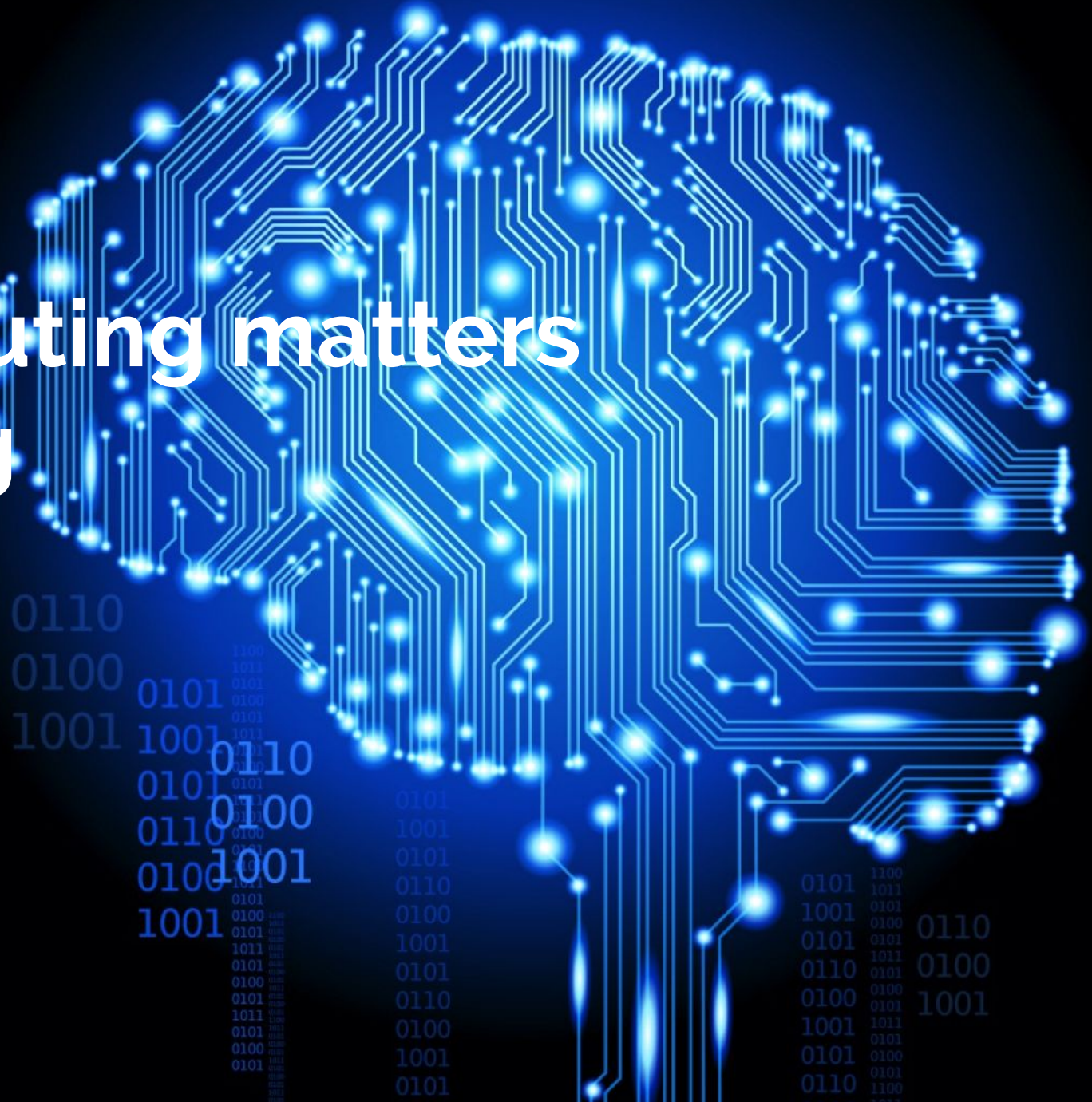


Why Supercomputing matters to Deep Learning

ESADE – MIBA (FALL 2017)

JORDI TORRES | FRANCESC SASTRE



Artificial Intelligence is changing our life





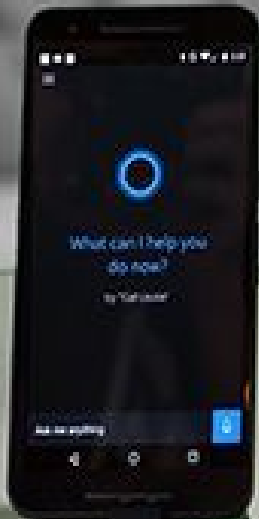
Quantum leaps in the quality of a wide range of everyday technologies thanks to Artificial Intelligence

Speech
Recognition

We are increasingly interacting with “our” computers by just talking to them



#1. Alexa
(Amazon Echo)



#2. Cortana
(Windows 10 Phone)

