

The Brain That Wouldn't Die

2021 한국뇌공학회 심포지엄 “디지털 브레인으로의 진화”
온라인 심포지엄, February 25, 2021

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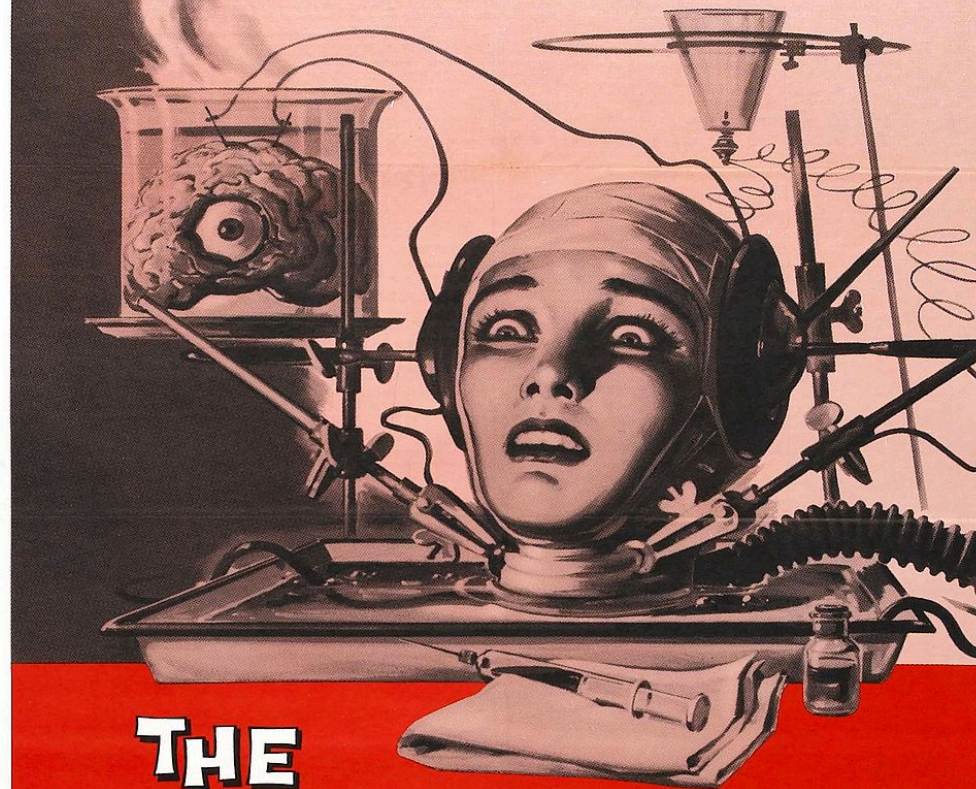


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▪ New AI, Self-improvement, Self-teaching, Self-reflection	
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1. The Brain in a Vat

ALIVE...WITHOUT A BODY...FED BY
AN UNSPEAKABLE HORROR FROM HELL!



THE BRAIN THAT WOULDN'T DIE



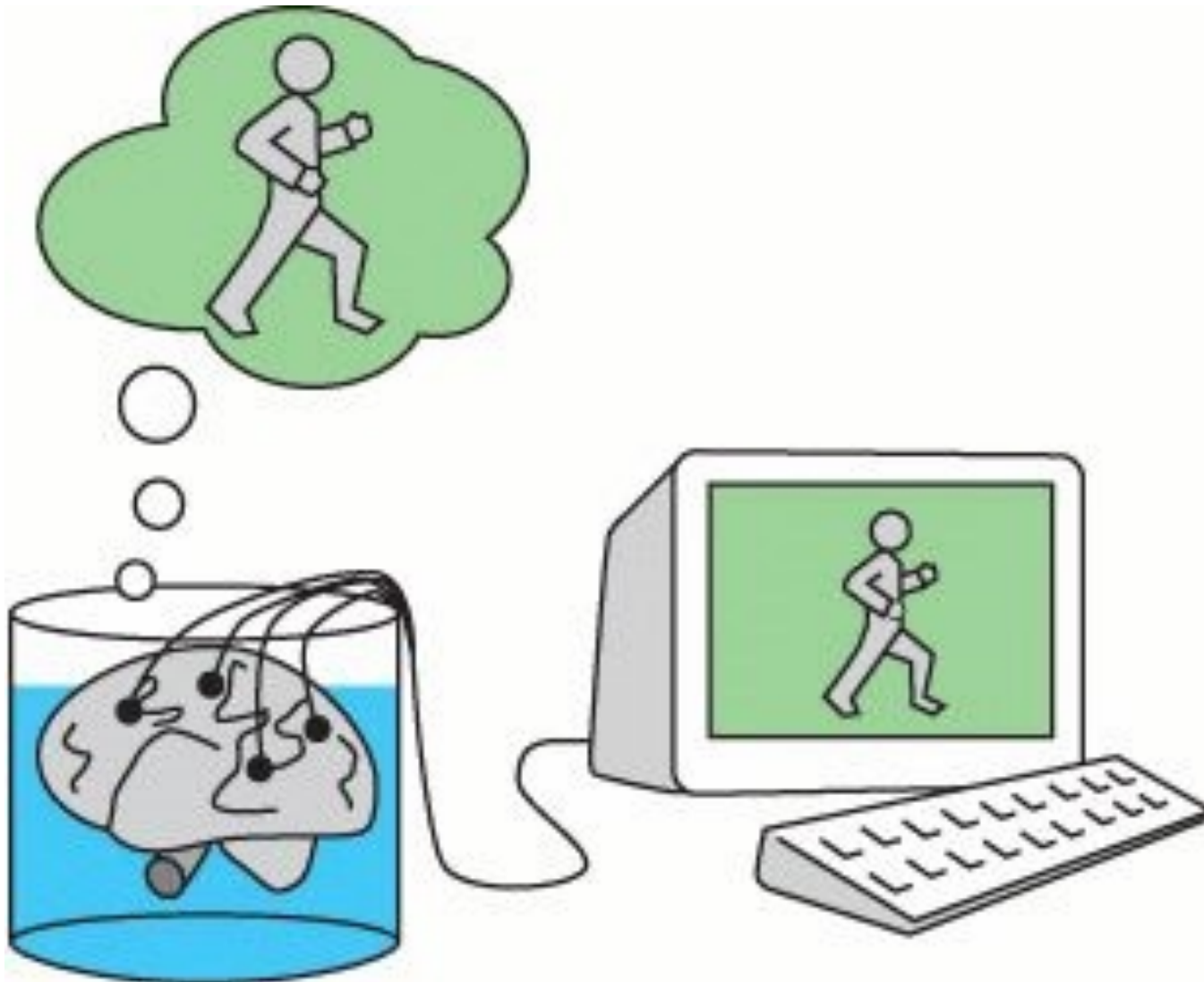
STARRING
HERB EVERS
VIRGINIA LEITH
LESLIE DANIEL

SCREENPLAY BY JOSEPH GREEN • DIRECTED BY JOSEPH GREEN • PRODUCED BY REX CARLTON • AN AMERICAN INTERNATIONAL RELEASE

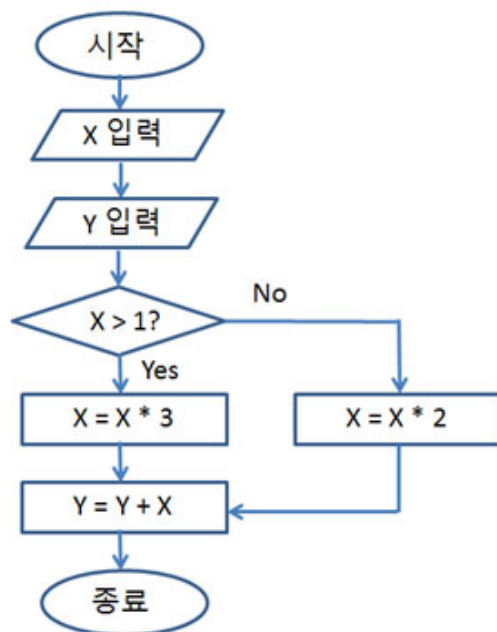
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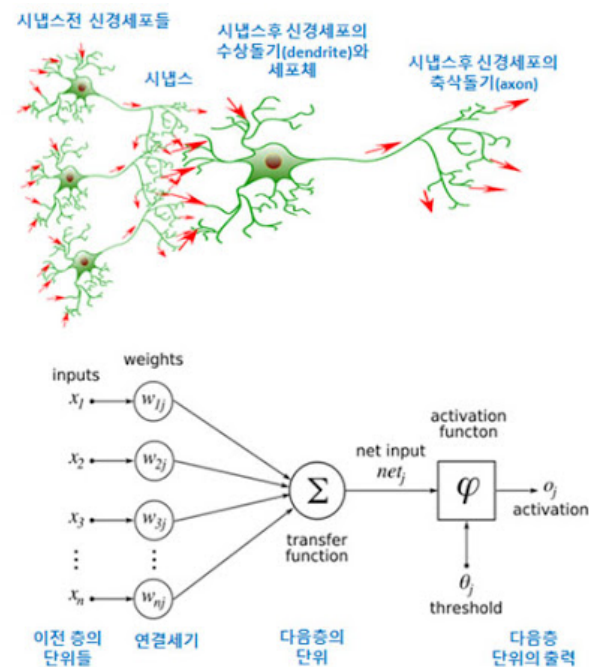
Brain in a Vat



