

Richard Feynman, Data-Intensive Science and the Future of Computing

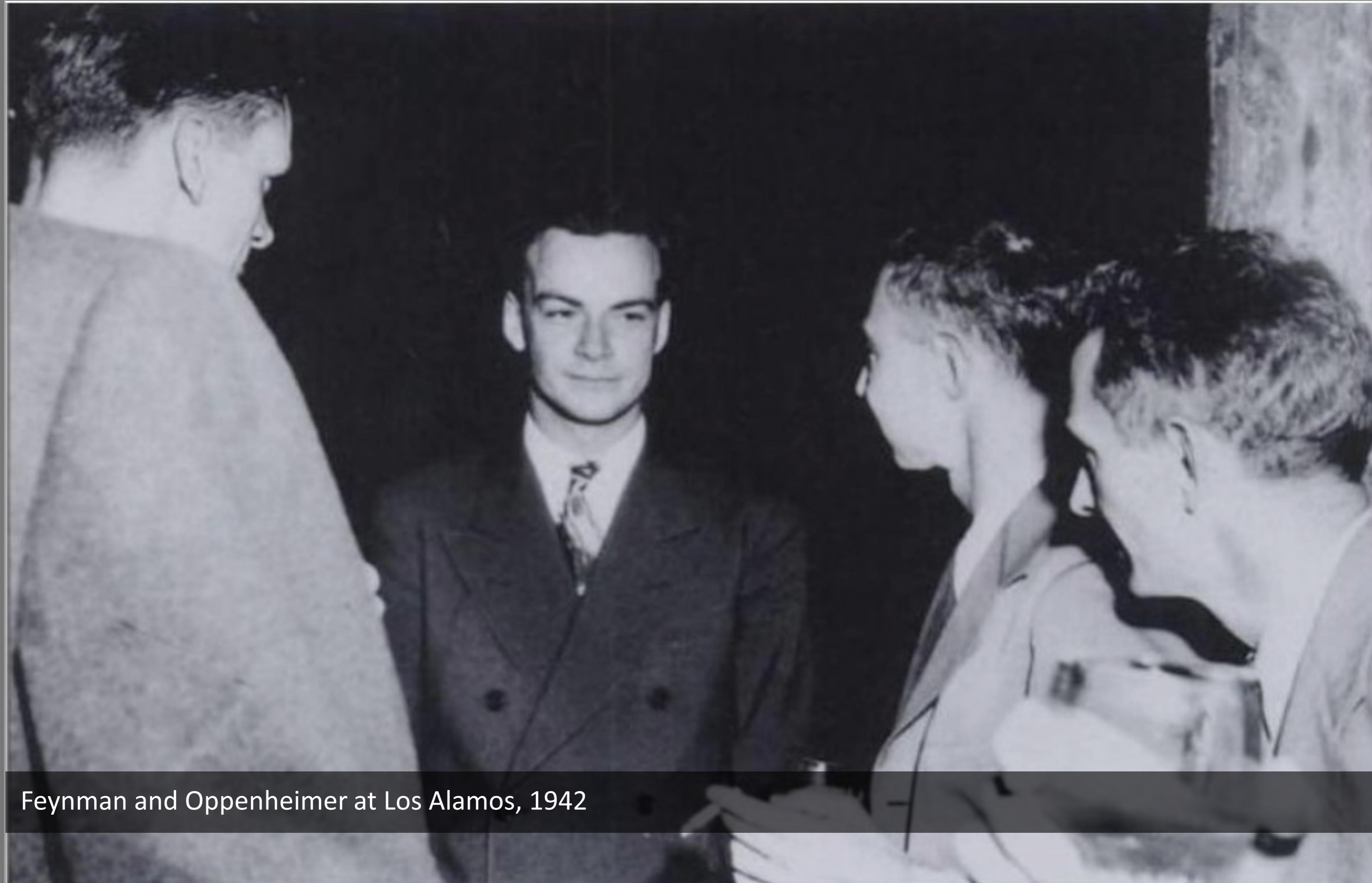


TONY HEY

Feynman and Computation



Parallel Computing without Computers



Feynman and Oppenheimer at Los Alamos, 1942

Feynman's Computational Toolkit

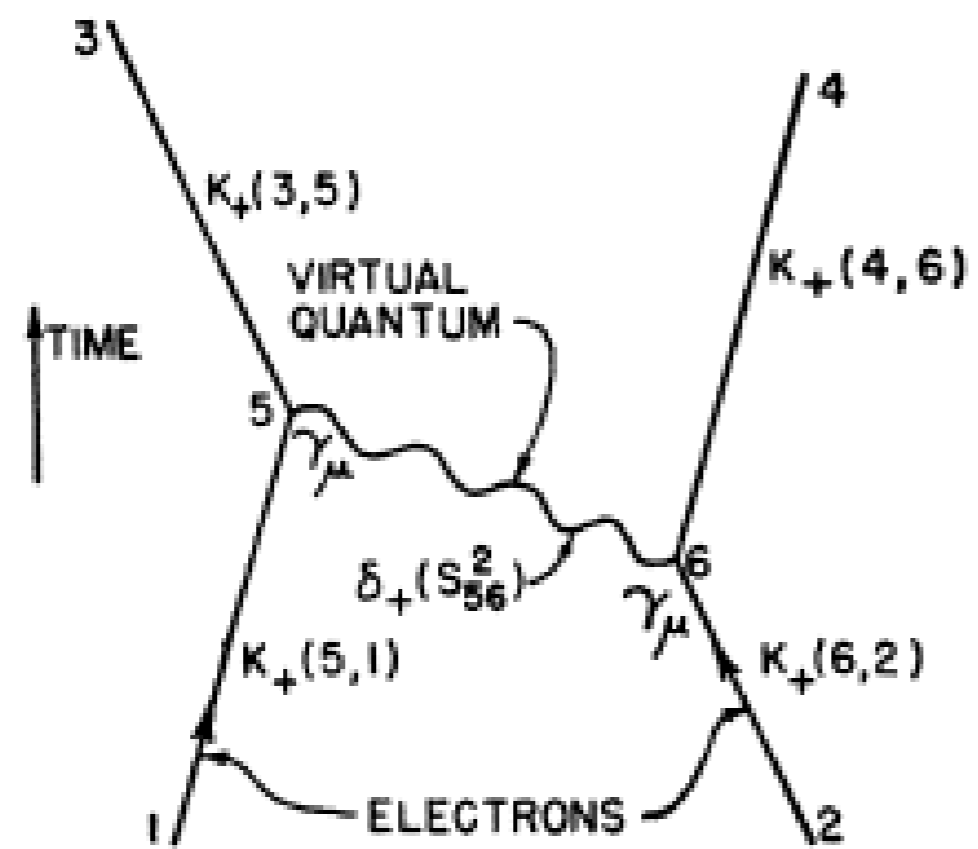


FIG. 1. The fundamental interaction Eq. (4). Exchange of one quantum between two electrons.

“Like the silicon chips of more recent years, the Feynman diagram was bringing computation to the masses”

Julian Schwinger

Caltech: Four Kings and the Joker



Carl Anderson, Murray Gell-Mann, Max Delbrück, Richard Feynman, George Beadle, 1969

Physics of Computation Conference, MIT 1981



Simulating Physics with Computers

- Can a universal classical computer simulate physics *exactly*?
- Can a classical computer *efficiently* simulate quantum mechanics?

