



Machine Learning and AI-IC *Applications of AI-ICs*

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Definition/Classification of AI



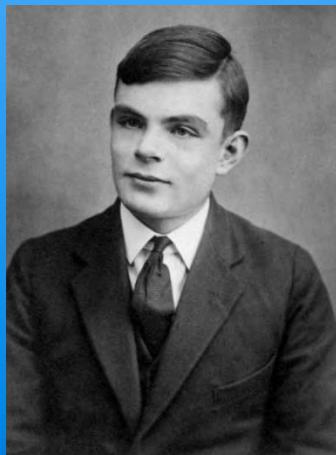
- What is AI? Merriam-Webster Dictionary:
 - “An area of computer science that deals with giving machines the ability to seem like they have human intelligence”
 - “The power of a machine to copy intelligent human behavior”
- How AI is classified?
 - Artificial Weak/Narrow Intelligence (**ANI**)
 - ◆ Focuses on improvement of individual ability, e.g. Siri
 - Artificial General Intelligence (**AGI**)
 - ◆ On humankind, human's brains, e.g. TrueNorth
 - Artificial Superintelligence (**ASI**)
 - ◆ Smarter than human brains, including innovation, recognition and social



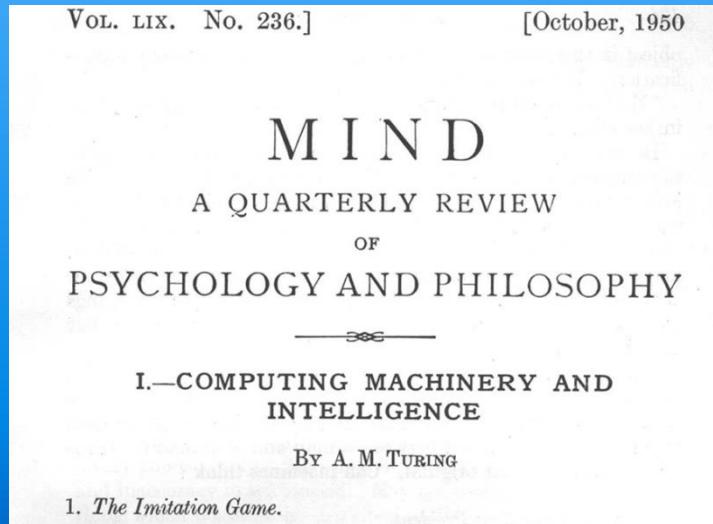
Alan Turing and AI

the "Nobel Prize of computing"

- ACM: A.M. Turing's Award By Year (since 1966)
- Turing is widely considered to be *the father of theoretical computer science and artificial intelligence.*



Turing aged 16



The Birth of AI (1952-56)

John McCarthy (Stanford)

Marvin Minsky (MIT)

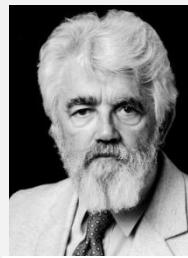
Trenchard More (IBM ret'd)

Ray Solomonoff (London)

Oliver Selfridge (MIT)



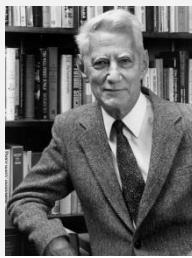
Dartmouth Summer Research Project on Artificial Intelligence 1956



John McCarthy,
"AI" 1955



Marvin Minsky,
MIT AI Lab



Claude Shannon,
MIT Boolean alg.



Ray Solomonoff,
Inductive Inference



Allen Newell,
Turing 1975



Herbert Simon,
Nobel78, Turing75



Arthur Samuel,
"ML" 1959



Oliver Selfridge,
Machine Perc.



Nat Rochester (IBM
701); Trenchard More



Julian Bigelow,
IAS/MANIAC

Source: https://en.wikipedia.org/wiki/Dartmouth_workshop

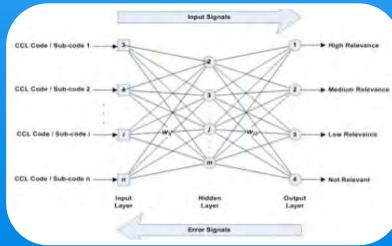
The Past 60+ Years of AI

“The First Wave
of AI (1956-76)”



*The Golden Years of AI (1956-74);
H. Simon & A. Newell 1975 Turing*

DNN Algorithm,
Backpropagation
(1986);



*PC Market ; IBM-Deep Blue
1997 & Jeopardy 2011*

“Winter Seasons
of AI (1976-05)”



Next Step:
BP? Capsule?

Machine Learning and AI-IC

Machine Learning Methods 

Machine Learning and Deep Learning 

CPU and GPU in AI-IC 

Applications of AI-IC 

Discussion 

Views of AI/ML/DL

Can AI Replace Human Beings?

Artificial
Intelligence
=
Machine Learning
=
Deep Learning

TO THE GENERAL PUBLIC

Artificial
Intelligence
Machine Learning
Deep Learning

TO AI RESEARCH STAFF

ML: Schools? Algorithms?

Schools of Machine Learning



- Symbolic (e.g. *Frank Rosenblatt, 1957*)
 - *logicism, aka logicism, psychology, or computerism*
- **Connectionism** (e.g. *Geoffrey Hinton, 1986 Nature*)
 - Aka *bionicsism or physiologism*
- Actionism
 - Aka *evolutionism or cyberneticism*
- **Probabilistic Graphical Models**
 - *E.g. Bayes network, Random mean forest*



Key Aspects of ML, Algorithms

● Types of ML Algorithm

- Supervised learning
 - Regression, KNN, SVM, Boosting (Ada, X-G), Decision Tree, Random Forest etc.
- Semi-supervised learning
- Unsupervised learning
 - Clustering (e.g. K-means, GMM), Dimensionality Reduction, PCA, ICA, etc.
- Reinforcement learning