Two Paradigms of AI



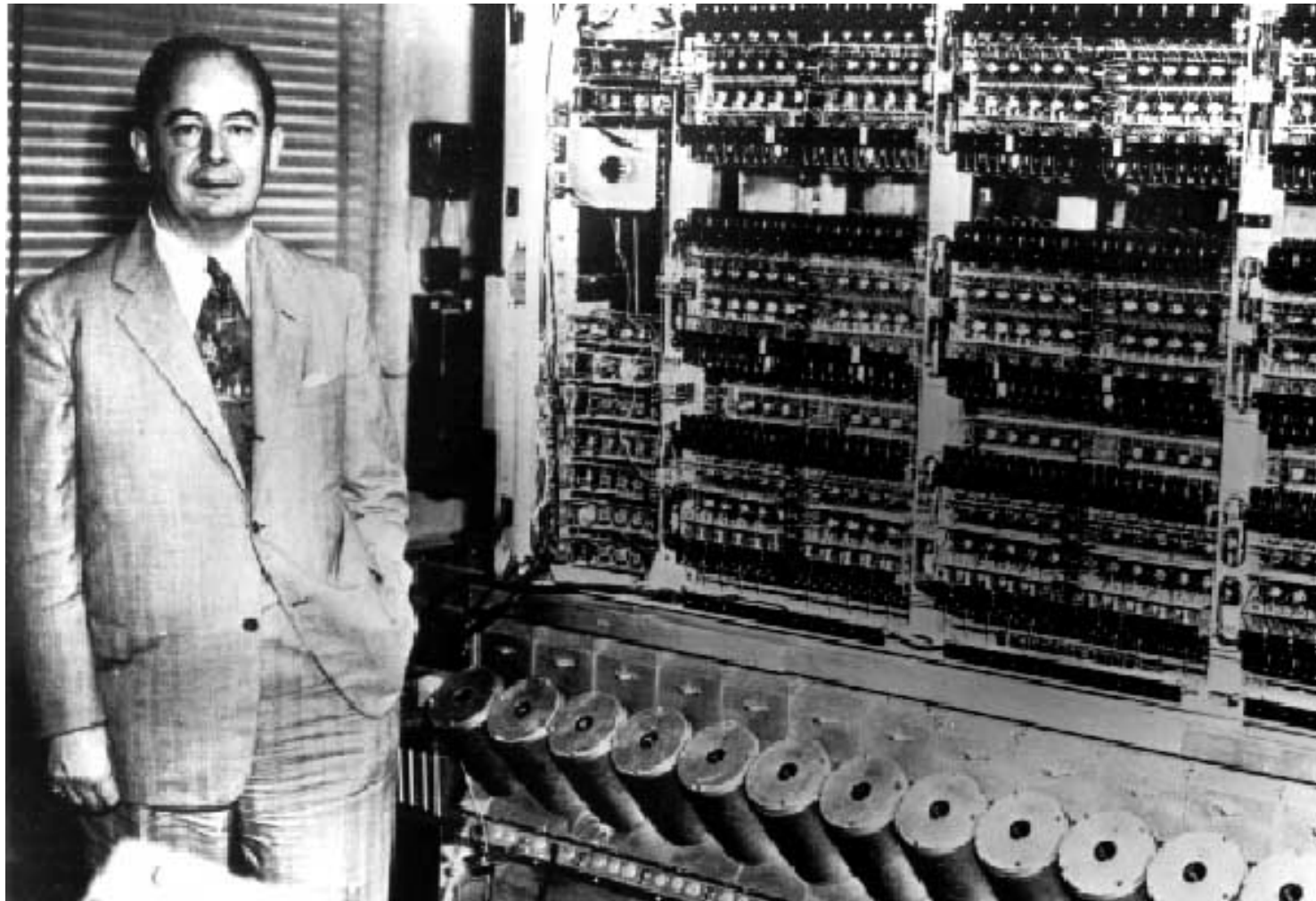
기호주의-논리규칙 모델



연결주의-뇌신경망 모델

출처: <http://scienceon.hani.co.kr/397536>

Von Neumann's *The Computer and the Brain* (1958)

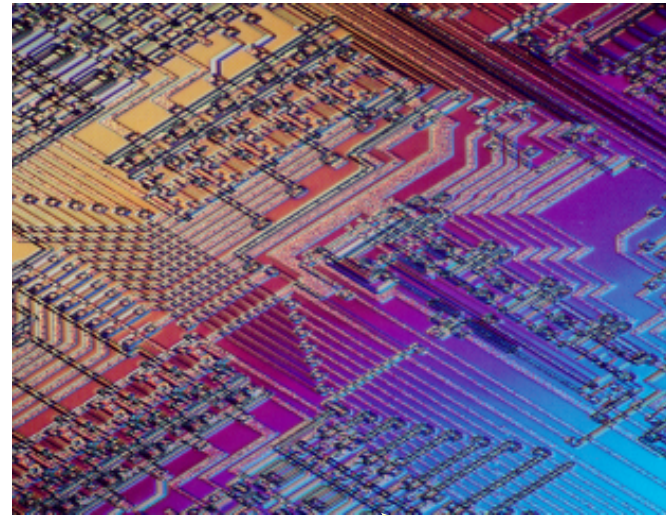


7 John von Neumann (1903-1957)

The Brain vs. Computer



1. 10 billion neurons
2. 60 trillion synapses
3. Distributed processing
4. Nonlinear processing
5. Parallel processing



1. Faster than neuron (10^{-9} sec)
cf. neuron: 10^{-3} sec
3. Central processing
4. Arithmetic operation (linearity)
5. Sequential processing

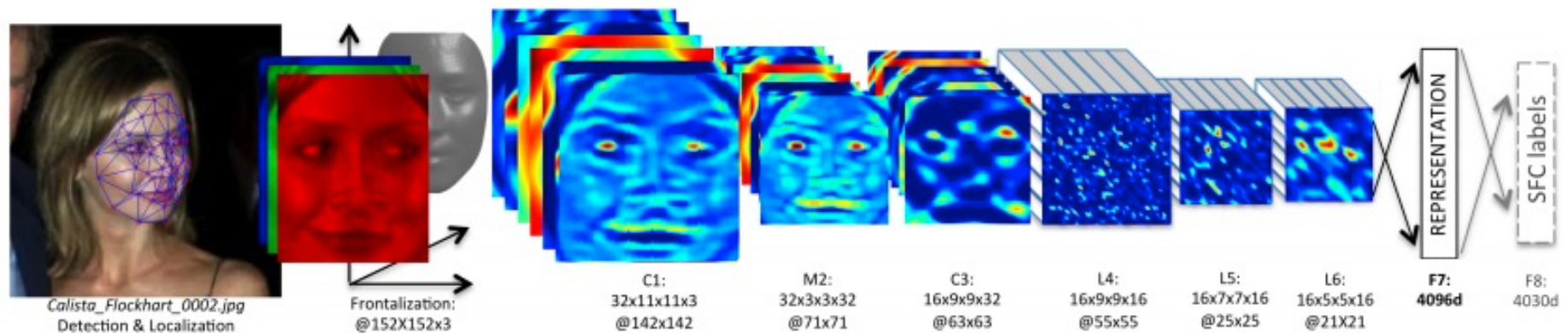
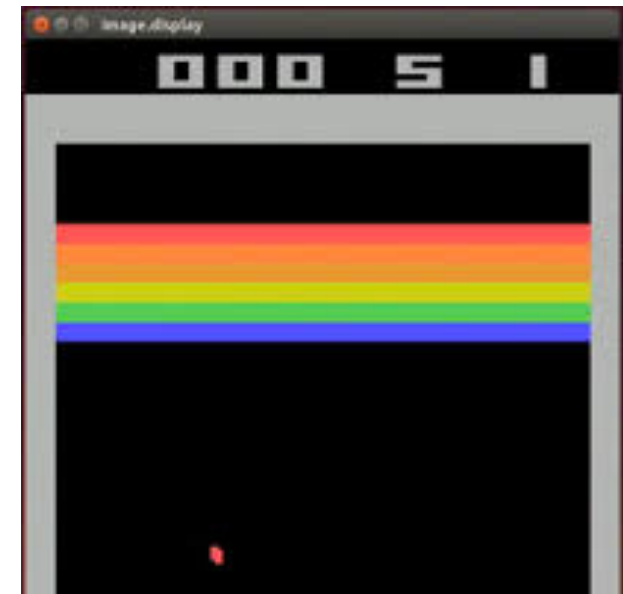
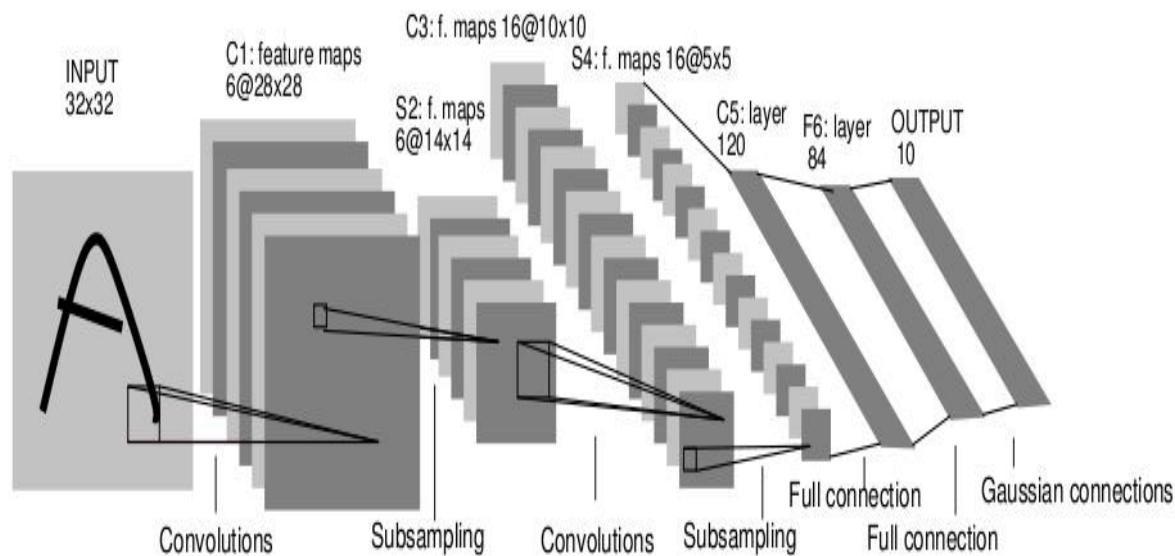
History of Brain-Style Computation (1/2)