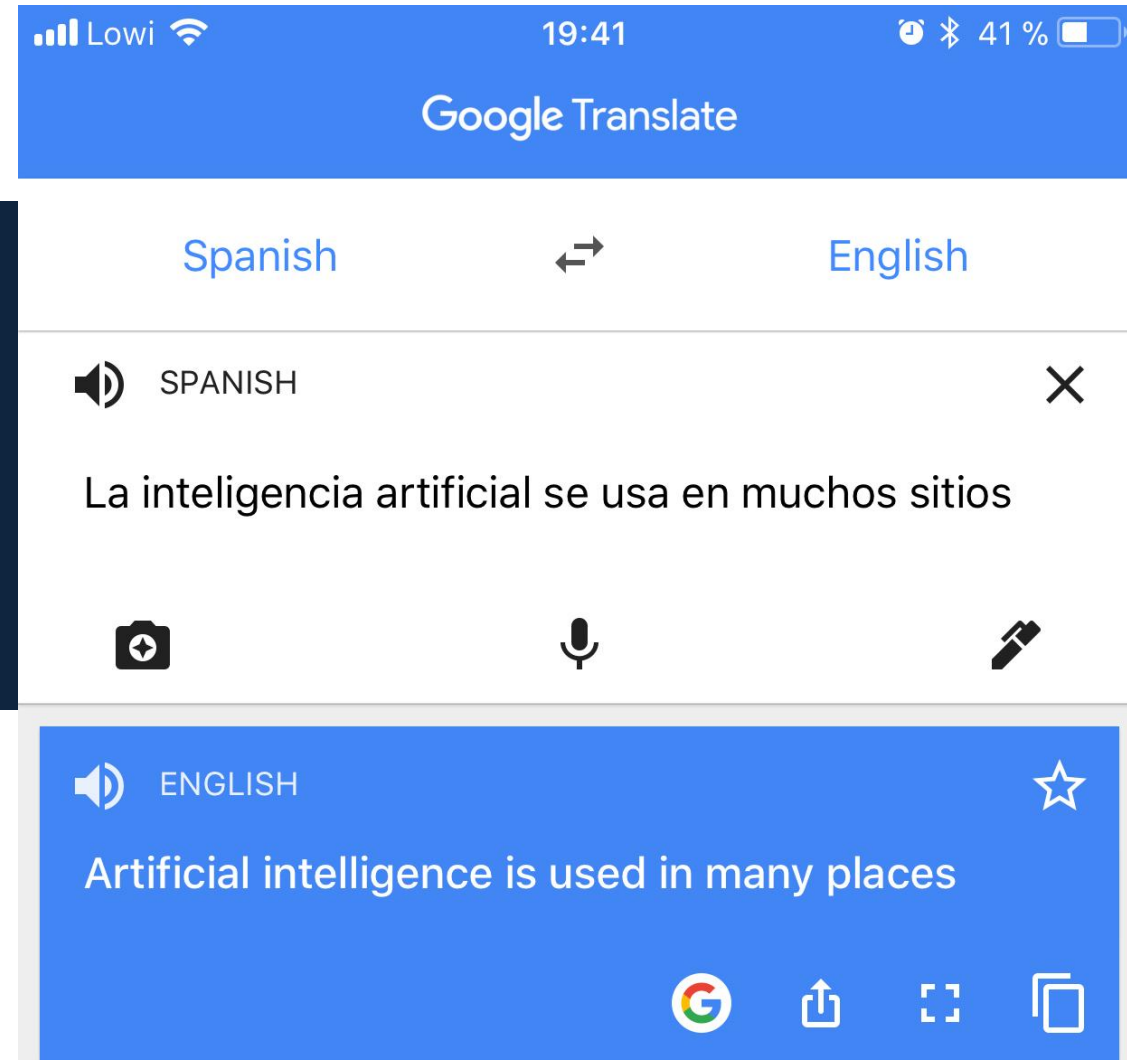


#3. Siri
(iPhone)



#4. Google Now
(Android)

Google Translate now renders spoken sentences in one language into spoken sentences in another, for **32 pairs** of languages and offers text translation for **100+ languages**.



Google Translate now renders spoken sentences in one language into spoken sentences in another, for **32 pairs** of languages and offers text translation for **100+ languages**.

Natural
Language
Processing



la inteligencia artificial se usa en muchos sitios



Artificial intelligence is used in many places.



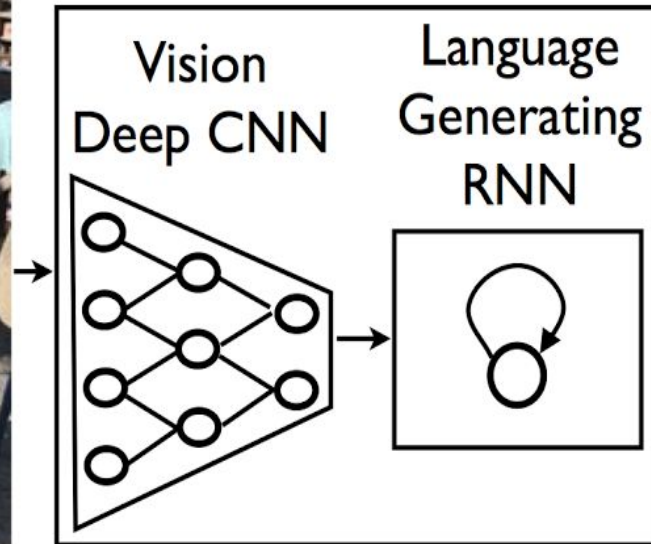
Preparing to speak...

español



English

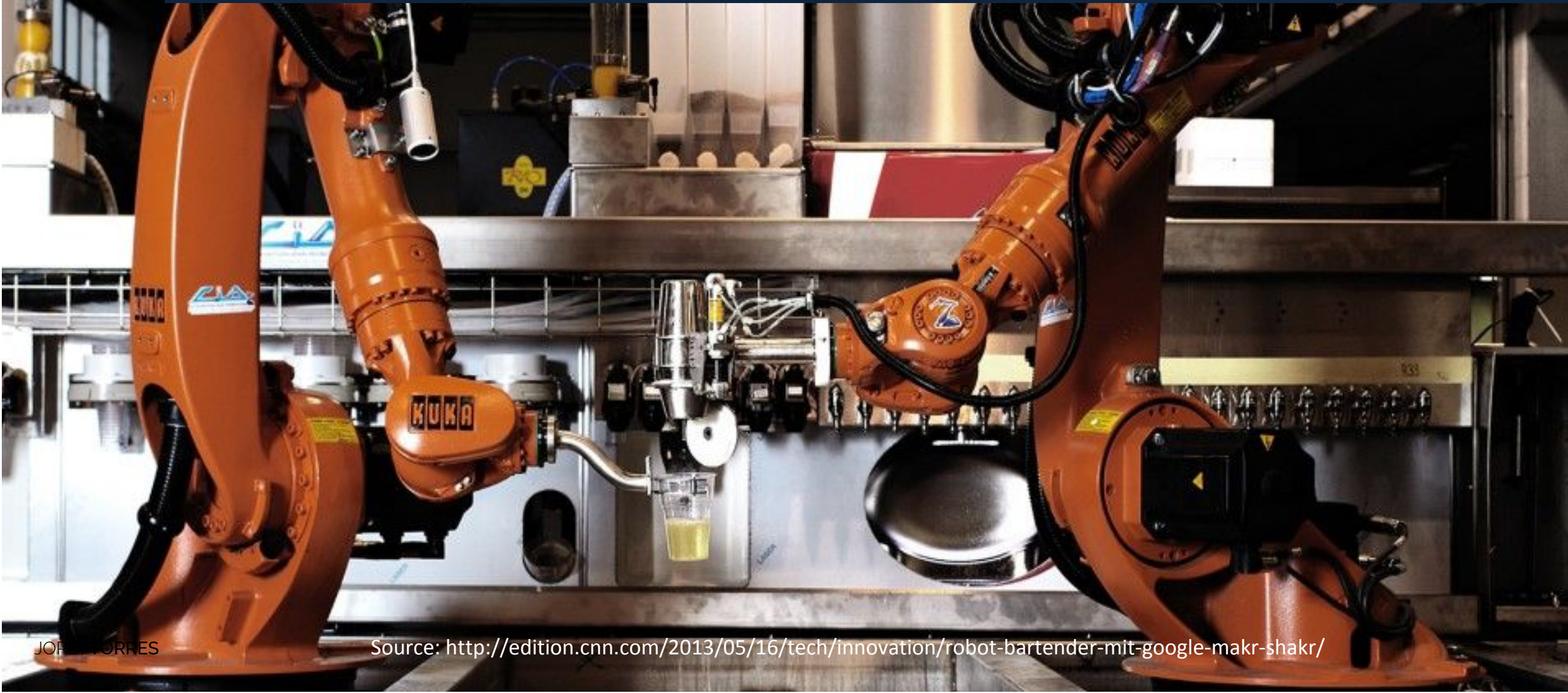
Now our computers can recognize images and generate descriptions for photos in seconds.



