

# Valeriu Beiu







# Structure of the presentation

Intro

• Part 1 Al from the early beginnings

Part 2 Follow the money

Part 3 Al hardware (growing land of the giant Al chips)

• GPU Nvidia <a href="https://www.nvidia.com/en-us/">https://www.nvidia.com/en-us/</a>

• TPU Google <a href="https://ai.google/">https://ai.google/</a>

• IPU Graphcore <a href="https://www.graphcore.ai/">https://www.graphcore.ai/</a>

• WSE Cerebras <a href="https://www.cerebras.net/">https://www.cerebras.net/</a>

Quantum approach (Google, IBM, and many others)

Fast ... and, hopefully ... green



### Kaissa

1974: first world computer chess champion



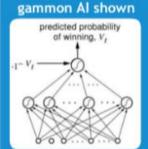
### Mac Hack

1967: chess Al beats person in tournament

History of Game Al By: Andrey Kurenkov

#### TD-Monte Carlo Go Gammon

1992: RL and neural 1993: first research on Go with stochastic net based backsearch



### **MCTS Go**

2006: French researchers advance Go Al with MCTS

### NeuroGo

1996: ConvNet with RL for Go, 13 kyu (amateur)

### Crazy Stone

2008: MCTS Go Al beats 4 dan player

### Zen19

2012: MCTS based Go Al reaches 5-dan rank

### Samuel's Checkers Al

Dartmouth

Conference

1956: the birth of Al

1956: IBM Che kers Al first demons rated

### Bernstein's Chess Al

1958: first fully functional chess Al developed

### Zobrist's Al

1968: First Go Al, beats human amateur

### Checkers Al Wins

1962: Samuel's program wins game against person



### CNN

1989: convolutional nets first demonstrated

### Backprop

1986: multi-layer neural net approach widely known

### CHINOOK

1994: checkers Al draws with world champion



### Deep Blue

1997: IBM chess Al beats world champion



### DeepMind

2014: Google buys deep-RL Al company for \$400Mil

### **AlphaGo**

2016: Deep Learning+MCST Go Al beats top human





From https://www.andreykurenkov.com

tory-of-game-ai/

# ARTIFICIAL INTELLIGENCE

IS NOT NEW

### **ARTIFICIAL INTELLIGENCE**

Any technique which enables computers to mimic human behavior



### MACHINE LEARNING

Al techniques that give computers the ability to learn without being explicitly programmed to do so



#### ARTIFICIAL INTELLIGENC

A program that can sense, reason, act, and adapt

#### MACHINE LEARNING

Algorithms whose performance improve as they are exposed to more data over time

### DEEP Learning

Subset of machine learning which multilayered neural networks learn from yast amounts of data

### **DEEP LEARNING**

A subset of ML which make the computation of multi-layer neural networks feasible



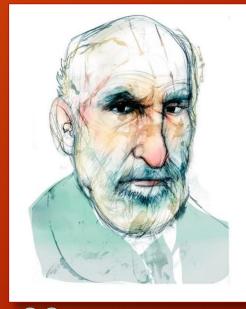
1950's 1960's 1970's 1980's 1990's 2000's 2010s

ORACLE"

66

As long as our brain is a mystery, the universe, the reflection of the structure of the brain will also be a mystery.

Mientras nuestro cerebro sea un arcano, el Universo, reflejo de su estructura, será también un misterio.



Chapter IX: Con tendencias a la literatura y al arte Chácharas de café: Pensamientos, anécdotas y confidencias Imprenta y Librería de Nicolás Moya, Madrid, Spain, 1920

Santiago Ramón y Cajal Nobel Prize in Physiology or Medicine (1906)