- 1943 McCulloch & Pitts Neuron
- 1948 Hixon Symposium on Cerebral Mechanisms in Behavior
- 1949 Hebbian Learning & Cell Assembly
- 1955 Session on Learning Machines (Selfridge's Pandemonium)
- 1957 Perceptron (Rosenblatt)
- 1958 The Computer and the Brain (von Neumann)
- 1960 Adalines & Madalines (Widrow)
- 1970's Kohonen, Amari, Grossberg
- 1979 First Issue of Cognitive Science Journal
- 1979 First Conference on Cognitive Science
- 1980 First Machine Learning Workshop (ML 1)
- 1982 Parallel Models of Associative Memory (Hinton & Anderson)
- 1982 Connectionist Models and Their Properties (Feldman & Ballard)
- 1982 Hopfield Network
- 1984 Kohonen Network (SOM)
- 1986 Machine Learning Journal (Langley)

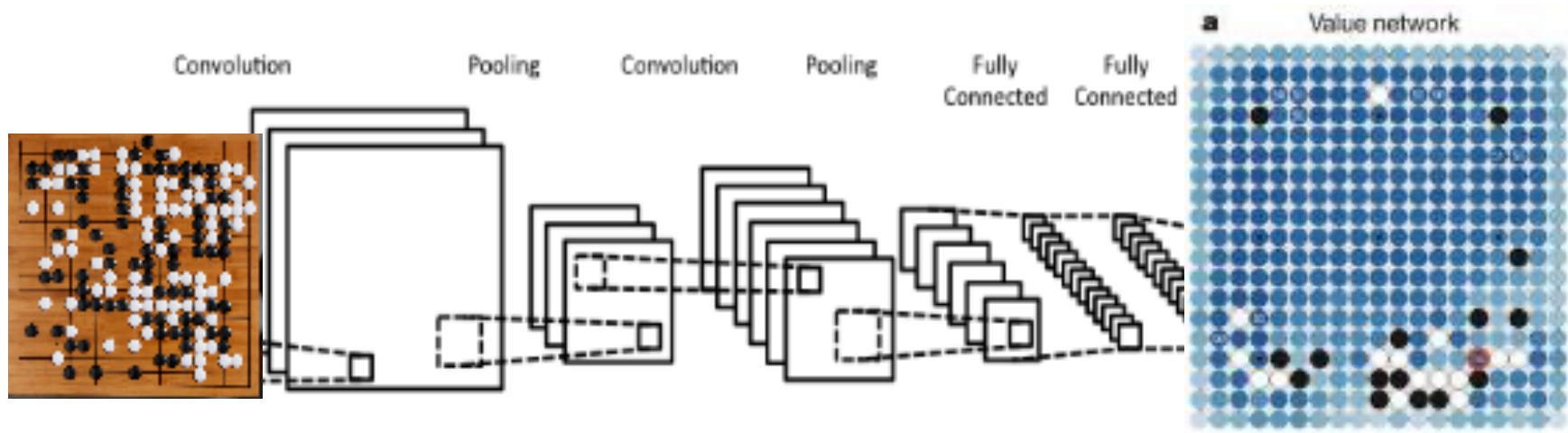
History of Brain-Style Computation (2/2)

- 1986 Parallel Distributed Processing (PDP) Models (Rumelhart & McClelland)
- 1987 International Neural Network Society (INNS)
- 1988 Connectionist Models Summer School (Touretzky, Hinton, Sejnowsky)
- 1988 Bayesian Networks (Pearl)
- 1989 Neural Computation Journal (Sejnowsky)
- 1989 First Neural Information Processing Systems Conference (NIPS 1)
- 1990 Brain Style Computation: Learning and Generalization (Rumelhart)
- 1992 First Computational Neuroscience Conference (CNS)
- 1992 Support Vector Machines (Vapnik)
- 1993 First ICML Conference (ML 10, Int. Conf. on Machine Learning)
- 1995 Helmholtz Machine (Hinton, Dayan, Neal)
- 2002 EU FP6 Artificial Cognitive Systems
- 2006 Deep Belief Networks (DBN)
- 2006 First Computational Cognitive Neuroscience Conference (CCN)
- 2008 IBM Cognitive Computing Project (SyNAPSE)
- 2009 MIT Intelligence Initiative (I²)

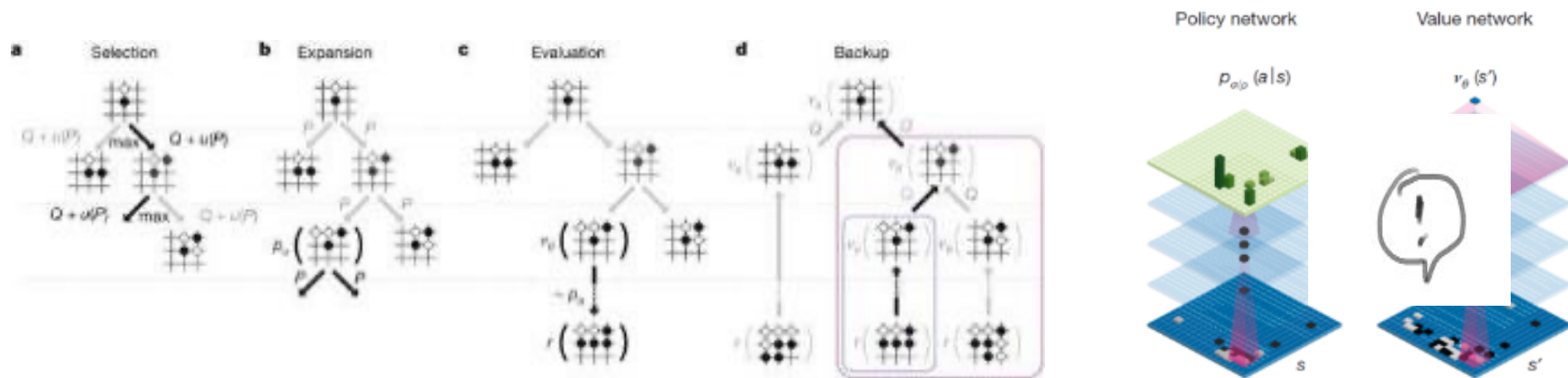
Deep Neural Networks



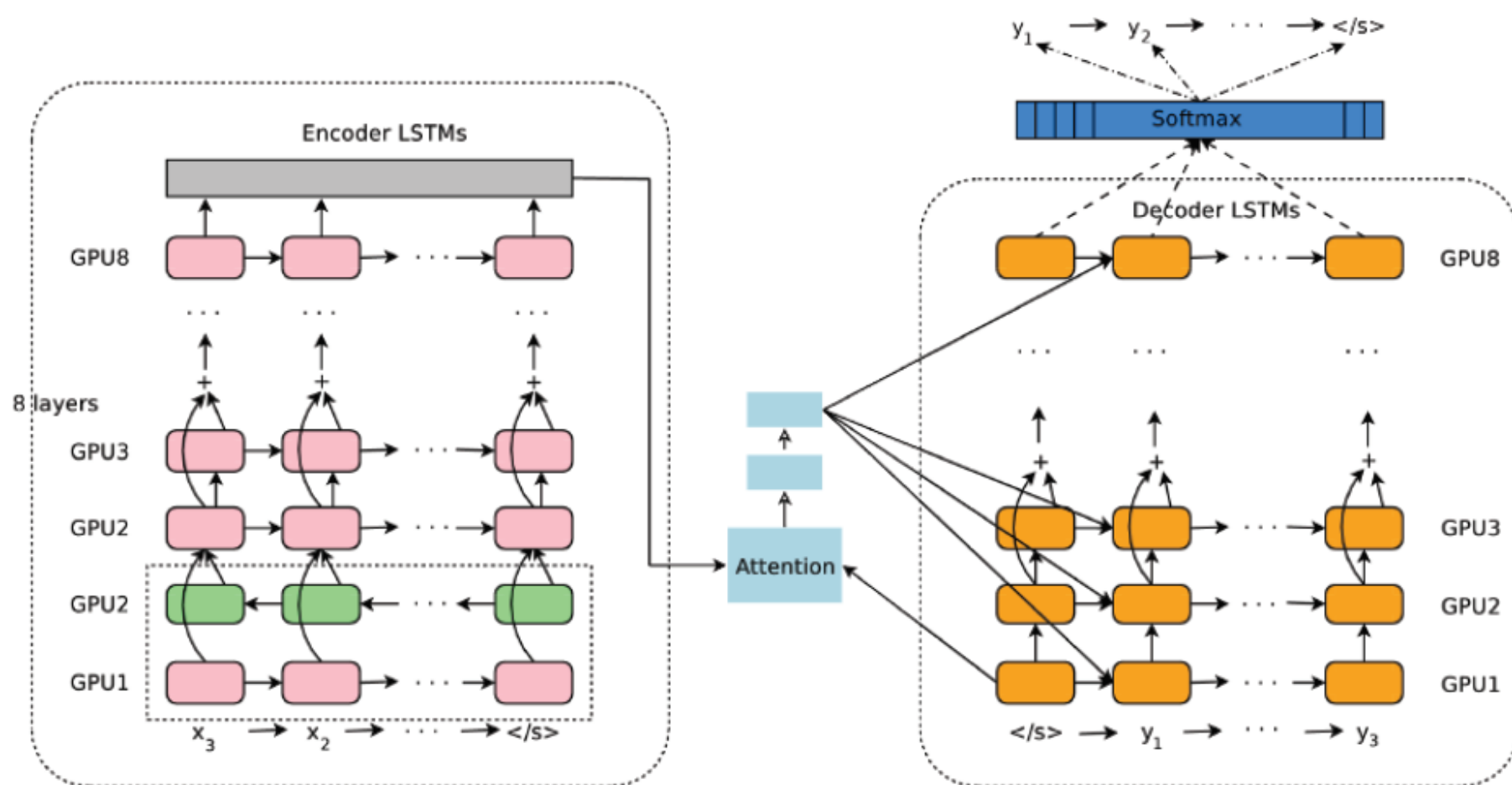
Gaming



AlphaGo (Google DeepMind)



Machine Translation



Google's Neural MT

Image Synthesis



Visual Storytelling (VIST)

GLAC Net: GLocal Attention Cascading Networks for Multi-image Cued Story Generation, T. Kim et al. *NAACL 2018 Workshop on Storytelling, 2018.*

● Visual Storytelling

- Visual story = photo sequence + sentence sequence

Input:



Output:

The local market was a lot of fun.