**A group of people
shopping at an
outdoor market.**

**There are many
vegetables at the
fruit stand.**

All these three areas are crucial to unleashing improvements in **robotics**, drones, self-driving cars, etc.



All these three areas are crucial to unleashing improvements in robotics, **drones**, self-driving cars, etc.



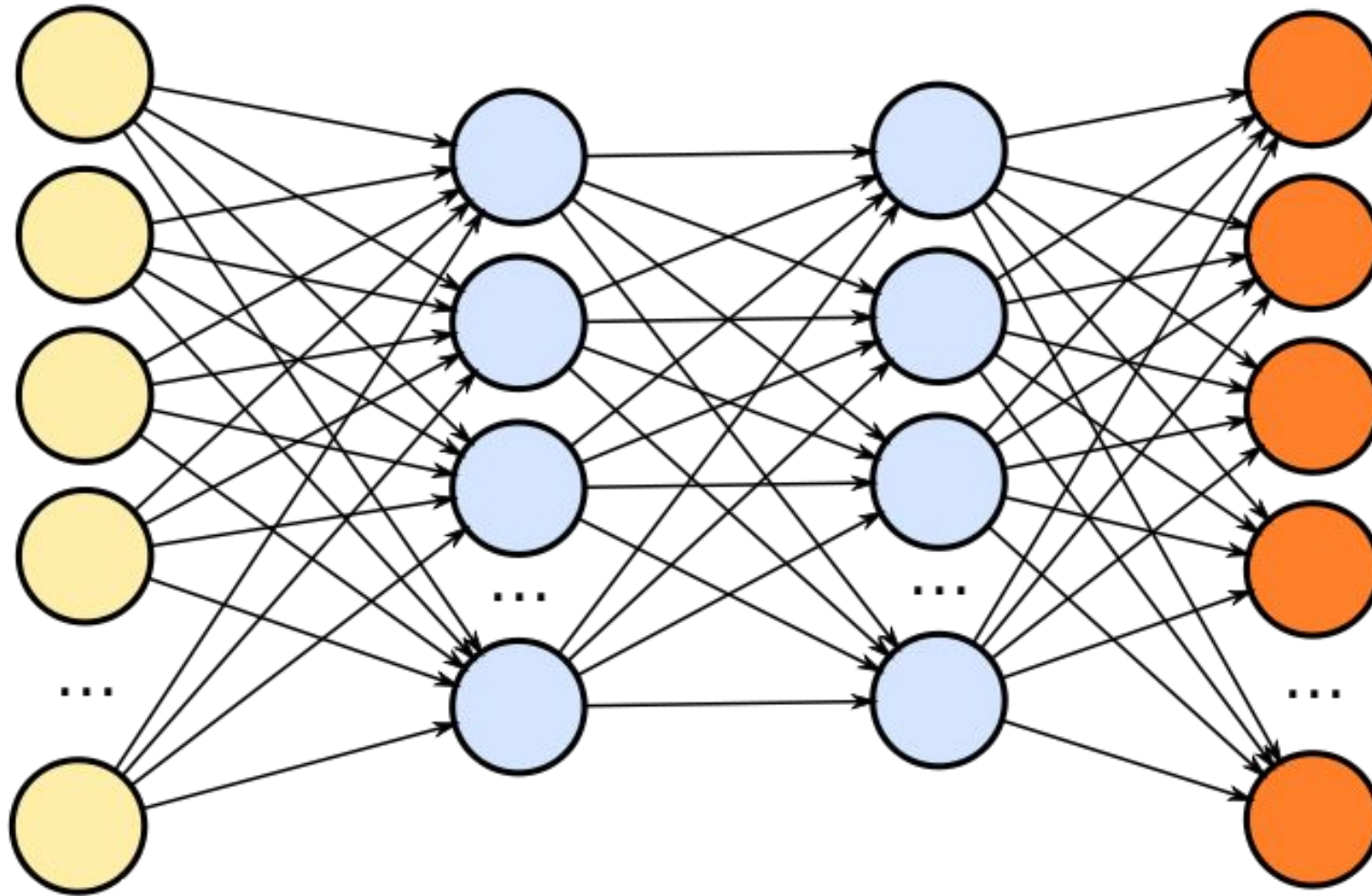
All these three areas are crucial to unleashing improvements in robotics, drones, **self-driving cars**, etc.

