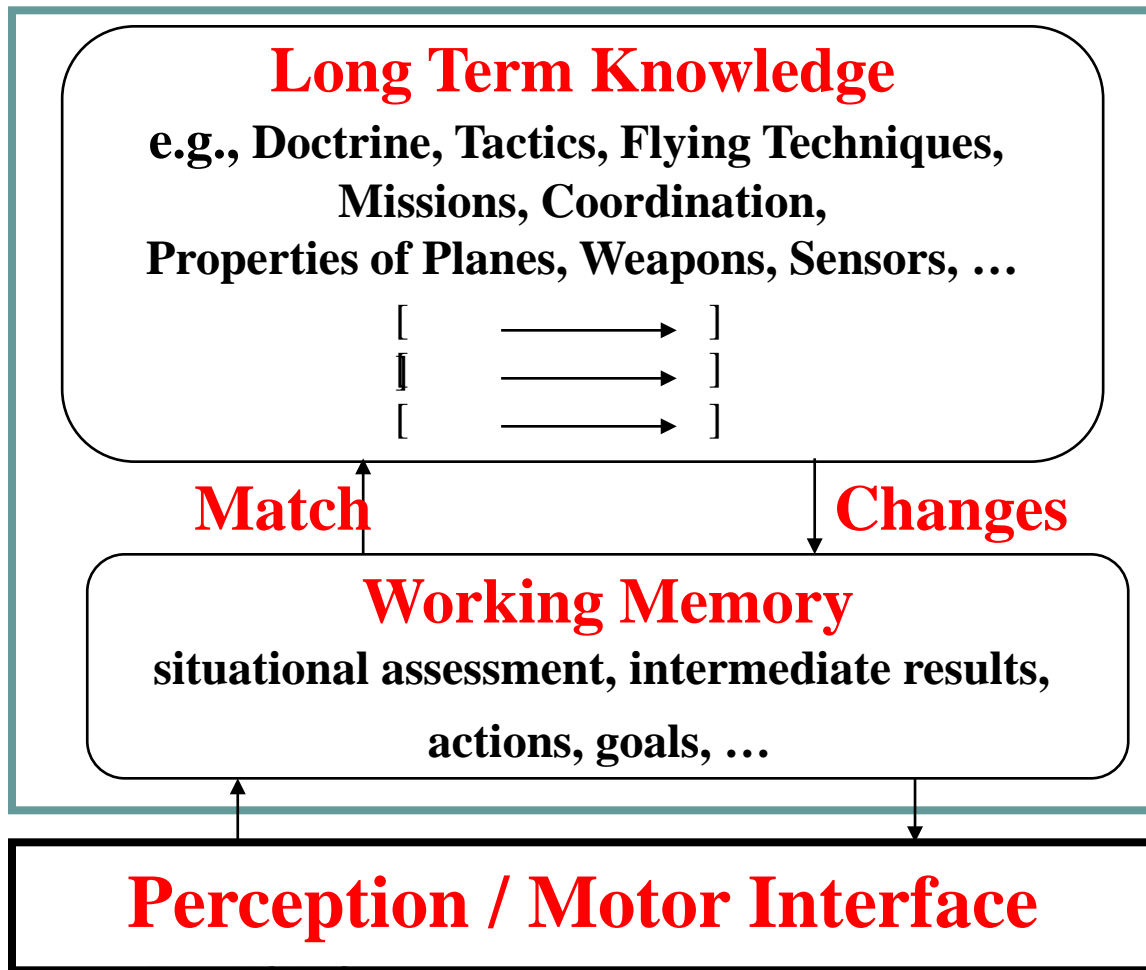
J.A. Anderson



Soar Architecture



Allen Newell

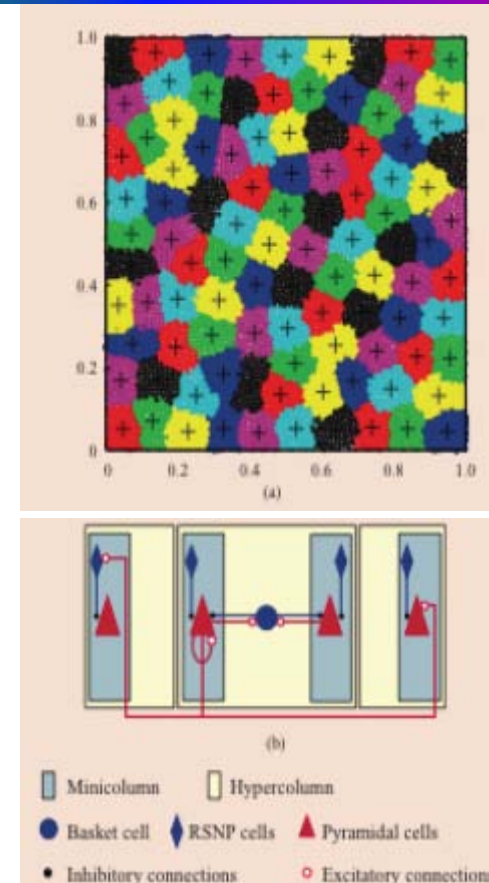


IBM Blue Brain Project

IBM Blue Brain Project(launched in 2005)

- IBM “cat brain scale” spiking neural net simulation
- 147,000 processors
- 144 terabytes of memory

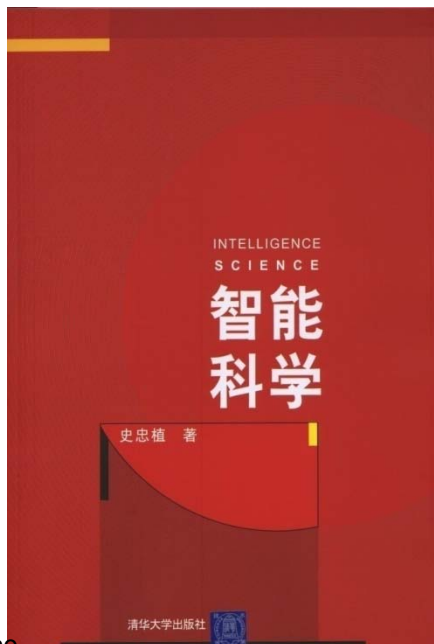
Dharmendra Modha



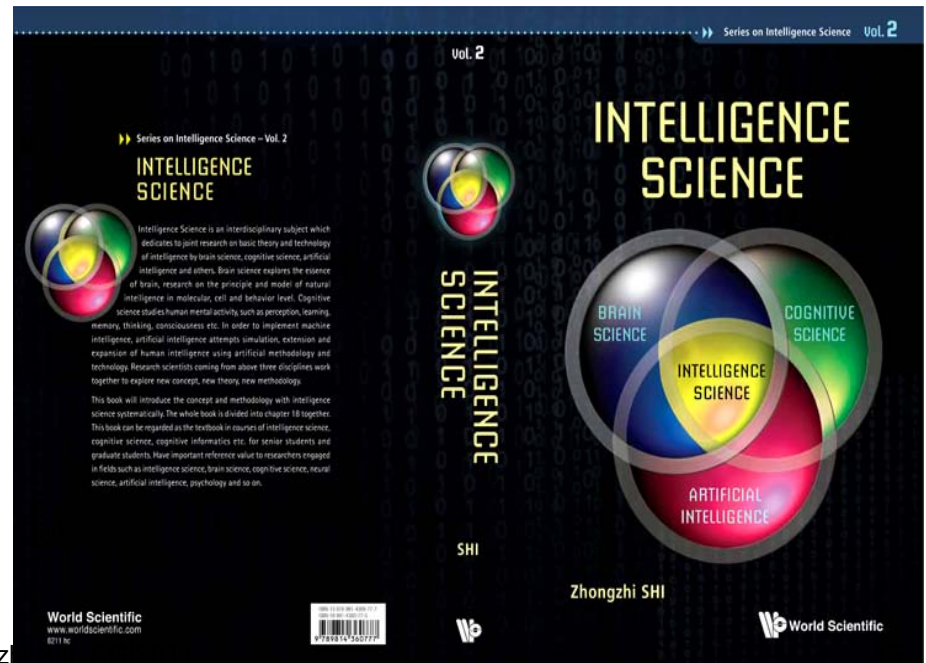
IBM Simulation

Intelligence Science

- An interdisciplinary subject that dedicates to joint research on basic theory and technology of intelligence by brain science, cognitive science, artificial intelligence and others.



2011-8-29



Zhongz

Series on Intelligence Science

World Scientific Publishing
will publish Series on
Intelligence Science. Prof.
Zhongzhi Shi is the Editor-in-
Chief.

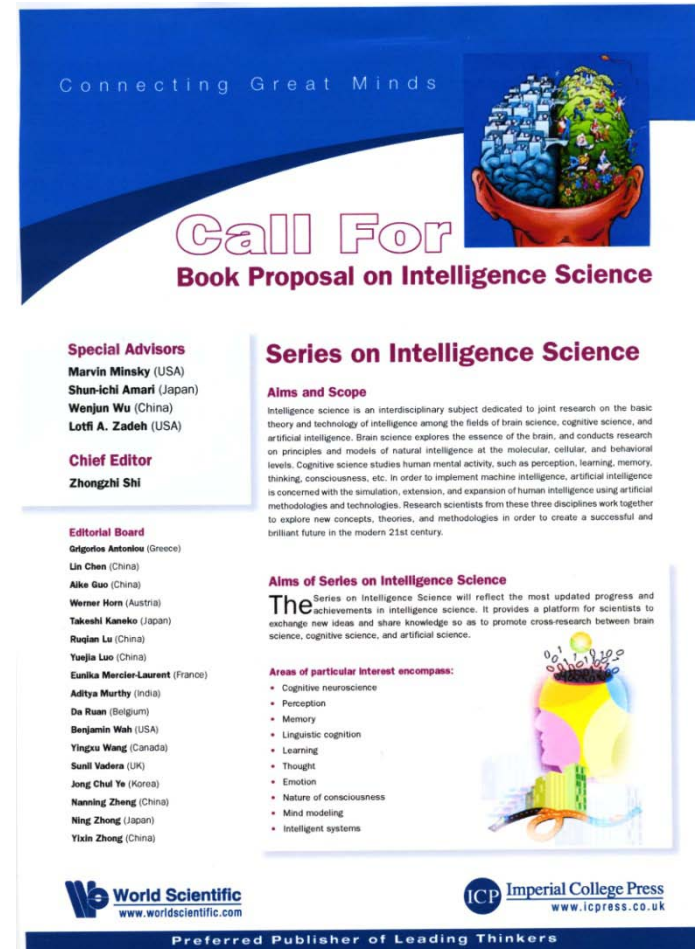
Special Advisors:

Marvin Minsky (USA)

Shun-ichi Amari (Japan)

Wenjun Wu (China)

Lotfi A. Zadeh (USA)



Connecting Great Minds

Call For
Book Proposal on Intelligence Science

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Marvin Minsky (USA)
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Zhongzhi Shi

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Series on Intelligence Science

Aims and Scope
Intelligence science is an interdisciplinary subject dedicated to joint research on the basic theory and technology of intelligence among the fields of brain science, cognitive science, and artificial intelligence. Brain science explores the essence of the brain, and conducts research on principles and models of natural intelligence at the molecular, cellular, and behavioral levels. Cognitive science studies human mental activity, such as perception, learning, memory, thinking, consciousness, etc. In order to implement machine intelligence, artificial intelligence is concerned with the simulation, extension, and expansion of human intelligence using artificial methodologies and technologies. Research scientists from these three disciplines work together to explore new concepts, theories, and methodologies in order to create a successful and brilliant future in the modern 21st century.

Aims of Series on Intelligence Science
The Series on Intelligence Science will reflect the most updated progress and achievements in intelligence science. It provides a platform for scientists to exchange new ideas and share knowledge so as to promote cross-research between brain science, cognitive science, and artificial science.