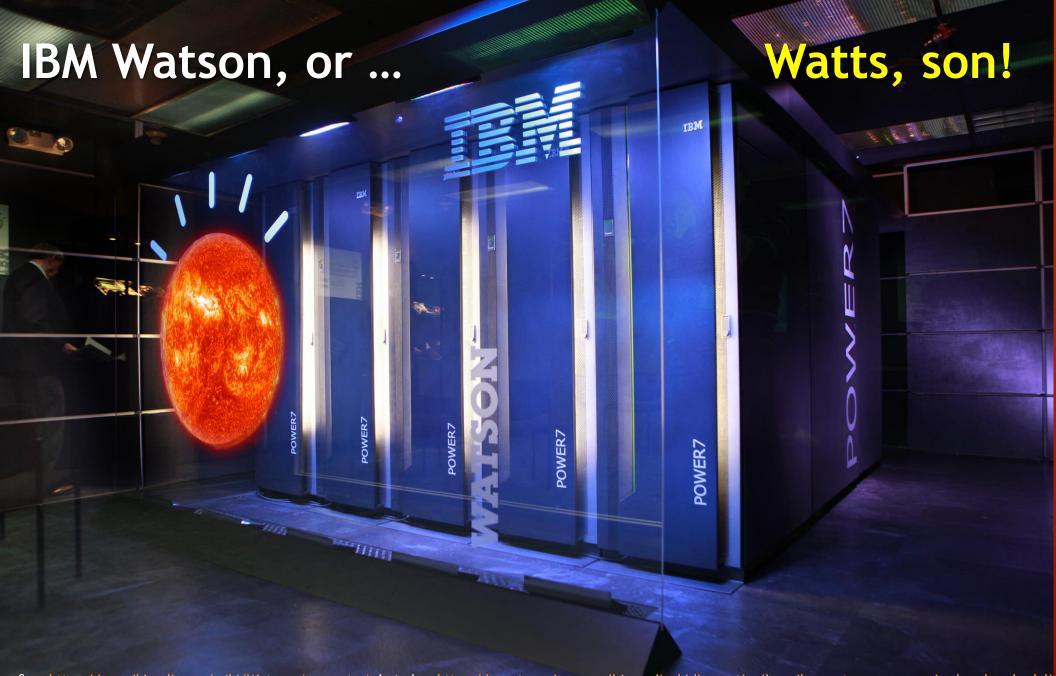


# **IBM Deep Blue**



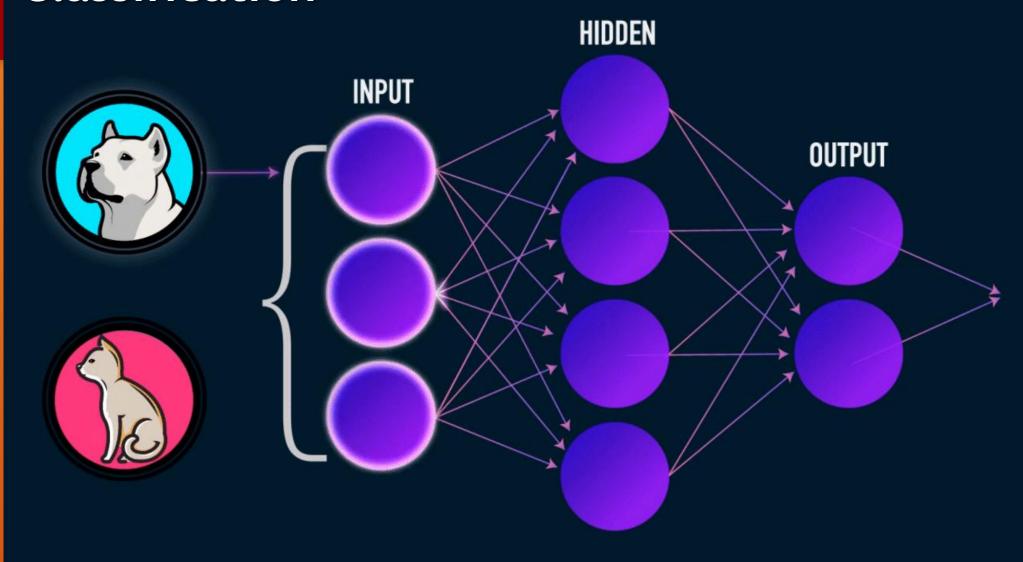




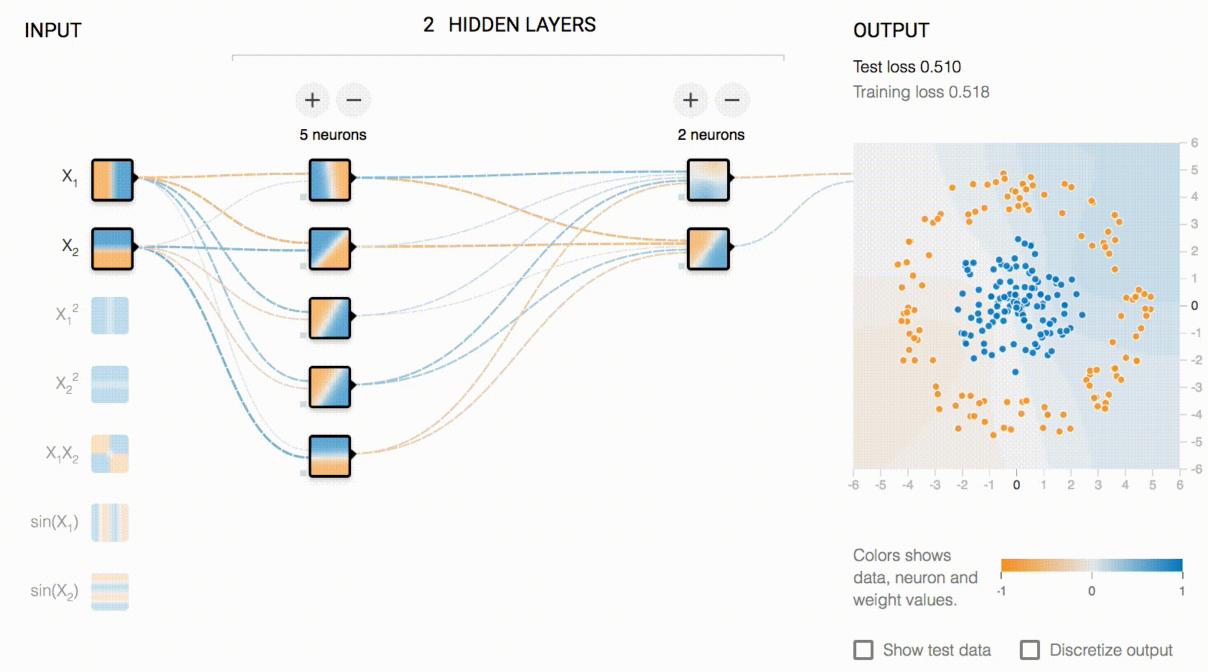


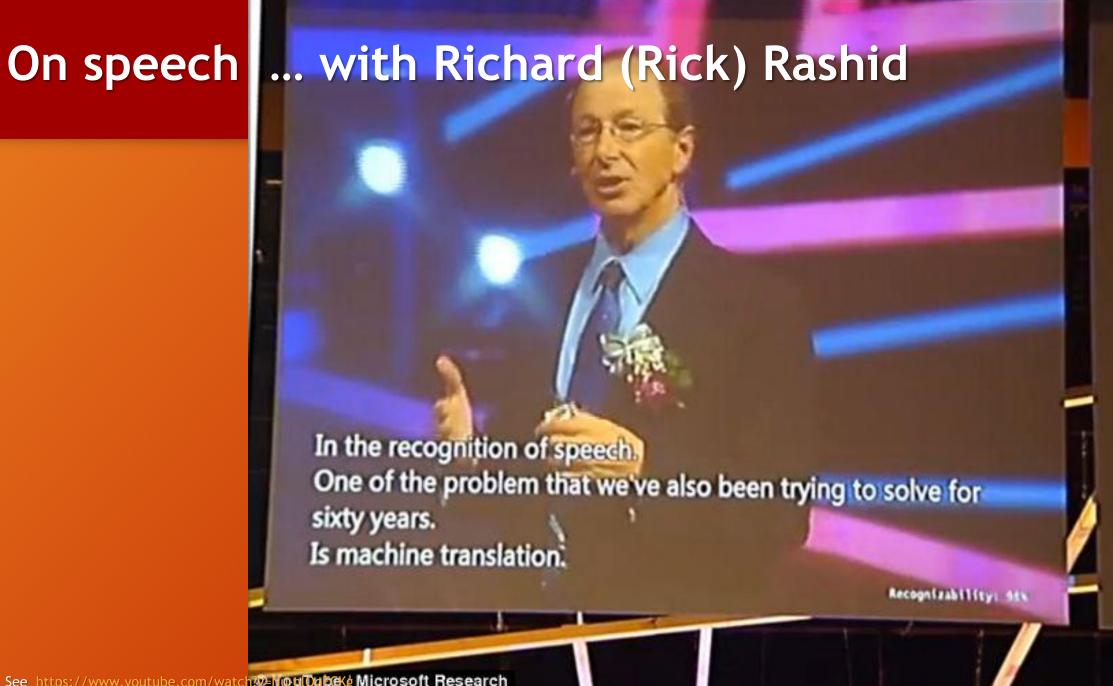


# Classification











# Deep Learning evolution



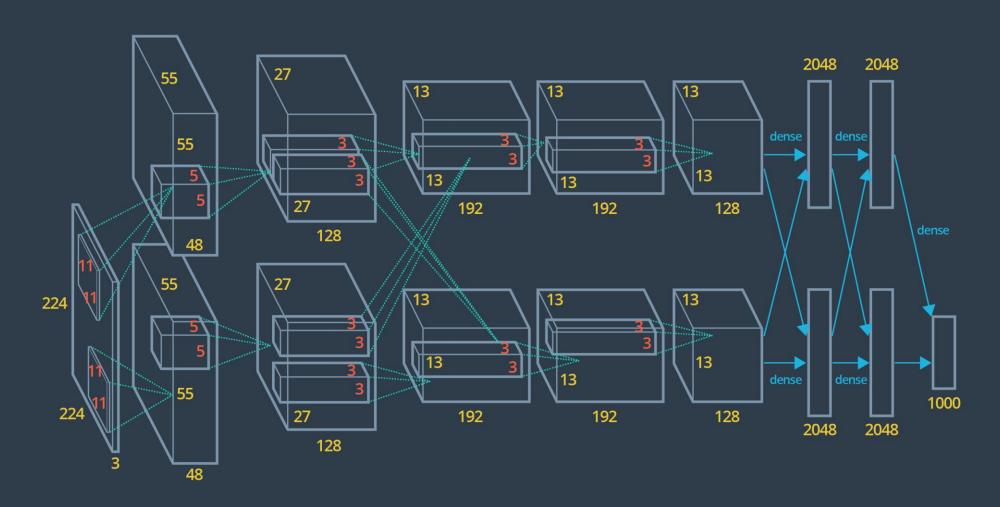
- N Neural Network
- P Probabilistic Model
- Supervised learning
- Unsupervised learning

[Salakhutdinov & Hinton 2009]

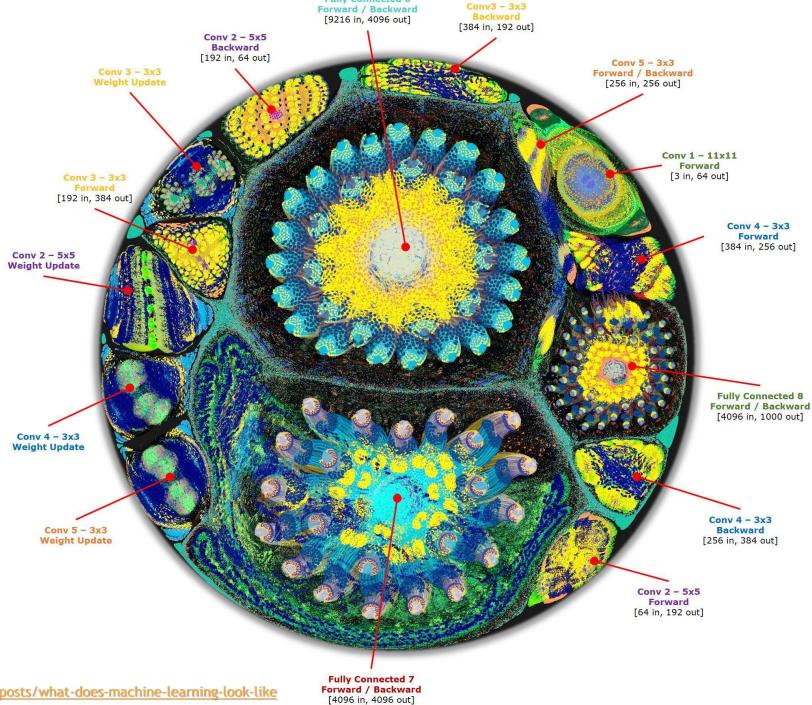
N Perceptron N RBM(P) SVM Boosting [Vapnik 1979] [Rosenblatt 1958] [Schapire 1990] [Hinton 1999] N AE Decision Tree(CLS) GMM (P) [Hunt 1966] [Hinton 1989] [Reynolds 1992] Traditional models Deep models N DBNP [Hinton et al 2006] N RNN N D-AE  $\Sigma\Pi(P)$ N Sparse Coding(P) [Olshausen 1996] [Grossberg 1973] [Vincent 2008] [Hornik 1989] N Conv. Net BayesNP(P) N Neural Nets [McCulloch & Pitt 1943] [Fukushima 1979] [Teh & Jordan 2009] N DBM (P)



# AlexNet



# AlexNet



**Fully Connected 6** 

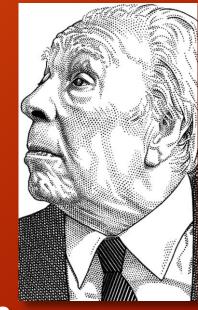


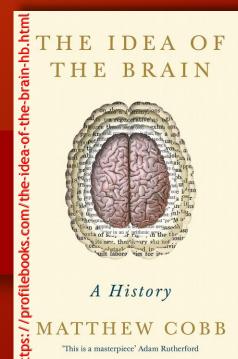


# "The solution to the mystery is always inferior to the mystery itself.

The path toward understanding the brain is long, winding, and littered with dead ends. ["..."] As The Idea of the Brain demonstrates, the mysteries of the mind may not just be stranger than we suppose; they may be stranger than we can suppose.