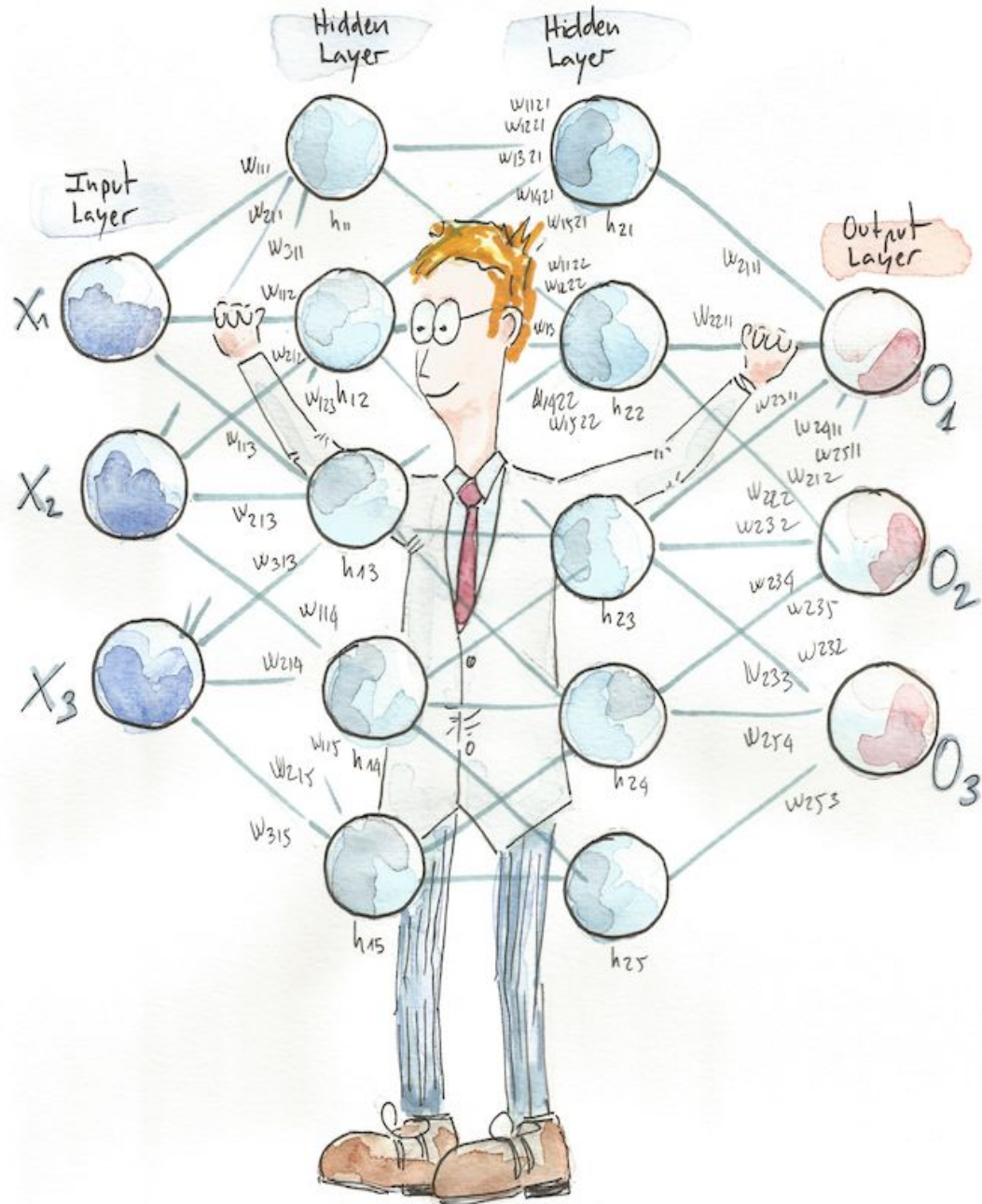


AI is at the heart of today's technological innovation.



Many of these breakthroughs have been made possible by a family of AI known as Neural Networks





Neural networks, also known as a **Deep Learning**, enables a computer to learn from observational data

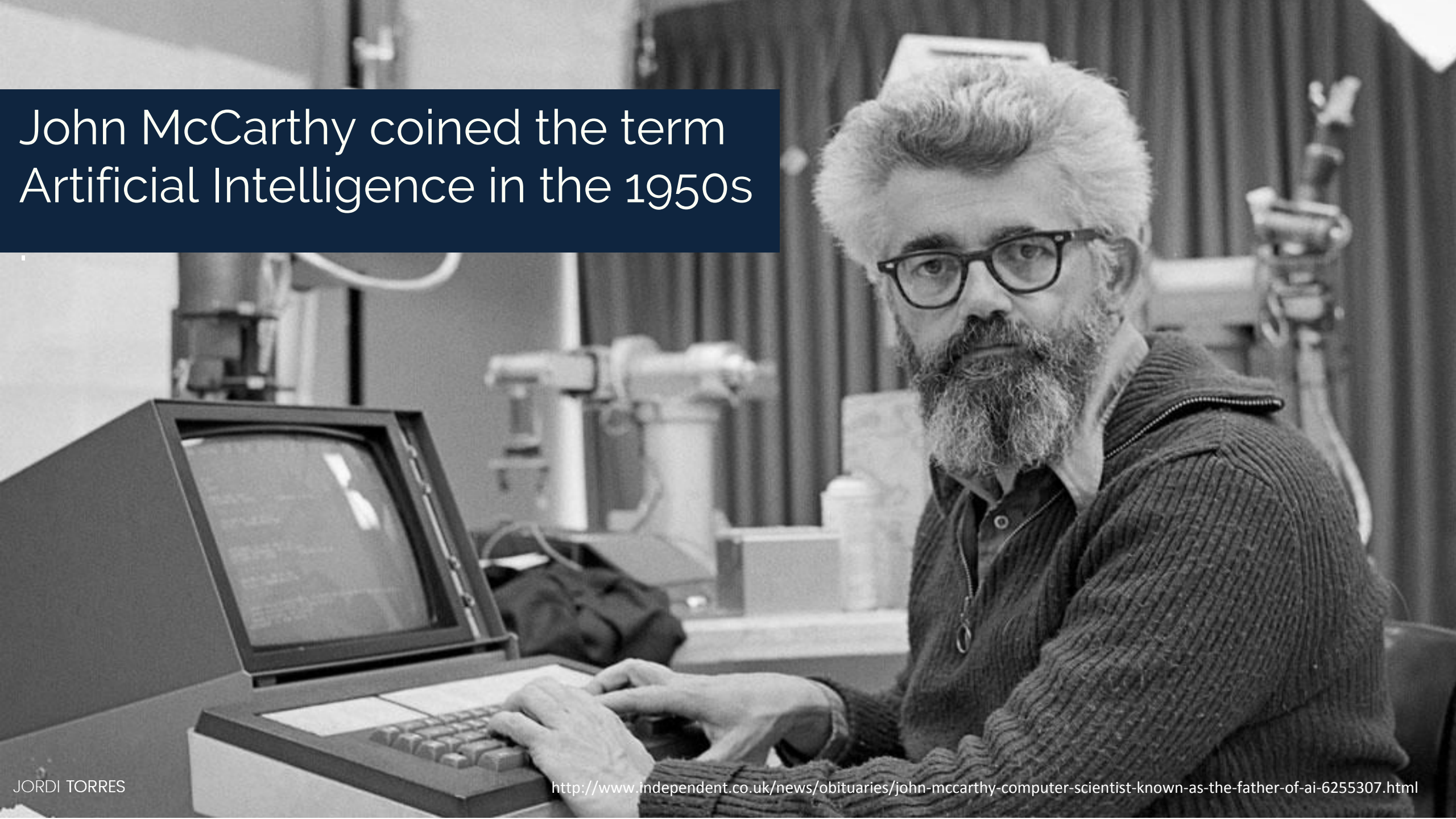
Although the greatest impacts of deep learning may be obtained when it is integrated into the whole toolbox of other AI techniques