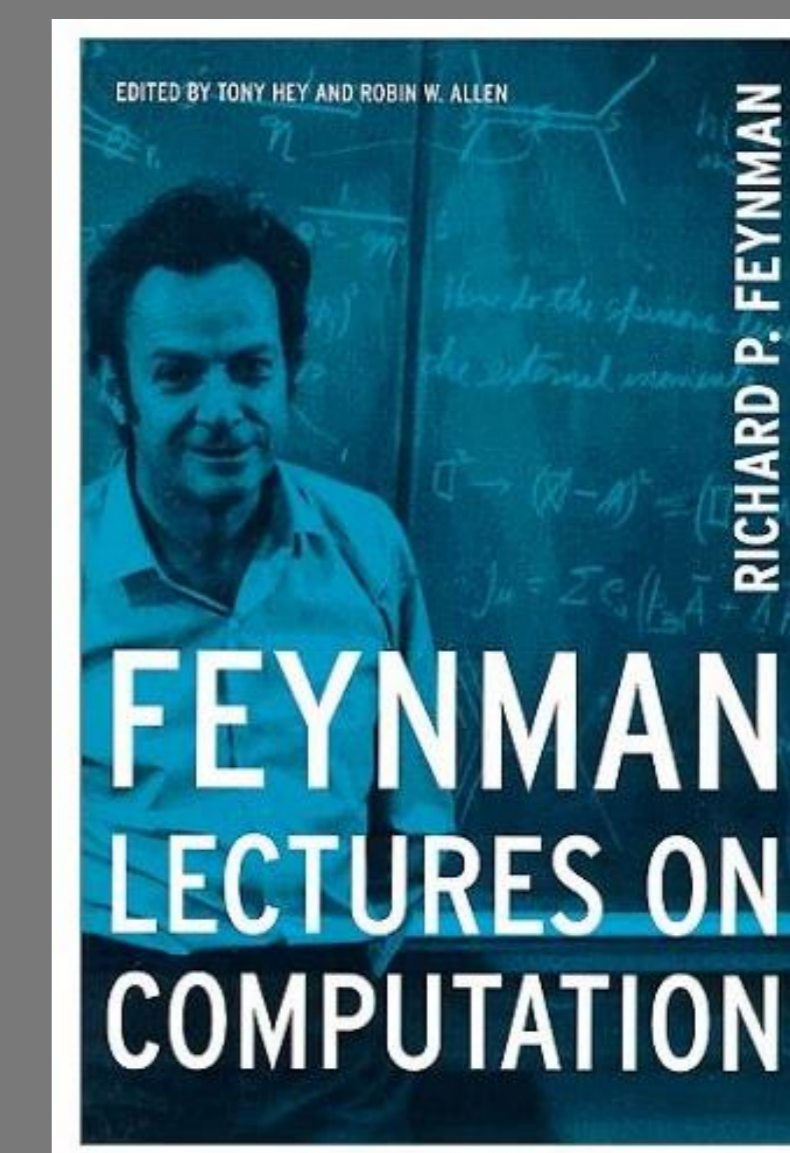
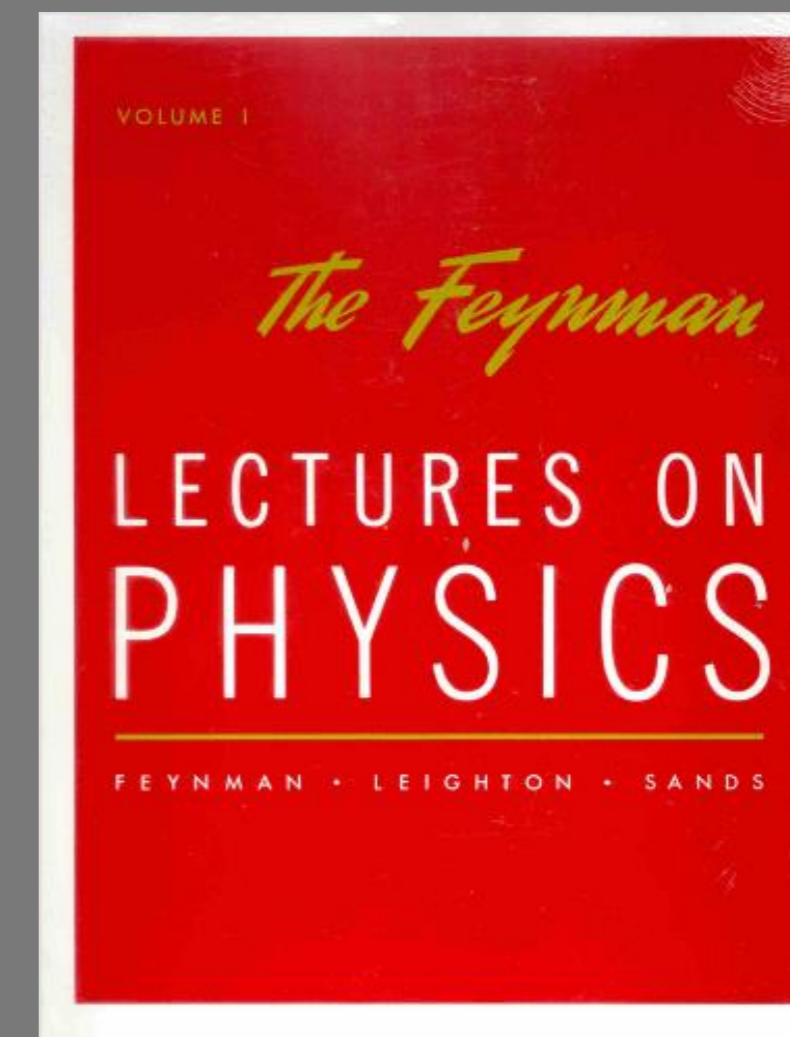
“I’m not happy with all the analyses that go with just classical theory, because Nature isn’t classical, dammit, and if you want to make a simulation of Nature, you’d better make it quantum mechanical, and by golly it’s a wonderful problem!”

“How can we simulate the quantum mechanics?....Can you do it with a new kind of computer - a quantum computer? It is not a Turing machine, but a machine of a different kind.”

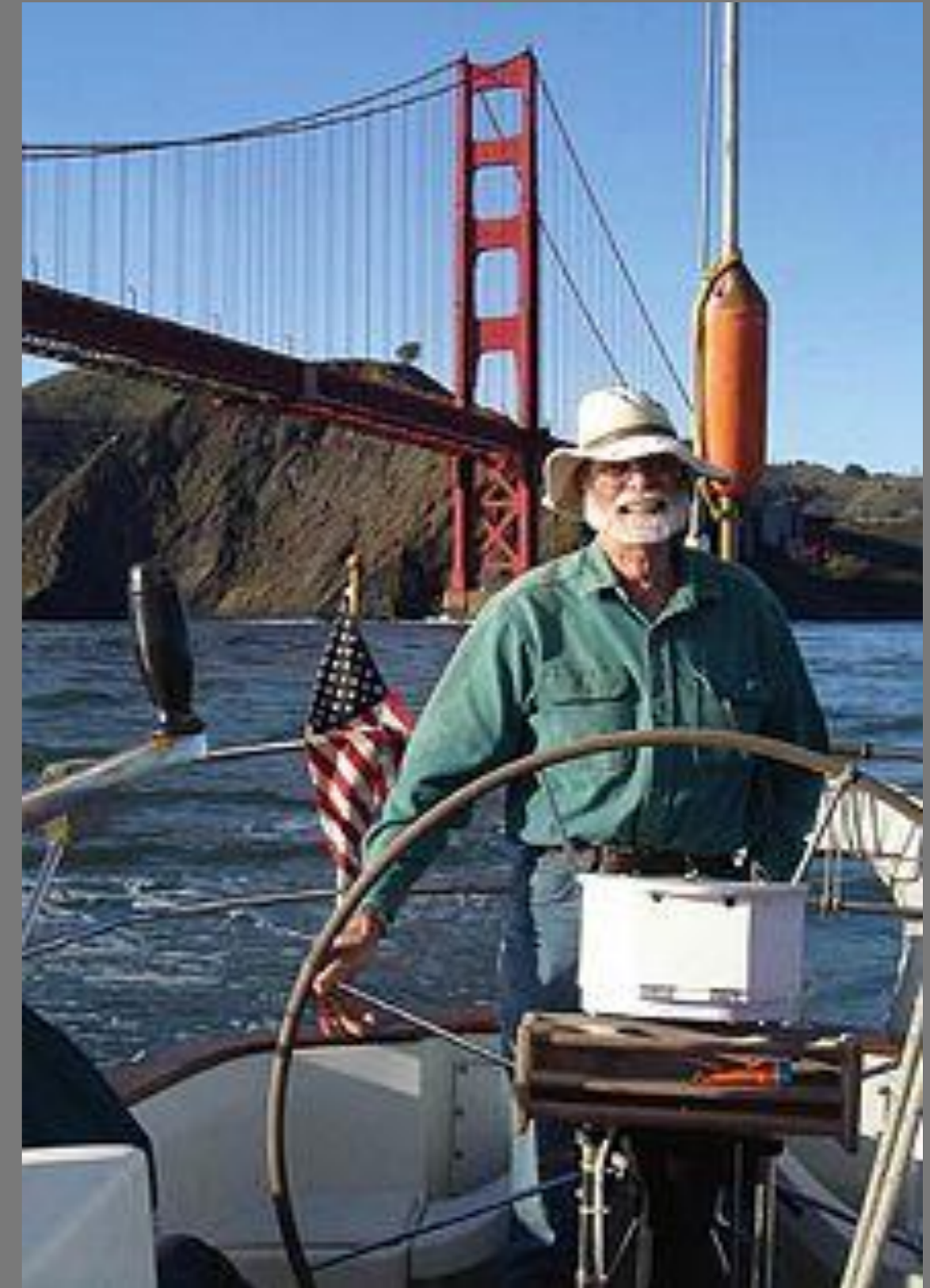
R P Feynman 1981

Fundamental Limits to Computation

- Feynman is famous for his Lectures on Physics but from 1981 to 1985 he lectured on computing
- Examined the fundamental limits to computation arising from:
 - Mathematics
 - Noise
 - Thermodynamics
 - Engineering in Silicon
 - Quantum Mechanics
- Complete set of reversible logic gates – CN, CNN, Fredkin, ...