Machine Learning and AI-IC

Machine Learning Methods 

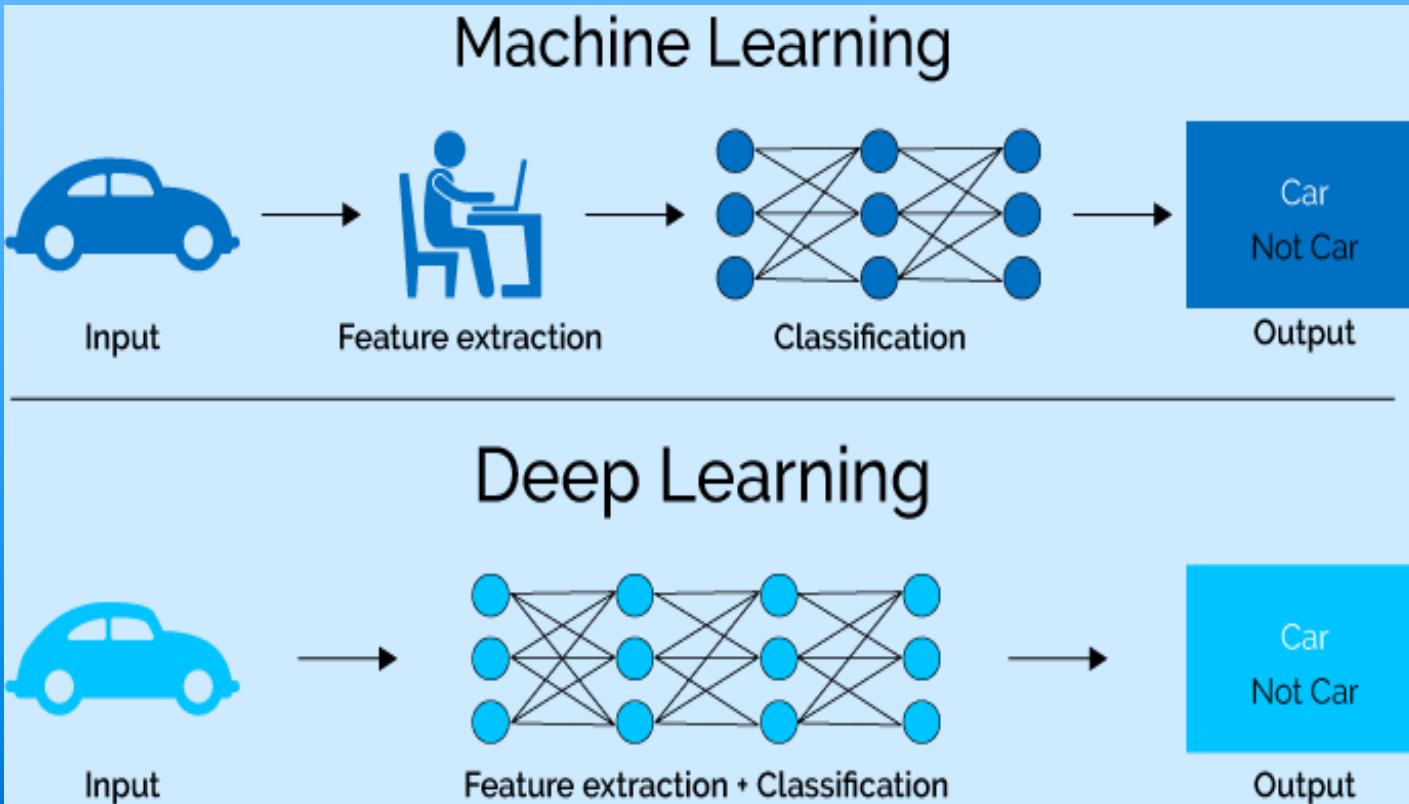
Machine Learning and Deep Learning 

CPU and GPU in AI-IC 

Applications of AI-IC 

Discussion 

Comparison of Learning Flow in ML and DL



Types of Deep Learning Algorithms

- Artificial Neural Network
 - ANN,
 - *Artificial Neural Network*
 - FNN,
 - Feedforward Neural Network
 - CNN,
 - Convolutional Neural Network
 - *Cellular Neural/Nonlinear Network*
 - RNN,
 - Recurrent Neural Network
 - LSTM, modified RNN
 - Transformer(s) from Google
- GAN,
 - Generative Adversarial Network

2018 ACM A.M. Turing Award

- Yoshua Bengio,
 - Professor at the University of Montreal and Scientific Director at Mila, Quebec's Artificial Intelligence Institute
- Geoffrey Hinton,
 - VP & Engineering Fellow of Google, Chief Scientific Adviser of The Vector Institute, and Univ. Prof. Emeritus at Univ. Toronto
- Yann LeCun,
 - Professor at New York University and VP and Chief AI Scientist at Facebook

ML/DL Applications

● ML Applications

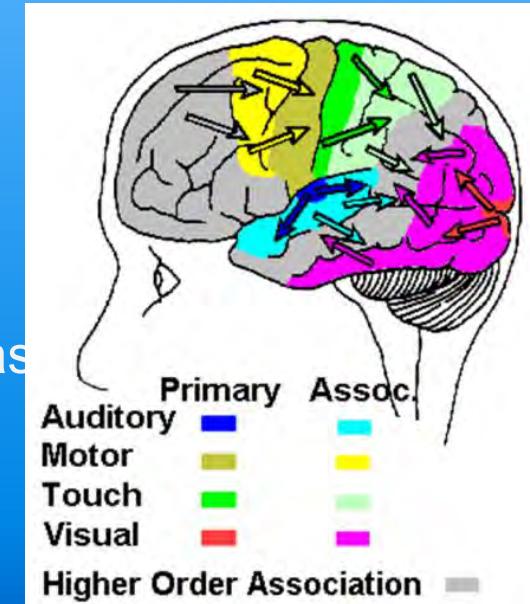
- Image Recognition. One of the most common uses of machine learning
- Speech Recognition. SR the translation of spoken words into text.
- Medical Diagnosis. ML provides methods, techniques, and tools that can help solving diagnostic...
- Statistical Arbitrage.

● DL Applications

- Self-driving cars
- Deep Learning in Healthcare
- Voice Search & Voice-Activated Assistants
- Automatic Colorization of Black and White Images.
- Automatically Adding Sounds To Silent Movies
- Automatic Machine Translation

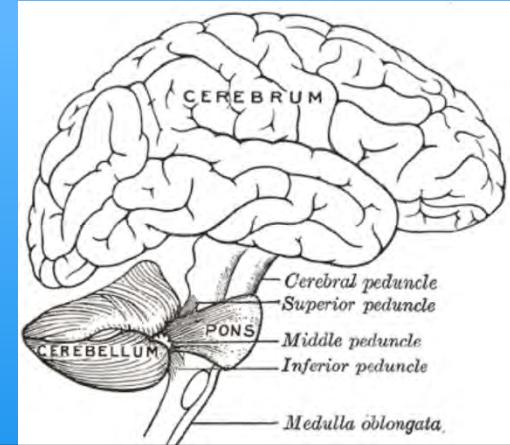
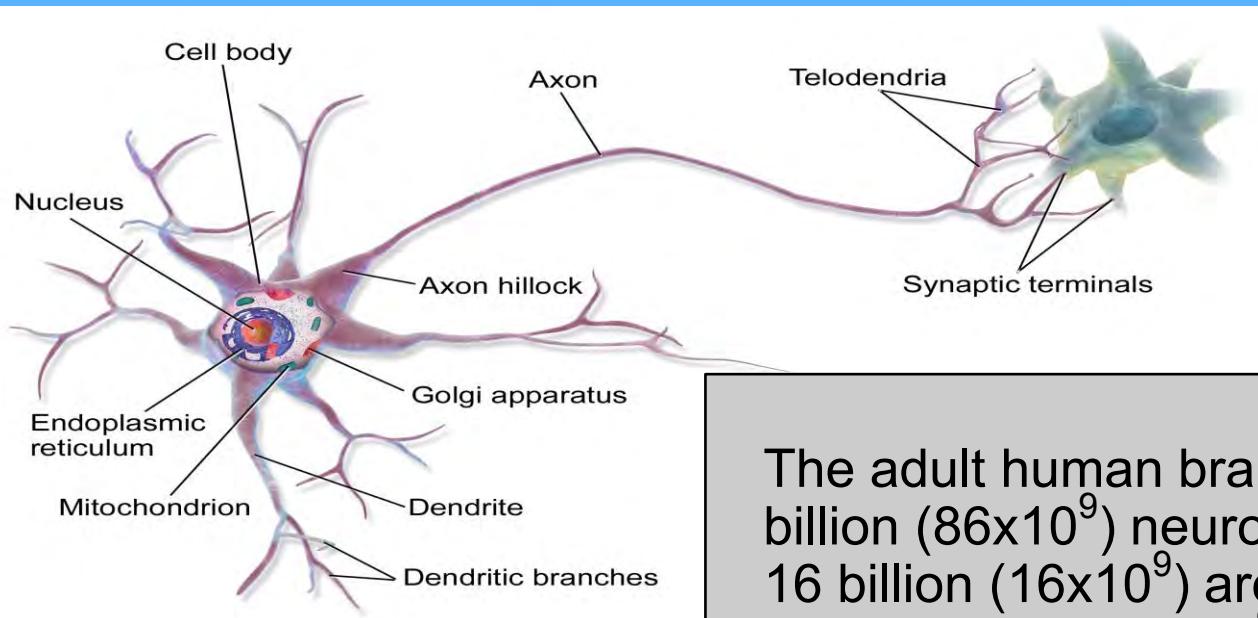
In Human Cerebrum