**Jorge Luis Borges** Received the first Prix Formentor (1961)







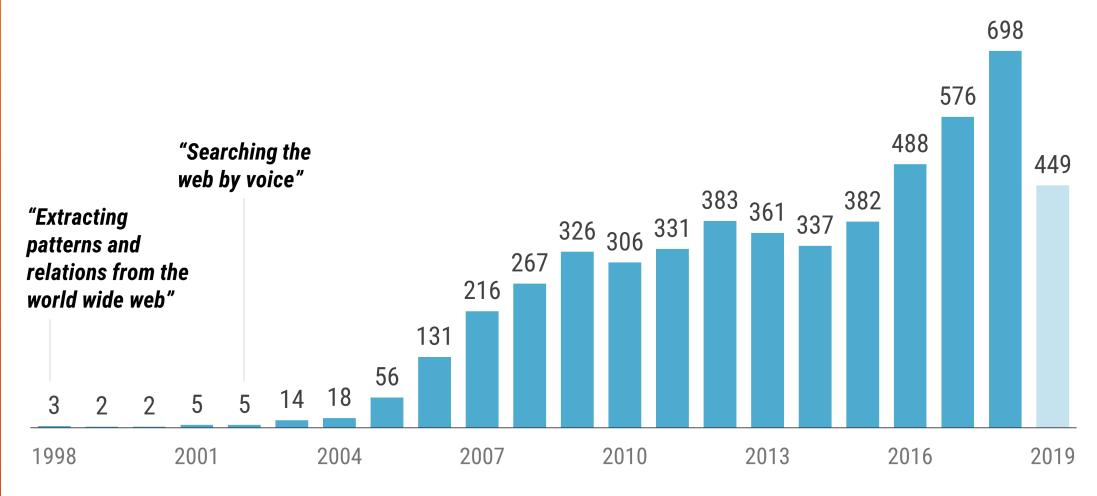






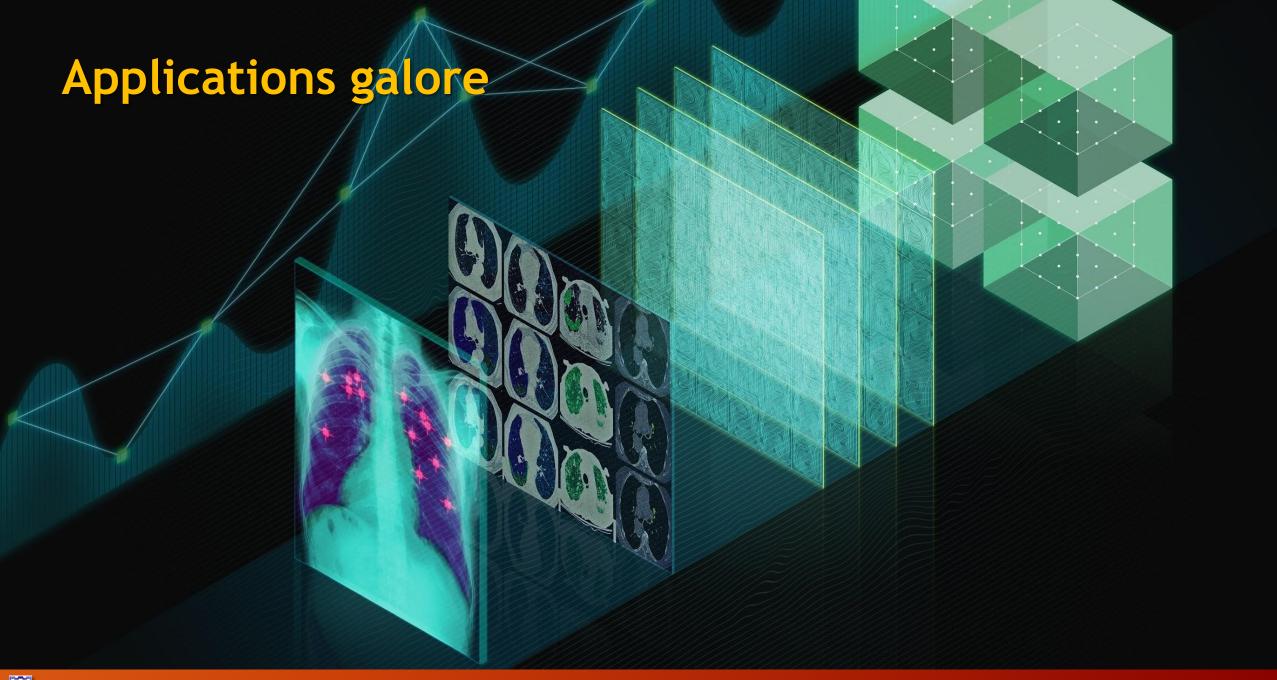
# Google's AI research dates back to its founding year

Al-related publications, 1998 - 2019 (as of Aug. 5)





**EE CBINSIGHTS** 



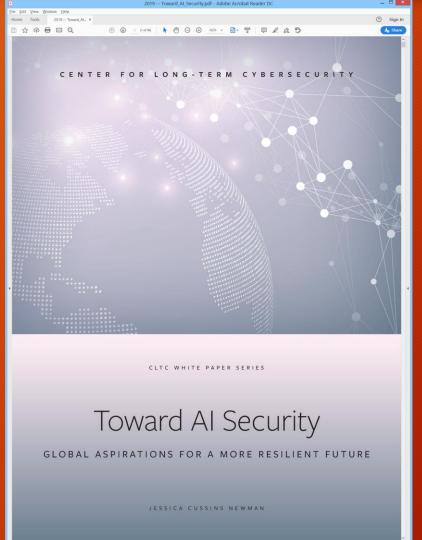


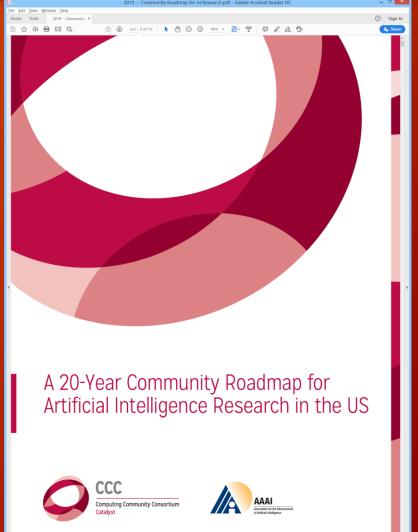
2017

# Feb. 2019

# Aug. 2019









### **60+ STARTUPS USING DEEP LEARNING**

#### **CORE AI: COMPUTER VISION**



#### **CORE AI: OTHER**



### **BI, SALES & CRM**













### **ROBOTICS & AUTO**







# $\times$ **DEEP LEARNING CBINSIGHTS**

### **SECURITY**

(24)



**E-COMMERCE** 



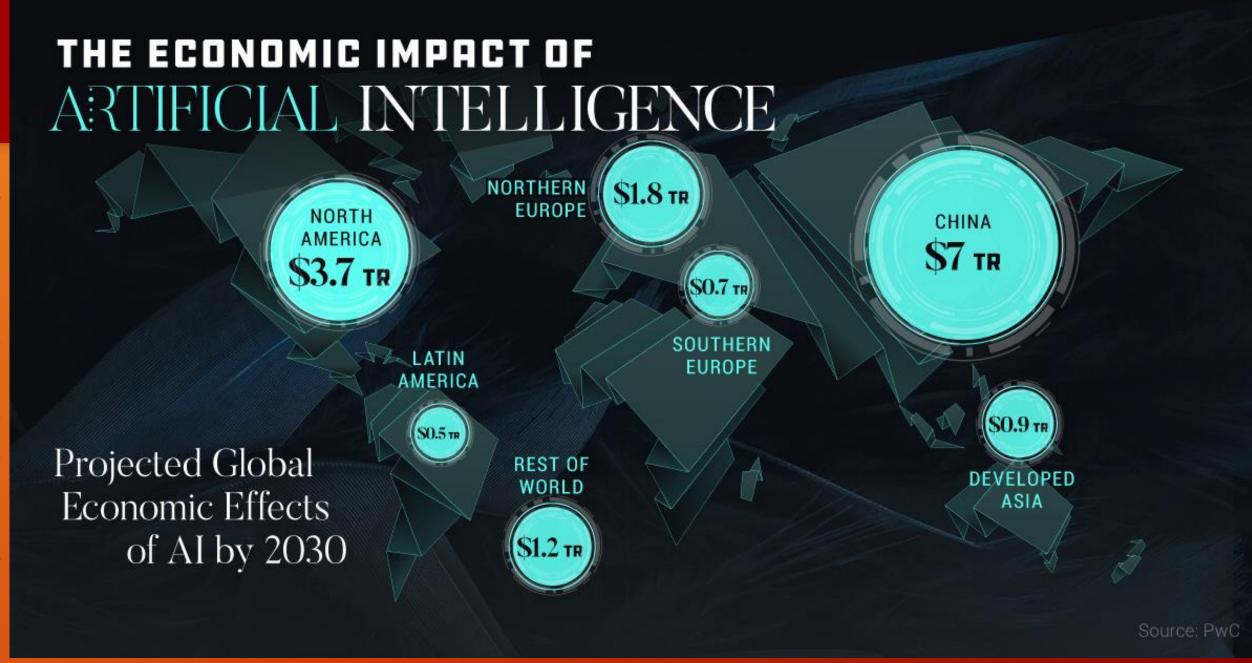






istock.com/a-image







# Europe

### STARTUP CONTINENT

### THE MOST WELL-FUNDED TECH STARTUPS IN EUROPE

The most well-funded VC-backed tech companies in Europe with new funding since 2014. Excludes countries whose most well-funded start-up has not raised more than \$1M.

Excludes debt and lines of credit. Data is as of 10/8/2019.











# ... Asia-Pacific

### STARTUP CONTINENT

### THE MOST WELL-FUNDED TECH STARTUPS IN THE ASIA-PACIFIC

The most well-funded VC-backed tech companies in Asia and the Pacific with new funding since 2014.

Excludes countries where most well-funded startup has not raised at least \$1M.

Excludes debt and lines of credit. Data is as of 9/3/2019.





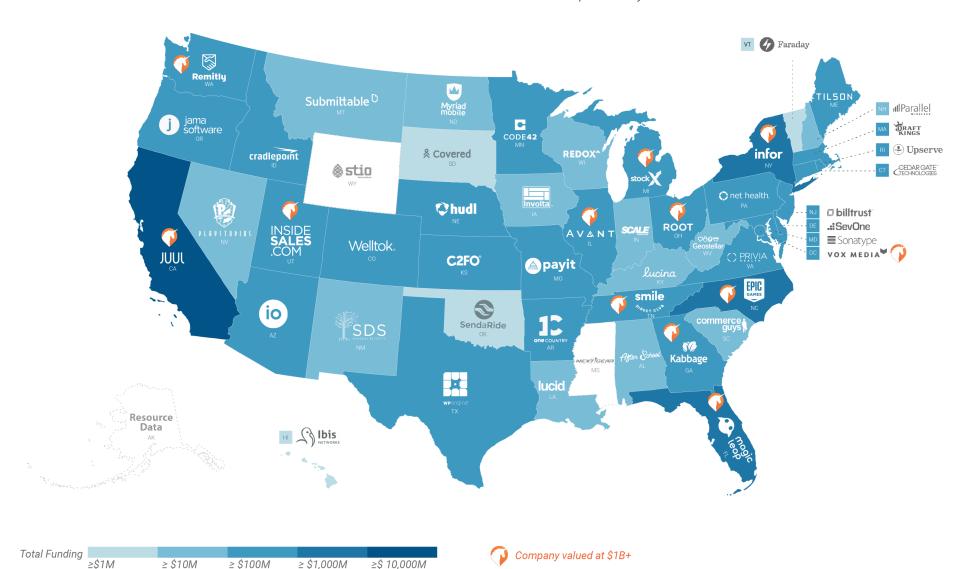






### THE UNITED STATES OF TECH STARTUPS

The most well-funded VC-backed tech startup in every US state.



Companies must have raised a minimum of \$1M and raised any amount of equity funding since January 2014 to be considered. Excludes debt and lines of credit. Alaska, Mississippi, and Wyoming had no companies meeting our full criteria.

Data is as of 7/24/2019.