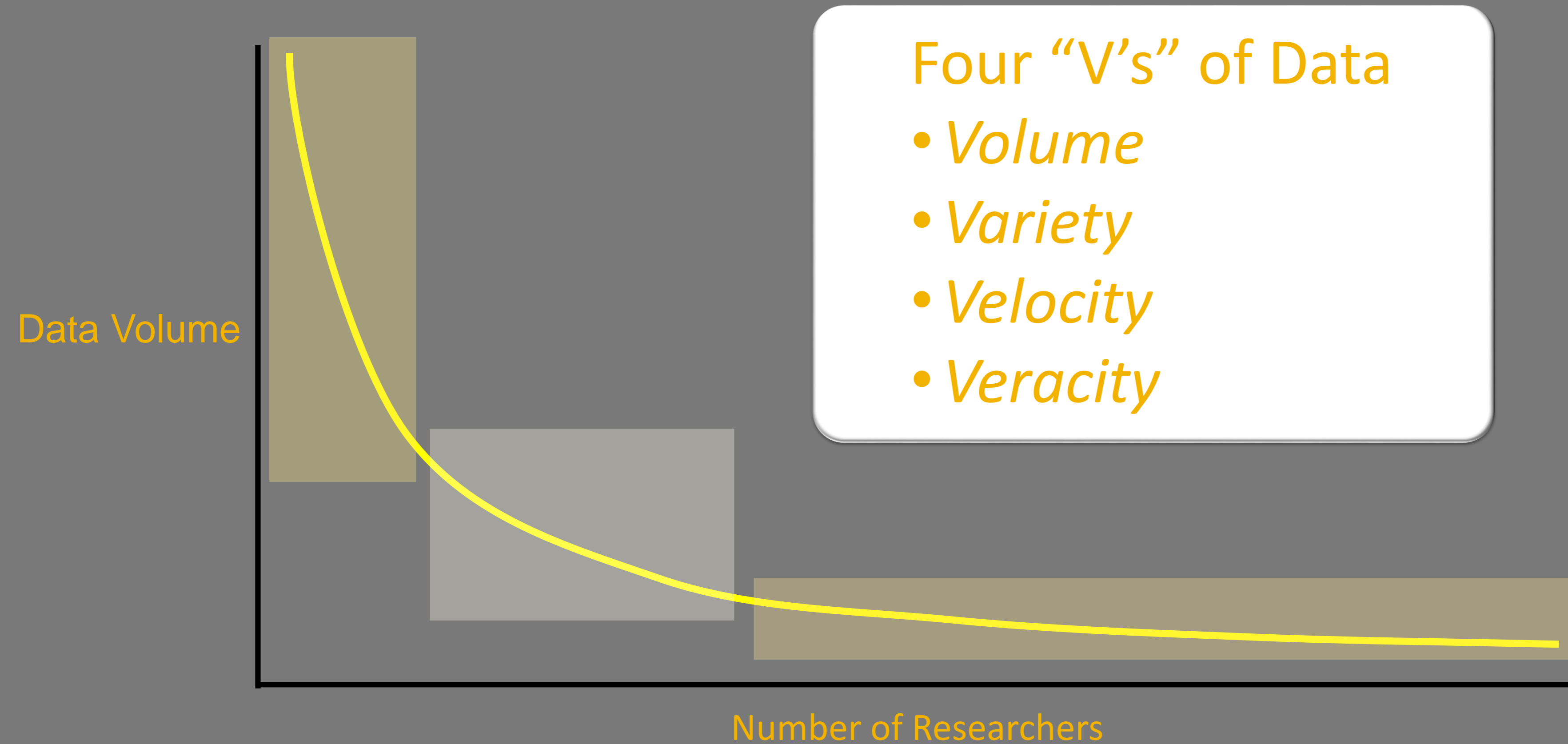


Data-Intensive Science



Jim Gray, Turing Award Winner

Much of Science is now Data-Intensive



‘The Long Tail of Science’

The Fourth Paradigm: Data-Intensive Science

Thousand years ago – **Experimental Science**

Description of natural phenomena

Last few hundred years – **Theoretical Science**

Newton's Laws, Maxwell's Equations...

Last few decades – **Computational Science**

Simulation of complex phenomena

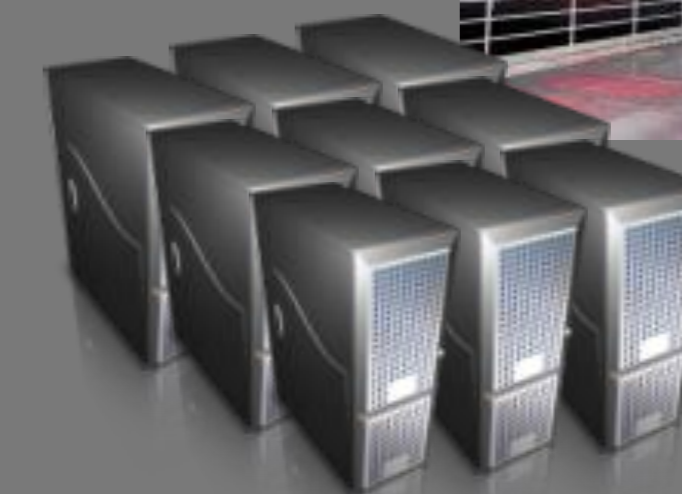
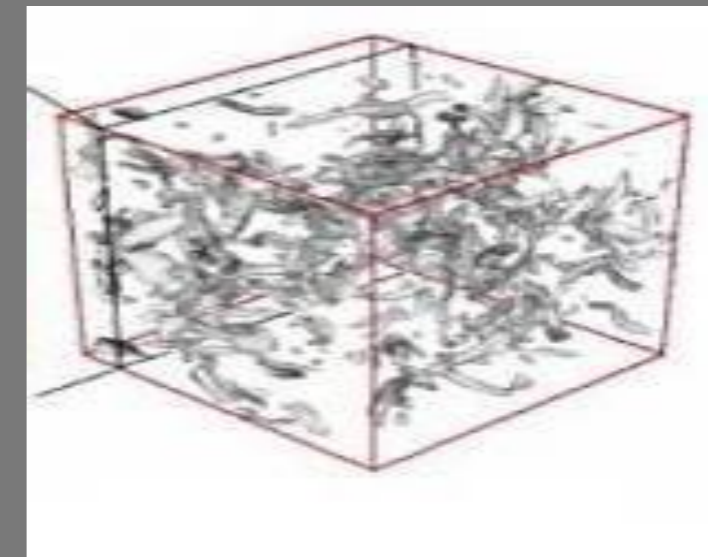
Today – **Data-Intensive Science**

Scientists overwhelmed with data sets from many different sources

- Data captured by instruments
- Data generated by simulations
- Data generated by sensor networks
- Data generated by satellites



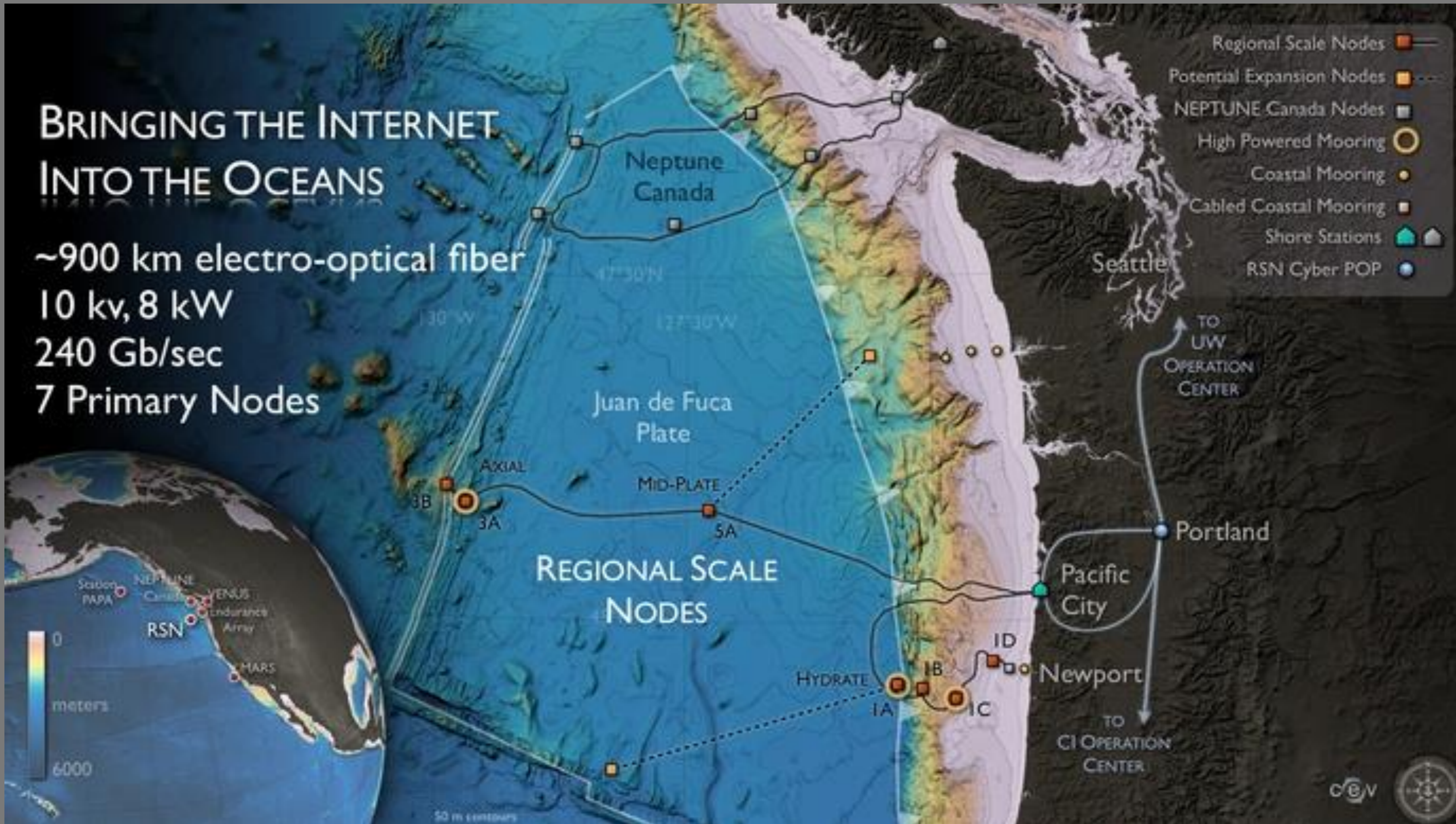
$$\left(\frac{\dot{a}}{a}\right)^2 = \frac{4\pi G\rho}{3} - K \frac{c^2}{a^2}$$



eScience is the set of tools and technologies to support data federation and collaboration