- Functional areas
 - Frontal lobe
 - Thinking/plan/short mem.
 - Parietal lobe
 - Touch/smell/taste
 - Occipital lobe
 - Visual activity
 - Temporal lobe
 - Memories
- Under cerebral cortex (*Primary, Association*)
 - Auditory,
 - Motor,
 - Touch,
 - Visual,
 - High order as



Neurons and Synapses (Neural Network)

● Human brain: Cerebrum, Cerebellum



The adult human brain contains about 85-86 billion (86×10^9) neurons,^{[38][39]} of which 16 billion (16×10^9) are in the cerebral cortex and 69 billion (70×10^9) in the cerebellum.^[39]

Machine Learning and AI-IC

Machine Learning Methods 

Machine Learning and Deep Learning 

CPU and GPU in AI-IC 

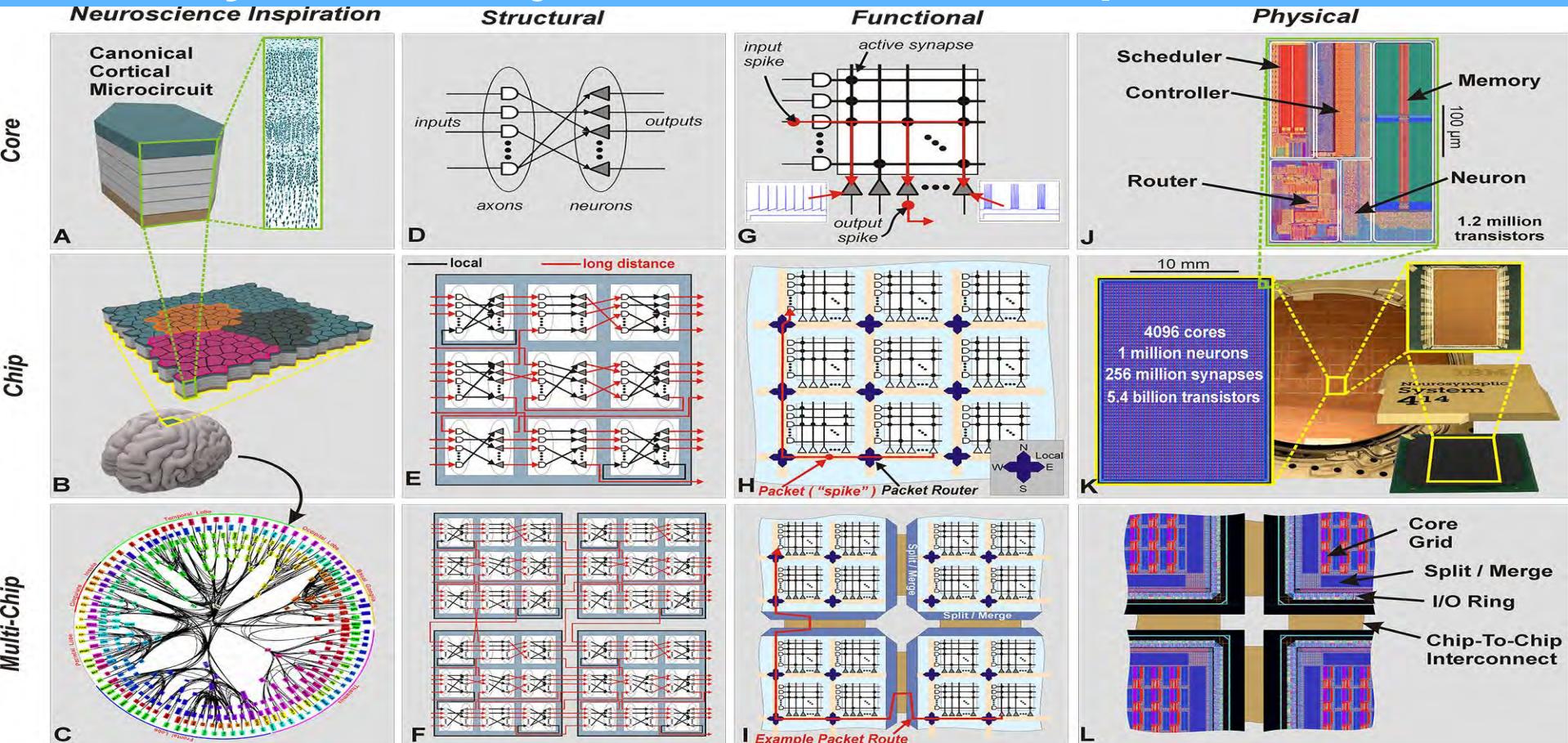
Applications of AI-IC 

Discussion 

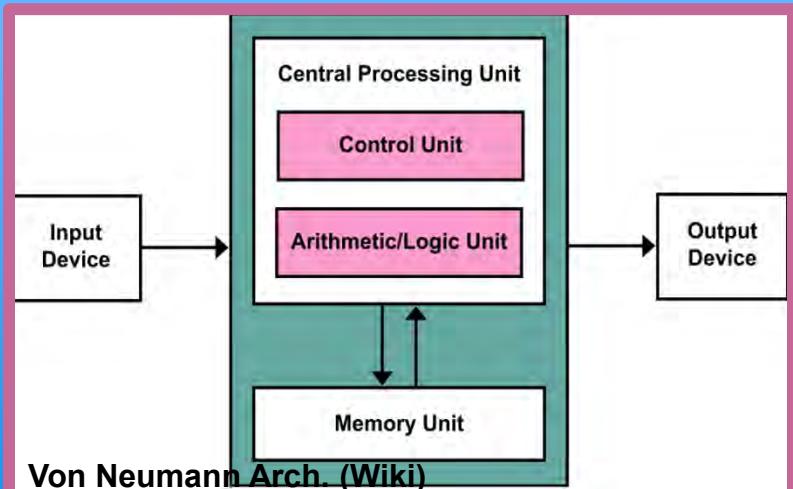
- A neuromorphic CMOS IC, TrueNorth chip
 - Many cores, 4096 cores, simulating a total $>10^6$ neurons
 - The programmable synapses is $>268 \times 10^6$ (2^{28})
- Contains 5.4×10^9 transistors (Sg28nm)
 - At low T, 70 mW, about 1/10,000th of conventional MPU
- Application
 - SyNAPSE 16 chips for DARPA