#### **GLOBAL UNICORN CLUB: 393 PRIVATE COMPANIES VALUED AT \$1B+** as of 8/27/2019

28

#### Cybersecurity



#### **Auto & transportation**



#### Hardware



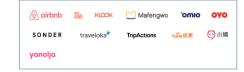
#### **Mobile & telecommunications**



#### Supply chain, logistics, & delivery



#### Travel



#### **Edtech**



#### **Data management & analytics**



#### Consumer & retail





#### Other



#### Health



#### E-commerce & direct-to-consumer



#### Fintech

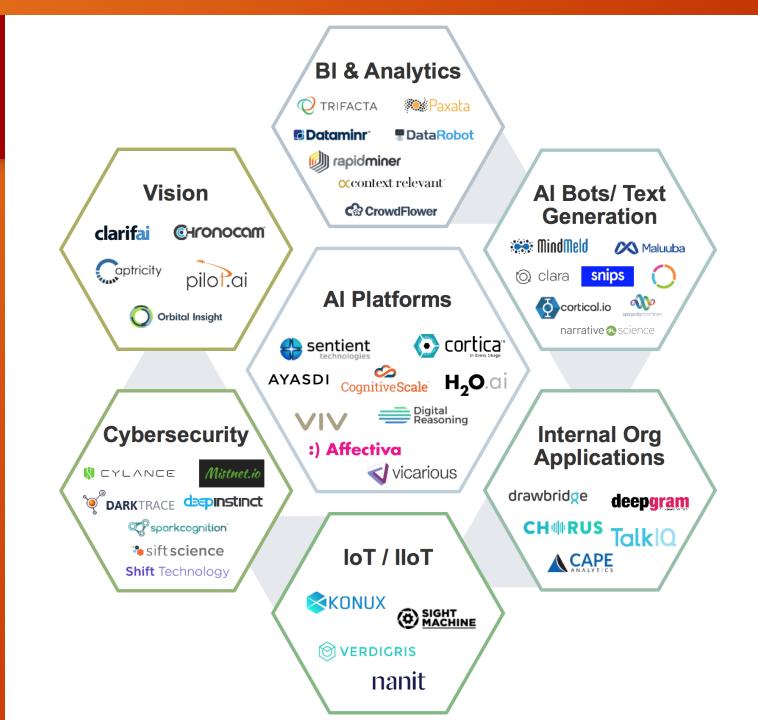


#### Internet software & services





# 



# THE COMING FLOOD OF DATA IN AUTONOMOUS VEHICLES

RADAR ~10-100 KB PER SECOND SONAR ~10-100 KB PER SECOND

GPS ~50KB PER SECOND

CAMERAS ~20-40 MB PER SECOND

AUTONOMOUS VEHICLES
4.00 GB
PER DAY... EACH DAY

~10-70 MB
PER SECOND





2020

2020

### Healthcare



### **Finance & Insurance**



### **Transportation**







### **Retail &** Warehousing









### **Media & Entertainment**



### **Education**



### **Telecom**





### **Manufacturing**











covariant

### Legal **Mining**











#### **CROSS-INDUSTRY TECH**





### **NLP, NLG, & Computer Vision**







**Energy** 

BEYOND

LIMITS

**■L INVENIA** 

**TACHYUS** 

### Sales & CRM





### **AI Model Development**



### **DevOps & Model Monitoring**

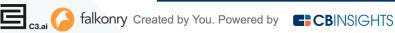


### **Cybersecurity**



### **BI & Ops Intel**





#### Other R&D

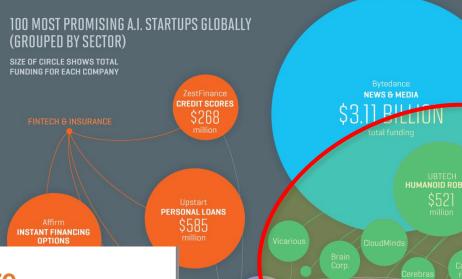






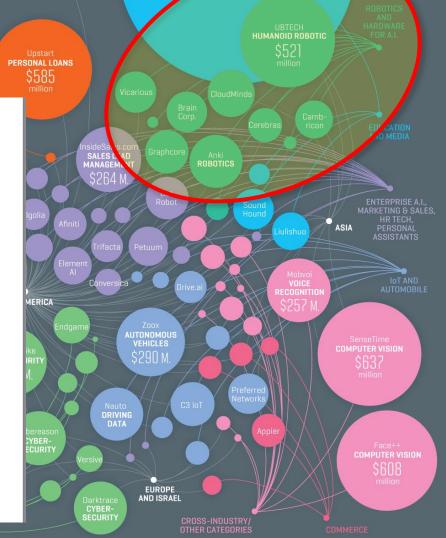
32

## ... recap



# Al – Big Data Driving a Renaissance of Hardware Development and Investment

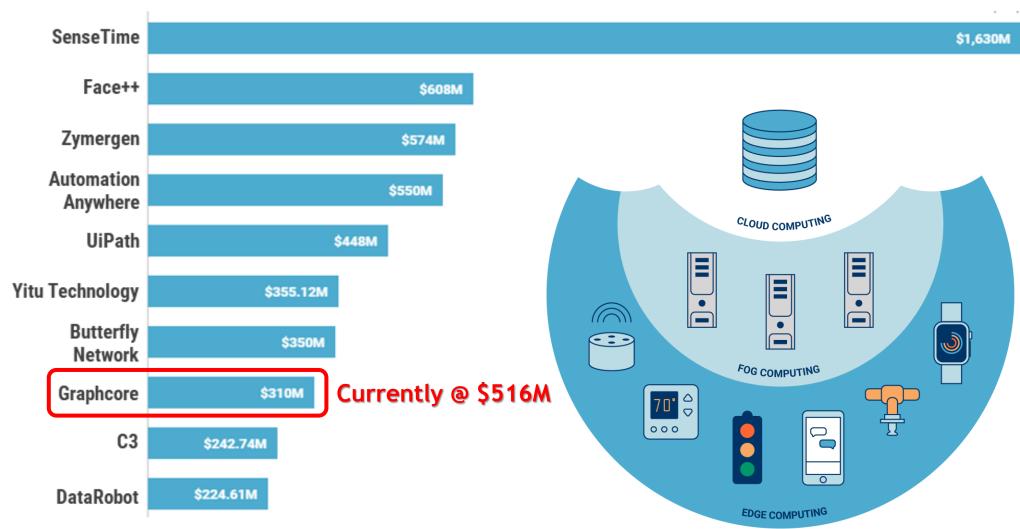
	INITIAL DEPLOYMENT	CLOUD	EDGE
Accelerators GPU, TPU, ASICS, FPGAs	Now	✓	Autos
Near Memory DDR, SRAM, HBM, NAND, SCM	Now to 2 years	✓	✓
New Memory MRAM, ReRAM, PCRAM, FERAM	Now to 5 years	√ ←	✓
In-Memory Compute Analog, ReRAM, PCRAM	2 to 5 years	✓ ←	<b>—</b> ✓
Novel HPC Quantum, Synaptic	5 to 10 years	✓	$\rightarrow$ $\checkmark$



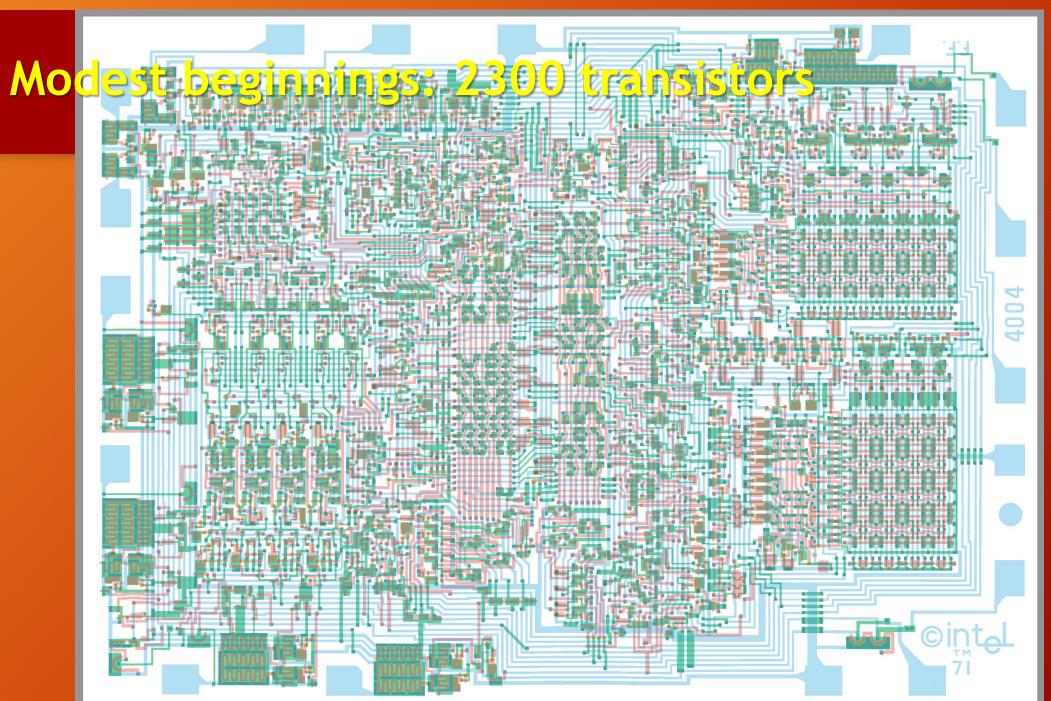


# **2019 AI 100: Most well-funded companies**

Based on total equity funding







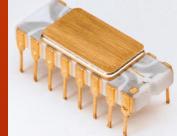
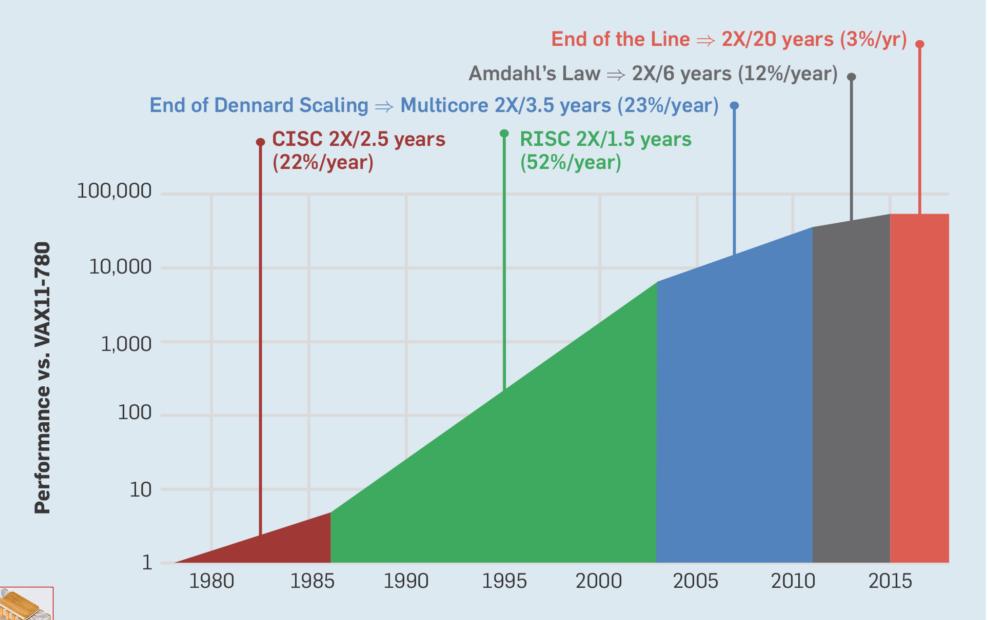


Figure 6. Growth of computer performance using integer programs (SPECintCPU).