Areas of particular interest encompass:

- Cognitive neuroscience
- Perception
- Memory
- Linguistic cognition
- Learning
- Thought
- Emotion
- Nature of consciousness
- Mind modeling
- Intelligent systems

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www.icpress.co.uk

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Basic Issues of Intelligence Science

- How Do Brain Neural Circuits Work?
- What is Perceptual Representation and Theory of Perception?
- How Are Memories Stored and Retrieved?
- What is the Neural Basis of Language?
- How Does the Brain Learn?
- How to Think in Human Brain?
- What is the Procedure of Intelligence Development?
- What is the Nature of Emotion?
- What is the Nature of Consciousness?
- How to Build Mind Model?

Zhongzhi Shi. Foundations of Intelligence Science. International Journal of Intelligence Science, PP.8-16, 2011

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Feature Binding

Understanding

World Wide BrainImage

Perspectives

Mind Model CAM

CAM-**C**onsciousness and **M**emory Model is

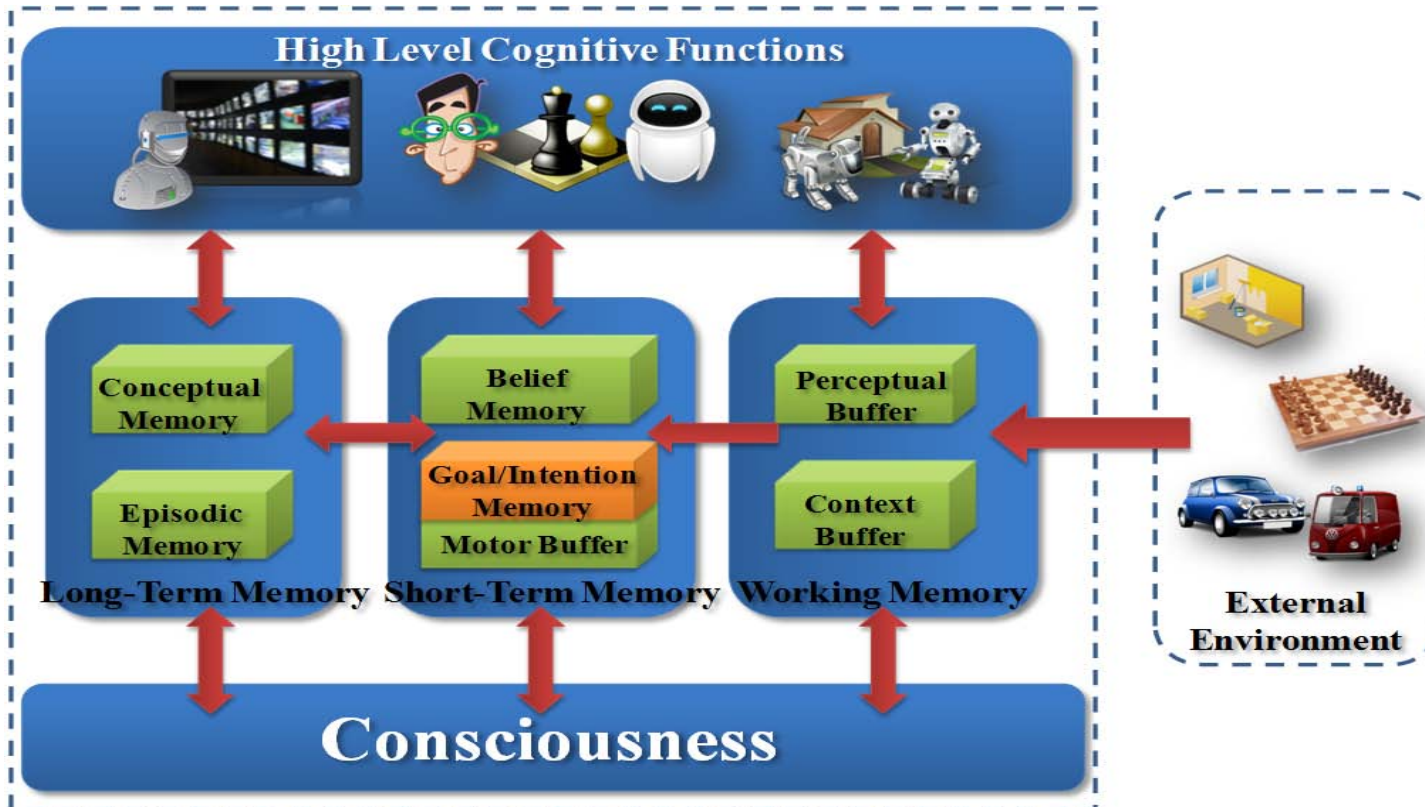
A new mind modeling for human cognitive activities

Characteristics of CAM

- Memory based Mind Model;
- Consciousness-centered Cognitive Activities;
- Action planning based way to arrange different cognitive function modules to perform cognitive activities;

CAM Architecture

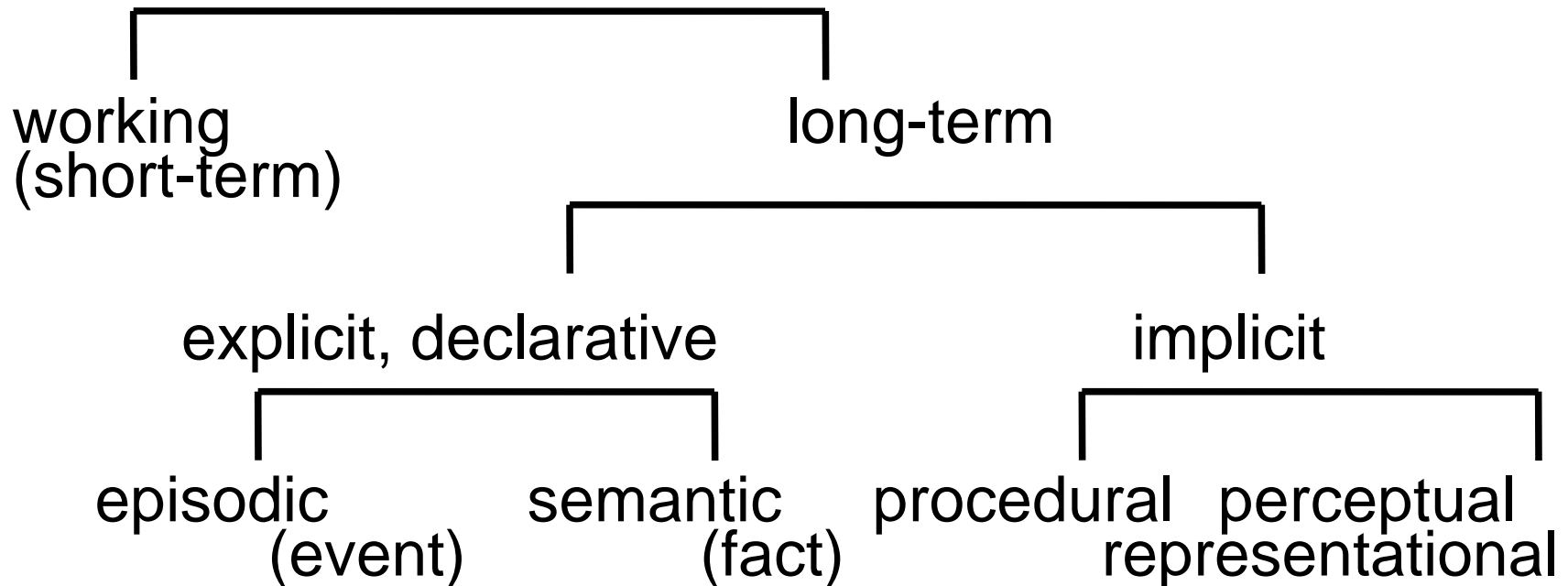
**CAM-Consciousness and Memory Model is
A new mind modeling for human cognitive activities**



Long-Term Memory

Tulving, Endel (1972). Episodic and semantic memory. Organization of memory. In E. Tulving and W. Donaldson. New York, Academic Press.