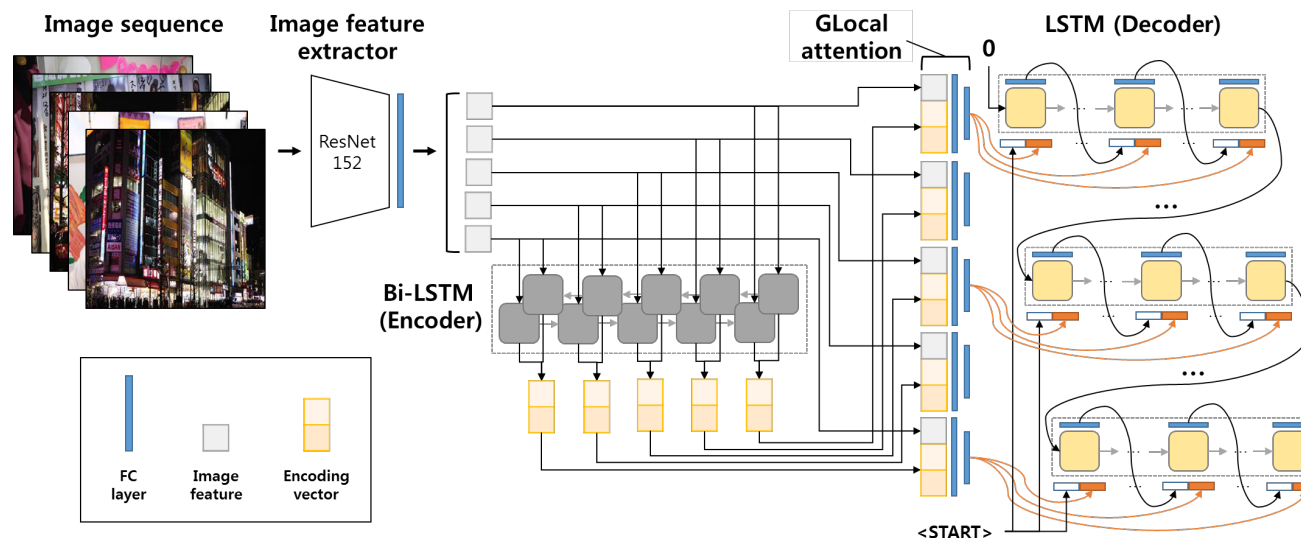
There were so many different kind of things to choose from.

Some of the shops looked very interesting.

I bought a few souvenirs for the kids

My favorite part was the bike ride.

- GLocal Attention Cascading Networks (**GLAC Net**): **1st Place in VIST Challenge**



(Provocative) Question

*Do deep neural networks learn
like a brain?*

2. The Embodied Brain in a Vat

Embodied Brain



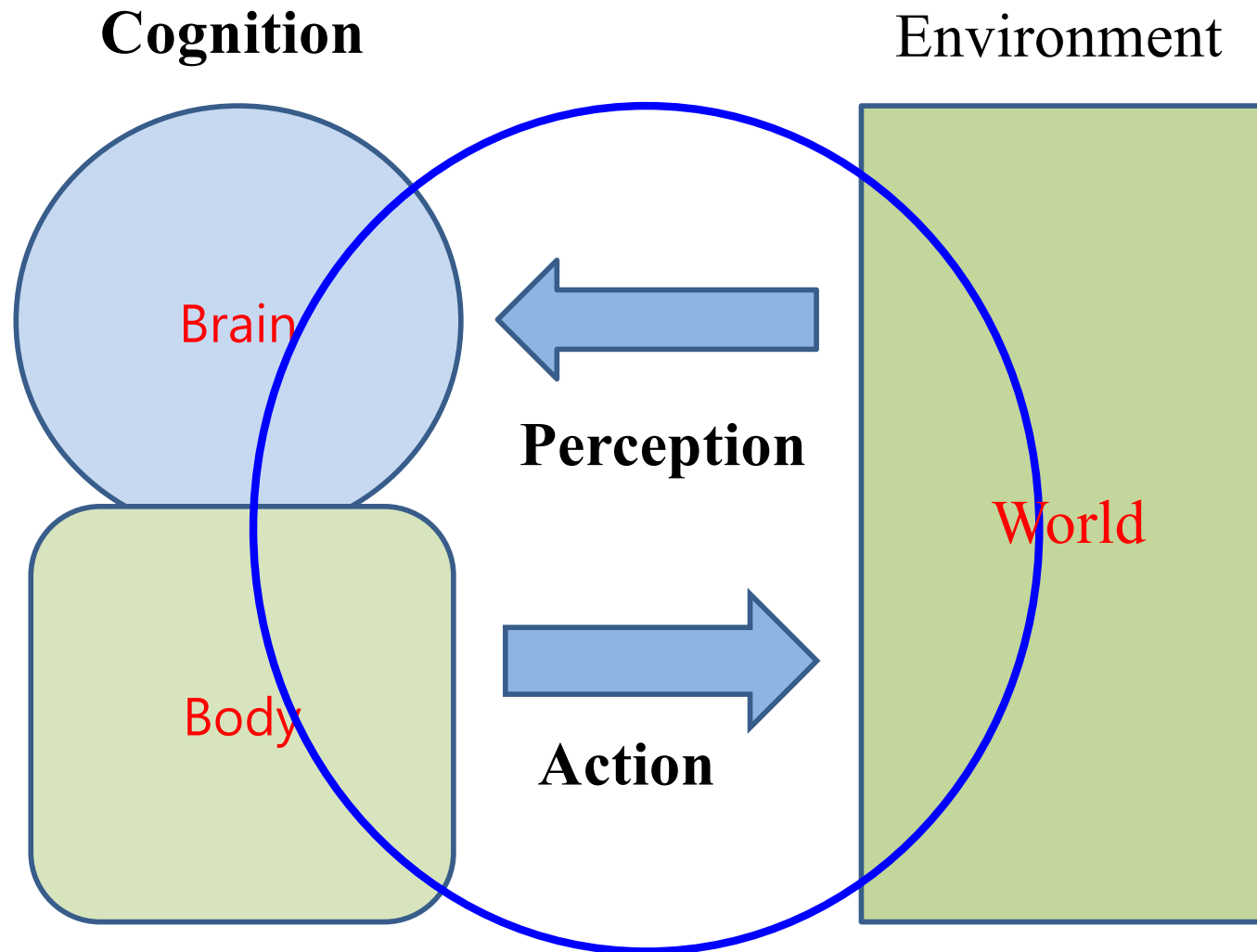
Embodied and Situated Brain



Agents in the Real World

Team Tidyboy
RoboCup@Home DSPL 2019

Brain, Body, and World



[Zhang, 2008]
[Zhang, 2012 a,b,c]
[Zhang, 2018]

- **Cognition grounded by dynamic perception-action cycle**
- **Knowledge construction by interaction with the world**

Why We Need Brain-Like Intelligence?

Smart Machines



[Super Smart TV]



[Cloud & Big Data]



[Smart Watch]



[Smart Glass / VR]



[Autonomous Vehicle]

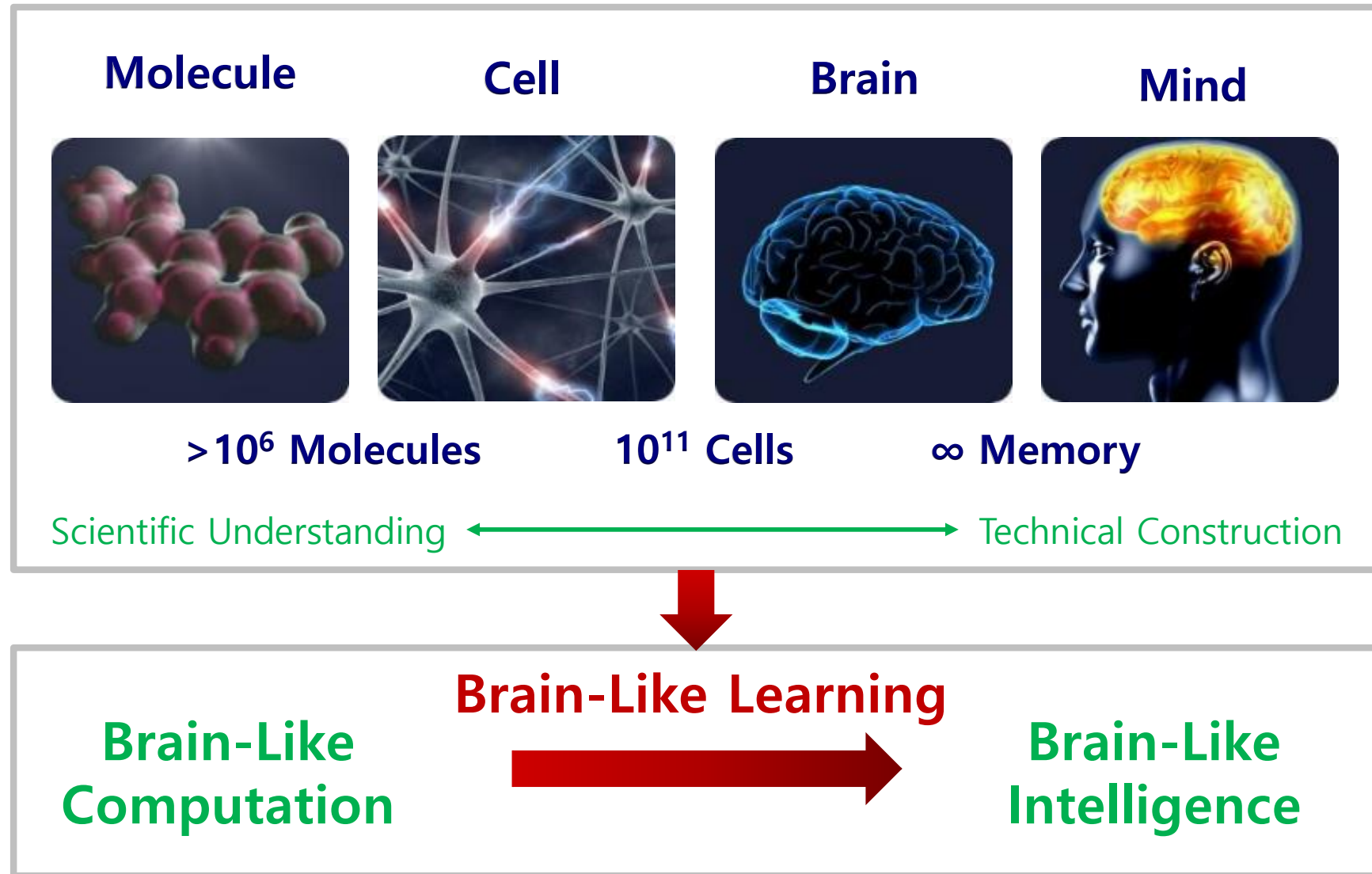


[Smart Home/Factory and IoT]