### **Late 2015**

# Create a new type of computer that can

- proactively interpret and learn from data,
- solve unfamiliar problems using what it has learned, and
- operate with the energy efficiency of the human brain



### A Nanotechnology-Inspired Grand Challenge for Future Computing

Announced 10/20/2015

Create a new type of computer that can proactively interpret and learn from data, solve unfamiliar problems using what it has learned, and operate with the energy efficiency of the human brain.

While it continues to be a national priority to advance conventional digital computing—which has been the engine of the information technology revolution—current technology falls far short of the human brain in terms of both the brain's sensing and problem-solving abilities and its low power consumption. Many experts predict that fundamental physical limitations will prevent transistor technology from ever matching these twin characteristics. This grand challenge will bring together scientists and engineers from many disciplines to look beyond the decades-old approach to computing based on the Von Neumann architecture as implemented with transistor-based processors, and chart a new path that will continue the rapid pace of innovation beyond the next decade.

To meet this challenge, major breakthroughs are needed not only in the basic devices that store and process information, but in the way a computer analyzes images, sounds, and patterns, interprets and learns from data, and identifies and solves problems. A human can do such tasks in ways that a conventional computer cannot, with a fault-tolerant, adaptive brain that uses less energy than it takes to power an incandescent light bulb. By combining innovations in nanotechnology, computer science, and neuroscience, radically new approaches to creating both hardware and software can be developed, enabling computers capable of efficiently interpreting images and speech, proactively spotting patterns and anomalies in data, learning from data as it is received, and solving unfamiliar problems using what has been learned.

Many of these approaches will require new kinds of nanoscale devices and materials integrated into three-dimensional systems. These nanotechnology innovations will need to be developed in close coordination with new computer architectures and informed by our growing understanding of the brain. Although it may take a decade or more, enabling these transformational computing capabilities will be essential for turning the rising deluge of data that surrounds us into useful information when and where it is needed. Efficiently interpreting and responding to this data will be crucial to solving important problems facing the Nation, from delivering individualized treatments for disease, to allowing advanced robots to work safely alongside people, to proactively identifying and blocking cyber intrusions.

### Read more about

- Nanotechnology-Inspired Grand Challenges
- A Federal Vision for Future Computing: A Nanotechnology-Inspired Grand Challenge (White Paper)
- · Statements of support for this challenge from Federal agencies (DoD, DOE, IARPA, NIST, NSF)
- · Statements of support for this challenge from other organizations (CCC, Moore Foundation, IEEE, Kavli
- Workshop reports and white papers relevant to this challenge
- Meetings and workshops relevant to this challenge

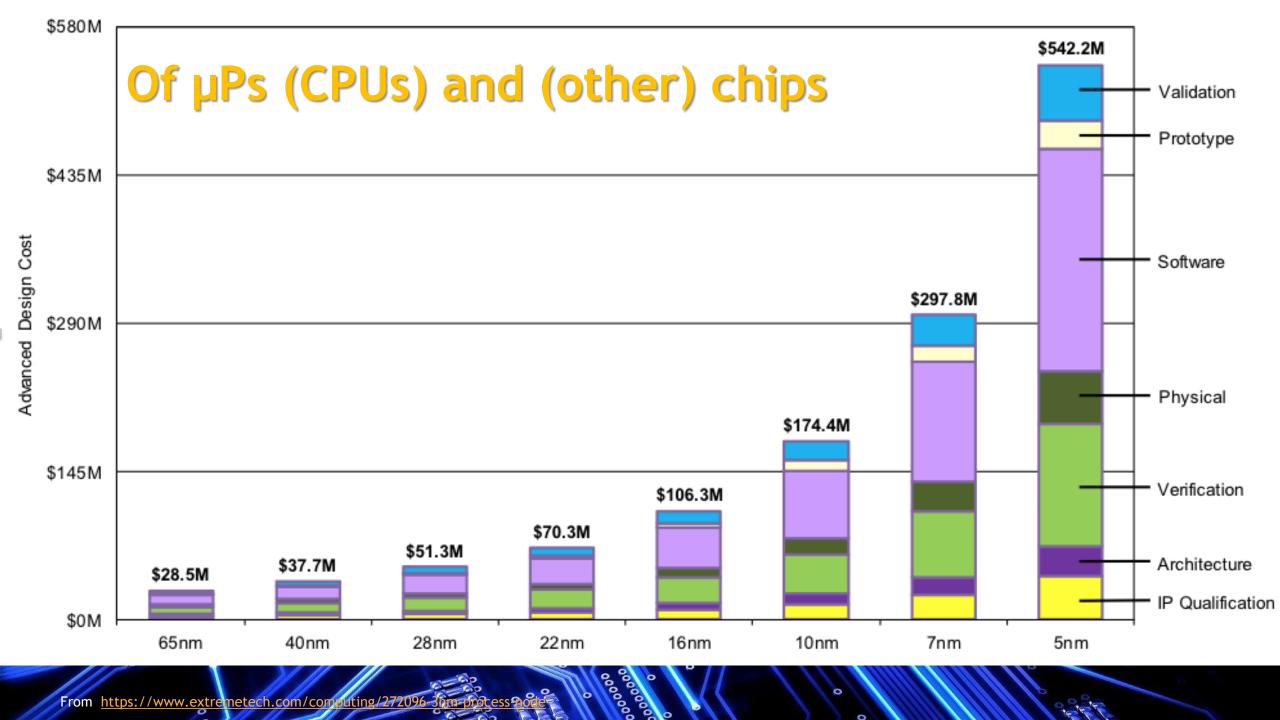


Until now we have been going the other way that is, in order to understand the brain we have to use the computer as a model of it. Perhaps it is time to reverse this reasoning: To understand where we should go with the computer, we should -Robert Noyce look to the brain for some clues. IEEE Centenary "The Next 100 Years" -1984 Robert Noyce Co-founder of Fairchild (1957) & Intel (1968)



"



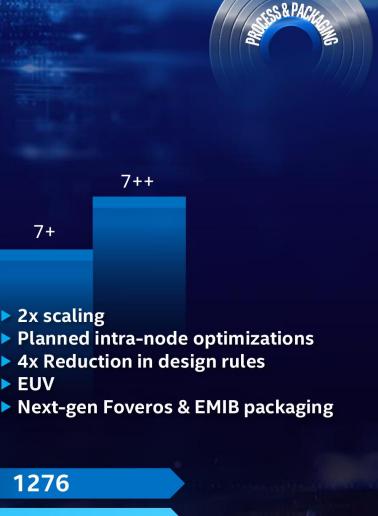


# RELENTLESS INNOVATION CONTINUES

Transistor efficiency

(Perf / W)

**WIGHINVESTOR MEETING** 



7+

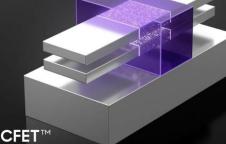
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10++









# 



OFP:3400B (standard)
ORION
UV-LS 2<sup>nd</sup> Gen

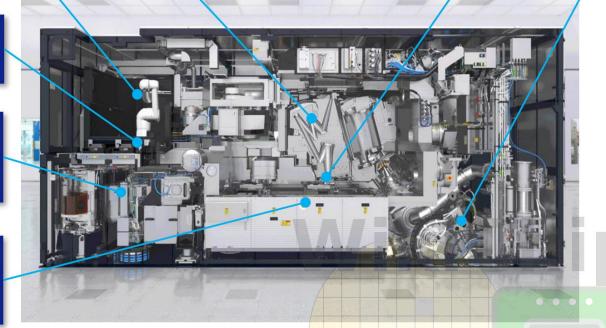
**Optics**Transmission improvement

DGL Membrane (optional)

Reticle Handler
Improved productivity

Wafer Handler @ ≥170WPH

Faster Reticle Align / reduced wafer overhead



Source with Modular Vessel collector swap <8hrs Inline tin refill

	NXE:3400B +OFP:3400	NXE:3400C
Resolution	13 nm	13 nm
Full wafer CDU	≤ 1.1 nm	<u>≤</u> 1.1 nm
DCO	≤ 1.4 nm	≤ 1.4 nm
ММО	≤ 1.5 nm	≤ 1.5 nm
XMMO	≤ 1.9nm	≤ 1.9 nm
Matched to	NXT:2000i	NXT:2000i
Productivity*	≥ 125 WPH	≥ 170 WPH
OPO** (M+3S)	≤ 2.4 nm	≤ 2.4 nm
Focus control**	≤ 60 nm	≤ 60 nm

<sup>\*\*</sup> On product overlay (OPO) and focus control are not ATP specs, but are performance targets for specific customer nodes to be achieved including Application and DUV configuration. Performance of these parameters is to be within population of NXE:3400B + OFP, but at the higher productivity



<sup>\*</sup> Productivity as ATP specs, 20mj/cm2, all ATP tests no DGL-membrane, no pellicles





Tech Giants/System

IC Vender/Fabless

(intel)

SAMSUNG

ON INVIDIA.

OHALCOWW.

**AMD** 

XILINX.

**<b>W** UNISOC

Rockchip

**Automated Driving** 

NP

RENESAS **TOSHIBA** 

47/

### Al Chip Landscape

V0.5 Aug

VVVE

Graphcore®

∴ habana

**H**\ILO

thinci

**MALRAY** 

grog

Tachyum?

**E**speranto

Energy-efficient AI devices

· Al tools for large-scale energy management

Startup Wo

Two distinct trends are emerging within AI and energy:

does not have the same power and resources as cloud computing. For instance, Kneron is one company that recently announced lowpower AI processors for edge devices.

First, energy efficiency will take center stage as AI comes to more

edge devices, such as phones and cameras, since edge computing

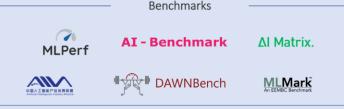
As another example, Apple acquired Xnor.ai, a startup that makes low-power edge AI tools, in Q1'20.