Universitat Politècnica de Barcelona

FACULTAT D'INFORMÀTICA
GUIA DOCENT

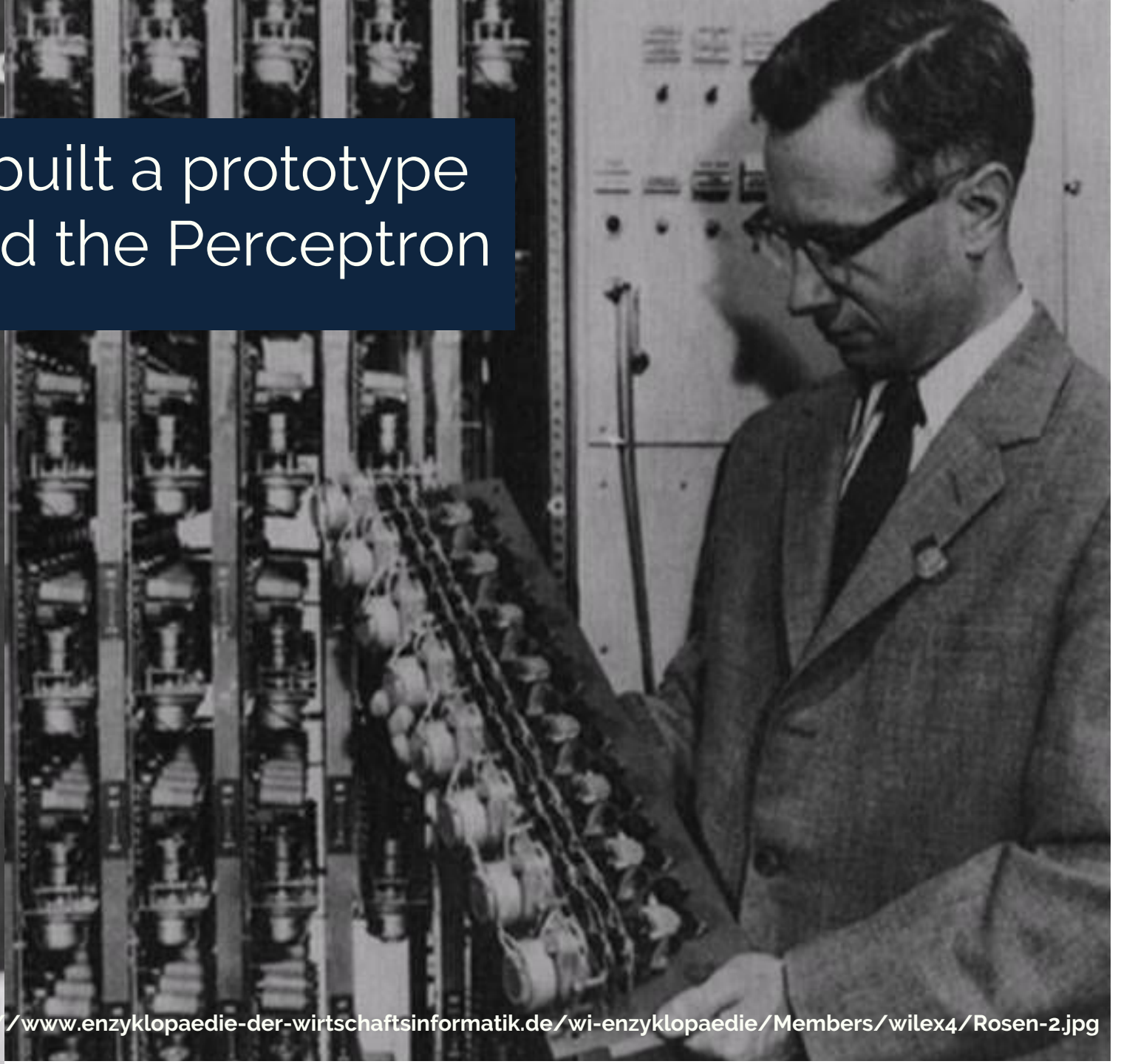
Curs 1982/83

John McCarthy coined the term
Artificial Intelligence in the 1950s



perceptro

In 1958 Frank Rosenblatt built a prototype neural net, which he called the Perceptron

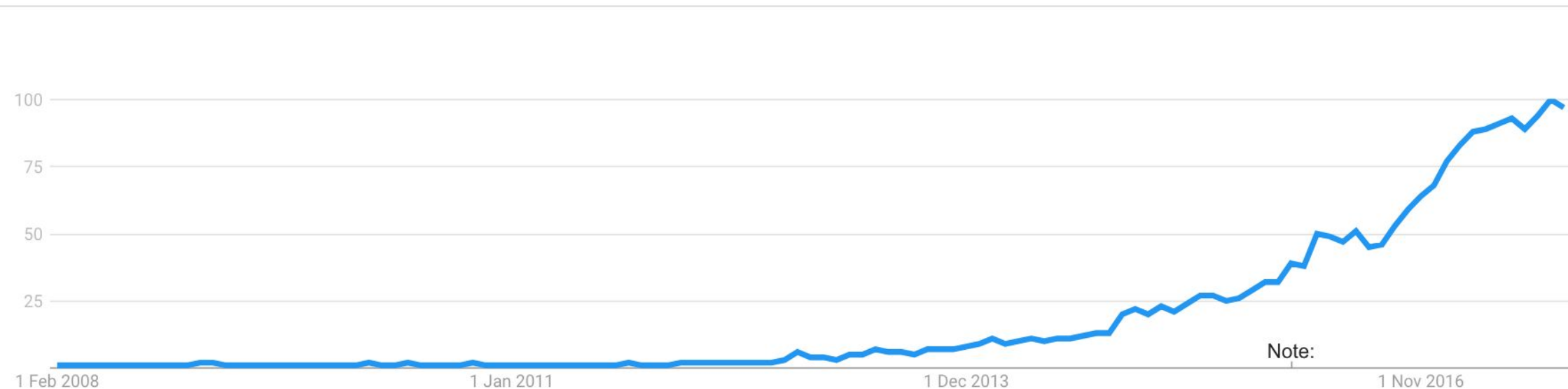


● "deep learning"
Search term

So why did Deep Learning only take off few years ago?

Worldwide ▼ 09/01/2008 - 01/10/2017 ▼ All categories

Interest over time ⓘ



Note:

One of the key drivers: The data deluge



One of the key drivers: The data deluge

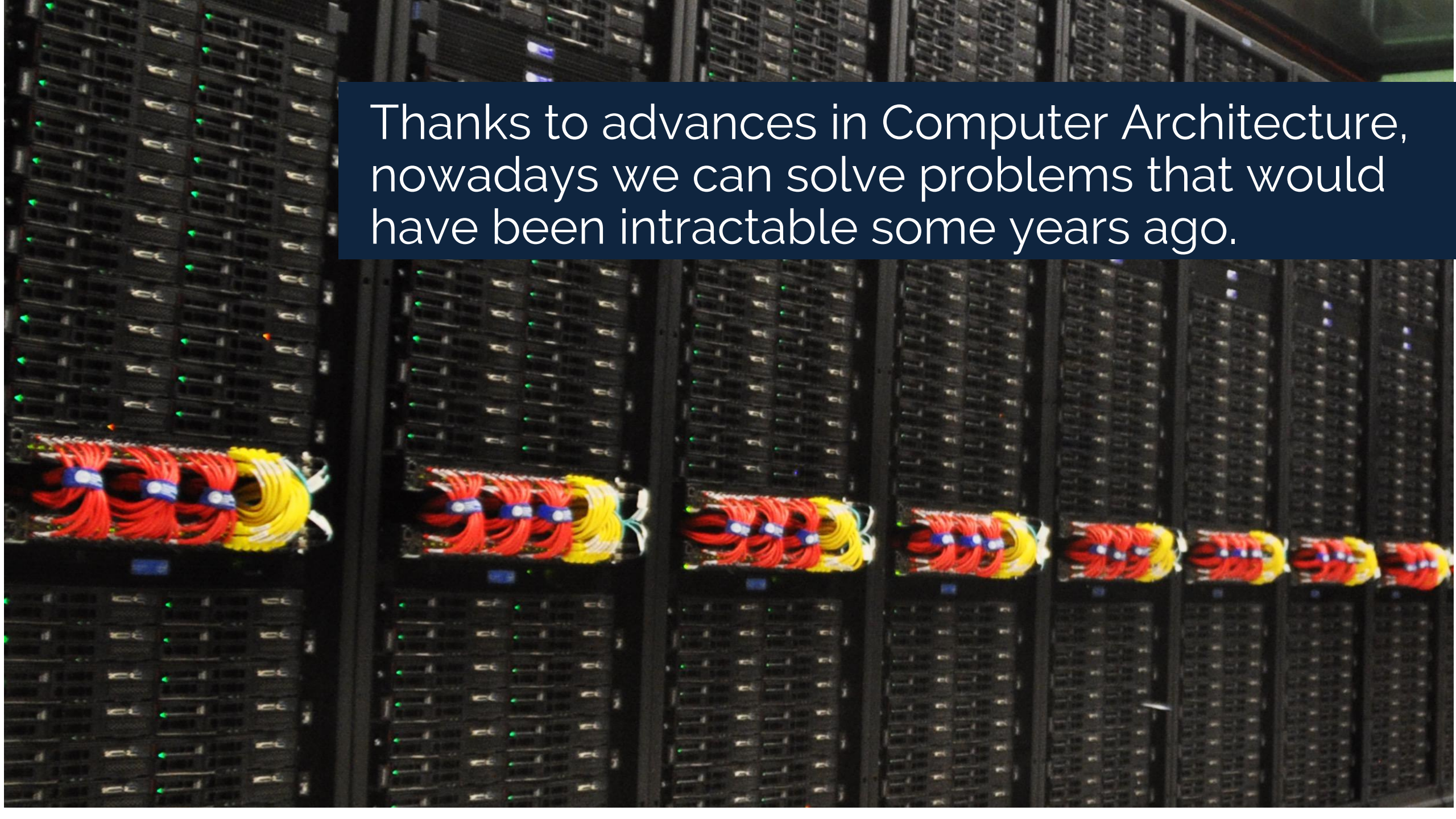
Thanks to the advent of Big Data
AI models can be “trained” by
exposing them to large data sets
that were previously unavailable.



Training DL neural nets has an insatiable demand for Computing



source: cs231n.stanford.edu/slides/2017/cs231n_2017_lecture15.pdf

A photograph of a server room. In the foreground, several bundles of network cables are neatly organized and plugged into server racks. The cables are color-coded, with red and yellow being the most prominent. The server racks are dark-colored and feature numerous green indicator lights. The background shows more server racks extending into the distance, creating a sense of depth. A dark blue semi-transparent box is overlaid on the upper portion of the image, containing white text.

Thanks to advances in Computer Architecture, nowadays we can solve problems that would have been intractable some years ago.

1982

FACOM 230 – Fujitsu

Instructions per second: few Mips * (M = 1.000.000)

Processors : 1



2012

MARENOSTRUM III - IBM

Instructions per second: 1.000.000.000 MFlops

Processors : 6046 (48448 cores)



2012

MARENOSTRUM III - IBM

Instructions per second: 1.000.000.000 MFlops

Processors : 6046 (48448 cores)

only 1.000.000.000 times faster

CPU improvements!

Until then, the increase in computational power every decade of “my” computer, was mainly thanks to CPU



CPU improvements!

Until then, the increase in computational power every decade of “my” computer, was mainly thanks to CPU

Since then, the increase in computational power for Deep Learning has not only been from CPU improvements . . .

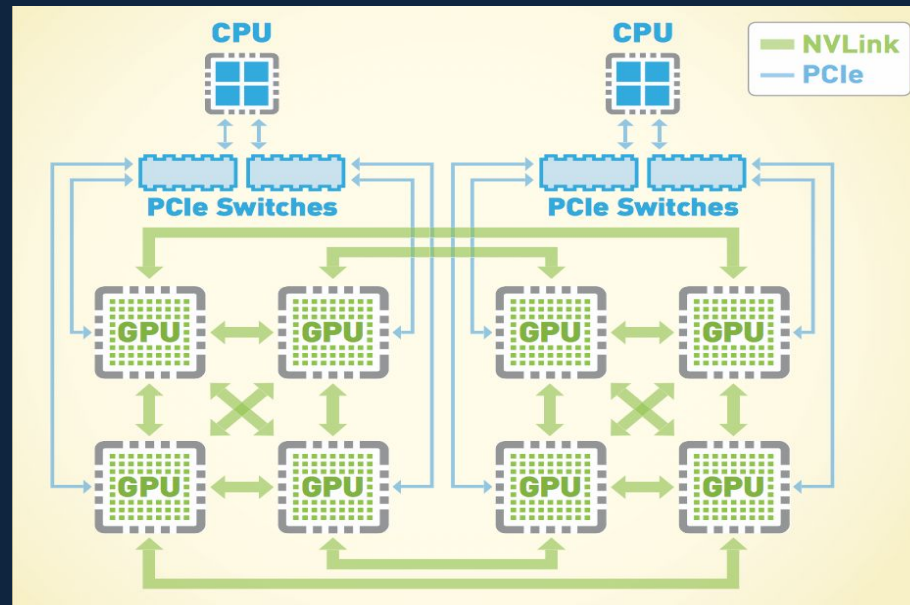


A server rack is shown with several GPU cards installed. The cards are illuminated with a bright green glow, highlighting their presence in the server. The server rack is dark, and the background is black, making the green light stand out.

but also from the realization that GPUs (NVIDIA) were
20 to 50 times more efficient than traditional CPUs.

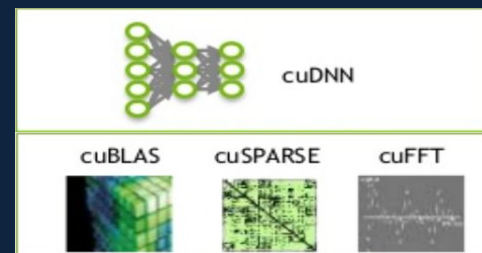
Deep Learning requires computer architecture advancements