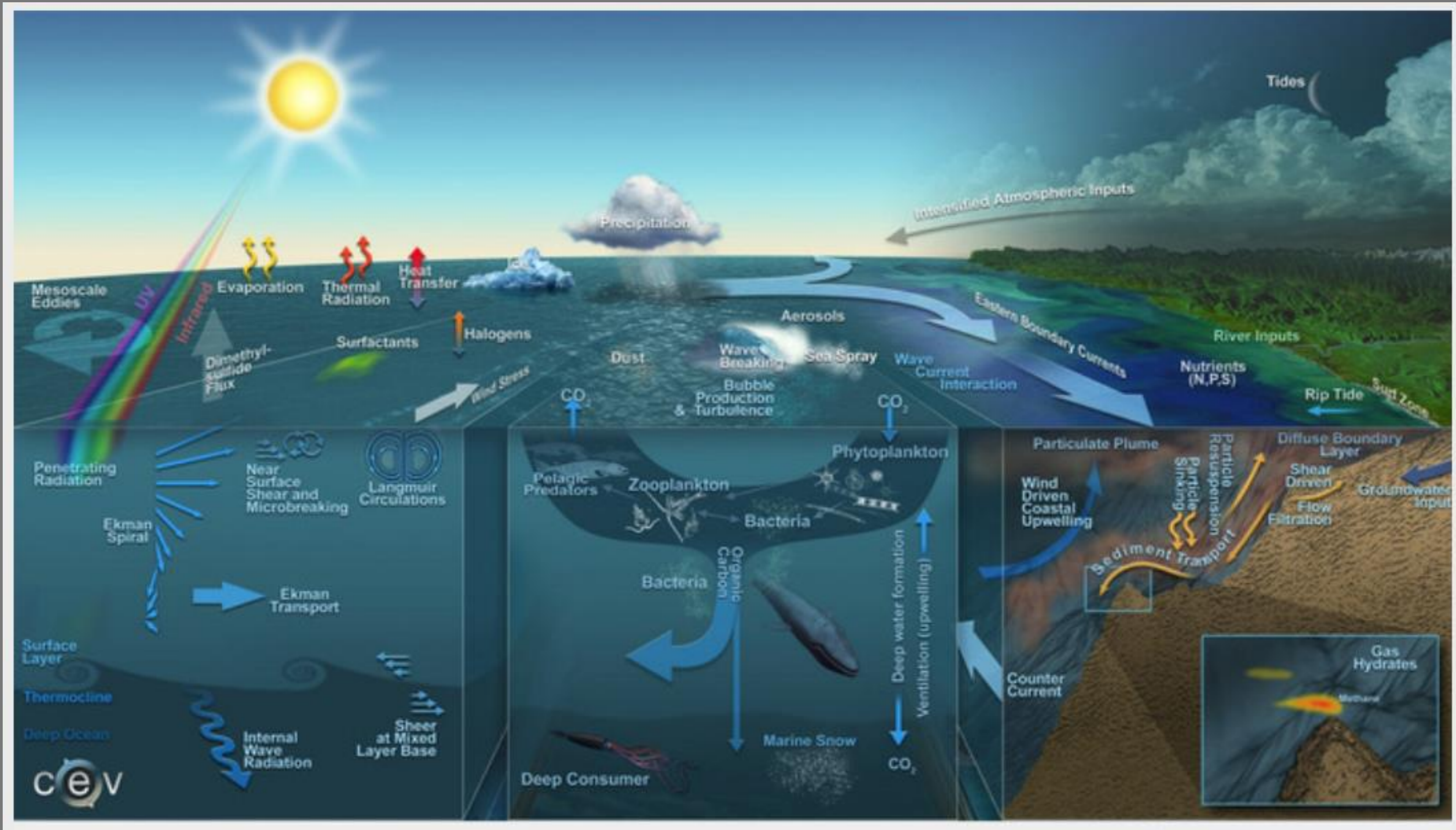
With thanks to Jim Gray

NSF's Ocean Observatory Initiative



Slide courtesy of John Delaney

Oceans and Life



Slide courtesy of John Delaney

PRINCETON SERIES IN MODERN OBSERVATIONAL ASTRONOMY

Statistics, Data Mining, and Machine Learning in Astronomy

A Practical Python Guide for the Analysis of Survey Data

Željko Ivezić, Andrew J. Connolly,
Jacob T. VanderPlas & Alexander Gray

Machine Learning Methods in the Environmental Sciences

Neural Networks
and **Kernels**

CAMBRIDGE

Machine Learning wins the Higgs Challenge

- Winner Gábor Melis, a graduate in software engineering and mathematics, developed an algorithm that is an ensemble of deep neural networks trained on random subsets of data provided with very little feature engineering and no physics knowledge
- Runner-up Tim Salimans, who has a PhD in Econometrics and works as a data science consultant, developed a solution he describes as a combination of a large number of boosted decision tree ensembles
- A Special High Energy Physics meets Machine Learning Award was presented to Tianqi Chen and Tong He of Team Crowwork. Their XG Boost algorithm was an excellent compromise between performance and simplicity, which could improve tools currently used in high-energy physics.



Winners of the Higgs Machine Learning Challenge: Gábor Melis and Tim Salimans (top row), Tianqi Chen and Tong He (bottom row).

The Third Age Of Computing

‘Computers For Embodiment’

‘Every 30 years there is a new wave of things that computers do. Around 1950 they began to model events in the world (*simulation*), and around 1980 to connect people (*communication*). Since 2010 they have begun to engage with the physical world in a non-trivial way (*embodiment*)’

Butler Lampson

1973: The Miracle of Xerox PARC



The Alto

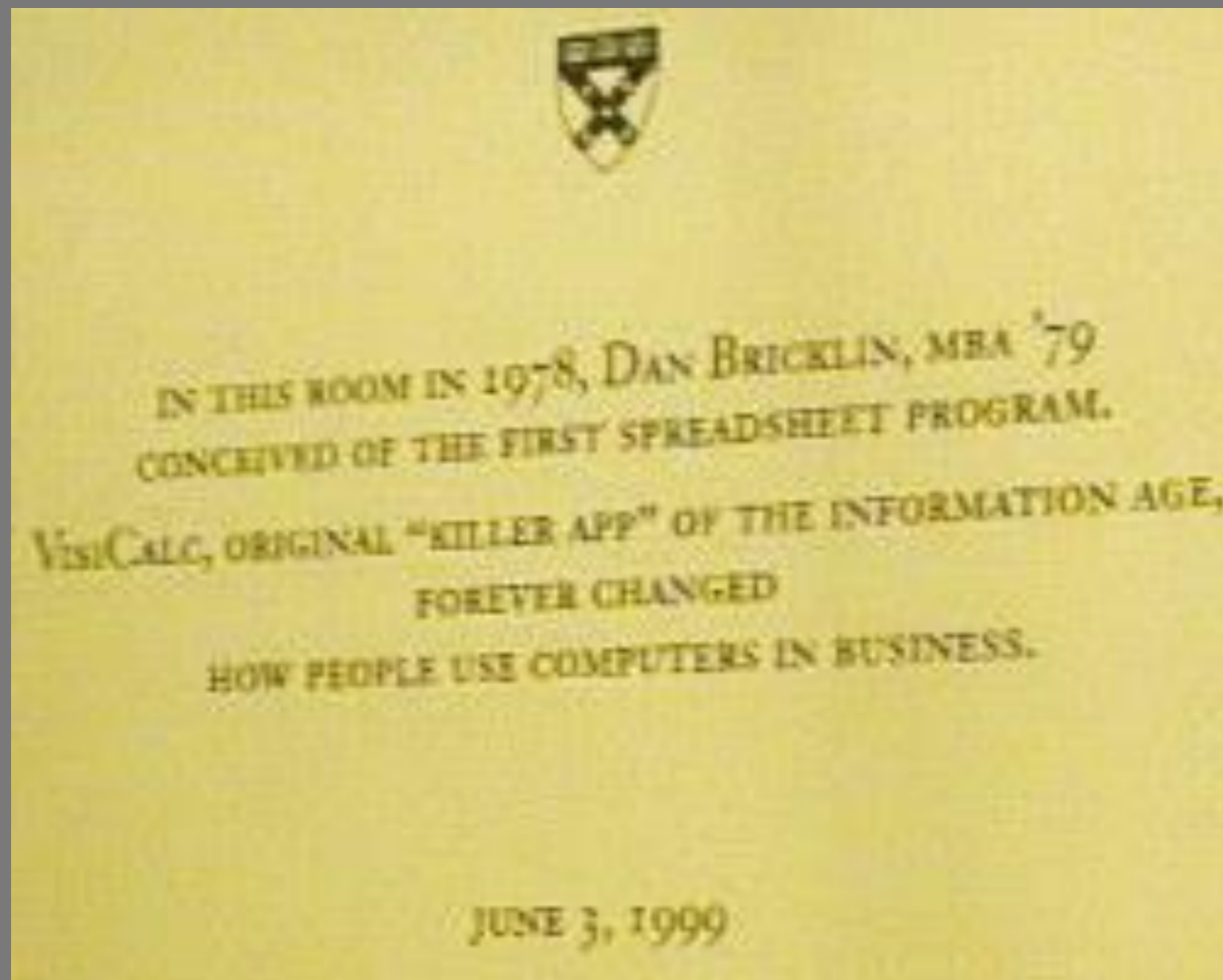
- WYSIWYG Word Processor
- Ethernet
- WIMP interface
- Laser printer