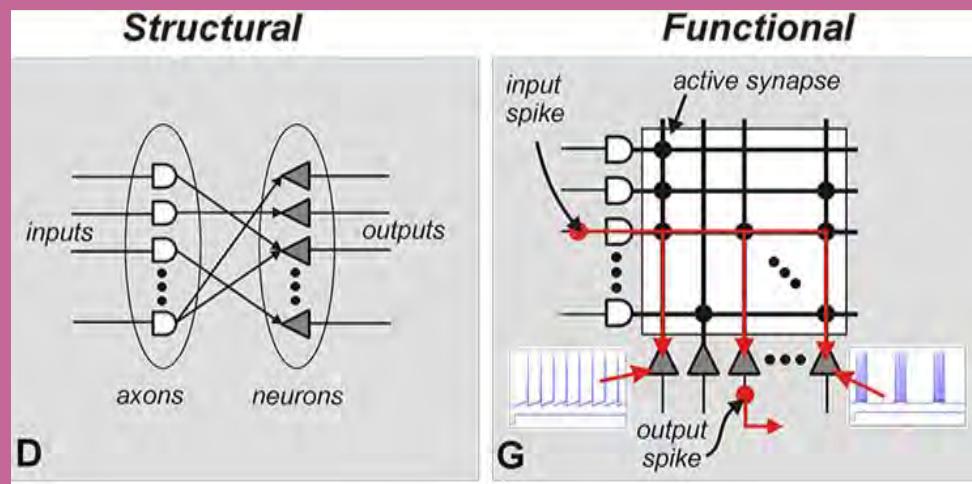
IBM SyNAPSE Project and TrueNorth Chip



CPU and von Neumann Bottleneck



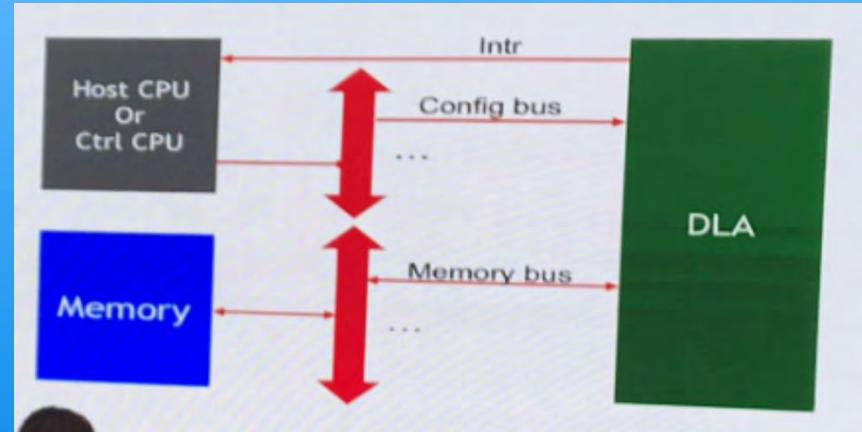
Von Neumann Computing
Serial Computing
Separate Memory
High Precision



Neuromorphic Computing
Highly Parallel
In Memory Computing
Tolerance to Low Precision

GPU for Deep Learning

- Nvidia, 1999 GeForce 256:
"the world's first GPU"
- a "single-chip processor with integrated transform, lighting, triangle setup/clipping, & rendering engines"
- GPU for DL and EC

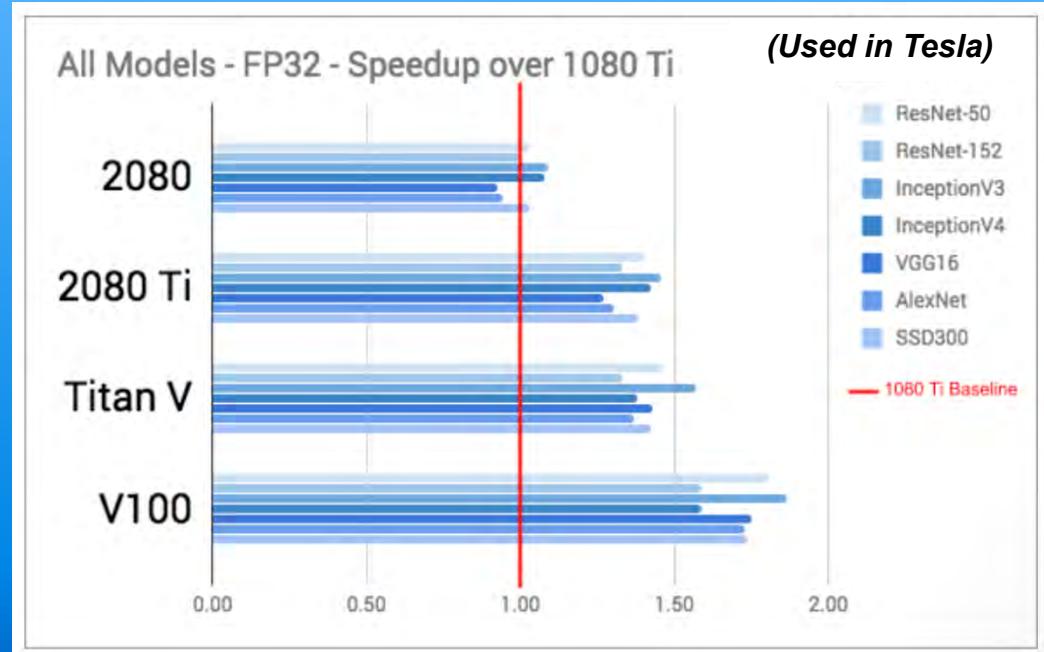


Nvidia DLA (Deep Learning Accelerator)
“DL Solution for Edge Computing”
(SiFive-RISC-V, 03/29/19 Shanghai)

GPU Benchmarks for DL

- ResNet-50
- ResNet-152
- Inception-V3
- Inception-V4
- VGG16
- AlexNet
- SSD300

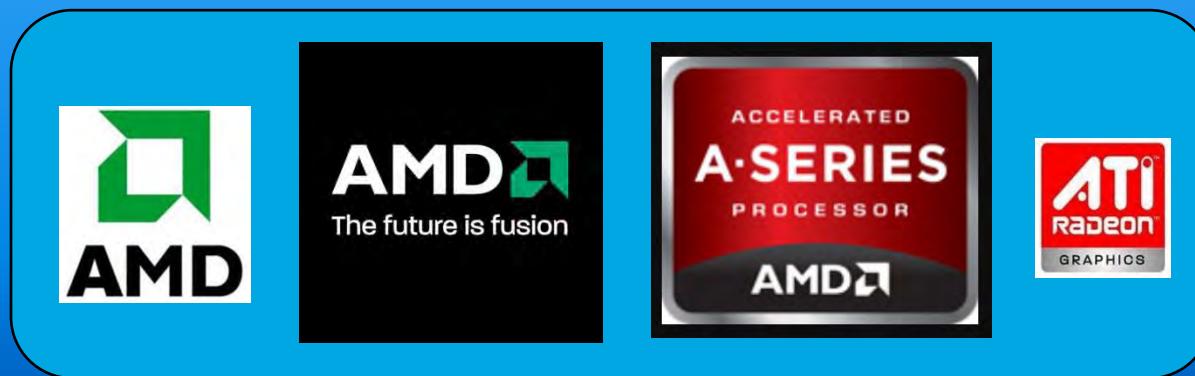
- GPU tested/used in *Tesla*



GPU from AMD



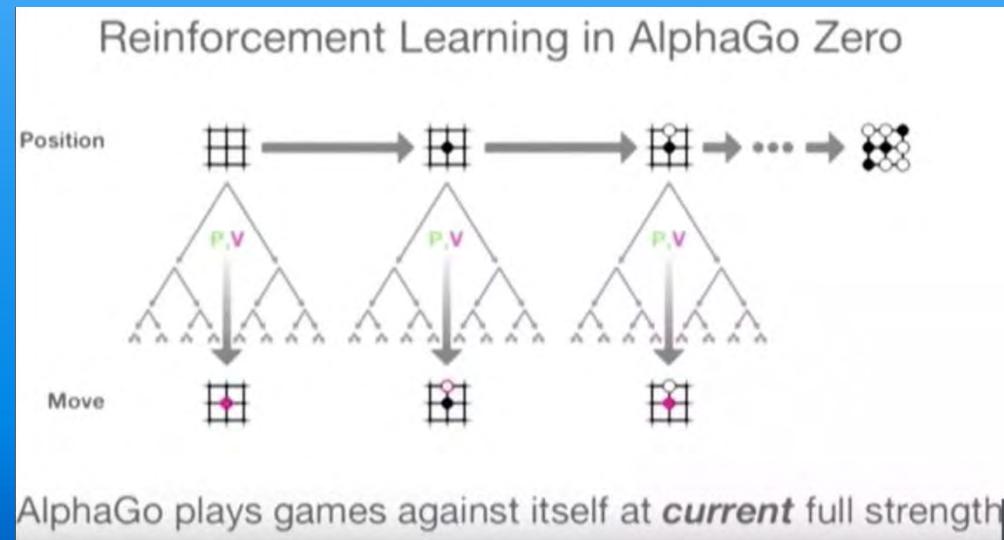
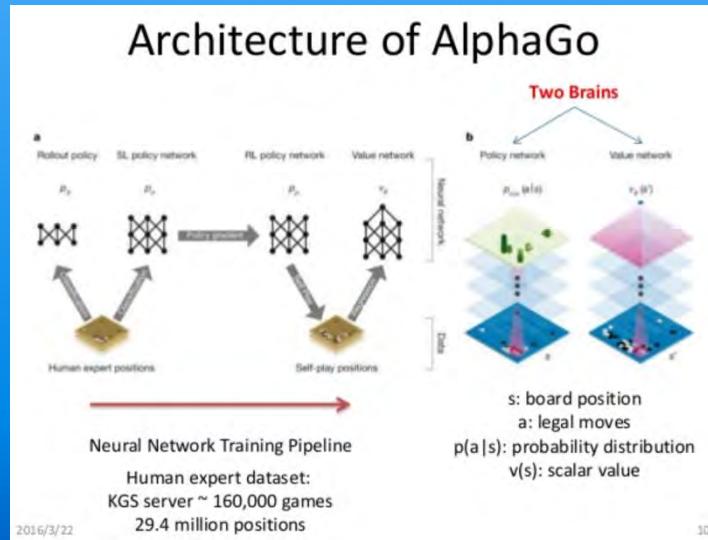
- Acquisition of ATI in 2006
 - CPU chips: A4, A6, A8
 - *Fusion* chip (CPU+GPU)
 - *Llano* (Accelerated) APU