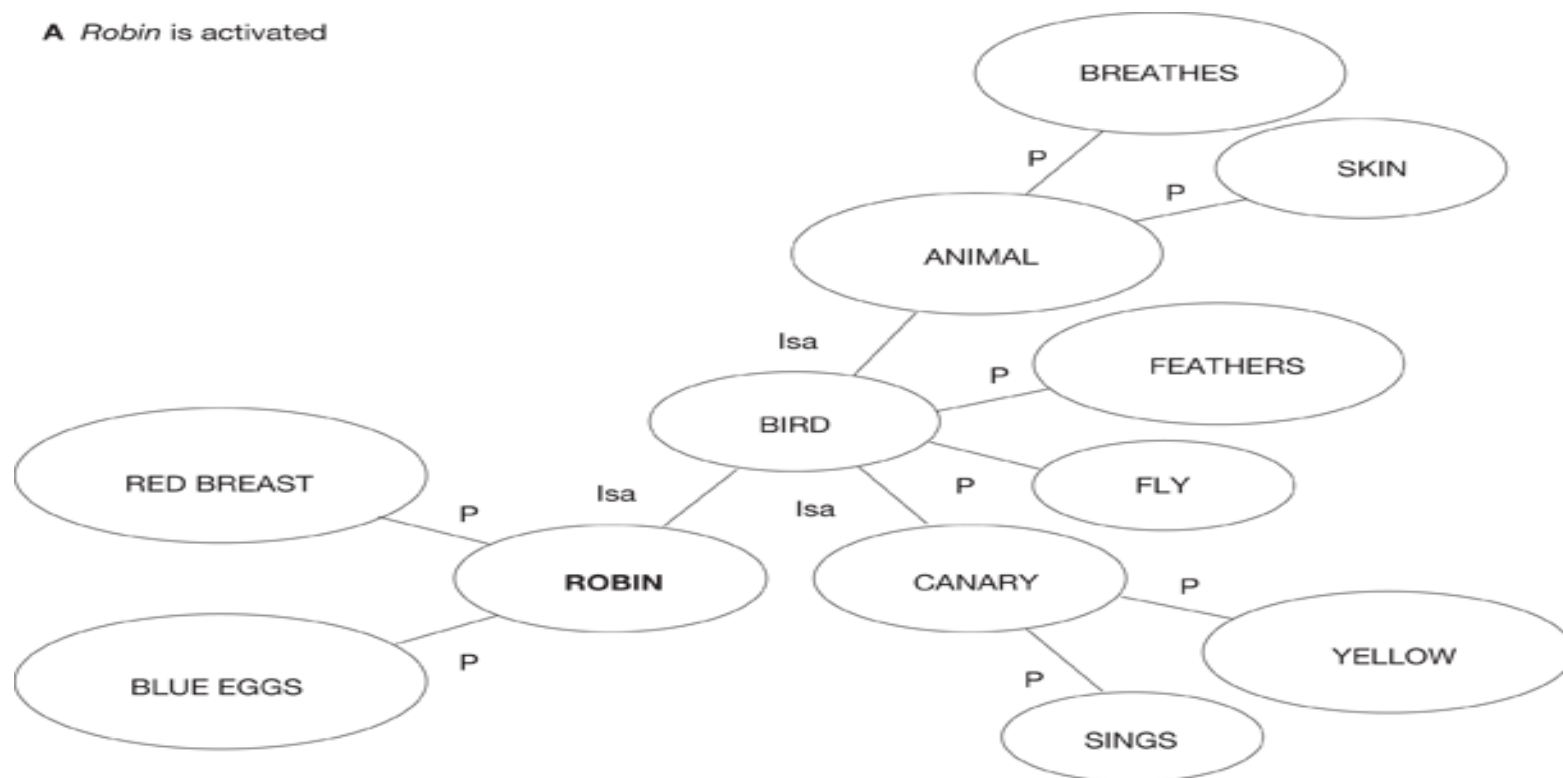
Types of memory



Semantic Memory

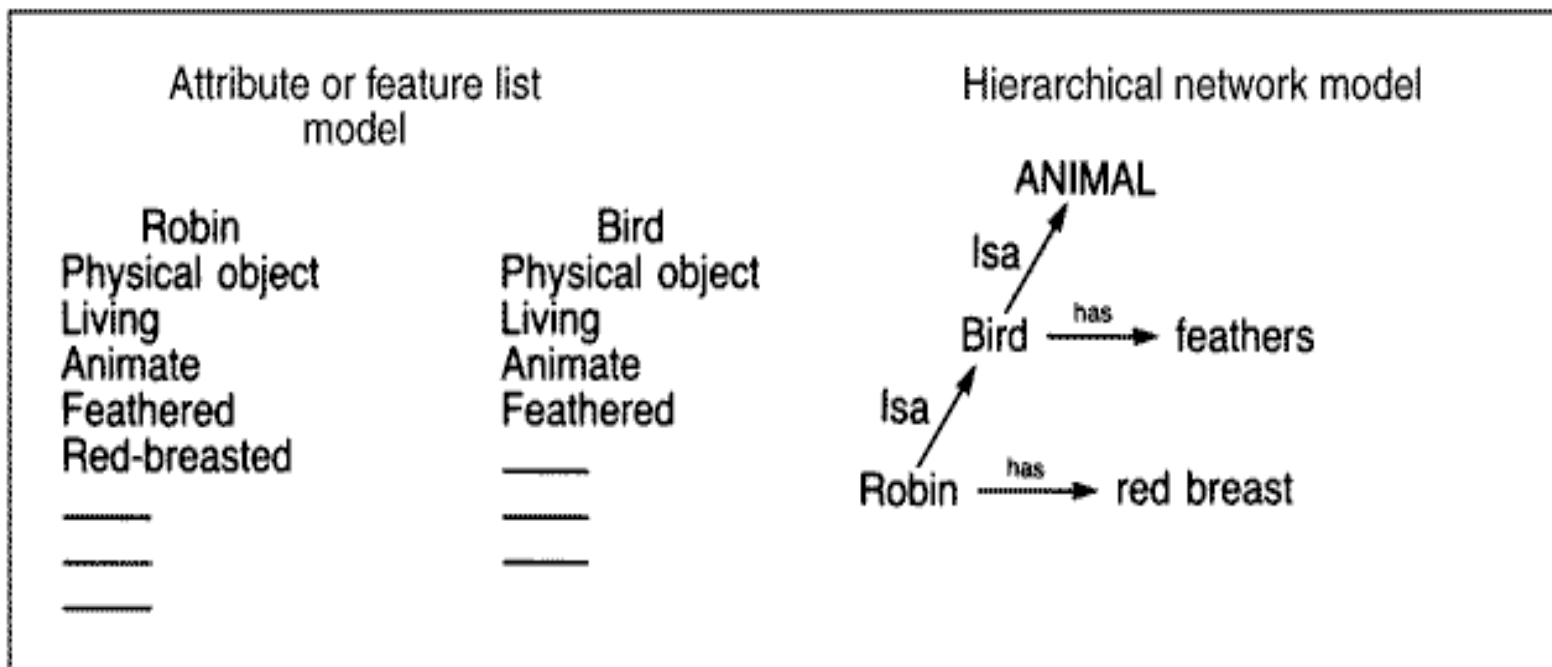
Collin and Quillian Model

A Robin is activated



Semantic Memory

Smith's feature overlap model



Dynamic Description Logic



In order to represent agent better, we add dynamic action to DL and propose Dynamic Description Logic

$$A(x_1, \dots, x_n) = (P_A, E_A)$$

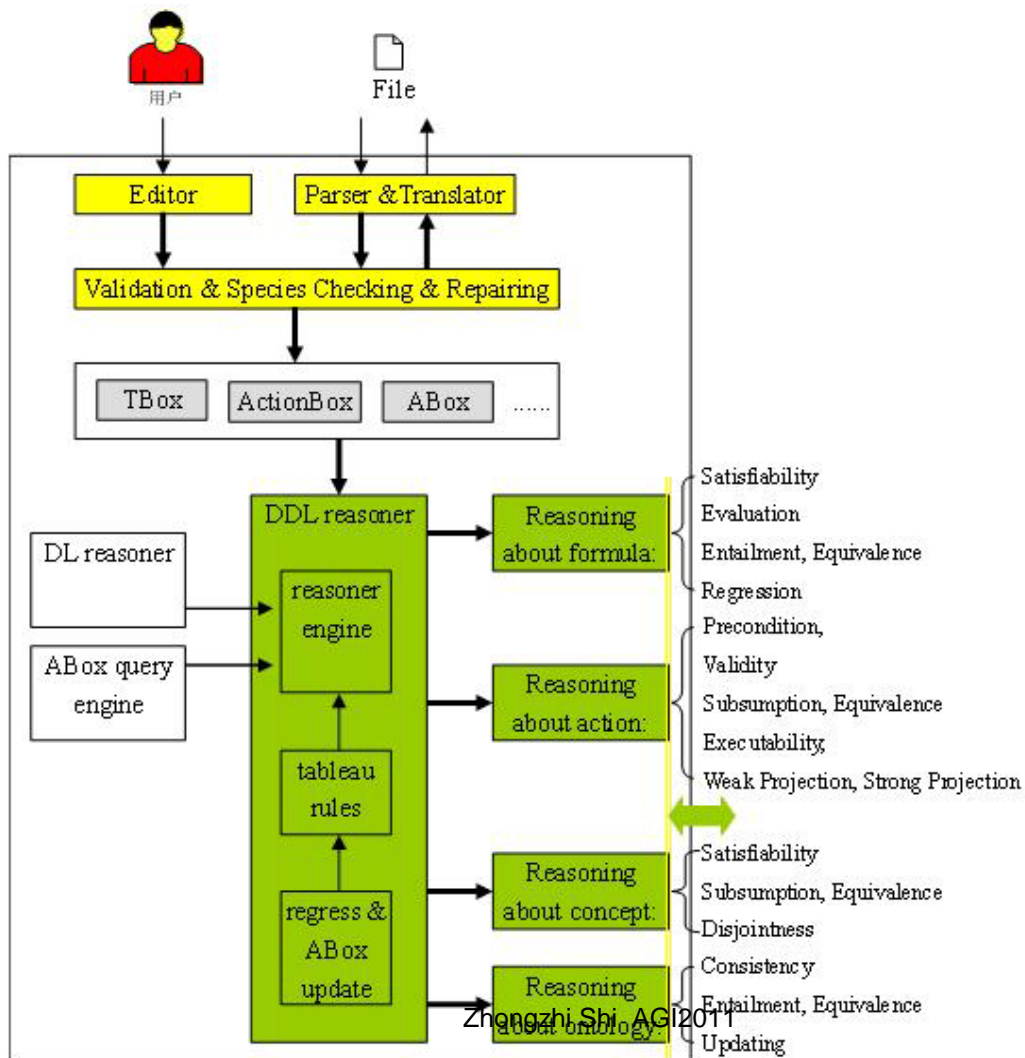
where

- (1) A is the action name.
- (2) x_1, \dots, x_n are individual variables, which denote the objects the action operate on.
- (3) P_A is the set of preconditions, which must be satisfied before the action is executed.
- (4) E_A is the set of results, which denote the effects of the action.

Dynamic Description Logic

- **Concept name:** $C_1, C_2, \dots;$
- **Role name:** $R_1, R_2, \dots;$
- **Individual constant:** $a, b, c, \dots;$
- **Individual variable:** $x, y, z, \dots;$
- **Concept operation:** $\neg, \sqcap, \sqcup, \exists, \forall;$
- **Axiom operation:** $\neg, \wedge, \rightarrow \forall;$
- **Action:** $A_1, A_2, \dots;$
- **Action-constraint** : ; (composition) , U
(alternation), * (repeat), ? (test);
- **Action variable:** $\alpha, \beta, \dots;$
- **Axiom variable:** $\varphi, \psi, \square, \dots;$
- **State variable:** $u, v, w, \dots;$

DDL Inference Engine



Episodic Memory

- Episodic Memory

The memory about past experience; contextualized store of specific events.

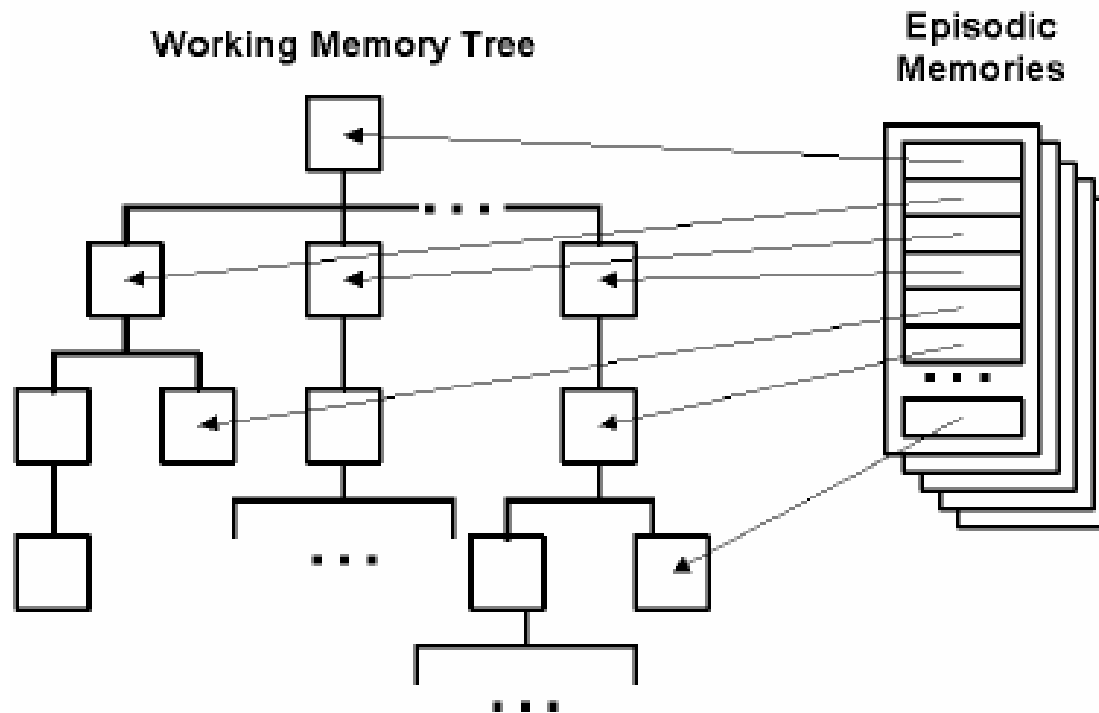
Example: a movie seen yesterday

- Element in Episodic Memory

episode: the stored past event;

episode in CAM is in form of possible world sequence.

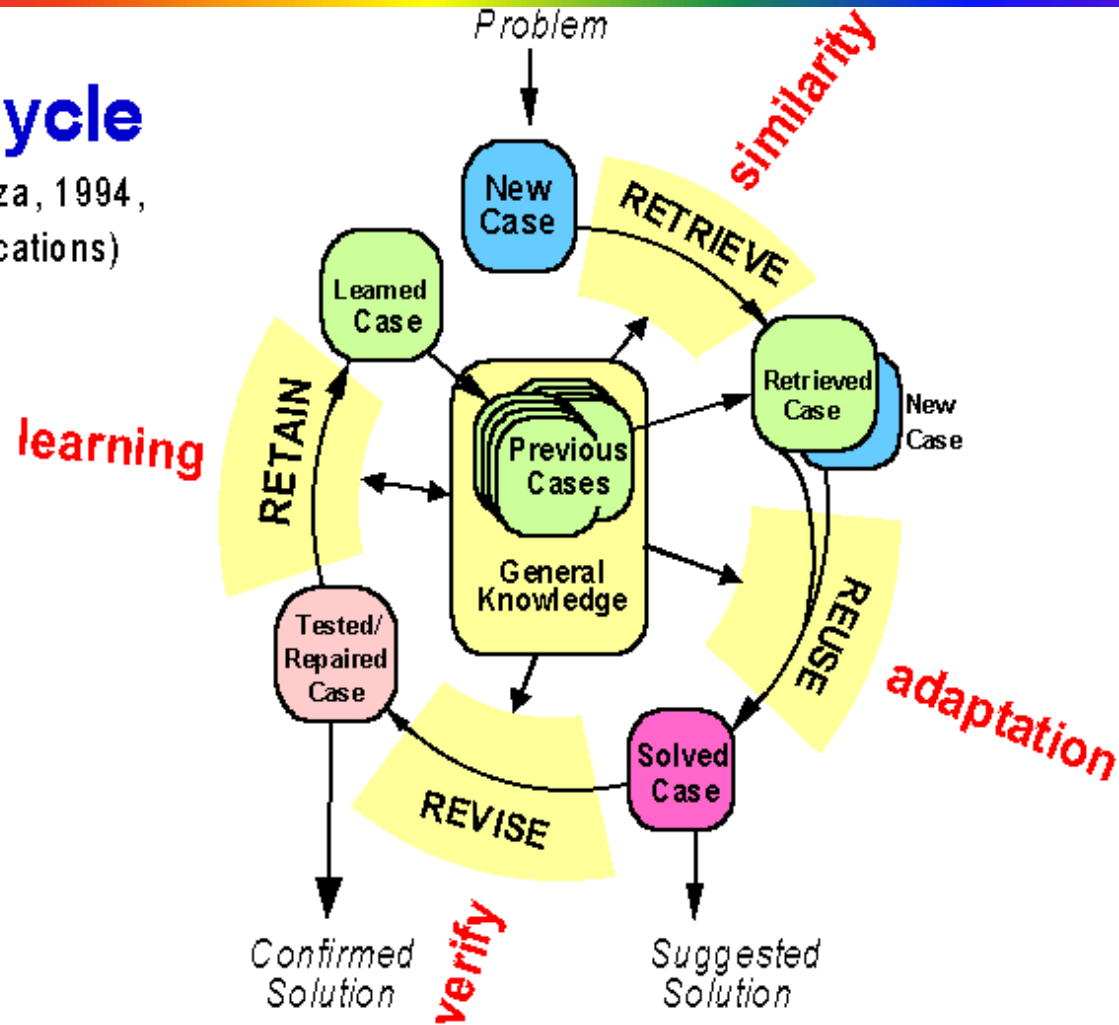
Episodic Memory



Episodic Memory

CBR Cycle

(Aamodt & Plaza, 1994,
AI Communications)