"...our hardware engineering and machine learning teams asked the audacious question, 'can we create a hardware, machine learning architecture capable of running deep learning models without a battery? That can be so low-power they can harvest ambient energy from the 🛚 sun?"

**PEZY Computing** Eta Compute GREENWAVES - XNOR.AI

More at https://basicmi.github.io/AI-Chip/











# ANNOUNCING NVIDIA DGX SUPERPOD

AI LEADERSHIP REQUIRES
AI INFRASTRUCTURE LEADERSHIP

Test Bed for Highest Performance Scale-Up Systems

- 9.4 PF on HPL | ~200 AI PF | #22 on Top500 list
- <2 mins To Train RN-50</li>

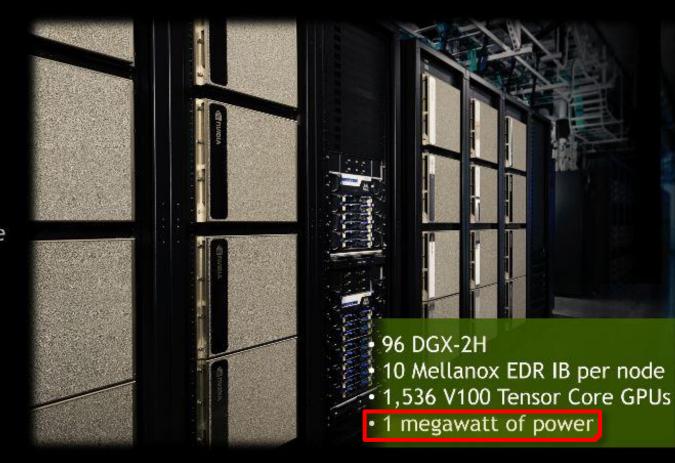
Modular & Scalable GPU SuperPOD Architecture

- Built in 3 Weeks
- Optimized For Compute, Networking, Storage & Software

Integrates Fully Optimized Software Stacks

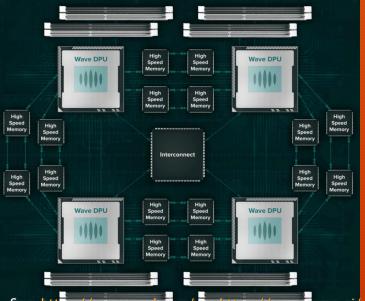
Freely Available Through NGC

Autonomous Vehicles | Speech AI | Healthcare | Graphics | HPC



# AMD, Wave (DPU), Habana







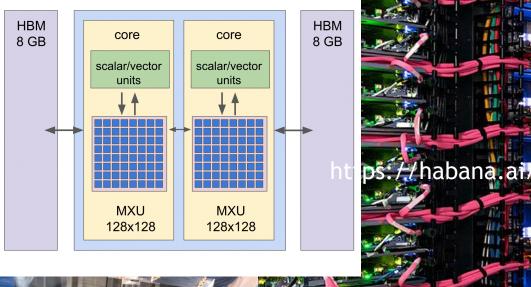


### TPUv2 Chip



- 16 GB of HBM
- 600 GB/s mem BW
- Scalar/vector units: 32b float
- MXU: 32b float accumulation but reduced precision for multipliers