Chuck Thacker and
Butler Lampson

1976: The 'Killer App' for the PC

Commemorative plaque on the wall of Aldrich 108 in Harvard Business School



Dan Bricklin, inventor of the spreadsheet

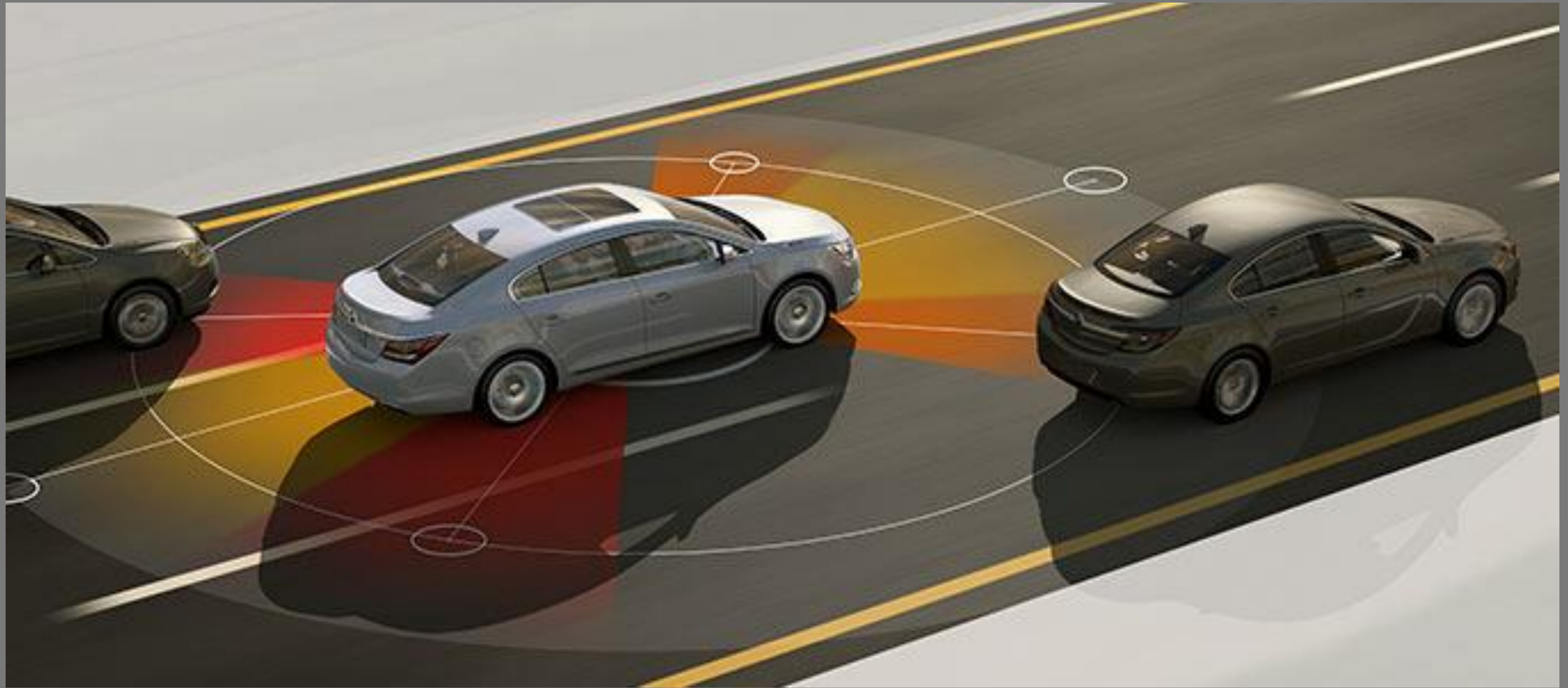
Social Computing



YOU DON'T
GET TO
500 MILLION
FRIENDS
WITHOUT MAKING
A FEW
ENEMIES

the social network

Smart Cars

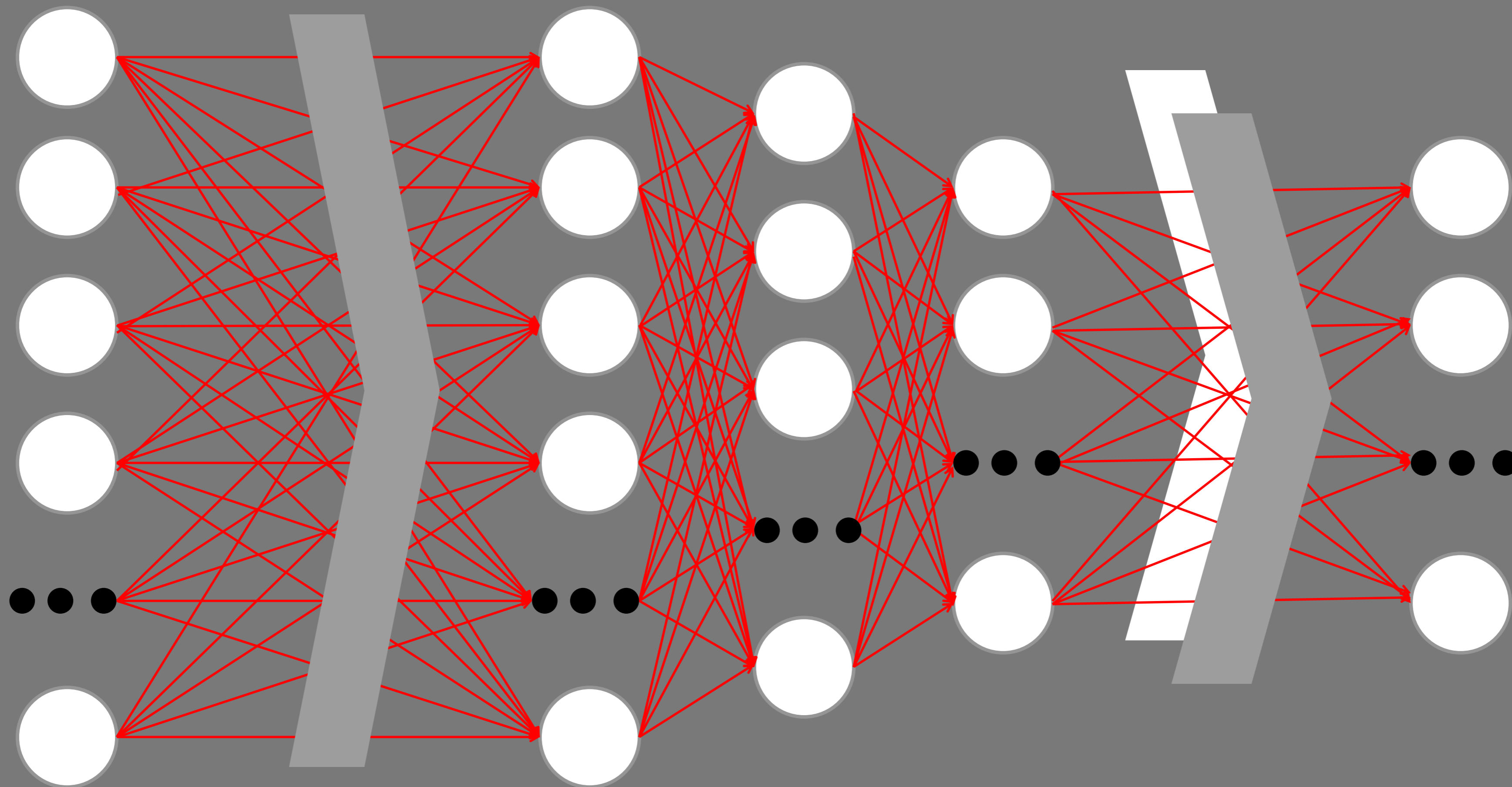


Artificial Neural Networks and Machine Learning

Input Layer

Hidden Layer

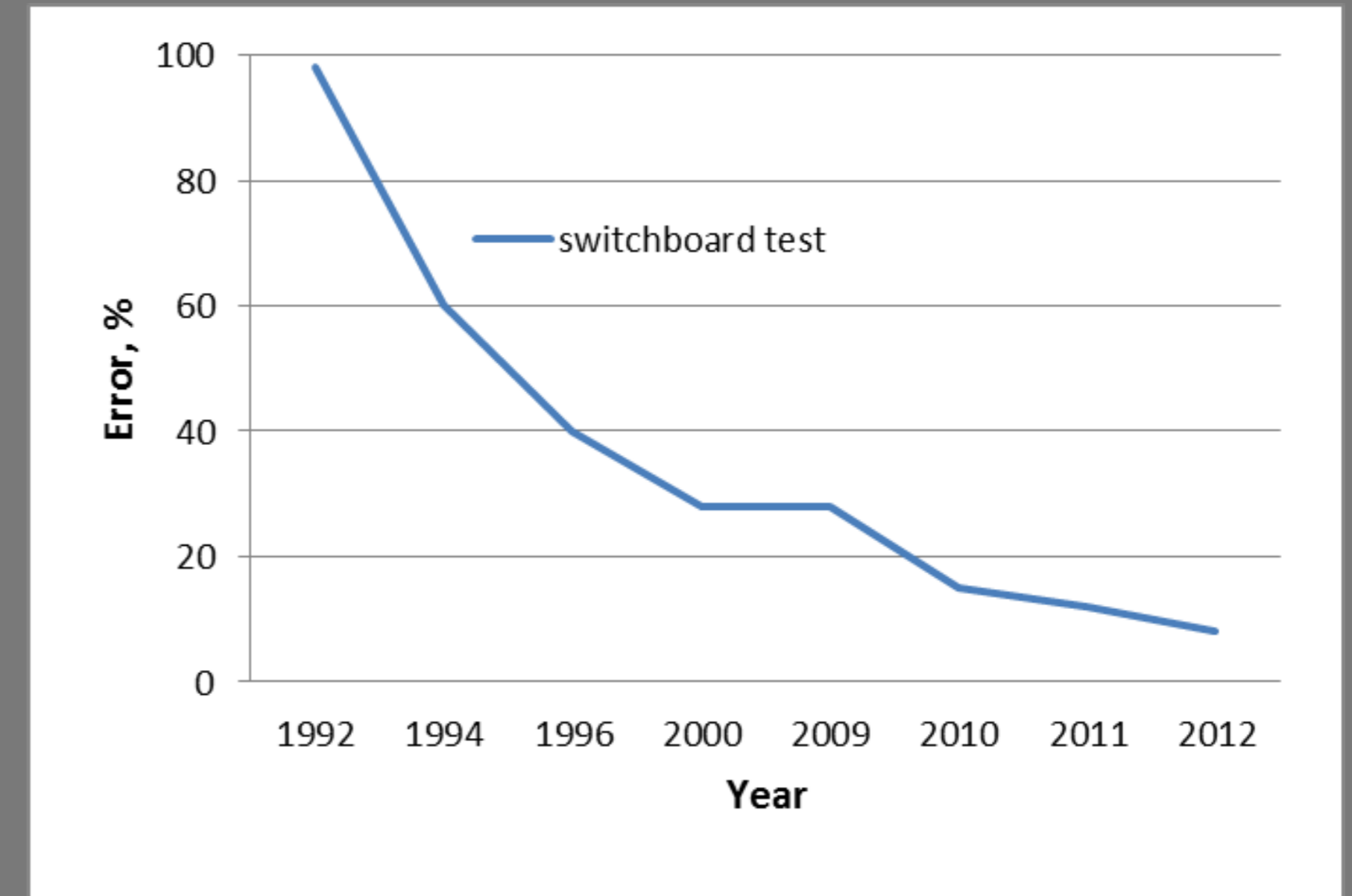
Output Layer



Machine Learning

Deep Neural Networks are now exciting the whole of the IT industry since they enable us to:

- Building computing systems that improve with experience
- Solve extremely hard problems
- Extract more value from Big Data
- Approach human intelligence
e.g. natural language processing

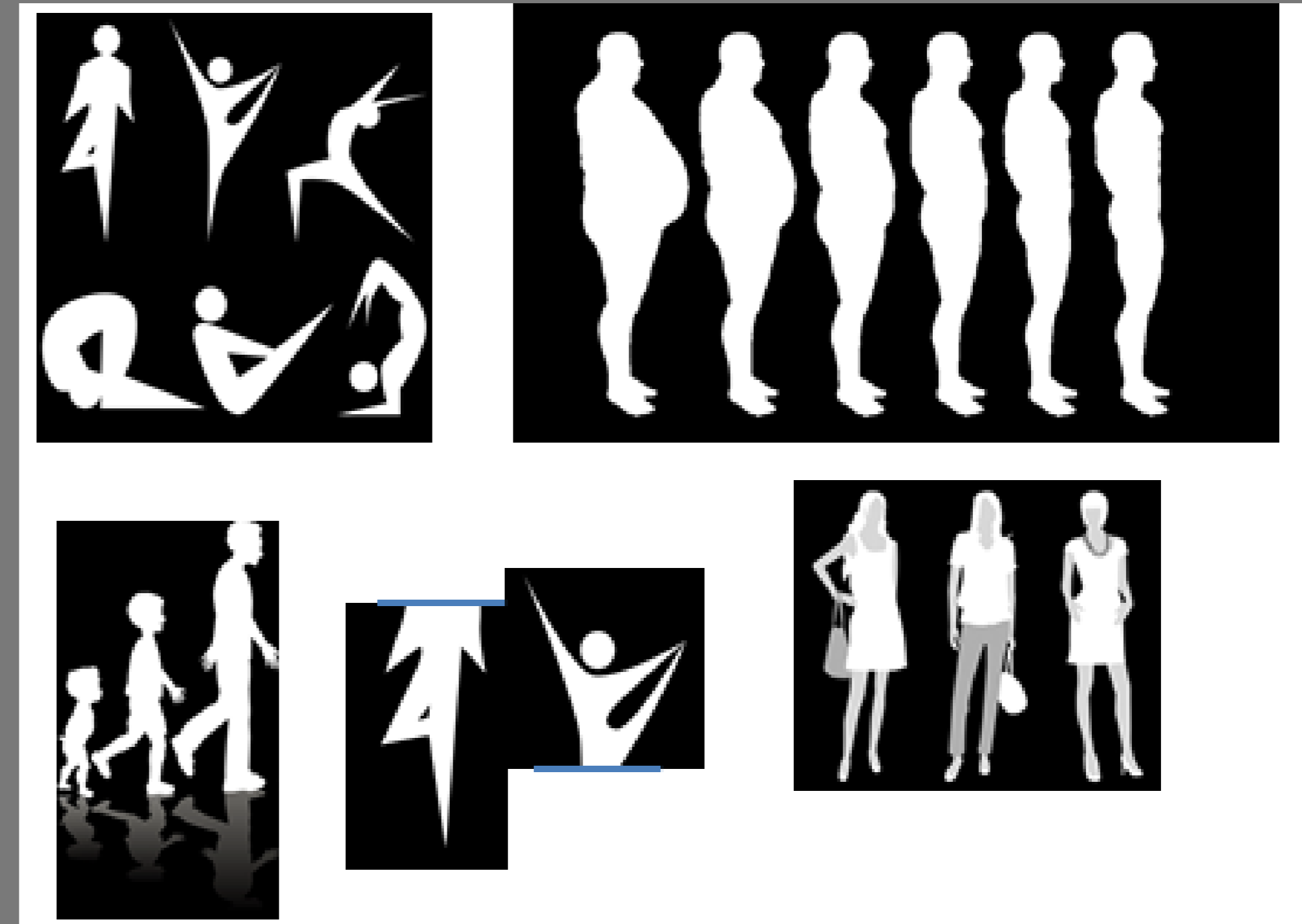


- Change in the Word Error Rate (WER) with time for the NIST "Switchboard" data.
- Dramatic improvement made in the last few years using Deep Neural Networks