Episodic Memory in CAM

- Episode Retrieval

- ☞ Retrieve episode by cue

- Cue is also in form of possible world sequence;

- Retrieve episode by matching the possible world sequence representing cue with the possible world sequence representing episode.

- ☞ Retrieve episode likes Case-based reasoning

Episodic Memory in CAM

- Matching Cue with Episode

$MatchPossibleWorld(w_1, w_2)$ is true iff $w_1 \models w_2$

$MatchAction(a_1, a_2)$ is true iff $P_{a_1} \models P_{a_2}, E_{a_1} \models E_{a_2}$

episode = $\{w_1, a_1, w_2, \dots, w_n\}$, cue = $\{c_1, b_1, c_2, \dots, c_k\}$,

$CueMatching(episode, cue)$ is true iff

exists sub sequence $\{w_{i+1}, a_{i+1}, \dots, w_{i+k}\} \subseteq$ episode satisfying

$MatchPossibleWorld(w_{i+j}, c_j)$ is true ($j = 1, 2, \dots, k$)

$MatchAction(a_{i+l}, b_l)$ is true ($l = 1, 2, \dots, k - 1$)

Cue Matching Algorithm

Algorithm 1: CueMatching(e, c)

```

input  : Episode  $e$ , Cue  $c$ 
output: whether  $c \preceq_p e$  hold
1  if  $length(e) < length(c)$  then
2  |   return false;
3  end
4   $n_e := first\_node(e)$  ;
5   $n_c := first\_node(c)$  ;
6  if  $MatchPossibleWorld(n_e, n_c)$  then
7  |    $\alpha_e := Null$  ;
8  |    $\alpha_c := action(n_c)$  ;
9  |   if  $\neg(Pre(\alpha_e) \rightarrow Pre(\alpha_c))$  unsatisfiable according DDL tableau
10 |   |   algorithm then
11 |   |   |    $n_{tmp} := n_e$  ;
12 |   |   |   while  $next\_node(n_{tmp}) \neq Null$  do
13 |   |   |   |    $\alpha_e := (\alpha_e; action(n_{tmp}))$  ;
14 |   |   |   |   if  $MatchAction(\alpha_e, \alpha_c)$  then
15 |   |   |   |   |   Let  $sub_e$  be the sub sequence by removing  $\alpha_e$  from  $e$  ;
16 |   |   |   |   |   Let  $sub_c$  be the sub sequence by removing  $\alpha_c$  from  $c$  ;
17 |   |   |   |   |   if  $CueMatching(sub_e, sub_c)$  then
18 |   |   |   |   |   |   return true;
19 |   |   |   |   |   end
20 |   |   |   |   end
21 |   |   |   |    $n_{tmp} := next\_node(n_{tmp})$  ;
22 |   |   |   end
23 |   end
24 |   end
25 |   end
26 end
27 Remove  $n_e$  from  $e$  ;
28 return  $CueMatching(e, c)$  ;

```



Match Possible World & Action

Function MatchPossibleWorld(w_i, w_j)

input : possible worlds w_i, w_j
output: whether $w_i \models w_j$ hold

```
1  $f_w := \text{Conj}(w_i) \rightarrow \text{Conj}(w_j)$  ;  
2 if  $\neg f_w$  is unsatisfiable according to DDL tableau algorithm then  
3 |   return true ;  
4 else  
5 |   return false ;  
6 end
```

Function MatchAction(α_i, α_j)

input : action α_i, α_j
output: whether $\alpha_i \models \alpha_j$ hold

```
1 if  $\alpha_i == \text{null}$  or  $\alpha_j == \text{null}$  then  
2 |   return false  
3 end  
4  $f_{pre} := \text{Conj}(\text{Pre}(\alpha_i)) \rightarrow \text{Conj}(\text{Pre}(\alpha_j))$  ;  
5  $f_{eff} := \text{Conj}(\text{Eff}(\alpha_i)) \rightarrow \text{Conj}(\text{Eff}(\alpha_j))$  ;  
6 if  $\neg f_{pre}$  and  $\neg f_{eff}$  are unsatisfiable according to DDL Algorithm then  
7 |   return true ;  
8 else  
9 |   return false ;  
10 end
```

What is Consciousness

Wikipedia:

Consciousness is a term that has been used to refer to a variety of aspects of the relationship between the mind and the world with which it interacts. It has been defined, at one time or another, as: subjective experience; awareness; the ability to experience feelings; wakefulness; having a sense of selfhood; or as the executive control system of the mind.

What is Consciousness

Stanford Encyclopedia Of Philosophy

- <http://www.informationisbeautiful.net/play/what-is-consciousness/>



A field that exists in its own parallel "realm" of existence outside reality so can't be seen (Substantive Dualism)



Consciousness and its states (belief, desire, pain) are simply functions the brain performs (Functionalism)



Is physical property of matter, like electromagnetism, just not one we know about (Property Dualism)



All matter has a psychic part. Consciousness is just the psychic part of our brain (Pan Physicalism)

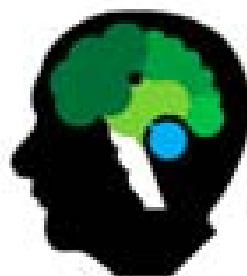
What is Consciousness



Simply, mental states are physical events that we can see in brain scans
(Identity Theory)



A sensation that "grows" inevitably out of complicated brain states
(Emergent Dualism)



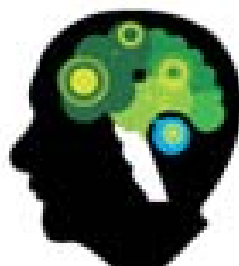
Literally just behaviour. When we behave in a certain way, we appear conscious
(Behaviourism)



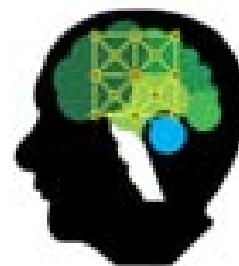
An accidental side-effect of complex physical processes in the brain
(Epiphenomenalism)



Not sure. But quantum physics, over classical physics, can better explain it
(Quantum Consciousness)



The sensation of your most significant thoughts being highlighted
(Cognitivism)

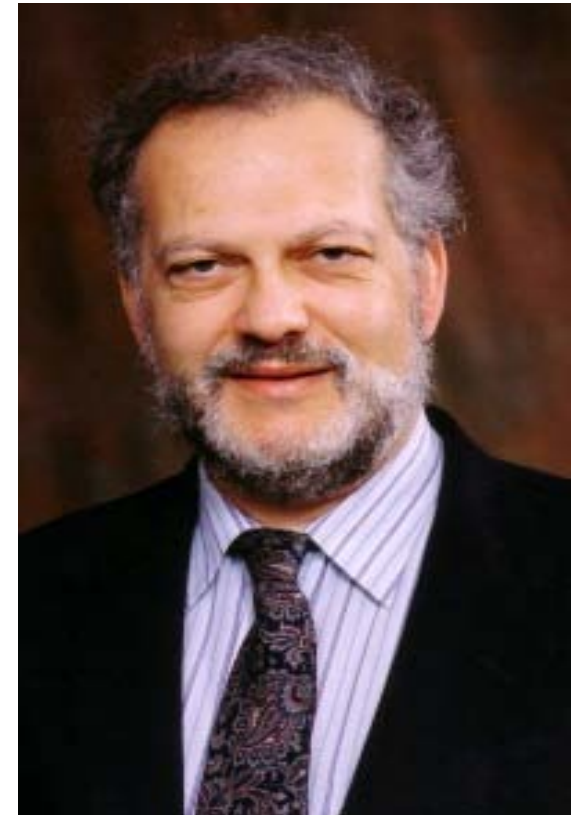
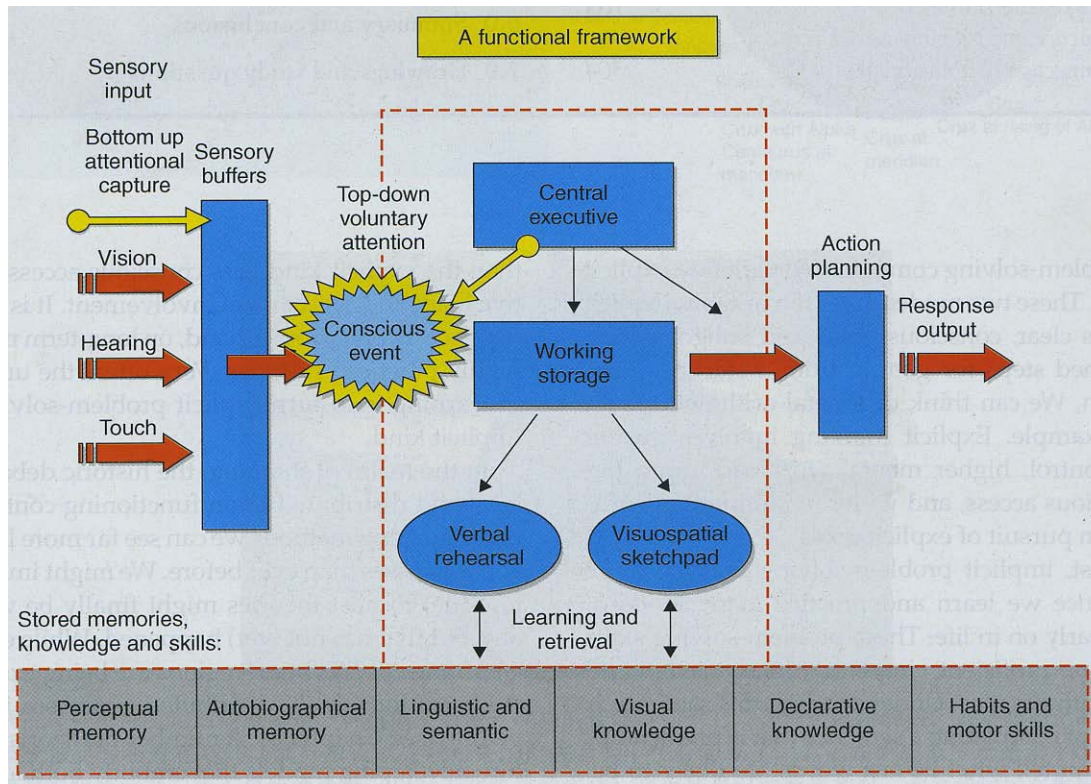


Consciousness is just higher order thoughts (thoughts about other thoughts)
(Higher Order Theory)