• 45 TFLOPS



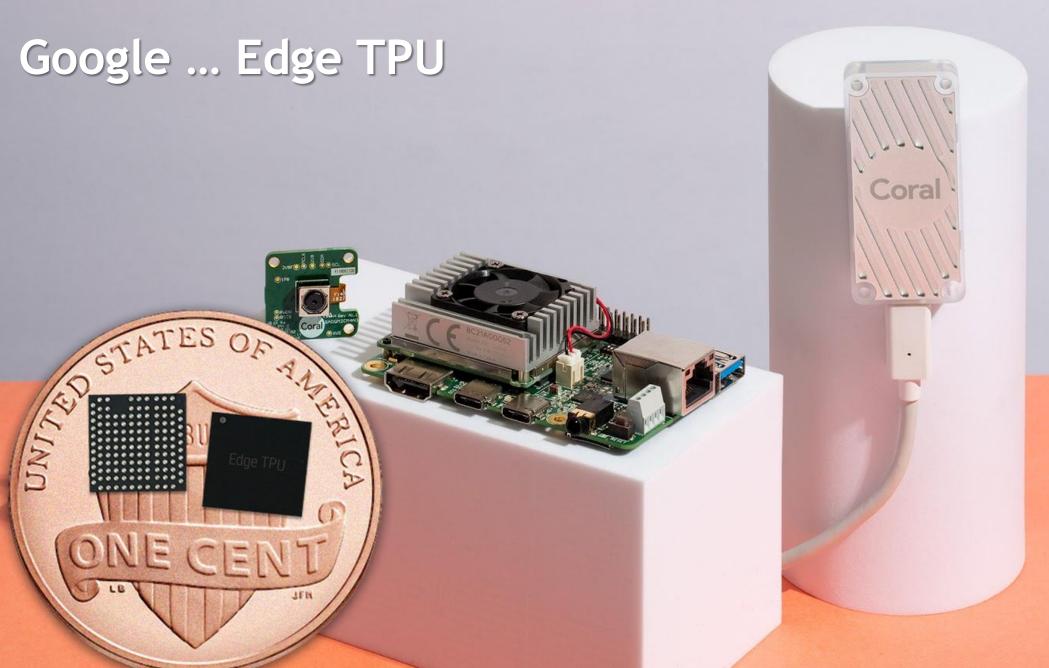


Google ... TPU v

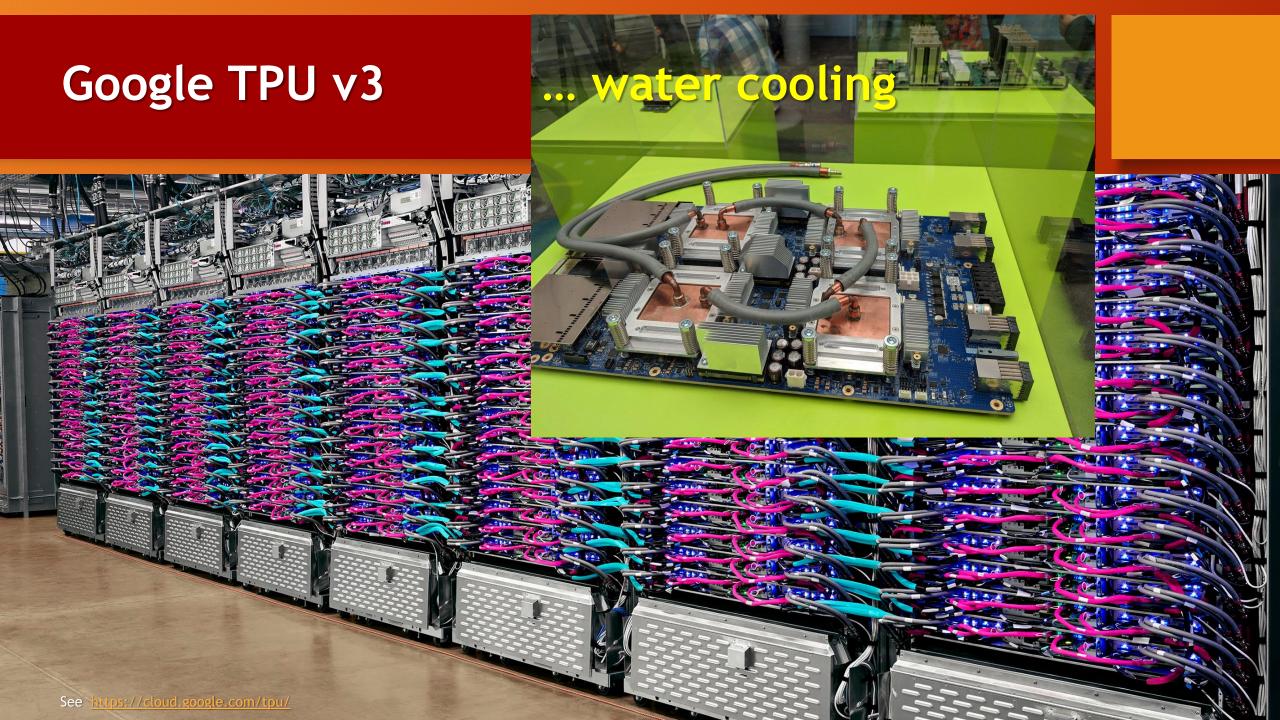


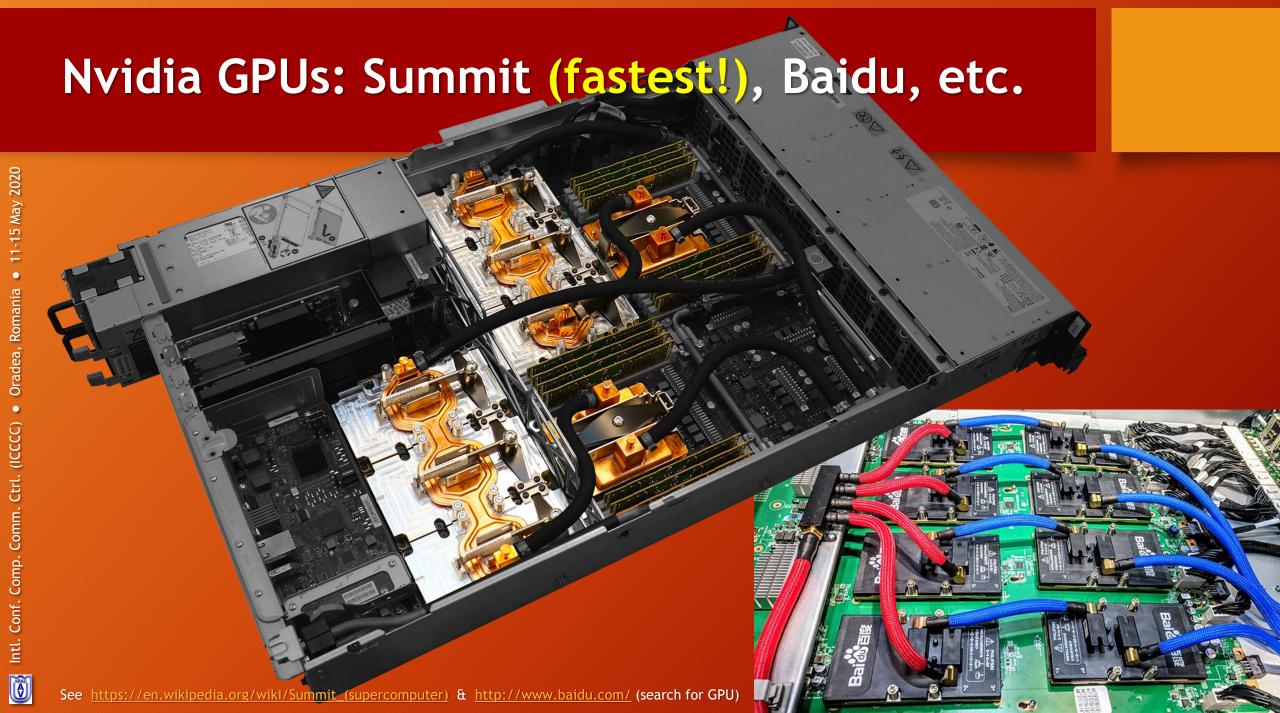
See https://cloud.google.com/tpu/











# 21.5×21.5 cm<sup>2</sup> ... 56× larger than the largest



### The world's largest chip

46,225 mm<sup>2</sup> chip

56x larger than the biggest GPU ever made

400,000 cores

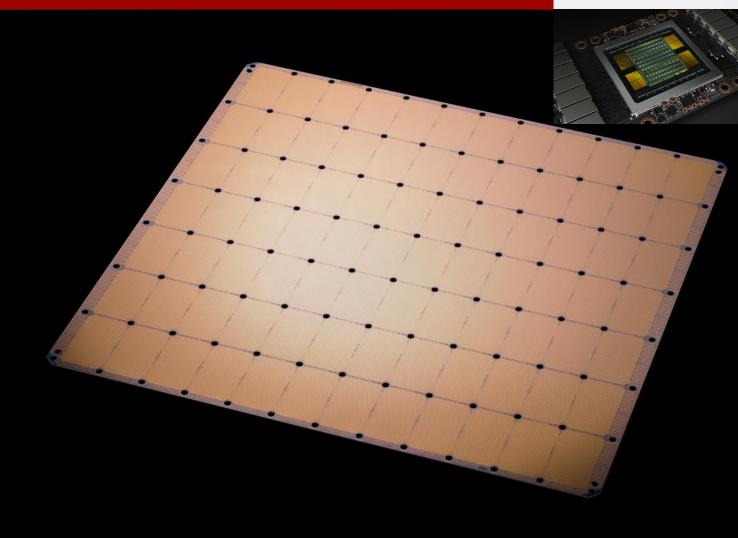
78x more cores

18 GB on-chip SRAM

3000x more on-chip memory

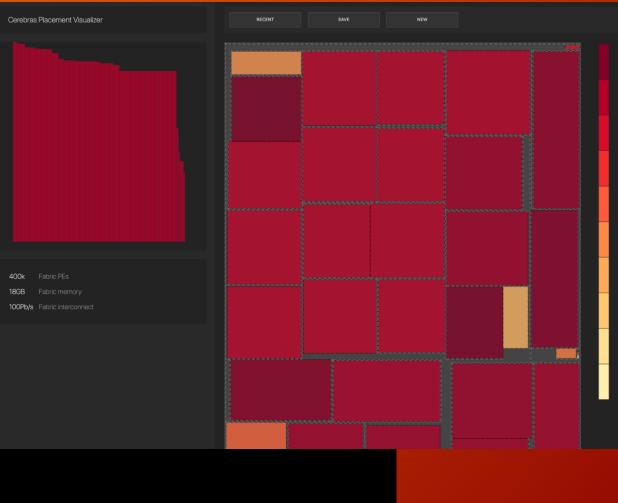
100 Pb/s interconnect

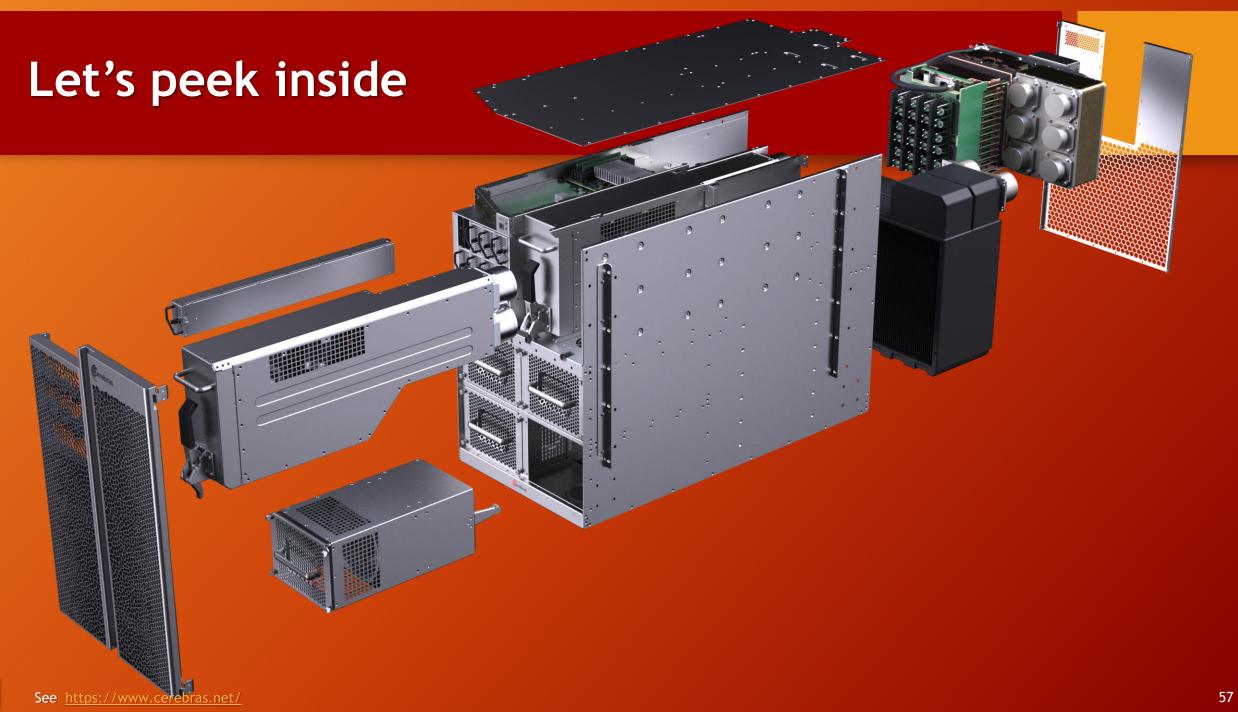
33,000x more bandwidth



### Cerebras workstation CS-1







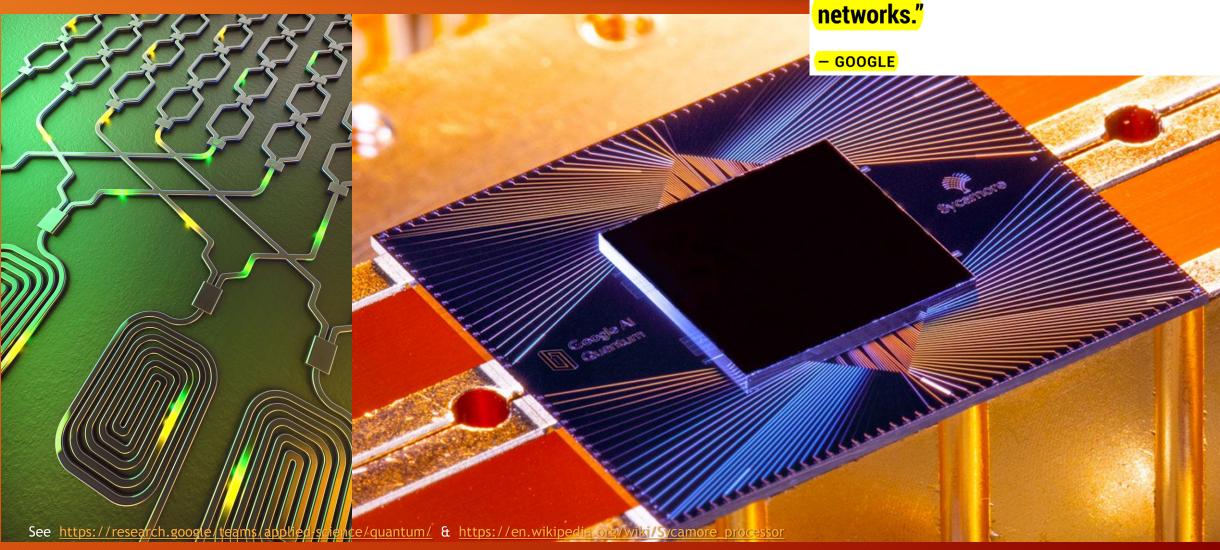
## Any other options?