AI-IC in AlphaGo/AlphaGo Zero

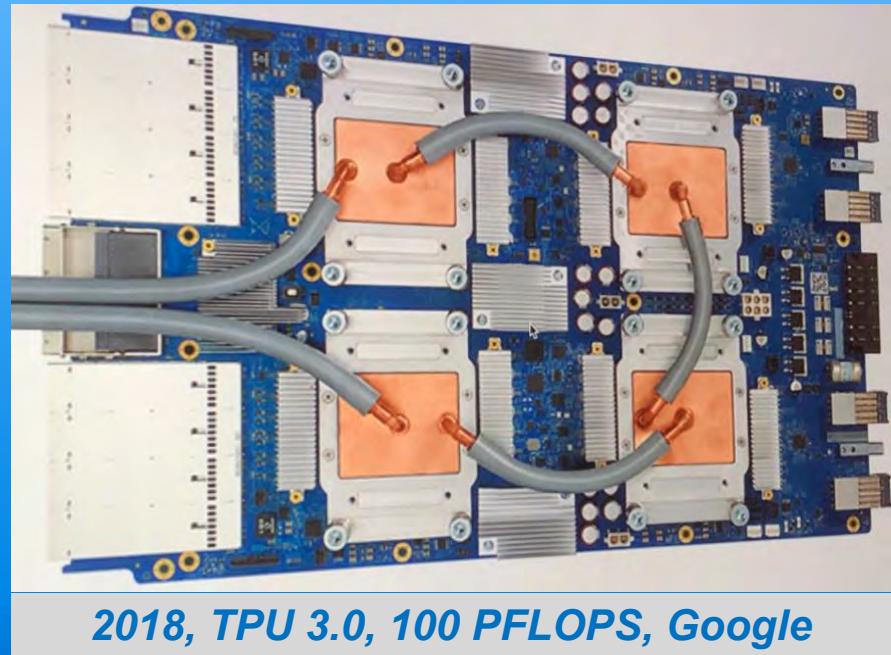


- Architecture based on CPU + GPU
 - AlphaGo (Oct. 15; Mar. 16; Mar. 17)
 - AlphaGo Zero (10/19/17)



TPU used in AlphaGo for ML

- Can be accessed from Cloud ● 2018 Edge TPU: IoT EC
- TPU in AlphaGo 2016 (Lee Sedol)
 - 2016 Gen1: CISC 8-bit, PCIe 3.0, 28nm, $\leq 331 \text{ mm}^2$,
 - 2017 Gen2: 4x16GB HBM 4x600GB/s, 180TFLOPS → 11.5PFLOPS
 - 2018 Gen3: 2x of Gen2, 1024 chips per pod, 28nm, 700MHz, 40W

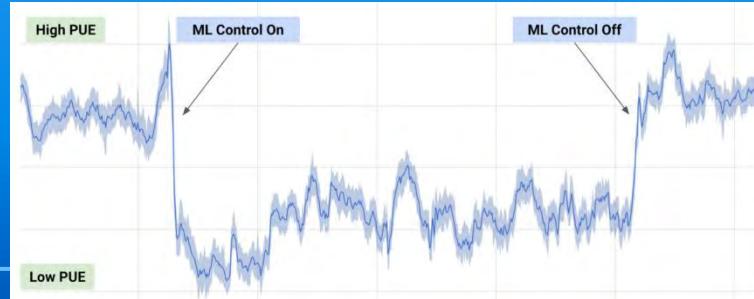


AlphaGo, AlphaStar, AlphaZero, AlphaFold

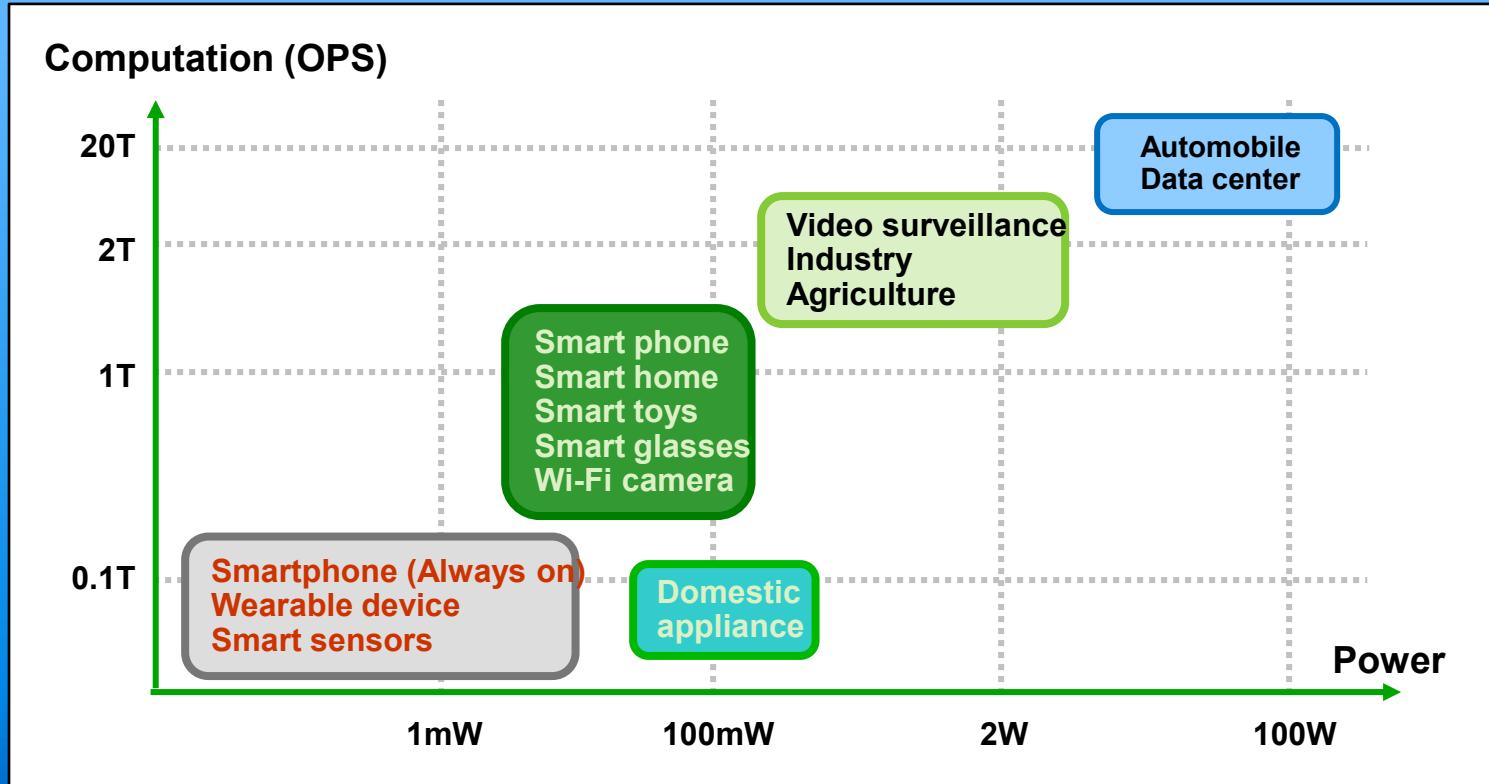
TPU used in TensorFlow



- TPU (announced at Google I/O, Mtn View))
 - TPU 1.0 (05/20/16); TPU 2.0 (05/23/17); TPU 3.0 (5/8-5/10/18)
- TensorFlow, ASIC (CPU+GPU)
 - <https://www.tensorflow.org/>
 - Feb 15, 2017, **TensorFlow 1.0** [09/27/16 → 11/06/16 → 02/15/17]
 - Nov 04, 2018, **TensorFlow 2.0**
- DeepMind,