A continuous stream of ever-recurring phenomena, pinched like eddies into isolated minds
(Buddhism)

Framework of Cognitive Functions

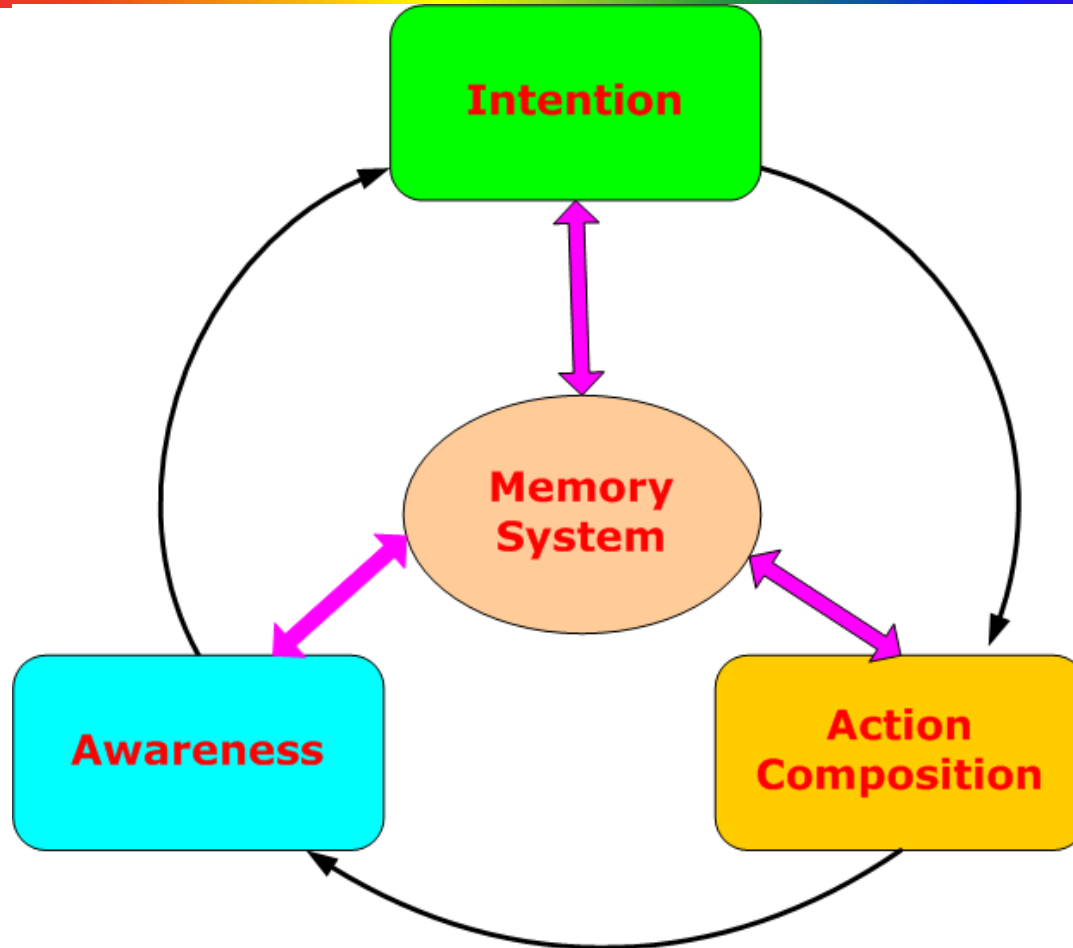


B. Baars, N. Gage. Cognition, Brain, and Consciousness: Introduction to Cognitive Neuroscience. Elsevier Ltd 2007

Computational Model of Consciousness

- Executive control system of the mind;
Farthing, G. W. (1992). *The Psychology of Consciousness*. Upper Saddle River, NJ: Prentice Hall.
- **Intentionality**
Intentionality refers to the feature of a desire to switch from one mental state to another mental state.
- **Global Workspace Theory**
GWT was developed based on robust evidence regarding conscious processes, combined with the artificial intelligence concept of a “blackboard architecture” that combined multiple sources of knowledge in order to identify an acoustical signal in a complex, noisy, and ambiguous environment

Cognitive Cycle of CAM



Contents Outline



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Mind Model CAM

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Perspectives

Nobel Prize in Physiology or Medicine in 1963



Alan Lloyd Hodgkin



Andrew Fielding
Huxley

- Combination of experiments, theoretical hypotheses, data fitting and model prediction
- Empirical model to describe generation of action potentials
- Published in the Journal of Physiology in 1952 in a series of 5 articles (with Bernard Katz)

The Hodgkin-Huxley Model

$$C_m \frac{dV}{dt} = -i_m + \frac{I_e}{A}$$

$$i_m = \bar{g}_L (V - E_L) + \bar{g}_K n^4 (V - E_K) + \bar{g}_{Na} m^3 h (V - E_{Na})$$

$$\tau_z(V) \frac{dz}{dt} = z_\infty(V) - z$$

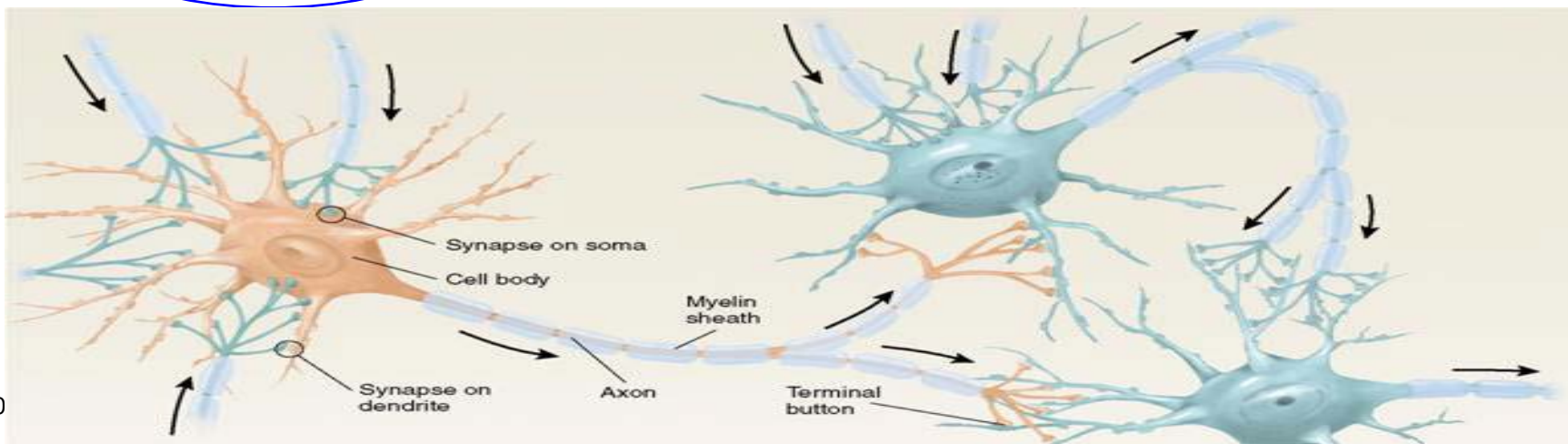
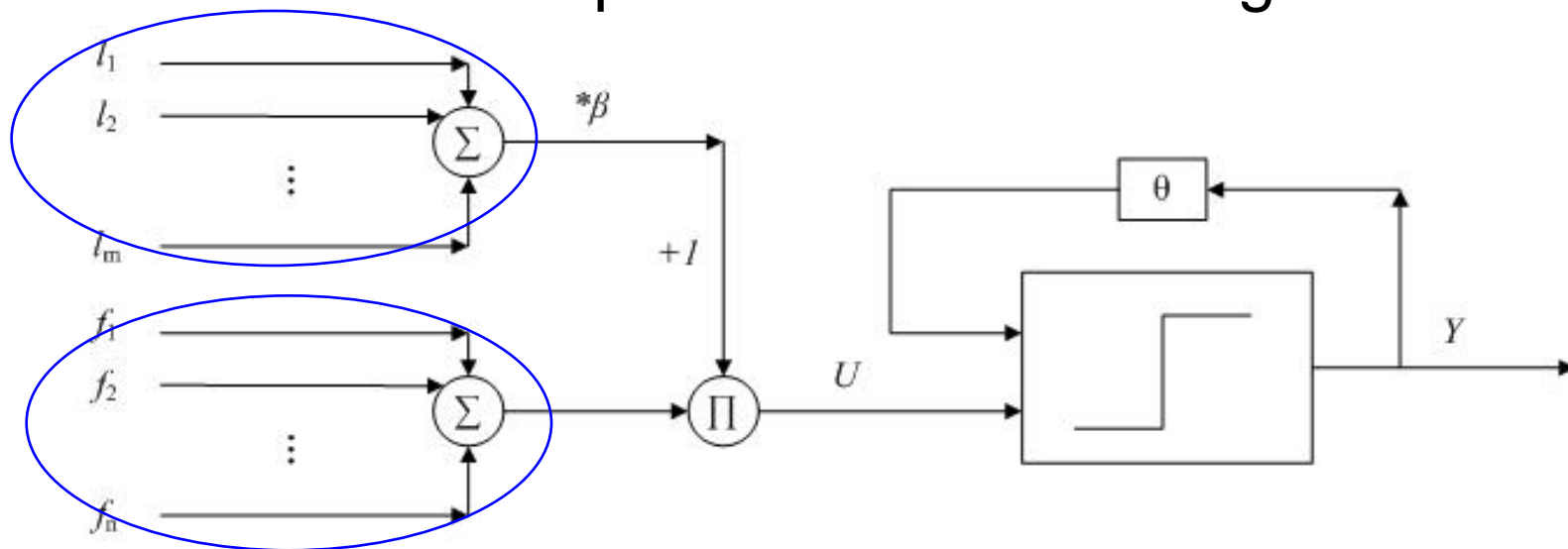
Gating equation

Feature Linking Network



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1990 Eckhorn Proposed Feature Linking Network

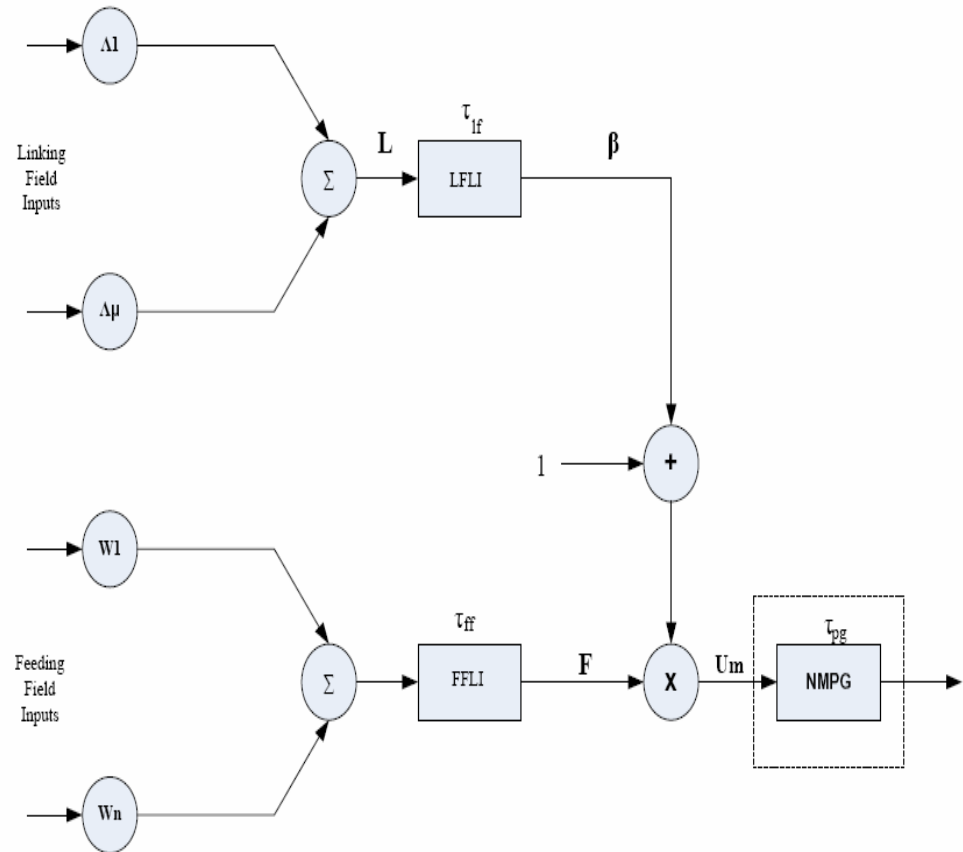


Pulse-Coupled Neural Network

The PCNN was originally presented by Eckhorn et al.

in order to explain the synchronous neuronal burst phenomena in the cat visual cortex.