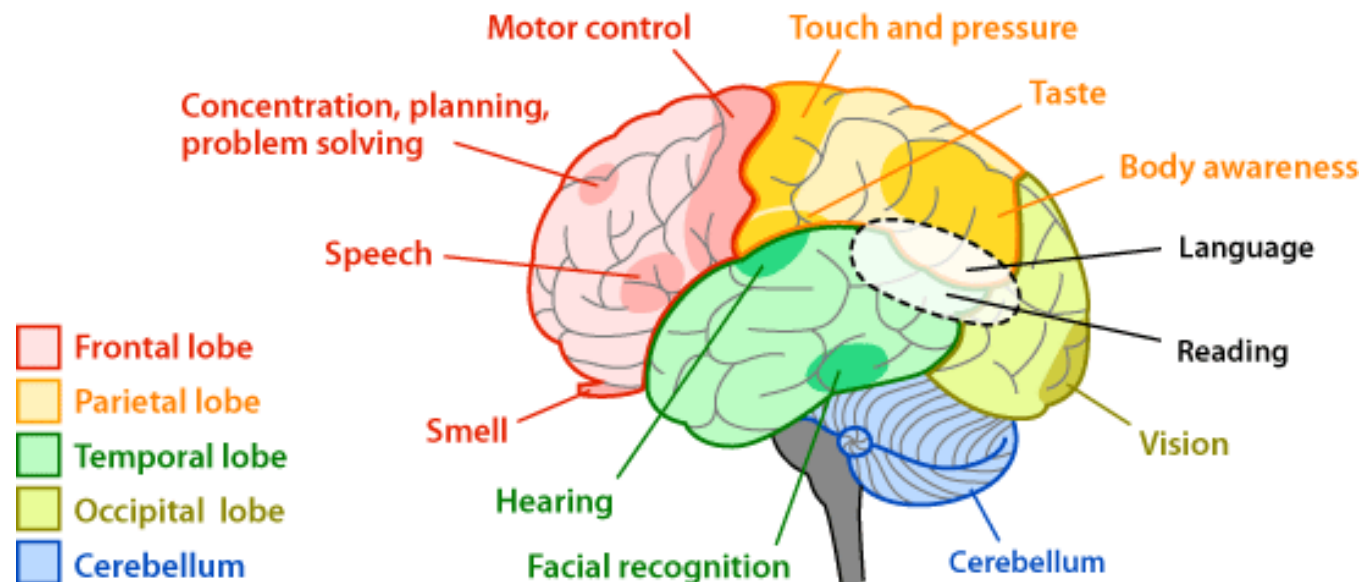
Artificial Intelligence

- **Real-World**
- **Real-Life**
- **Real-Time**
- **Multi-Sensor Streams**
- **Real-Life Context**
- **Open-Endedness**
- **Non i.i.d. Big Data**
- **Non-Stationary**
- **Uncertainty**
- **Rapid, Flexible, Robust**
- **Brain-Like Cognitive AI**

Brain-Like Intelligence



Brain Computation: Why Fast, Flexible, and Robust?



Three Levels of Analysis

- **Structural**

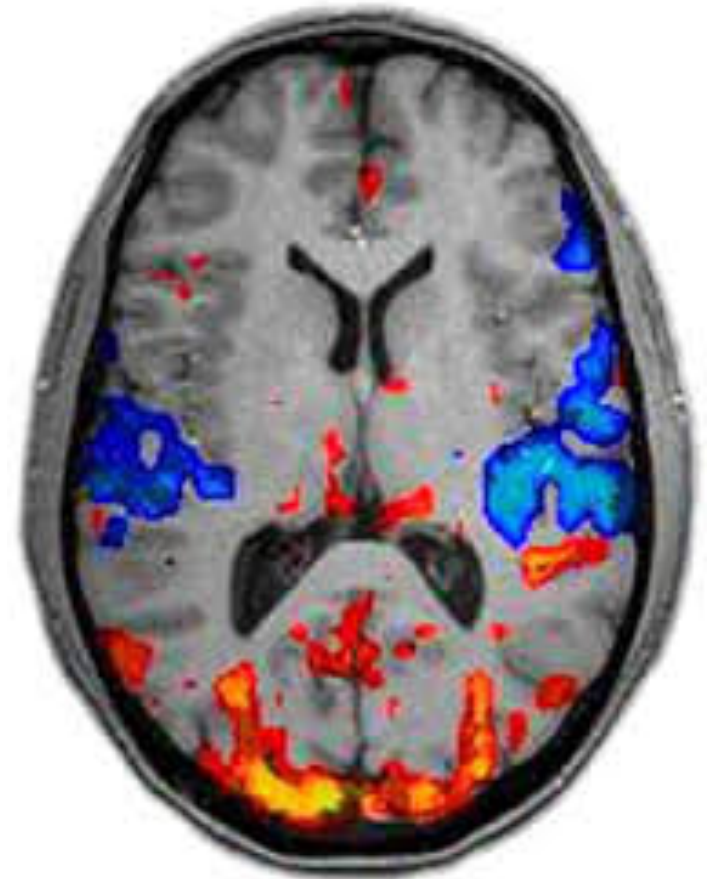
- Containers & substrate
- Representation
- Anatomy

- **Organizational**

- Contents & information
- Encoding
- Development and learning

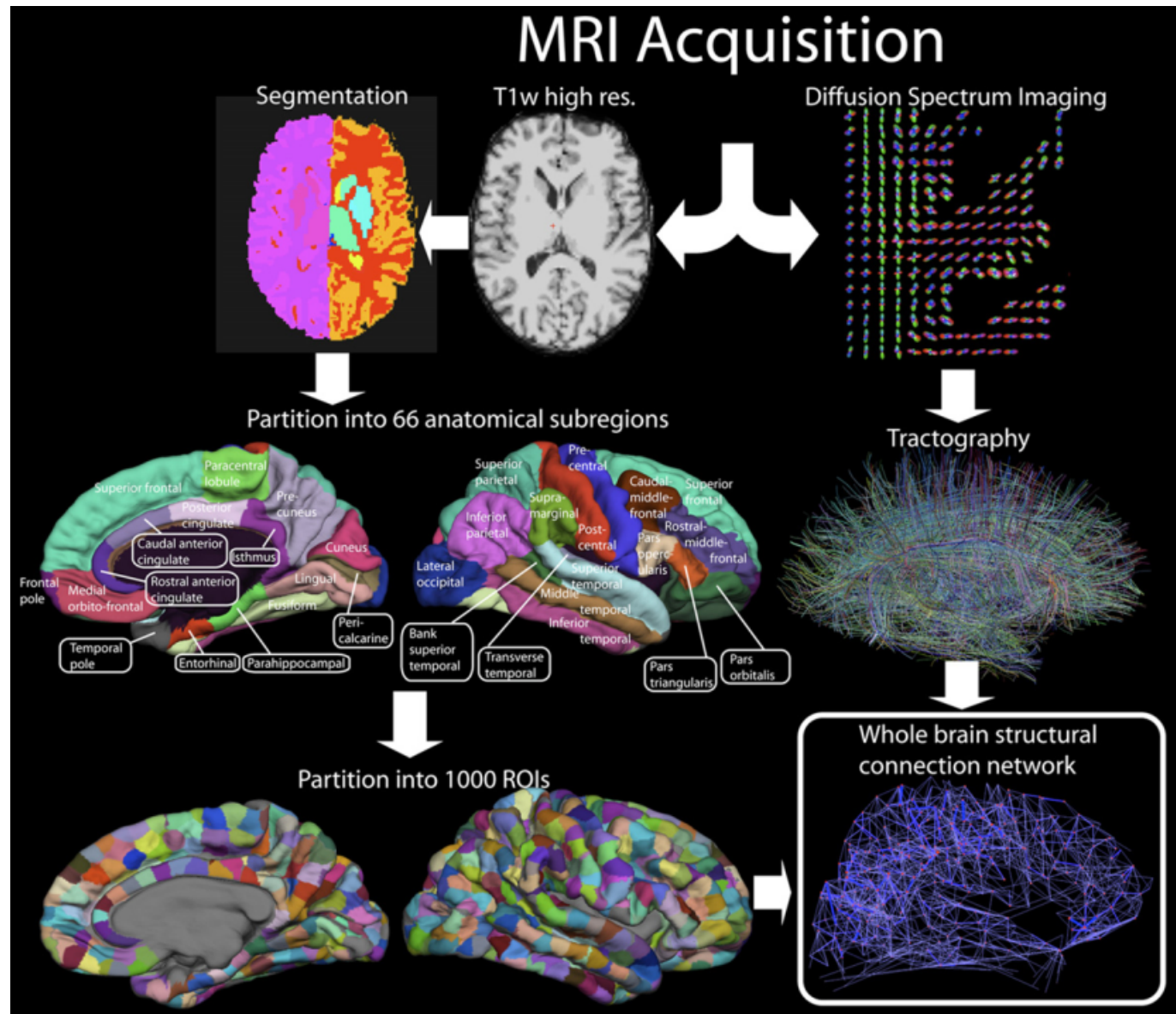
- **Functional**

- Computation & processing
- Decoding
- Behaviors



Cognitive Computing I, II, III (in Korean), B.-T. Zhang,
*Communications of the Korean Institute of Information Scientists
and Engineers*, **30**(1):75-111, 2012. [[PDF](#)]