Computer Vision and Machine Learning



Images from the Kinect 3D camera

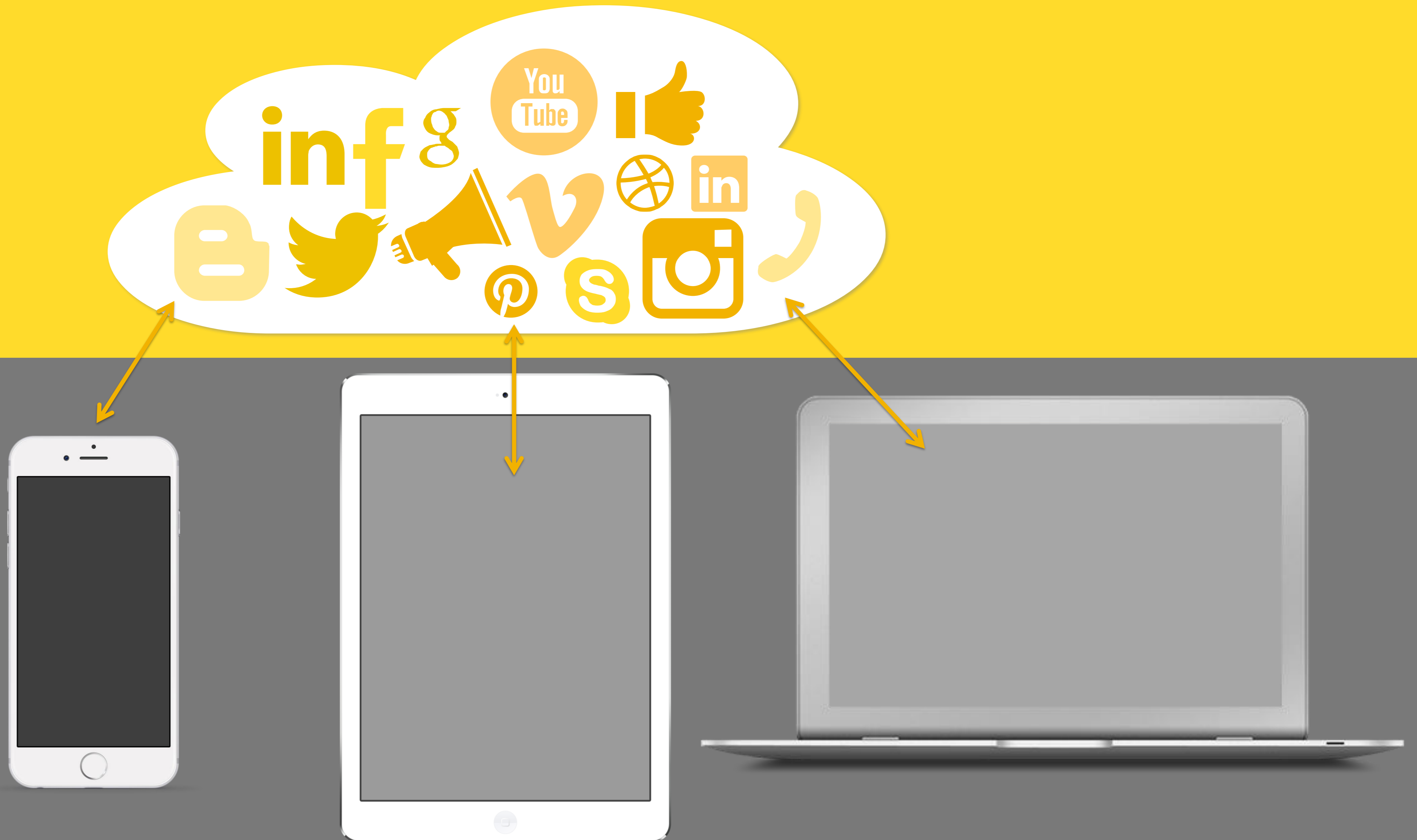


Learns from a training set containing millions of synthetically generated images

Master Chief and Cortana



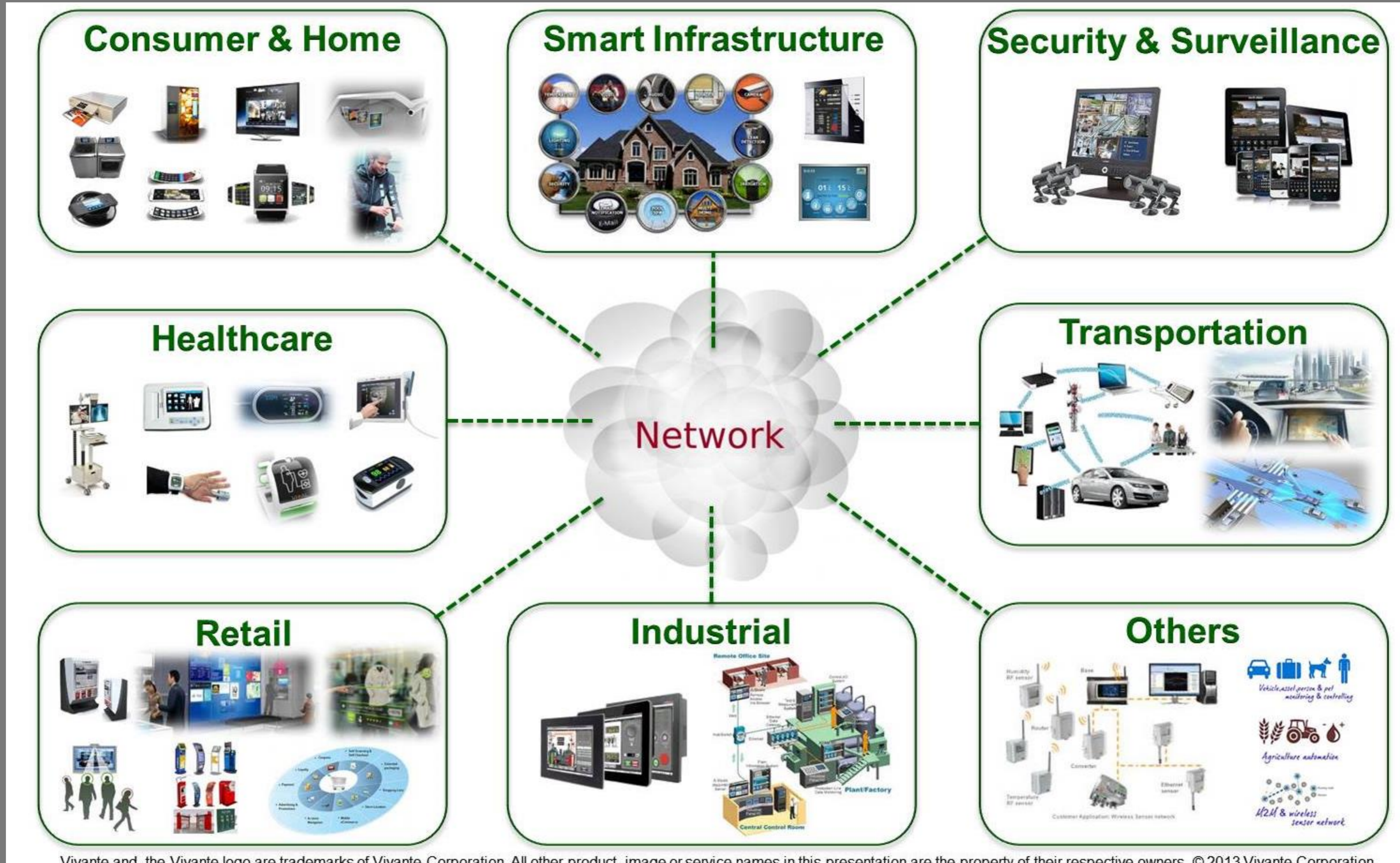
The Cloud Transforms Mobile Devices



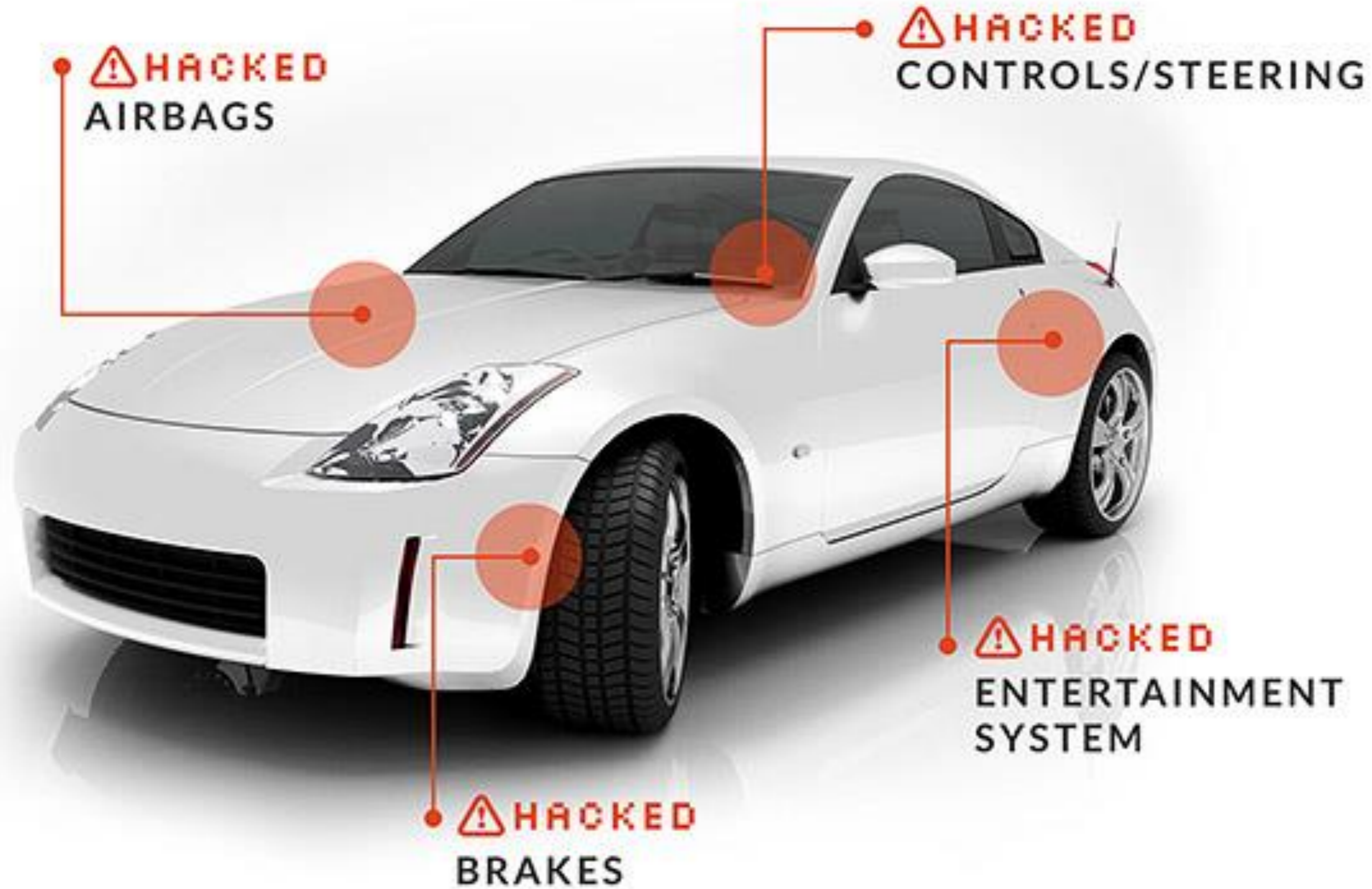
'Intelligence' everywhere ...



The Internet of Things



Car Hacking



Artificial Intelligence: Weak or Strong AI?

‘The assertion that machines could act **as if** they were intelligent is called **weak AI** hypothesis by philosophers, and the assertion that machines that do so are actually thinking (not just simulating thinking) is called the **strong AI** hypothesis’

Stuart Russell & Peter Norvig, 2010

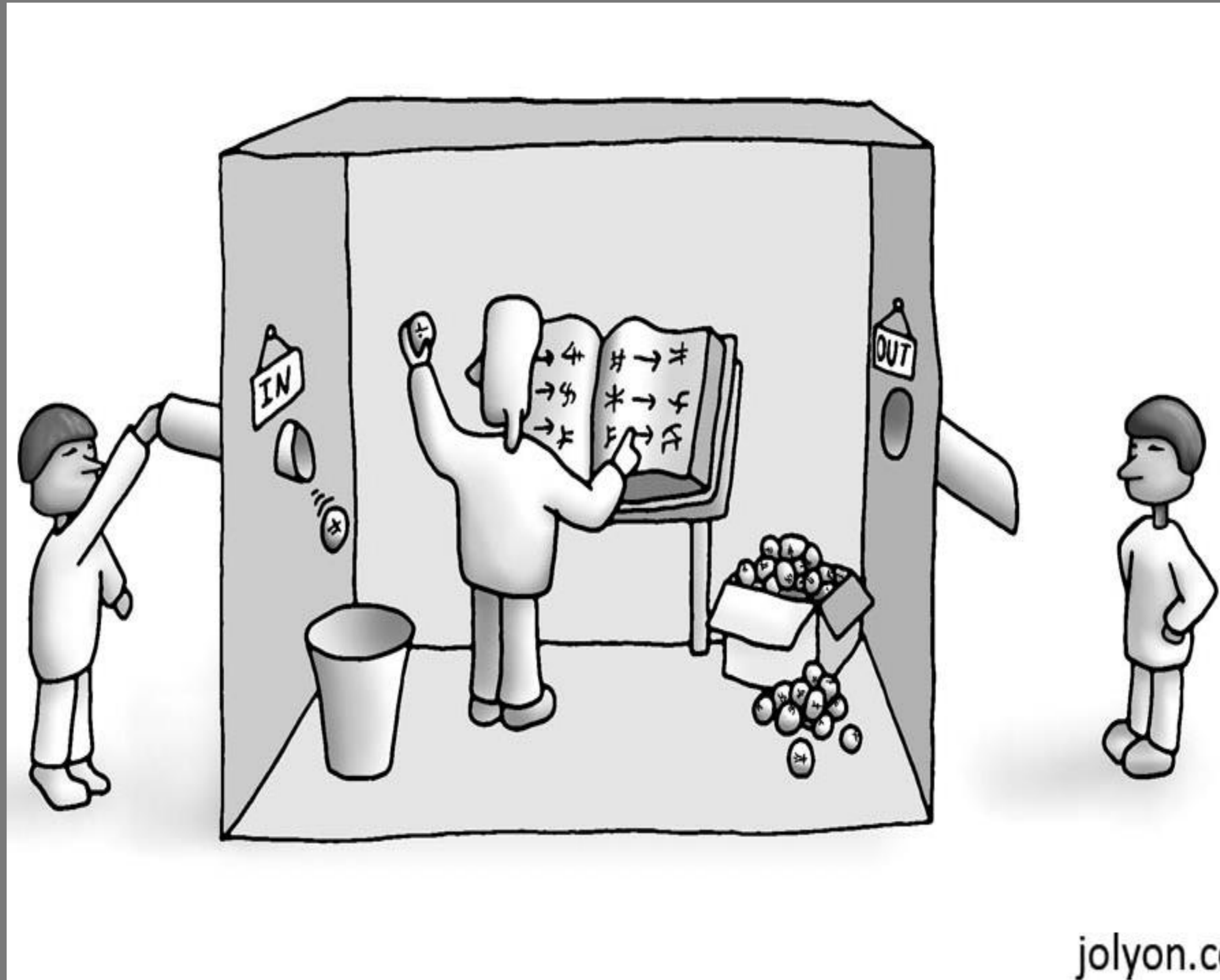
Computer Chess



IBM Watson and Jeopardy!