Fast tightly coupled network interfaces



Dense computer hardware

AI specific processors



Optimized libraries and kernels

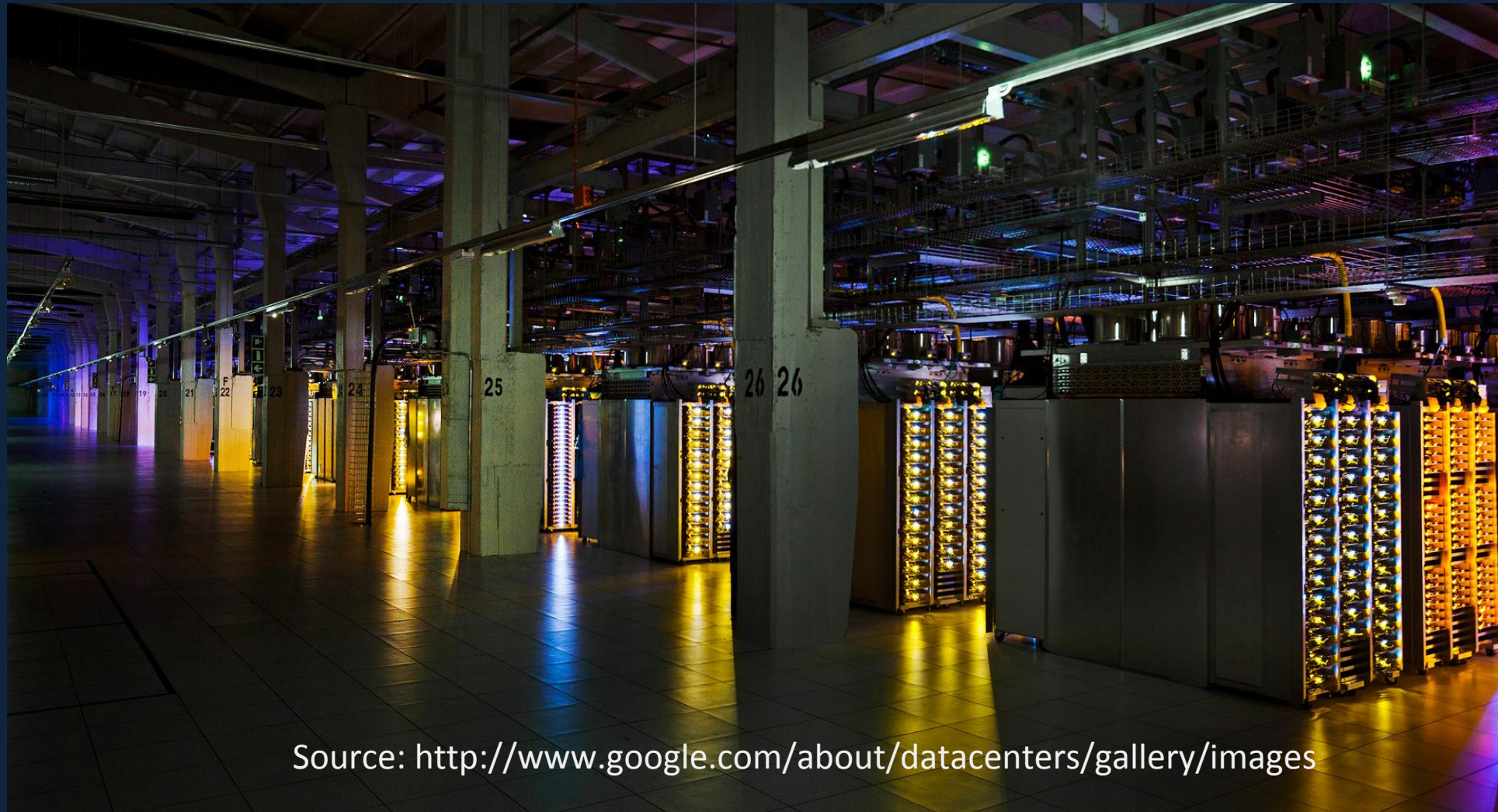


COMPUTING POWER
is the real enabler!

**Now we are entering into an era
of computation democratization
for companies !**

And what is “my/your” computer like now?

And what is “my/your” computer like now?



Source: <http://www.google.com/about/datacenters/gallery/images>

And what is “my/your” computer like now?



Source: <http://www.google.com/about/datacenters/gallery/images>

Huge data centers!



28.000 m2



Foto: Google

28.000 m²



Foto: Google

28.000 m²



Foto: Google

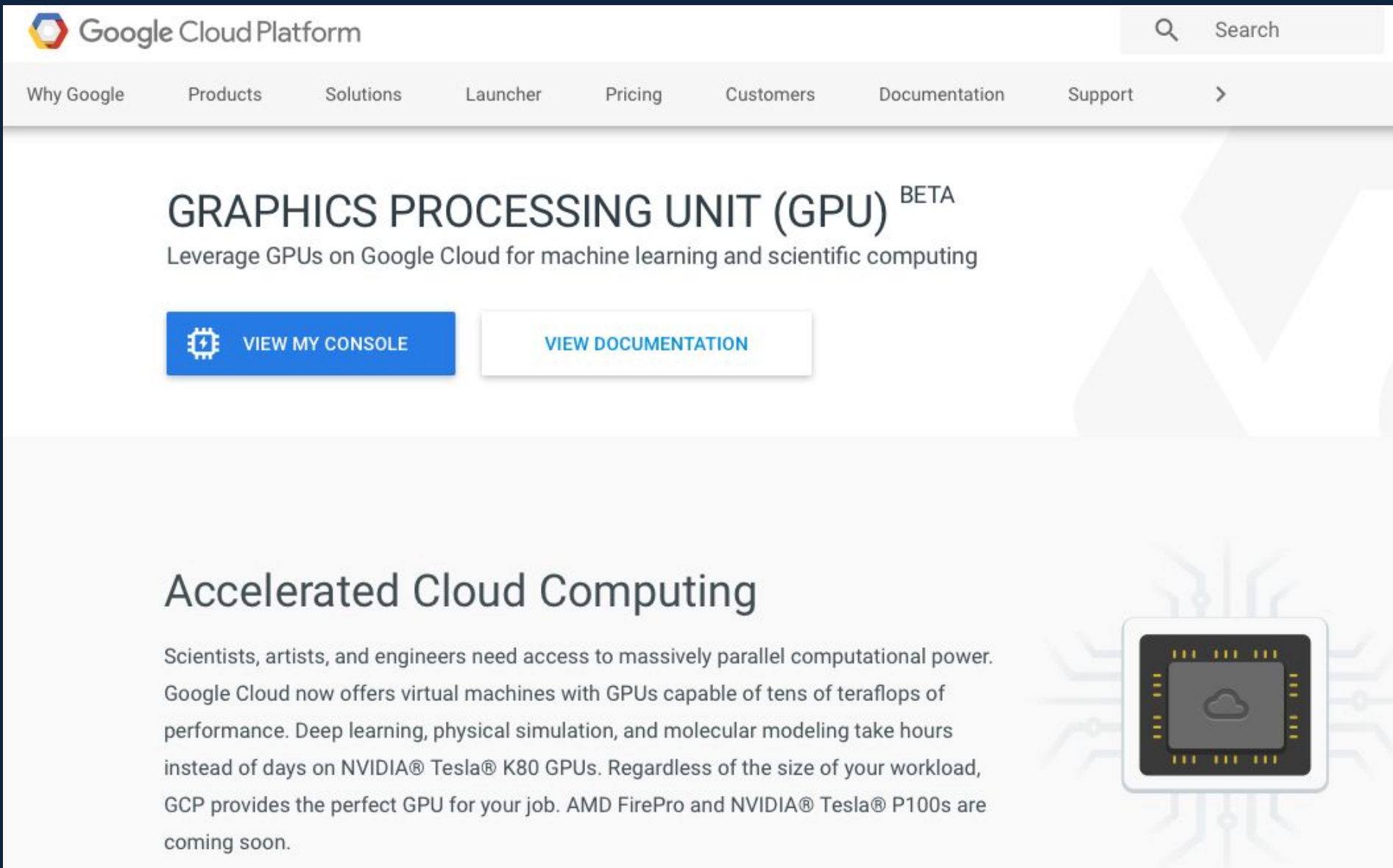
28.000 m²

For those (*experts*) who want to develop their own software, cloud services like Amazon Web Services provide GPU-driven deep-learning computation services