The model neuron consists of three parts: the dendritic tree, the linking modulation, and the pulse generator.



Bayesian Linking Field Model

Principle ideas

- It utilizes noisy neural model, where the inputs and outputs of neurons are firing probabilities, but not pulse
- Each neuron has two parts of inputs, namely feeding inputs and linking inputs.
- Weights of connections between neurons reflect the statistical relation between them, and can be learnt via learning.
- The outputs of neurons are determined by both their own inputs and the outputs of their neighbors.

Neuron Model

- Dynamic activity

$$P(X) = \sum_i w_i P(f_i)$$

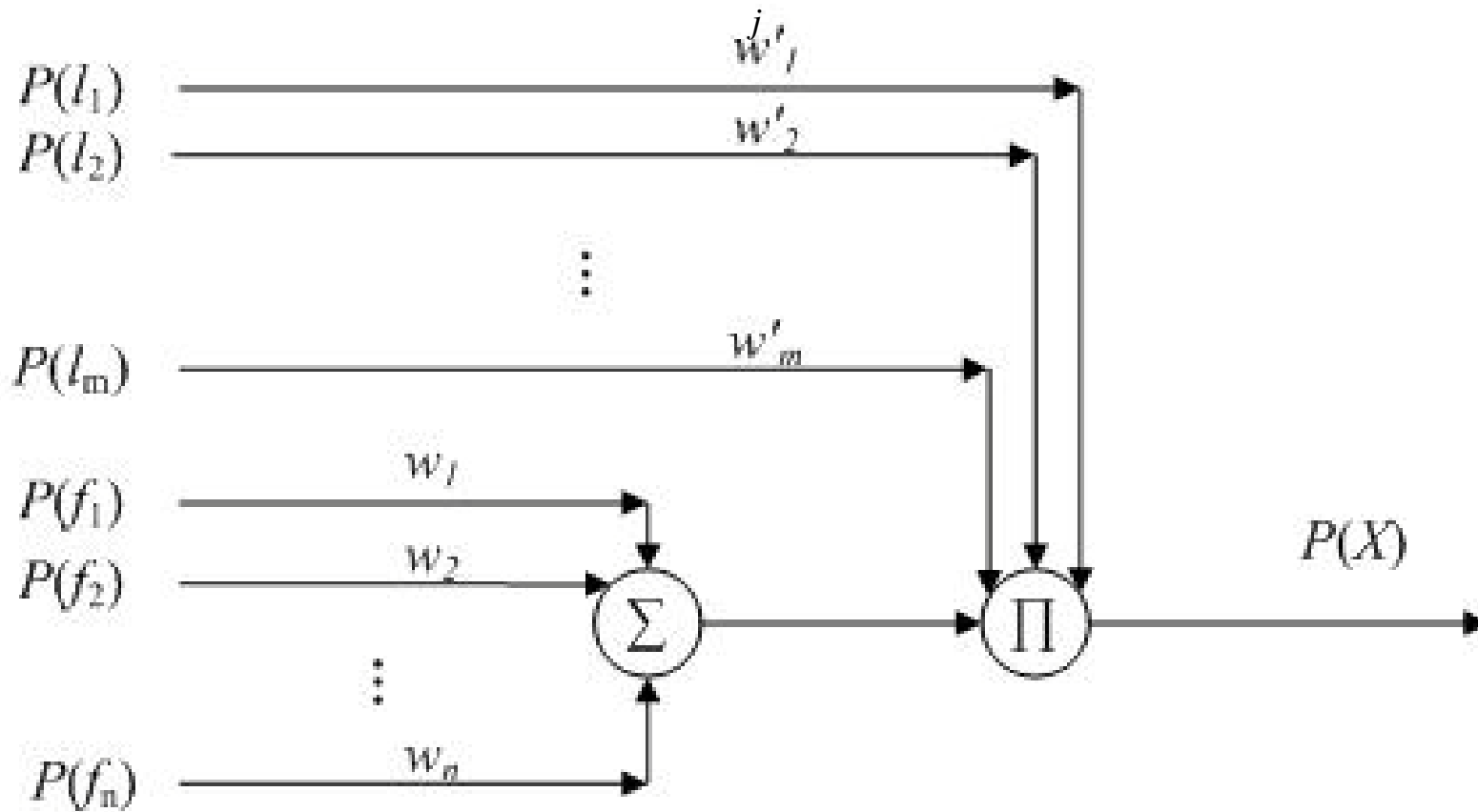
Feeding
input

$$P(X | l_1, l_2, \dots) = P(X) \cdot \prod_j w'_j P(l_j)$$

Linking
input

Bayesian Linking Field Model

$$P(X | l_1, l_2, \dots) = P(X) \cdot \prod w_j' P(l_j)$$



$$P(X) = \sum_i w_i P(f_i)$$

Competition Among Neurons

- Let $X1$ and $X2$ be two different neurons; $F1$ and $F2$ be the set of their feeding pre-synaptic neurons respectively. Then there exists competitive relation between $X1$ and $X2$ if and only if at least one of the two conditions below holds.
 - $F_1 \cap F_2 \neq \emptyset$
 - Exist $f1 \in F1$ and $f2 \in F2$, and $f1$ and $f2$ are competitive

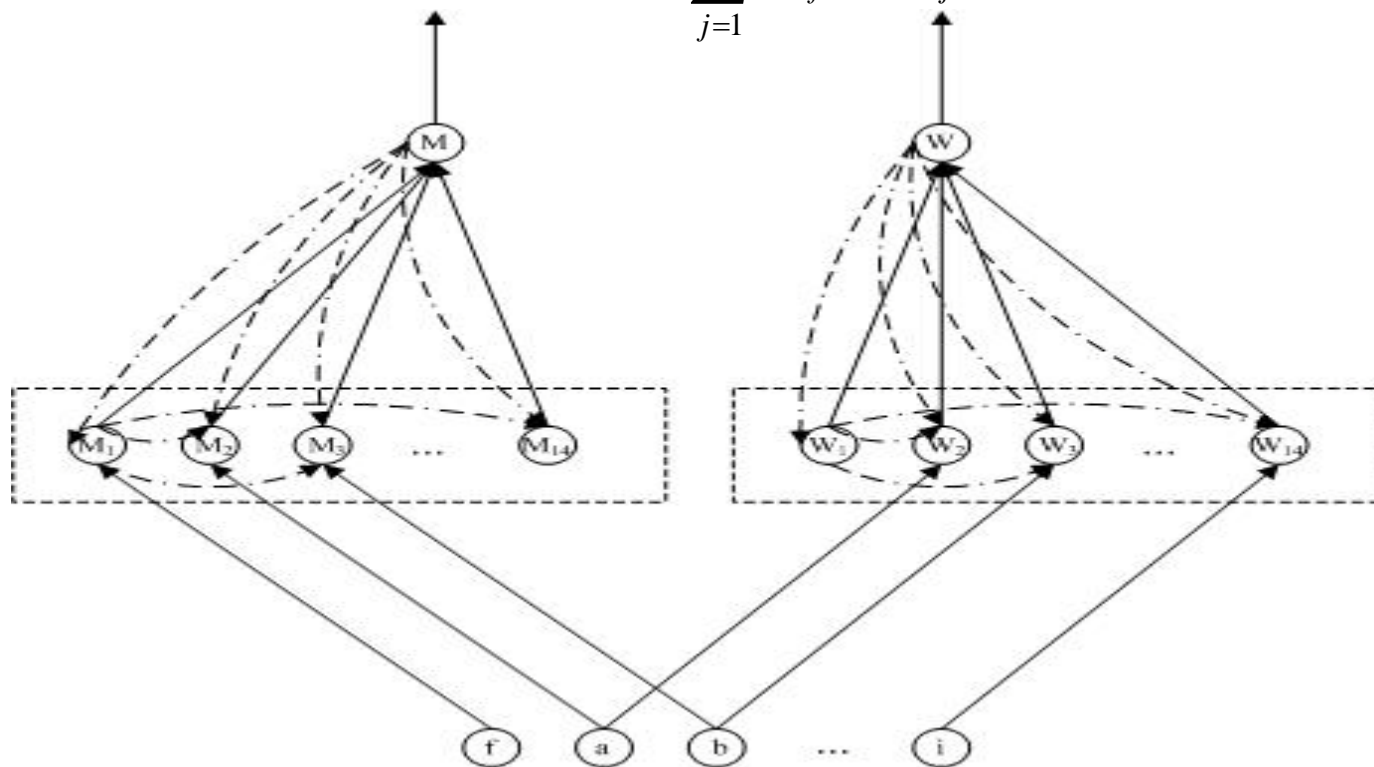
Competition Among Neurons

- To implement competitive relations, we normalize the firing probabilities of the neurons that are competitive each other.
- Let X_1, X_2, \dots, X_n be n neurons that are competitive each other; $P_{before}(X_i)$ is the firing probability of X_i before competition. Then the firing probability of X_i after competition is:

$$P_{after}(X_i) = \frac{P_{before}(X_i)}{\sum_{j=1}^n P_{before}(X_j)}$$

Bayesian Linking Field Model

$$P_{after}(X_i) = \frac{P_{before}(X_i)}{\sum_{j=1}^n P_{before}(X_j)}$$



Perceptual Problem Description



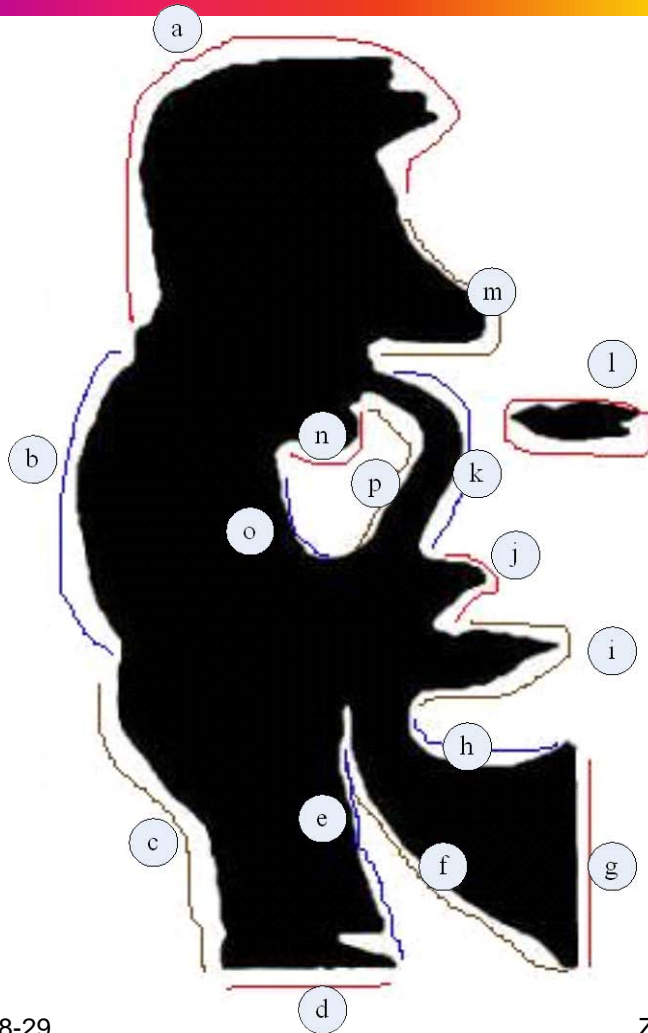
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The picture can be interpreted as