Brain-Scale Structural and Functional Networks



Hagmann et al., 2010

Structural Principles: HiMACS

- **Hierarchical**
 - Layers, multiscale, deep
- **Molecular**
 - Chemical switch, distributed, massive associative
- **Aggregate**
 - Assemblies, microcircuits, cascades
- **Complex**
 - Heterogeneous, hubs, recurrent, convolutional
- **Sparse**
 - Distributed, modular, sparse population coding

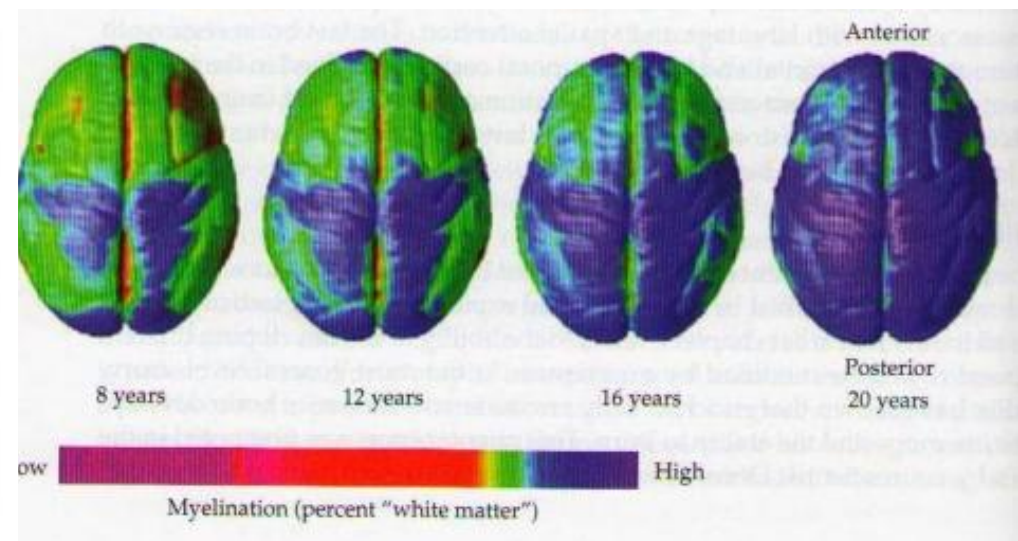
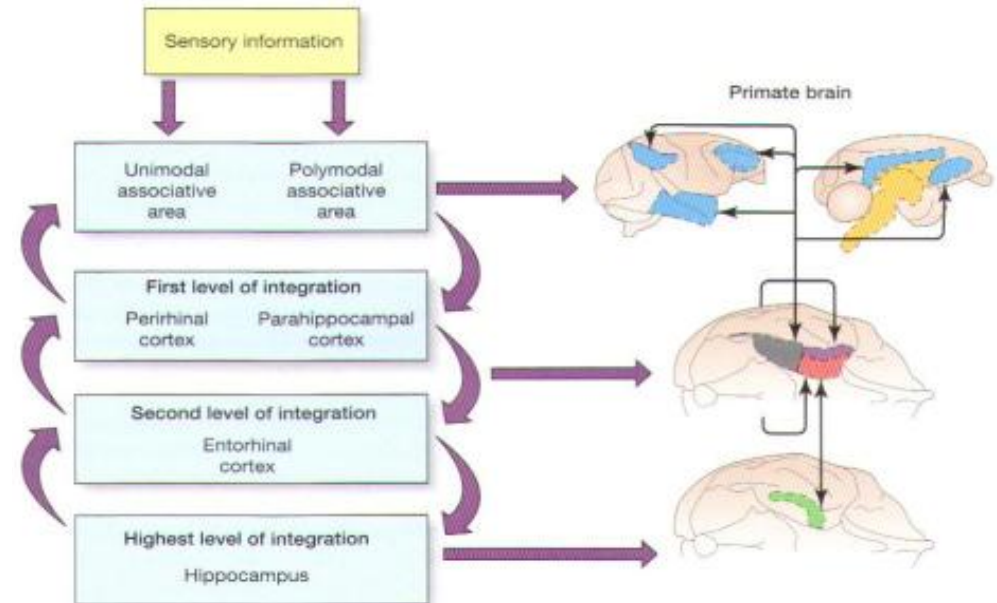
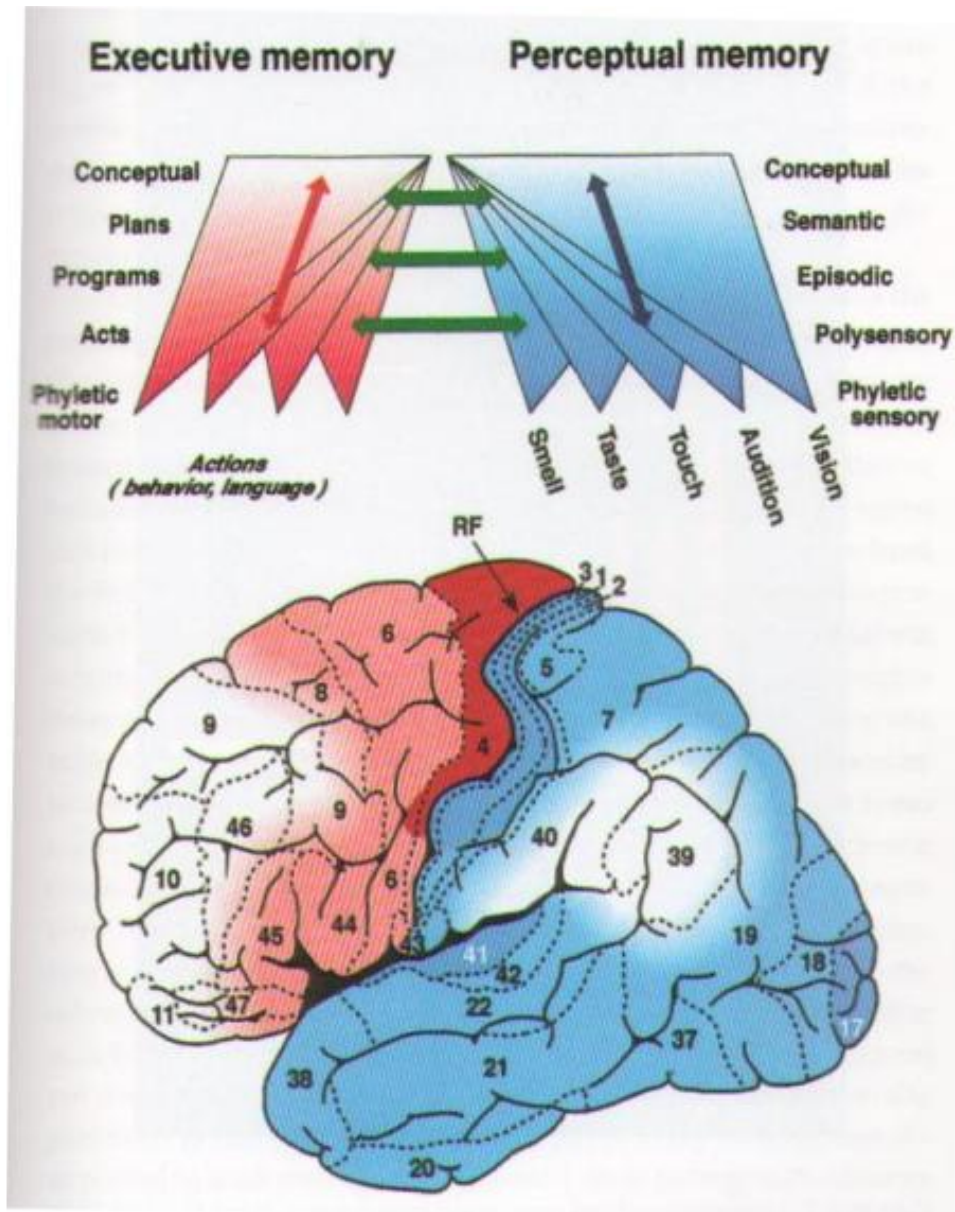
Organizational Principles: DevMIND

- **Developmental**
 - Growing, self-reproducing, adaptive
- **Mapping**
 - Topology-preserving map, transformation
- **Incremental**
 - Continual change, online
- **Nonequibratory**
 - Far from equilibrium, edge of chaos, asynchronous
- **Decompositional**
 - Disassembly, reassembly, reorganization

Functional Principles: DynaLIFE

- **Dynamic**
 - Perception-action cycle, interaction, feedback
- **Long**
 - Learning continually over time
- **Integrative**
 - Spatiotemporal integration, multisensors
- **Futuristic**
 - Anticipatory, internal model, goal, desire
- **Emergent**
 - Collective, population decoding

Brain as Widely Distributed, Parallel, Interactive, Overlapping, Dynamic Relational Memory Networks