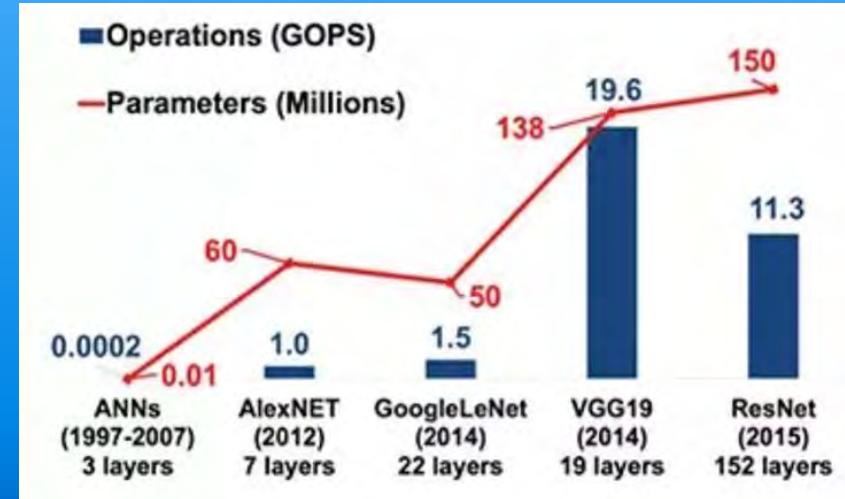


Computation Needs vs Power Limits

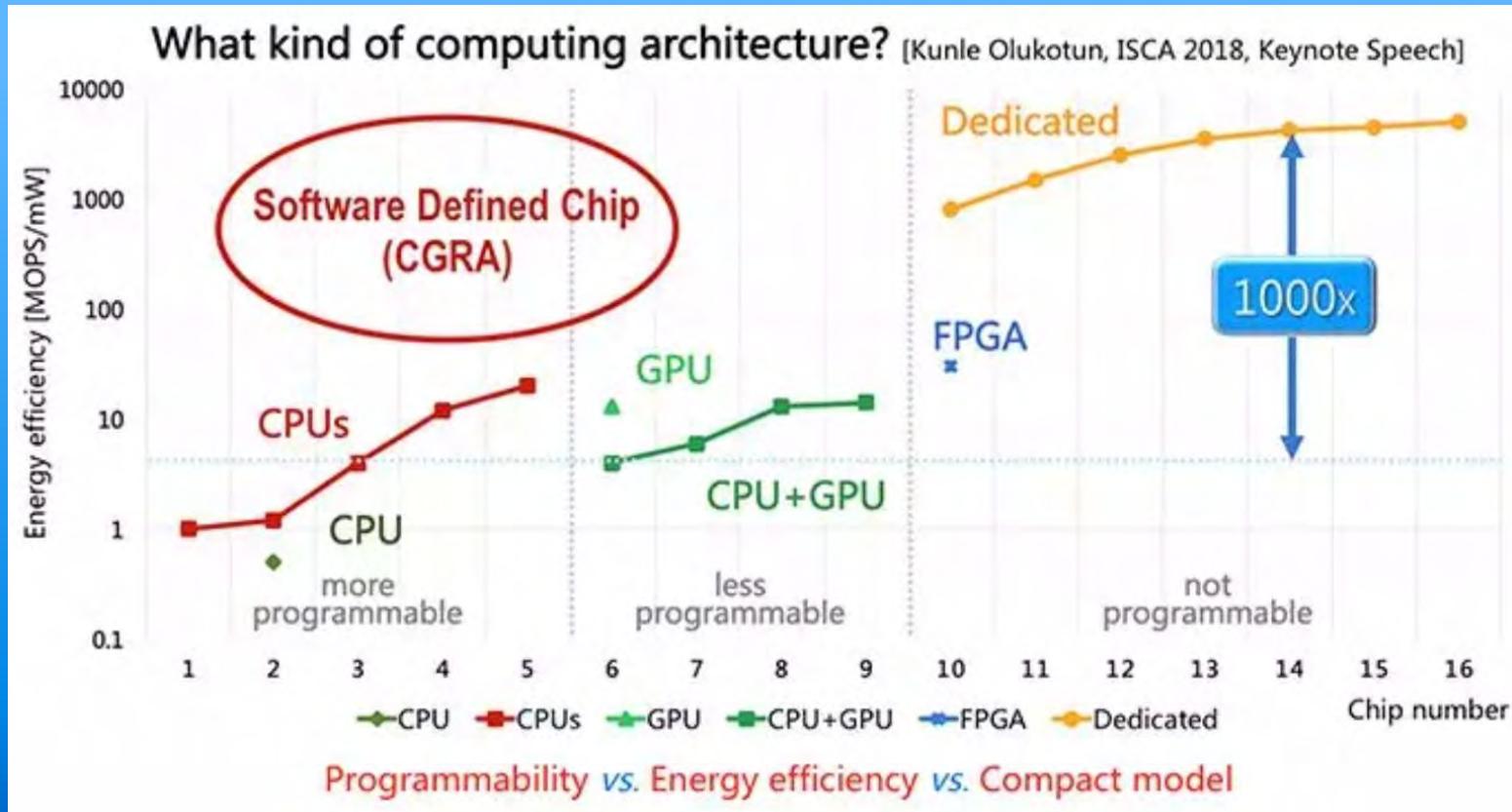


Features of AI-IC

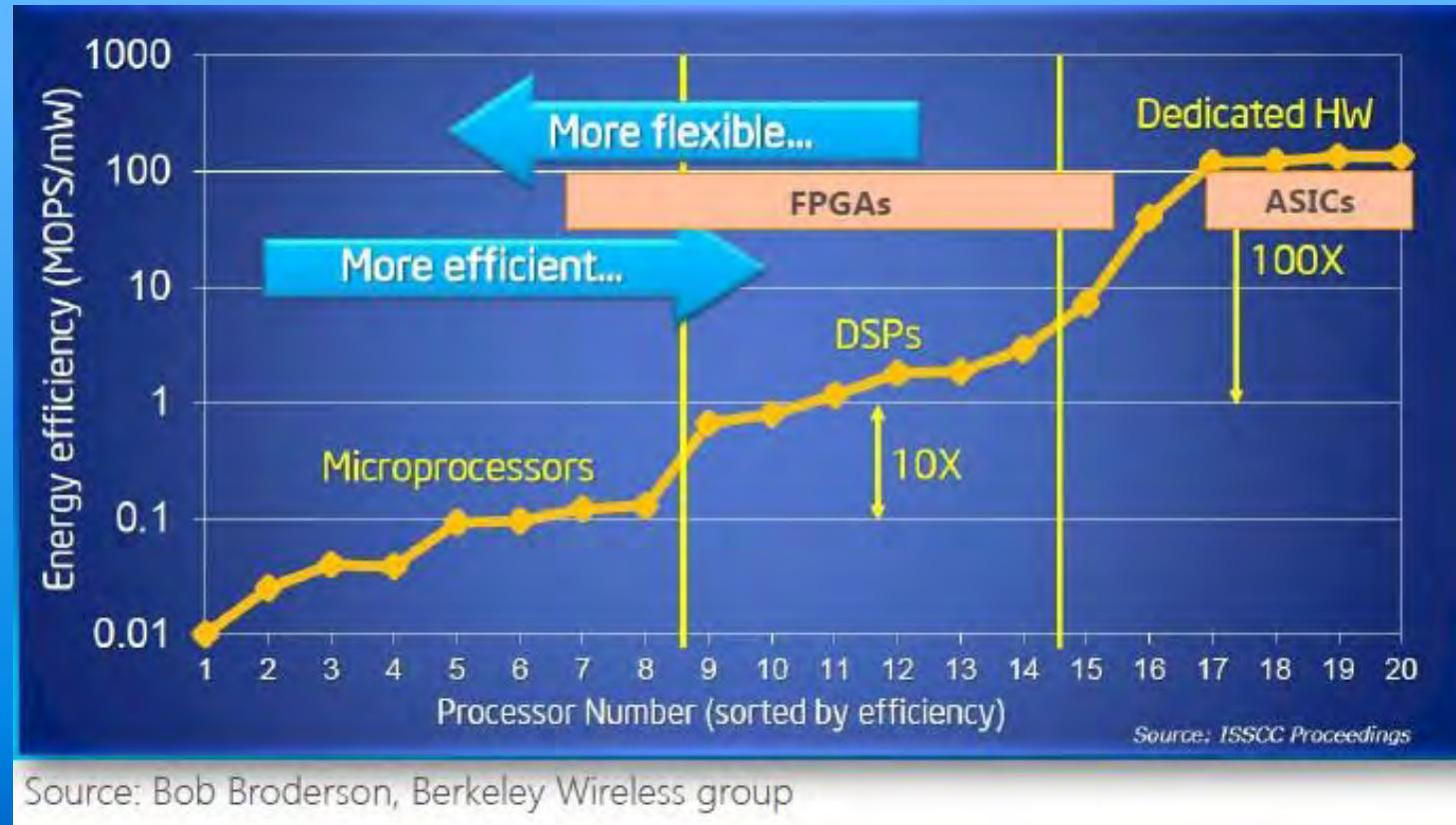
- Requirements for an AI-Chip
 - Programmability
 - **High Energy Efficiency**
 - Common use for DL
- CNN becomes most popular
 - Pattern (LeNet)
 - Image (AlexNet)
 - Vision (LRCN net)
- To meet **Edge Computing** with high Energy Efficiency
- OPS and Parameters



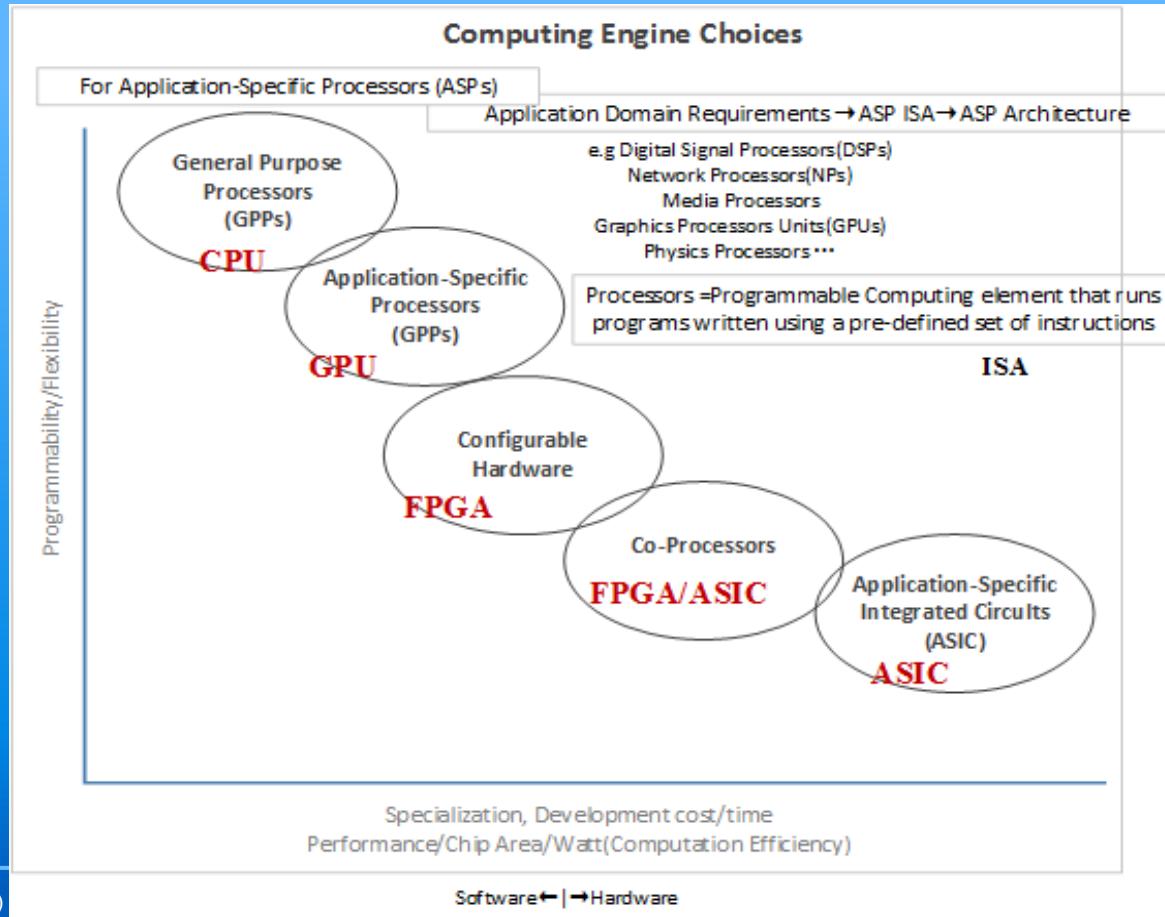
Progress of Energy Efficiency (OPS/W)