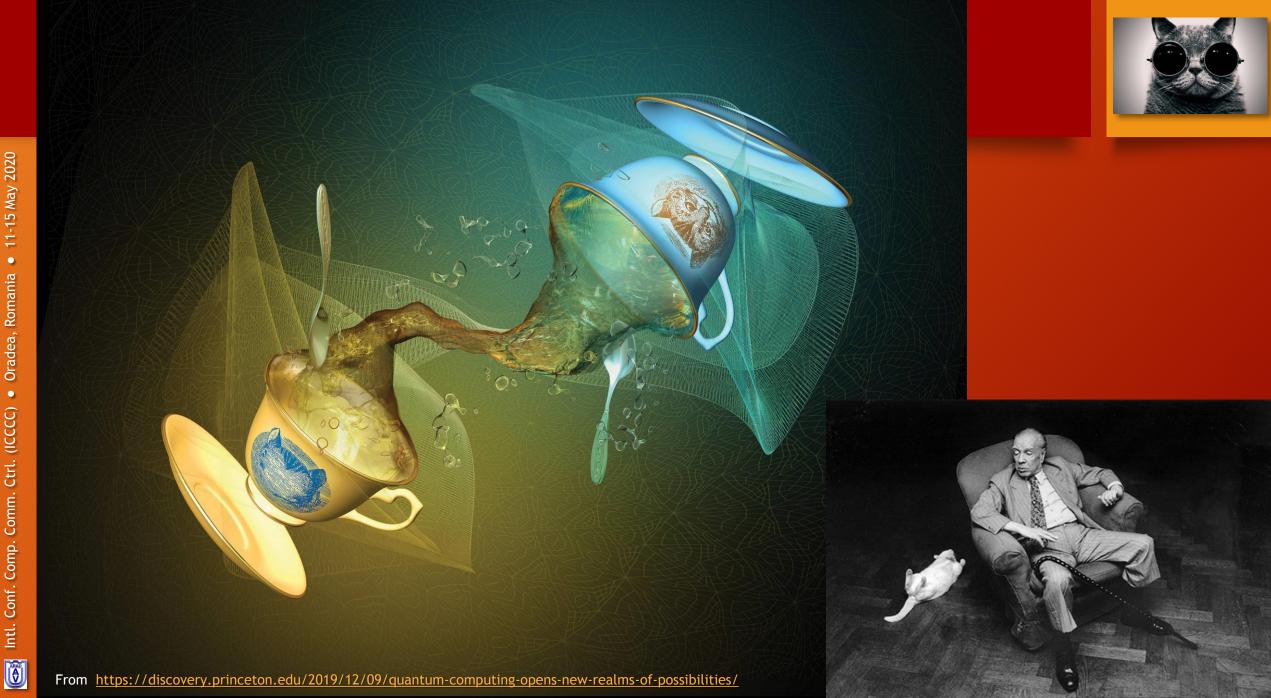
Oradea, Romania



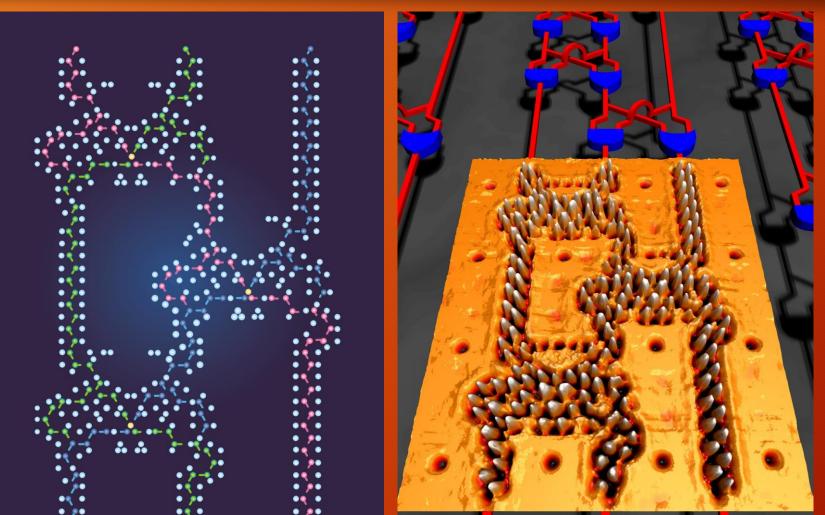
Reserve your seat by 1 October 2019 for early registra

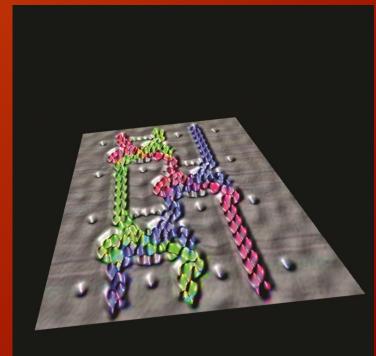
"Traditional machine learning took many years from its inception until a general framework for supervised learning was established. We are at the exploratory stage in the design of quantum neural networks."





## Early IBM "quantum" 3-bit sorter (1993 ...)







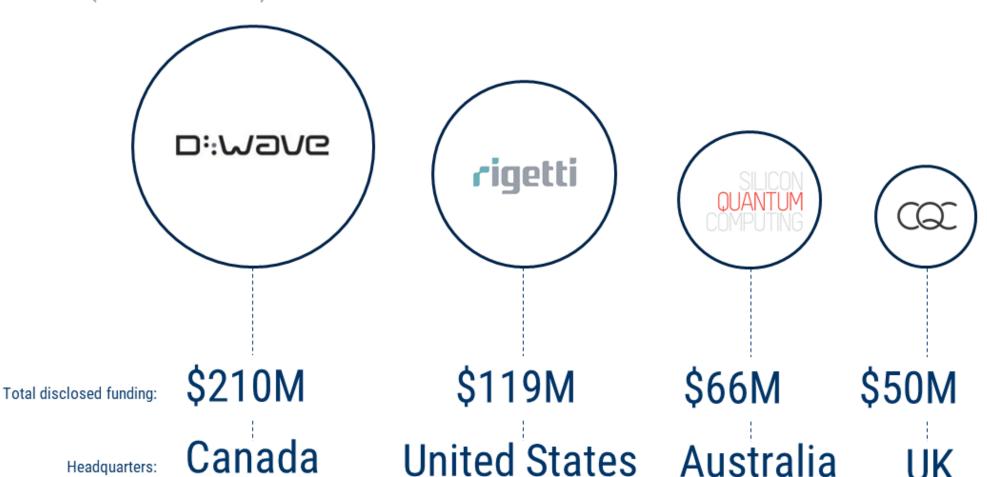
### No small effort

100 Estimated annual spending on non-classified 50 National quantum-technology research, 2015, €m 10 spending **European Union\*** 550 **Netherlands** 27 Denmark 22 Britain Russia 30 Canada Sweden 15 105 100 Japan Finland 12 63 Germany Poland 12 China 220 France 120 52 **United States** South 360 Spain Korea 13 25 Italy 36 Switzerland Austria 35 Singapore 67 Australia 44 75 Brazil 11 World 1,500 (estimate) \*Combined estimated budget of EU countries Source: McKinsey



### Quantum computing startups with ≥ \$50M raised

(as of 1/7/2019)



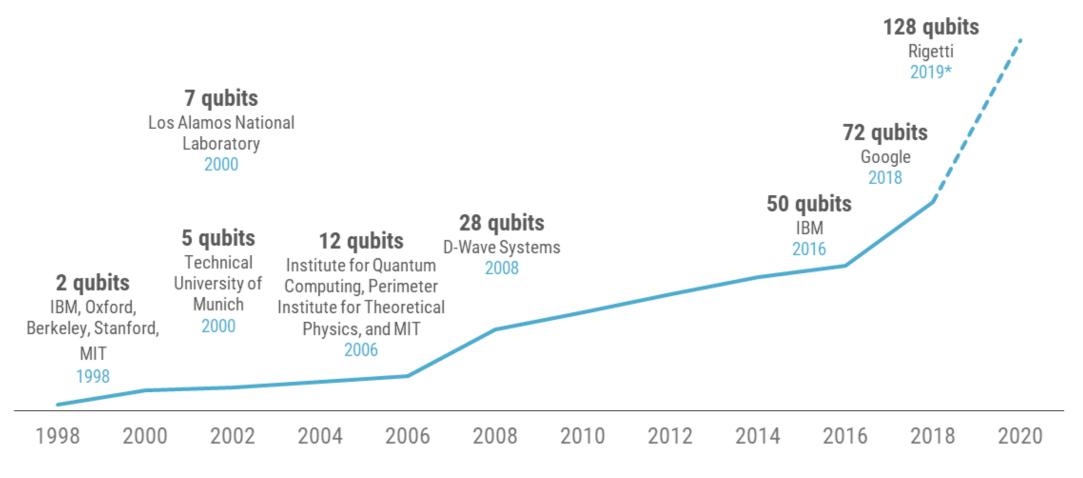


Source: cbinsights.com

### Quantum computers are getting more powerful

Number of qubits achieved by date and organization 1998 - 2020\*







Source: MIT, Qubit Counter. \*Rigetti quantum computer expected by late 2019.

### The three known types of quantum computing and their applications, generality, and computational power.



A very specialized form of quantum computing with unproven advantages over other specialized forms of conventional computing.

DIFFICULTY LEVEL





The most likely form of quantum computing that will first show true quantum speedup over conventional computing. This could happen within the next five years.

DIFFICULTY LEVEL





### **Quantum Annealer**

The quantum annealer is least powerful and most restrictive form of quantum computers. It is the easiest to build, yet can only perform one specific function. The consensus of the scientific community is that a quantum annealer has no known advantages over conventional computing.

APPLICATION Optimization Problems

GENERALITY Restrictive

COMPUTATIONAL POWER
Same as traditional computers

### **Analog Quantum**

The analog quantum computer will be able to simulate complex quantum interactions that are intractable for any known conventional machine, or combinations of these machines. It is conjectured that the analog quantum computer will contain somewhere between 50 to 100 qubits.

### APPLICATIONS

Quantum Chemistry Material Science Optimization Problems Sampling Quantum Dynamics

GENERALITY Partial

COMPUTATIONAL POWE

### **Universal Quantum**

The universal quantum computer is the most powerful, the most general, and the hardest to build, posing a number of difficult technical challenges. Current estimates indicate that this machine will comprise more than 100,000 physical qubits.

### APPLICATIONS

Secure computing Machine Learning Cryptography Quantum Chemistry Material Science Optimization Problems Sampling Quantum Dynamics Searching

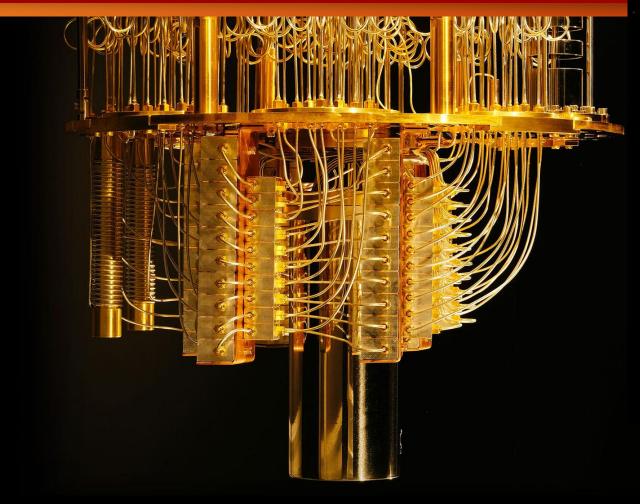
Complete with known speed up

COMPUTATIONAL POWER Very High





# IBM Q





## Fast ... and, hopefully ... green (low power)

### Noblesse oblige (alternatives)

- With great power comes great responsibility\*
- With great size comes great power consumption



With larger and larger chips comes ... lower and lower power consumption
 1 MW → 250 kW → 15 kW → 10 kW



... great, just that we still have a lot of work ahead (a long way to go)
 → 1000 W → 100 W → Brain

Solution	Power (W)
Nvidia GPU	1,000,000
Google TPU	250,000
Graphcore IPU	25,000
Quantum	25,000
Cerebras WSE	15,000
Pending	1,000
Pending	100
Brain	40