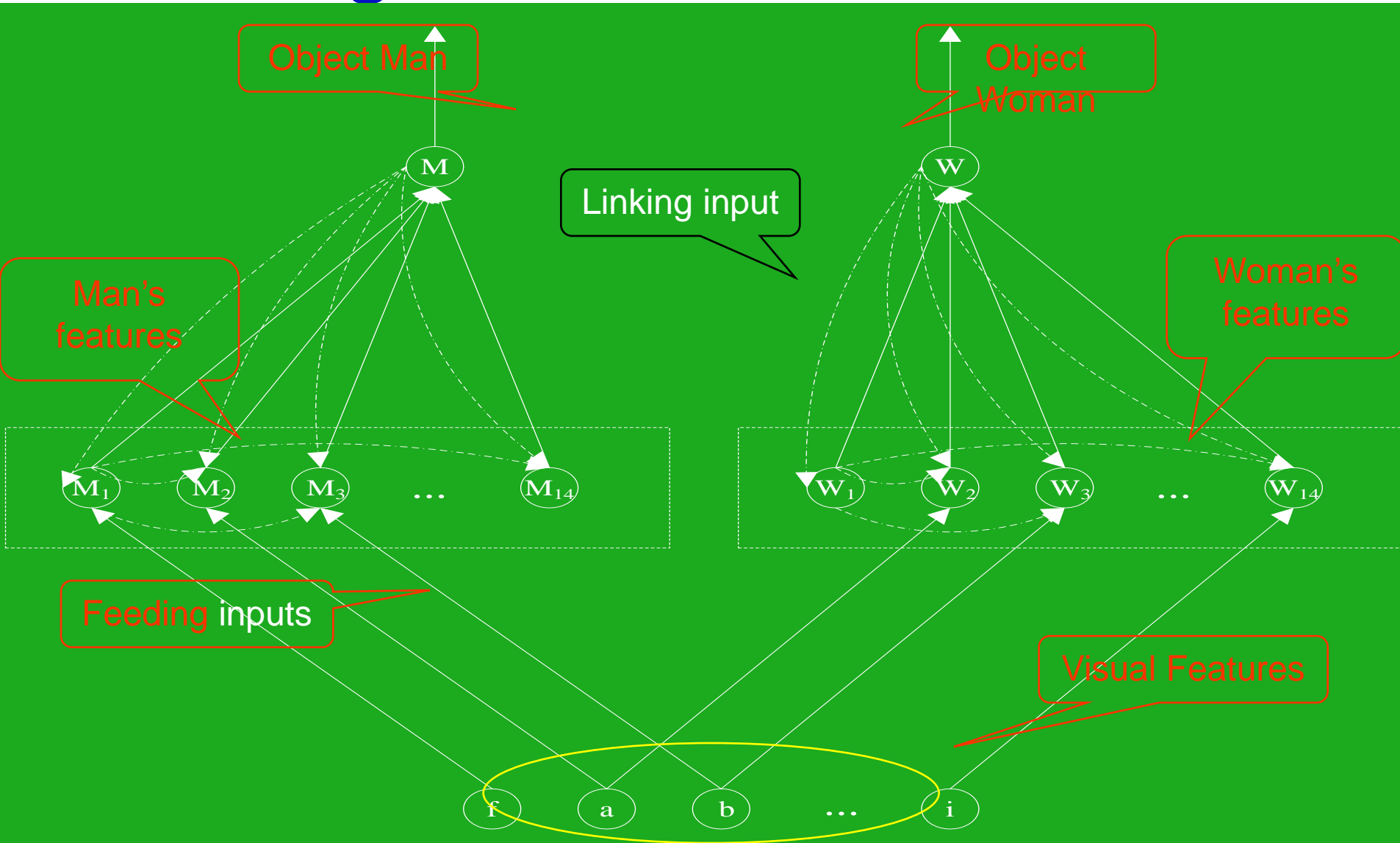
- a man playing a saxophone or
- a woman face

Simplification



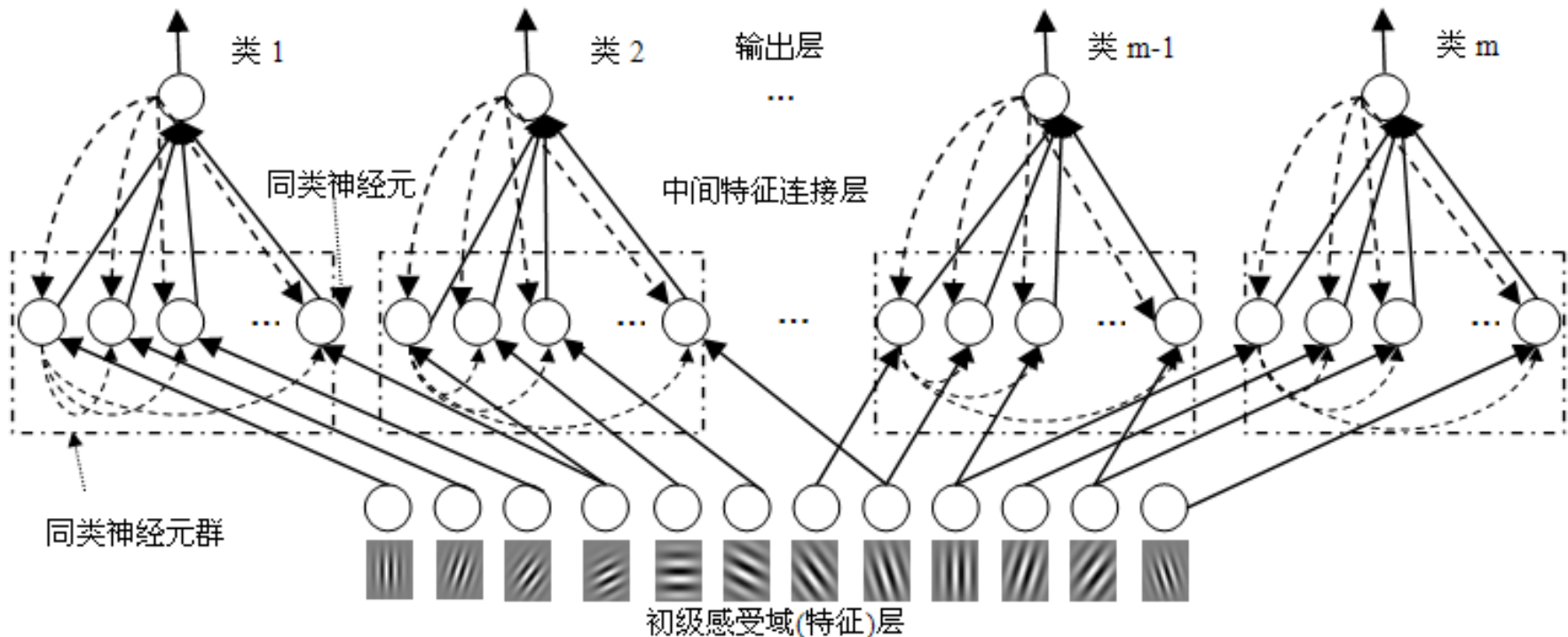
Entire picture is
partition into 16
compartments