Comparison of Energy Efficiency (Processors)



Comparison of Computing Engines' Flexibility



Machine Learning and AI-IC

Machine Learning Methods 

Machine Learning and Deep Learning 

CPU and GPU in AI-IC 

Applications of AI-IC 

Discussion 

Applications of ML and HW

- Biomedical informatics

Computer vision

- Customer relationship management
- Data mining
- Email filtering
- Inverted pendulum

AlexNet
VGG
ResNet

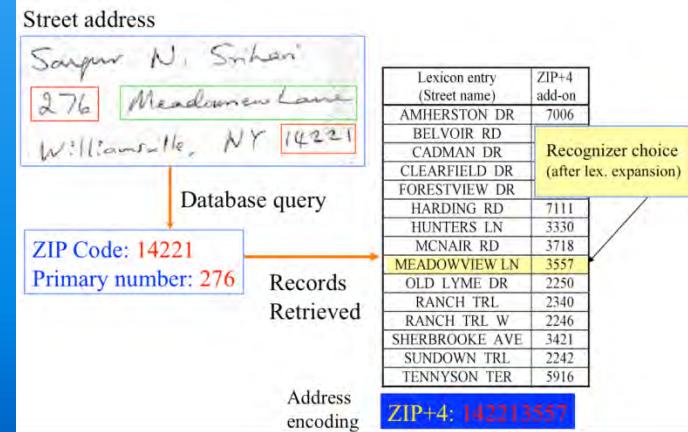
Natural language processing (NLP)

- Automatic ...
- translation ...