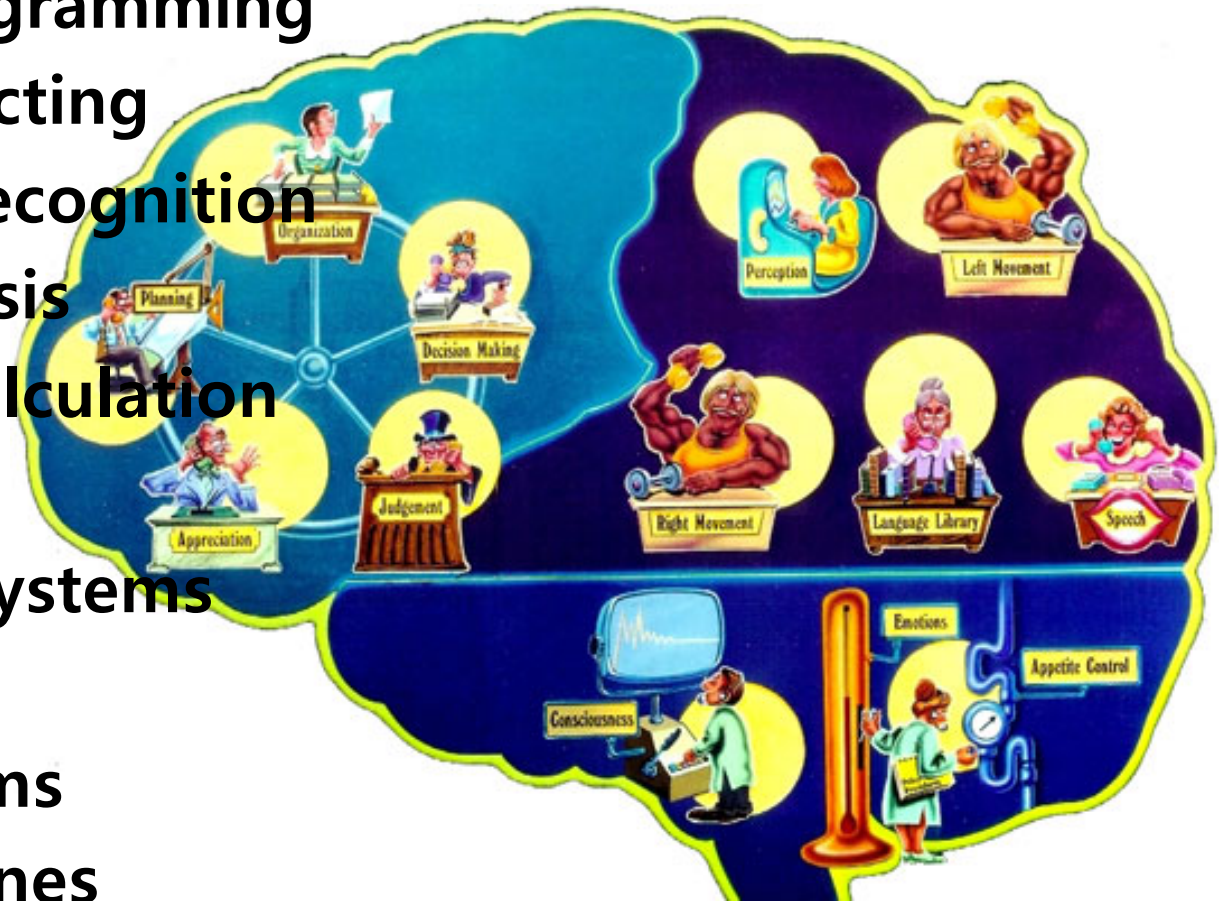


[Fuster, 2004]

Brain-Like AI

- **Organizing** vs. programming
- **Teaching** vs. instructing
- **Regeneration** vs. recognition
- **Synthesis** vs. analysis
- **Construction** vs. calculation

- Lifelong learning systems
- Growing machines
- Anticipatory systems
- Imagination machines
- Intentional systems



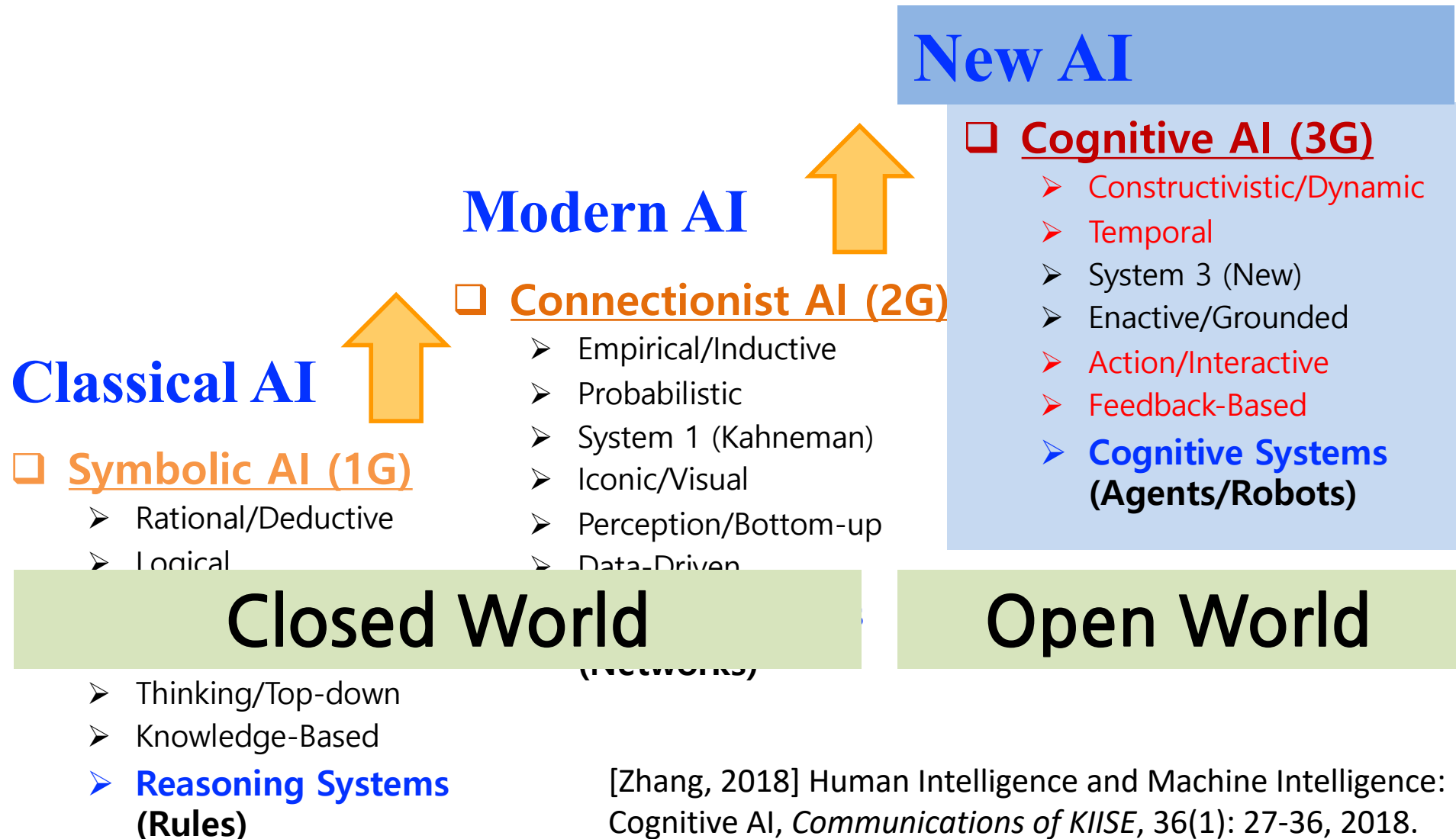
Cognitive Computing I, II, III (in Korean), B.-T. Zhang,
*Communications of the Korean Institute of Information
Scientists and Engineers*, 30(1):75-111, 2012. [[PDF](#)]

3. Brain-Like AI

Human-Level AI and the Brain

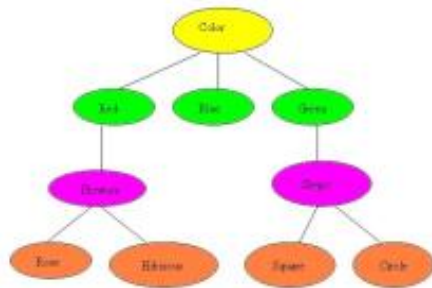
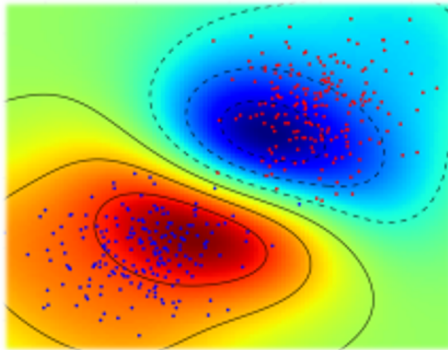
- **Human intelligence** involves solving problems by sequentially acting in an uncertain world to achieve a long-term goal. *It's not just a single perception-to-action mapping.*
- **Human-level AI** requires goal-directed autonomous cognitive capability to continually perceive, plan, act, and learn about the world.
- This is exactly what the brain does best.

The New AI



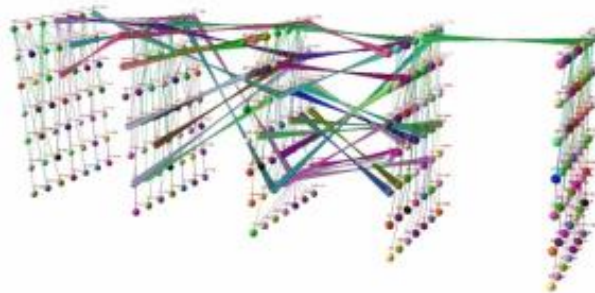
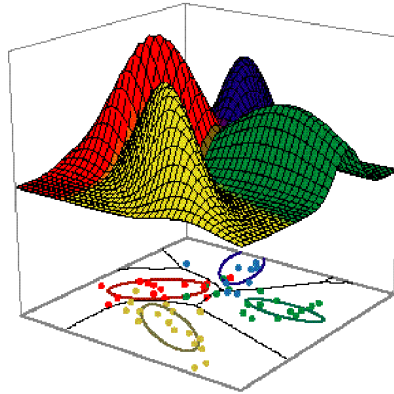
Autonomous Learning

1G: Supervised Learning (1980~2000)



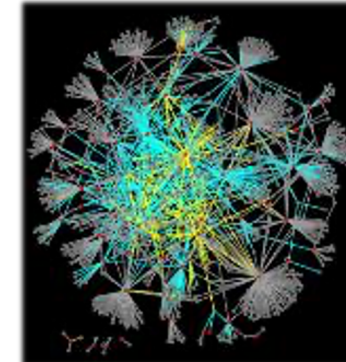
- Decision Trees
- Kernel Methods
- Multilayer Perceptrons

2G: Unsupervised Learning (2000~2020)