The architecture of Bayesian Linking Field Model



Computational Model for Feature Binding

Bayesian Link field Networks Model



Contents Outline



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Mind Model CAM

Feature Binding

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World Wide Brain

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Image Understanding

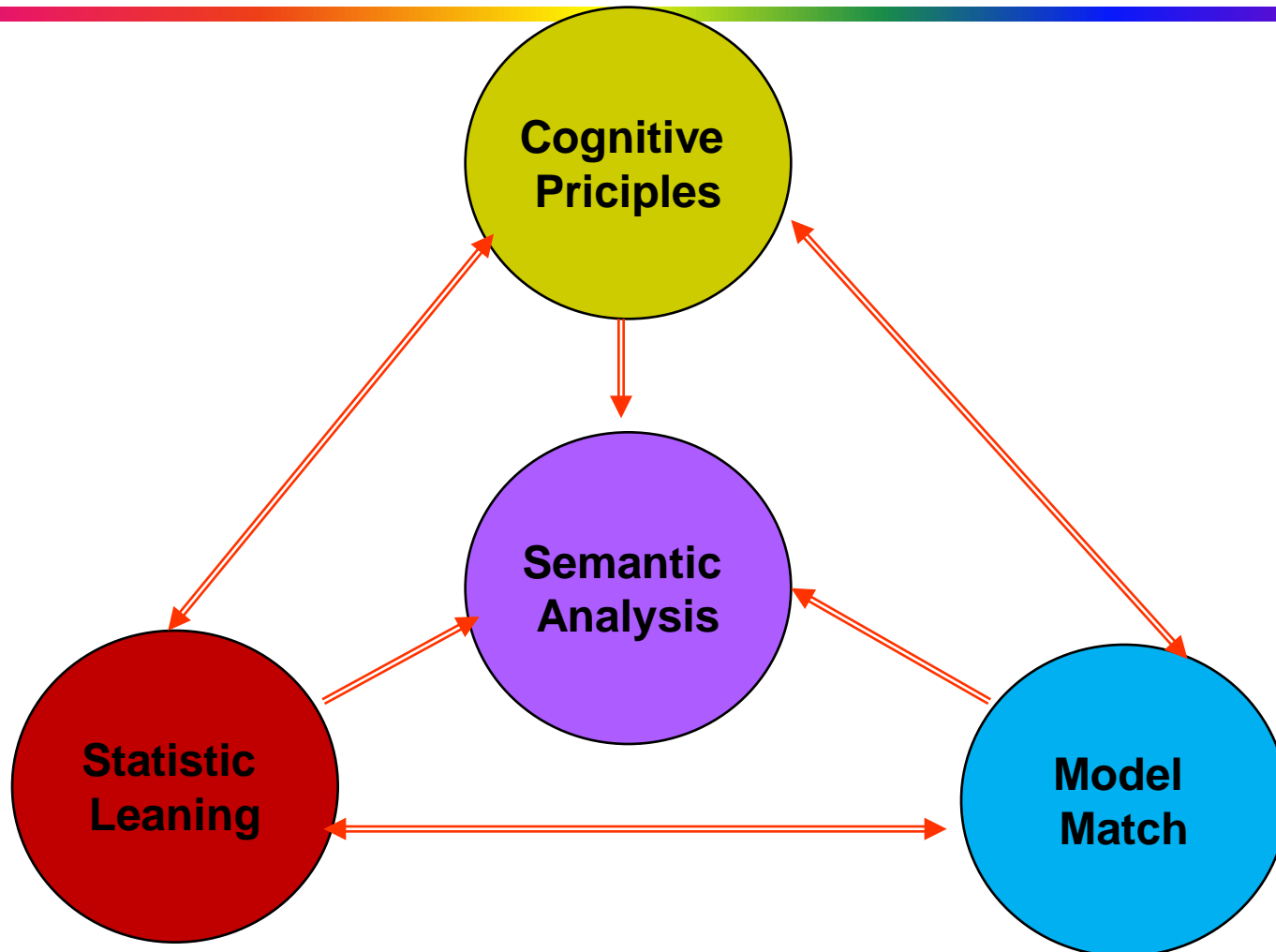


Suzhou Garden

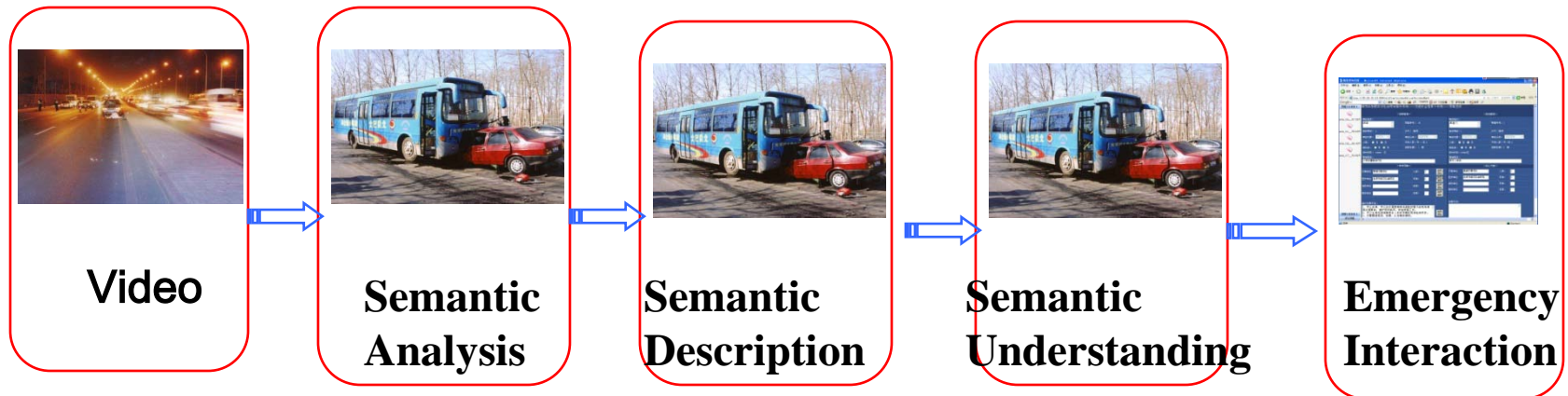


Suzhou Building

Semantic Analysis



Traffic Surveillance



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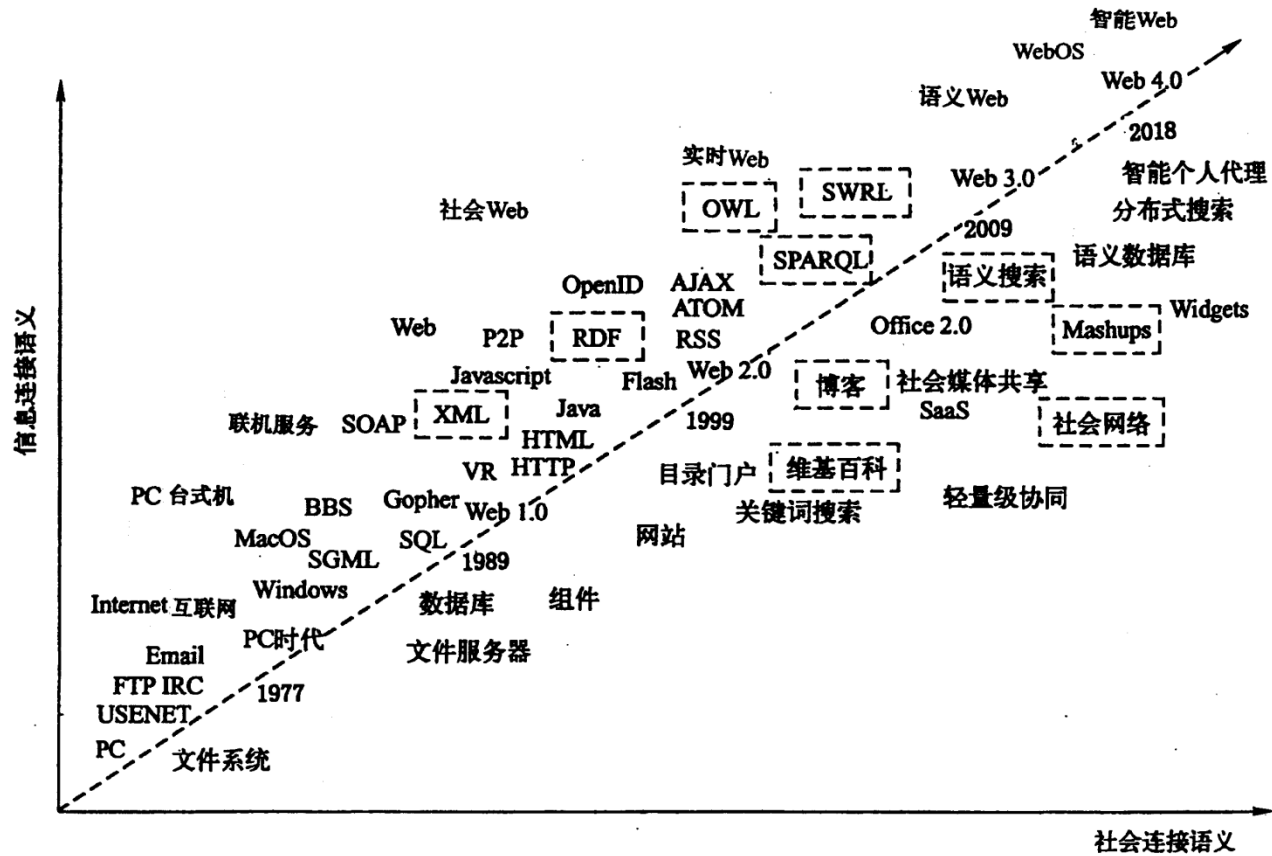
Feature Binding

Image Understanding

World Wide Brain

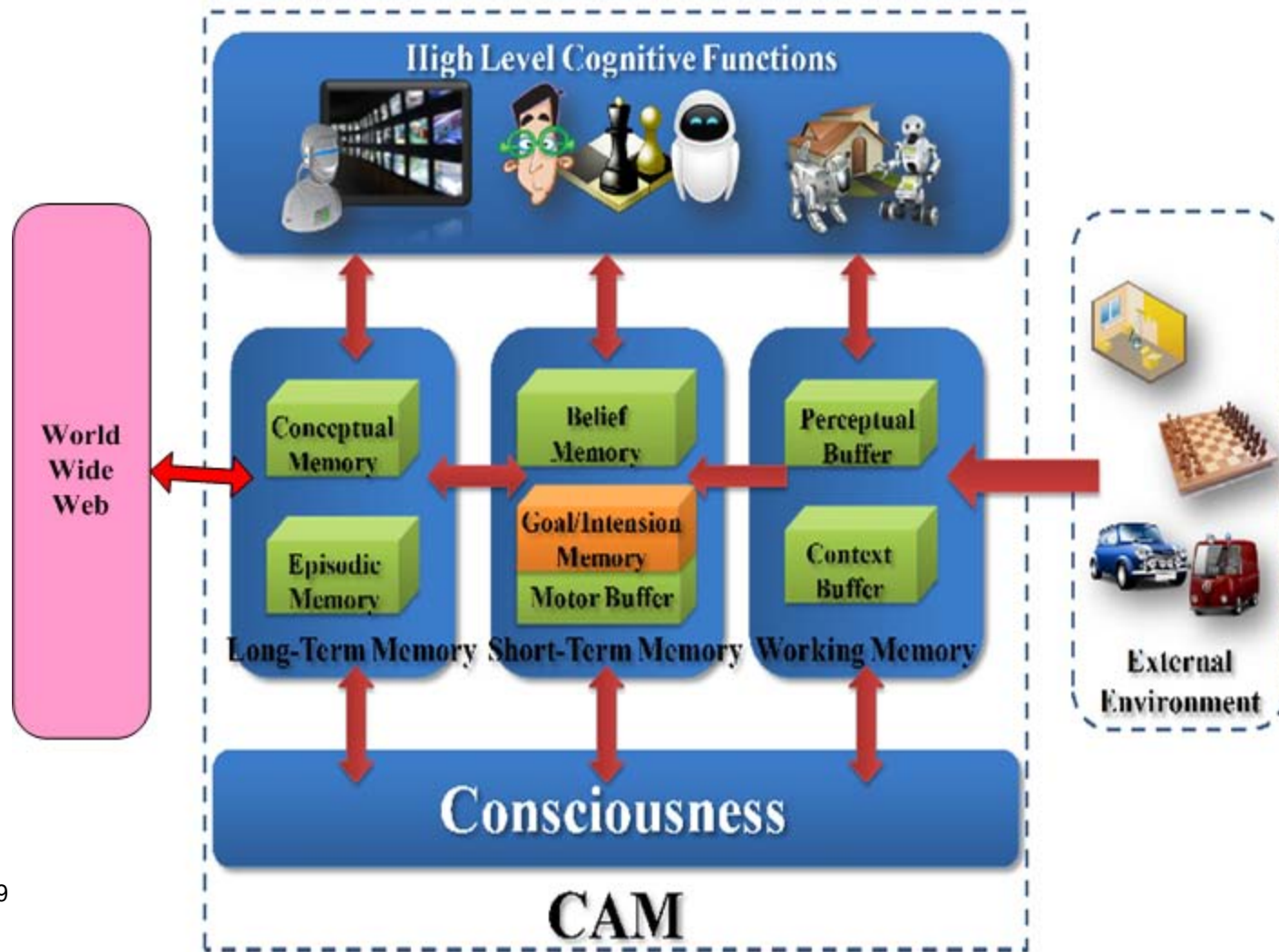
Perspectives

Internet Development Roadmap

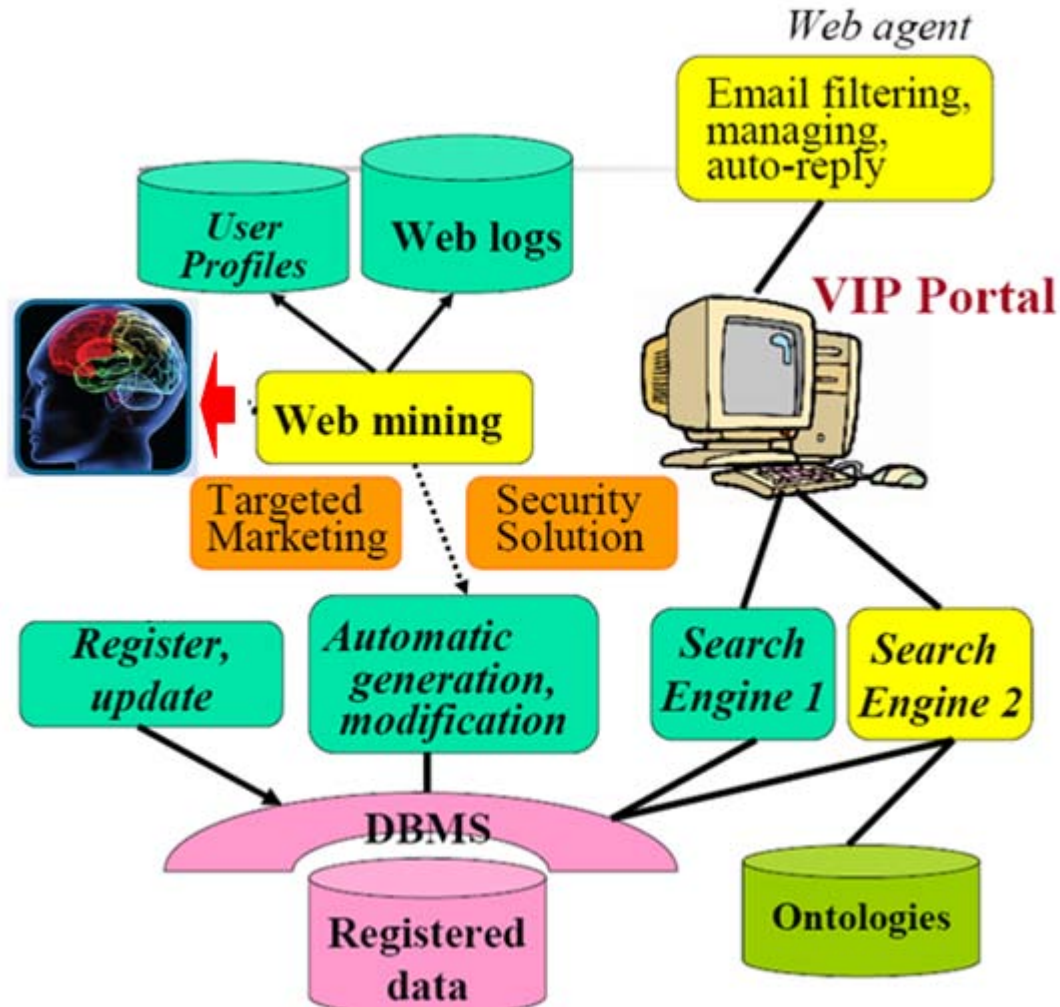


Spivack 2008

World Wide Brain



World Wide Brain



Conclusions

- **Human level intelligence needs new research paradigm.**
- **Intelligence Science is a new research paradigm**
- **The mind model CAM has obvious features and potential applications**

Thank You

Question!

Intelligence Science
<http://www.intsci.ac.cn>