Pattern recognition

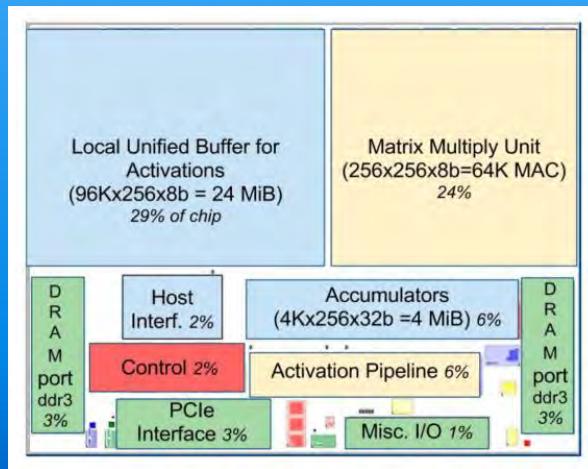
- Facial recognition system

- Handwriting recognition
- Image recognition
- Optical character recognition
- Speech recognition
- Recommendation system
- Search engine
- Social engineering

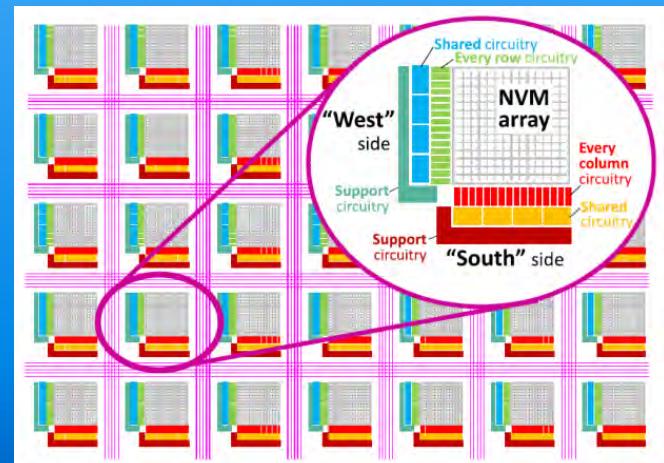
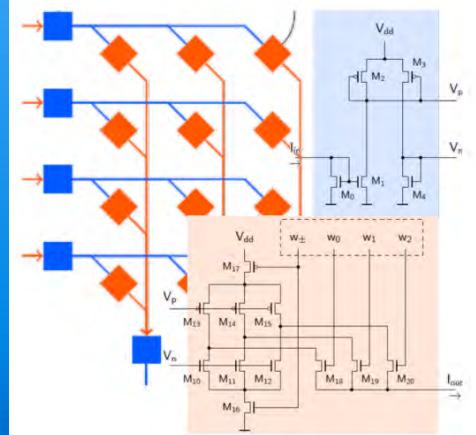


New Hardware for Inference and Training

- Digital
 - GPU, TPU, FPGA etc.
 - $10\text{-}10^3$ speedup
 - Power hungry, large
- Analog
 - Neuromorphic
- Beyond-Si
 - RRAM, STT etc.
 - $>10^3$ speedup
 - Small footprint, E efficient



Ref. Mohanty et al.
IEDM 2017



Apple Chip A Series

Year	2017	2018	2019	2020	2021
Generation	A11	A12	A13	A14	A15?
Process, nm	10	7	N7	5	
Chip area, mm ²	1.83	5.8	1.16		
TOPS	0.6	5	6	11	

WSE by Cerebras

	Cerebras WSE	Largest GPU	Cerebras Advantage
Chip size	46,225 mm ²	815 mm ²	56.7 X
Cores	400,000	5,120	78 X
On chip memory	18 Gigabytes	6 Megabytes	3,000 X
Memory bandwidth	9 Petabytes/S	900 Gigabytes/S	10,000 X
Fabric bandwidth	100 Petabits/S	300 Gigabits/S	33,000 X

Cerebras CS-1: A 15 RU System for Training and Inference in the Data Center



- Accelerates all deep learning models: CNN, RNN, LSTM, etc.
- Powered by an array of 400,000 Cerebras' AI optimized processor cores
- 18 GB on chip memory → 3,000 times more than a graphics processing unit
- > 9PB/s on-die memory bandwidth → 10,000 more than a graphics processing unit
- 100 Pb/s total interconnect bandwidth → 33,000 times more than a graphics processing unit
- System IO: 12 x 100 GbE
- System power: 20 KW;
- Programmed with TensorFlow, PyTorch, Mxnet, Caffé2, Theano, CNTK



More Compute than Up To 1,000 GPUs
1/40th the space, 1/50th the power



Cerebras WSE-2

- WSE Gen1 16nm, Aug 2019
- WSE Gen2 7nm, Q3 2021
 - 850,000 AI Cores
 - 2,600B transistors, 56 mTr/mm²
 - Memory 40GB, 20 PB/s
 - Fabric 220 Pb/s



WSE-2 vs Largest GPU

IC	WSE	WSE-2	Nvidia A100
Area	46,255mm ²	46,255mm ²	826mm ²
Transistors	1.2 trillion	2.6 trillion	54.2 billion
Cores	400,000	850,000	6,912 + 432
On-chip memory	18GB	40GB	40MB
Memory bandwidth	9PB/s	20PB/s	1,555GB/s
Fabric bandwidth	100Pb/s	220Pb/s	600GB/s
Fabrication process	16nm	7nm	7nm

Machine Learning and AI-IC

Machine Learning Methods 

Machine Learning and Deep Learning 

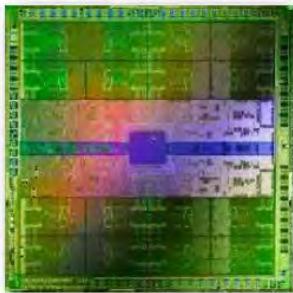
CPU and GPU in AI-IC 

Applications of AI-IC 

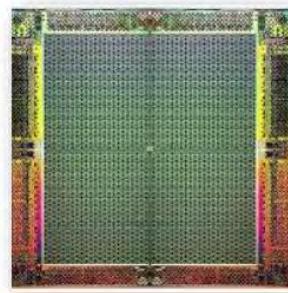
Discussion 

Hardware Acceleration Platforms

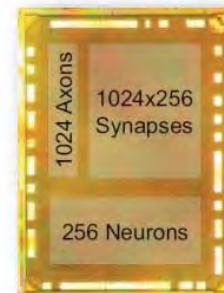
- CPU, 1X
- GPU, FPGA, ASIC, Beyond CMOS



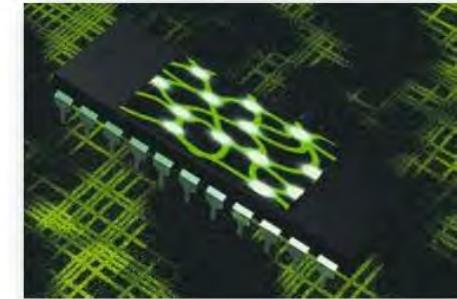
GPU
10 – 30 X



FPGA
10 – 50 X



CMOS ASIC
 $10^2 – 10^3 X$



Beyond CMOS
 $>10^3 X$

Comparison of AI-Chip Features

GPU, FPGA, ASIC, Neuromorphic

Type Feature	GPU	FPGA	ASIC	Brain-inspired
Customization	General	Semi-custom	Customized	Neuromorphic
Programma- bility	No	Easy	Difficult	No
App scenario	Cloud train.	Acc., D Ctr, Infer.	Widely used	Comp. recog. Envir.
Vendor	Nvidia	Xilinx, Altera	Google, Cambricon	IBM
Advantages	Peak comp., mature	Perf. Power, prog., fast	Av. Perf., power, size	Power, comm., recog
Disadv.	Effic., prog., power	Cost, peak comp.	NRE, R&D cycle, risk	Immature