- Deep Networks
- Markov Networks
- Bayesian Networks

3G: Autonomous Learning (next generation)



- Learning by Doing
- Perception-Action Cycle
- Recursive Self-improvement

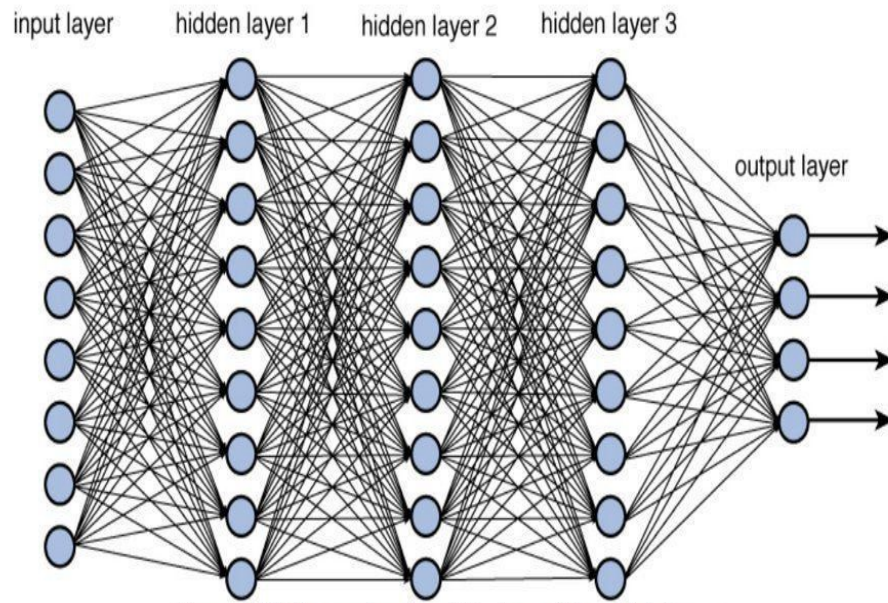
AI의 단계: L1~L6

L1	인간 프로그래밍 (Human Programming)	Tool	
L2	자동 프로그래밍 (Automatic Programming)	Oracle	We are here
L3	자기 교사 (Self-teaching)	Genie	} Brain-like AI
L4	자기 반성 (Self-reflection)	Sovereign	
L5	인간수준 학습 (Human-Level Learning)	AGI (HLAI)	
L6	초인간 학습 (Superhuman Learning)	Superintelligence	

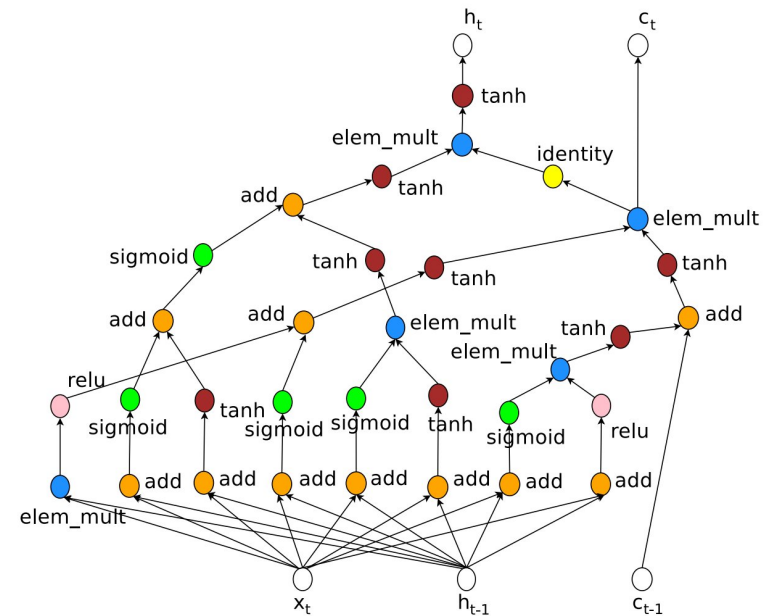
[장병탁, 인공지능의 미래-슈퍼인텔리전스, 2021-1-23]

AI Level 2

- ❖ AI가 스스로 알고리즘을 작성한다.
- ❖ 어떤 질문에든 대답해주는 **오라클(Oracle)**
- ❖ 자동 프로그래밍(Automatic programming)
- ❖ Automated deep learning (AutoDL)



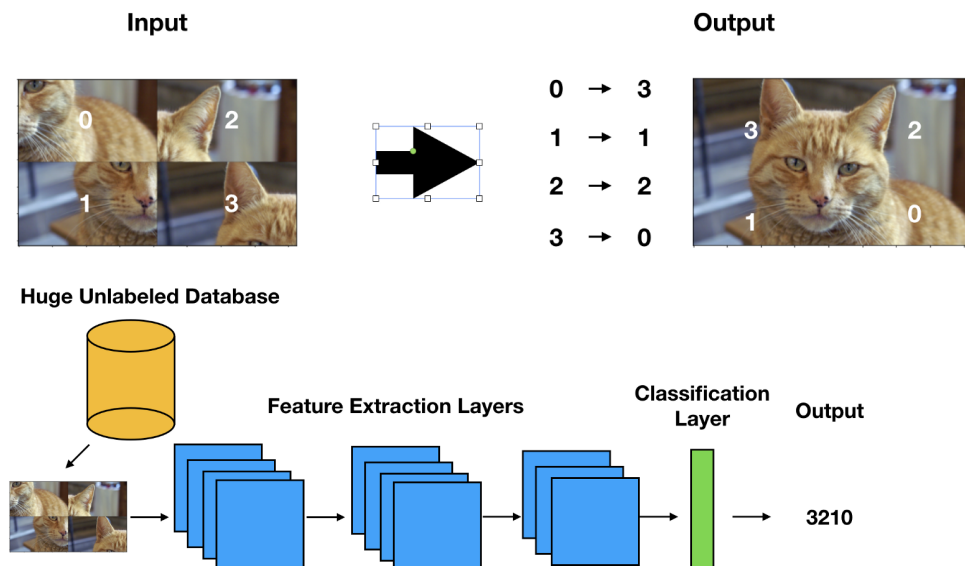
Deep Neural Networks



Neural Architecture Search

AI Level 3

- ❖ AI가 스스로 학습 데이터를 생성한다.
- ❖ 어떤 명령이든 실행하는 지니(Genie)
- ❖ 자기감독/자기교사 학습(Self-supervised learning, Self-teaching)
- ❖ 지속적 학습(Continual learning)



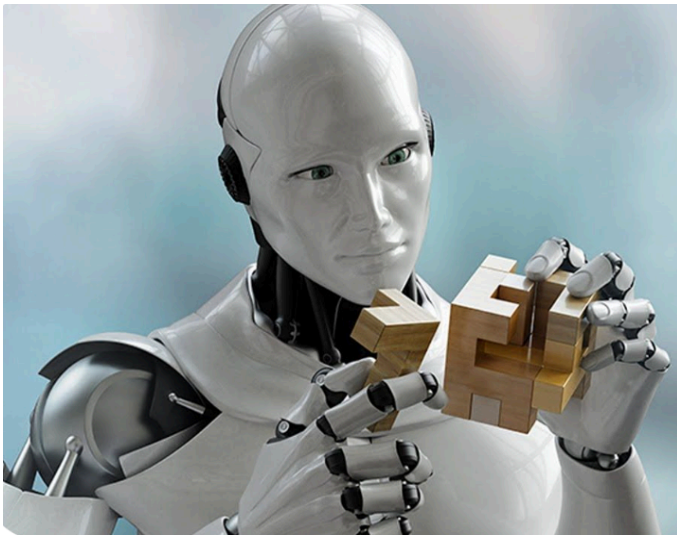
Self-supervised Learning



Self-teaching

AI Level 4

- ❖ AI가 스스로 목표함수를 수정한다.
- ❖ 자주적 개방형 작동 시스템 **소버린(Sovereign)**
- ❖ 자기반성 학습(Self-reflective learning)
- ❖ 재귀적 자기개선 시스템 (Recursively self-improving systems) **“Seed AI”**



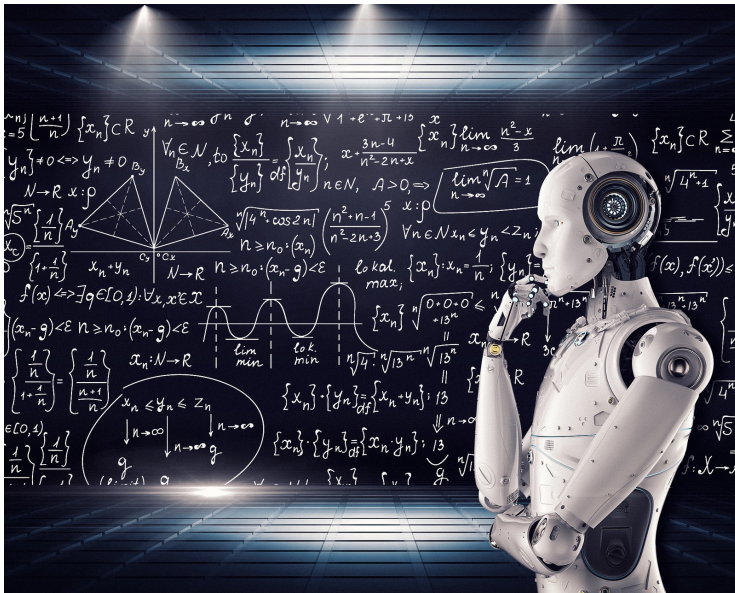
Autonomous Learning



BabyMind Project

AI Level 5

- ❖ AI가 스스로 목적과 미션을 조정한다.
- ❖ 인간수준 인공지능(Human-level AI)
- ❖ 인공일반지능(AGI)



AGI



SNU Aupair 홈로봇

AI Level 6

- ❖ L5의 수행에 인간을 능가한다.
- ❖ 수퍼지능(Superintelligence)
- ❖ 초인간 학습 기계(Superhuman learning machines)



영 화 Robot & Frank (2012)

4. Future of Digital Brain

Human-Level AI & Autonomous Cognitive Systems



Future of Digital Brain