THE DINOSAURS	NOTABLE WOMEN	OXFORD ENGLISH DICTIONARY	NAME THAT INSTRUMENT	BELGIUM	COMPOSERS BY COUNTRY
\$200	\$200	\$200	\$200	\$200	\$200
\$400	\$400	\$400	\$400	\$400	\$400
\$600	\$600	\$600	\$600	\$600	\$600
\$800	\$800	\$800	\$800	\$800	\$800
\$1000	\$1000	\$1000	\$1000	\$1000	\$1000

John Searle's Chinese Room







ange without prior notice

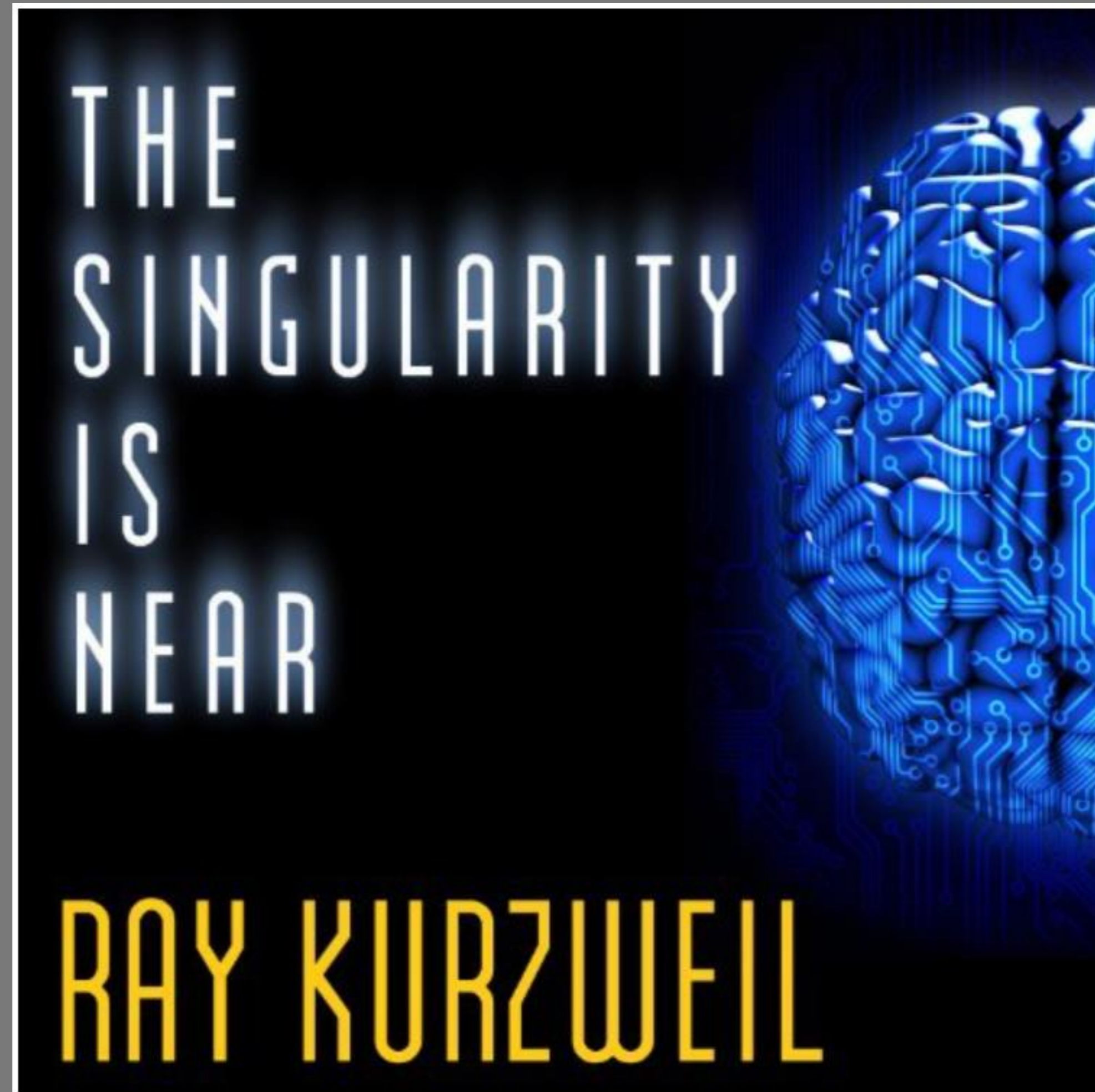
JOAQUIN PHOENIX AMY ADAMS ROONEY MARA
OLIVIA WILDE — SCARLETT JOHANSSON

her

A SPIKE JONZE LOVE STORY

CASTING BY JONATHAN HAYES. COSTUME DESIGNER: JESSICA WATSON. HAIR: JESSICA WATSON. MAKEUP: JESSICA WATSON. PRODUCTION DESIGNER: JONATHAN HAYES. EXECUTIVE PRODUCERS: JONATHAN HAYES, JONATHAN HAYES. PRODUCED BY JONATHAN HAYES. WRITTEN BY SPENCER COOPER. DIRECTED BY SPIKE JONZE. MUSIC BY DAVID JULYAN. EDITOR: JONATHAN HAYES. EXECUTIVE PRODUCERS: JONATHAN HAYES, JONATHAN HAYES. PRODUCED BY JONATHAN HAYES. WRITTEN BY SPENCER COOPER. DIRECTED BY SPIKE JONZE. MUSIC BY DAVID JULYAN. EDITOR: JONATHAN HAYES.

A Fourth Age of Computing ...

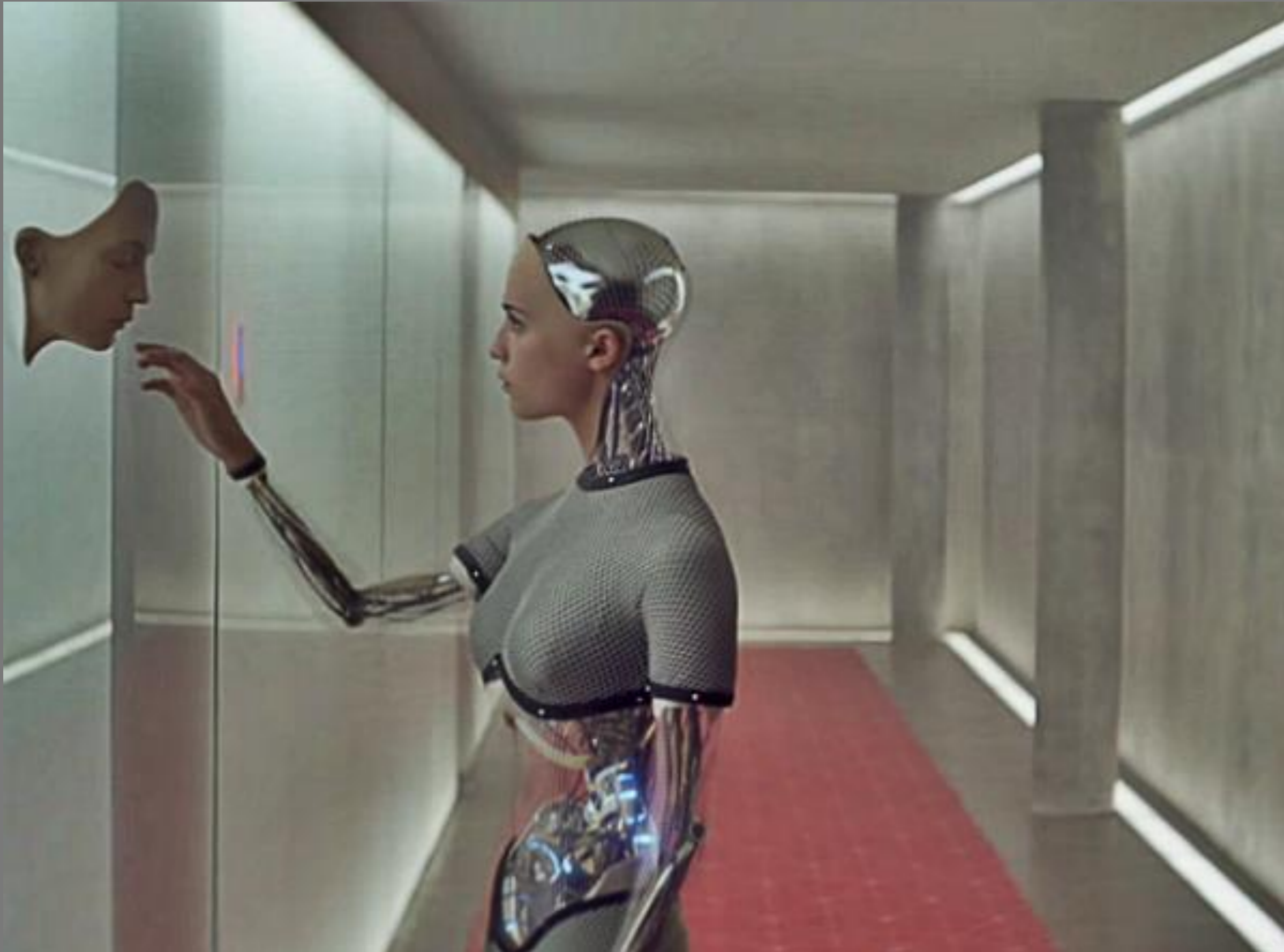


... or Feynman's dumb file clerk?

‘The inside of a computer is as dumb as hell but it goes like mad!’



Sentient Computers – The Singularity?



Ex Machina

‘Human consciousness is just about the last surviving mystery’

Daniel Dennett