The screenshot shows a web browser window with the URL <https://aws.amazon.com/blogs/aws/new-p2-instance-type-for-amazon-ec2-up-to-16-gpus/>. The page header includes the Amazon Web Services logo, a navigation menu with 'Live AWS re:Invent', 'Products', 'Solutions', 'Pricing', 'Software', and 'More', and a 'Menu' button. The main content area features the title 'New P2 Instance Type for Amazon EC2 – Up to 16 GPUs' by Jeff Barr, dated 29 SEP 2016. Below the title is a table with the following specifications:

Instance Name	GPU Count	vCPU Count	Memory	Parallel Processing Cores	GPU Memory	Network Performance
p2.xlarge	1	4	61 GiB	2,496	12 GiB	High
p2.8xlarge	8	32	488 GiB	19,968	96 GiB	10 Gigabit
p2.16xlarge	16	64	732 GiB	39,936	192 GiB	20 Gigabit

And Google ...




The image shows a screenshot of the Google Cloud Platform website. At the top left is the Google Cloud Platform logo. To the right is a search bar with a magnifying glass icon and the text "Search". Below the logo is a navigation menu with links: "Why Google", "Products", "Solutions", "Launcher", "Pricing", "Customers", "Documentation", "Support", and a right-pointing chevron. The main content area features a large heading "GRAPHICS PROCESSING UNIT (GPU) ^{BETA}" followed by the subtext "Leverage GPUs on Google Cloud for machine learning and scientific computing". Below this are two buttons: a blue button with a gear icon and the text "VIEW MY CONSOLE", and a white button with a blue border and the text "VIEW DOCUMENTATION". The lower section has a heading "Accelerated Cloud Computing" and a paragraph of text. To the right of the text is an illustration of a GPU chip with a cloud icon on it, set against a background of circuit lines.

Google Cloud Platform

Why Google Products Solutions Launcher Pricing Customers Documentation Support >


GRAPHICS PROCESSING UNIT (GPU) ^{BETA}

Leverage GPUs on Google Cloud for machine learning and scientific computing

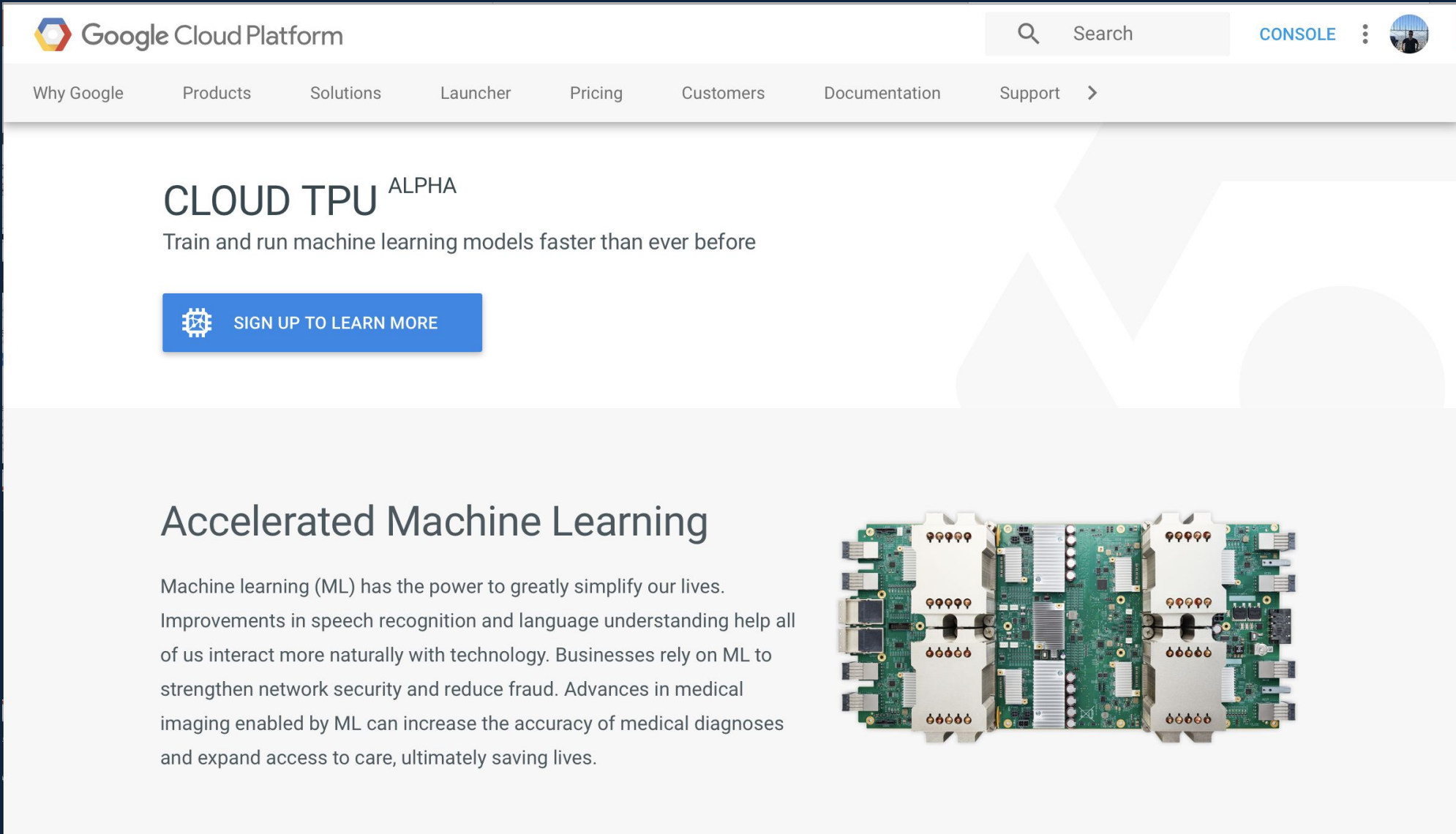
 [VIEW MY CONSOLE](#) [VIEW DOCUMENTATION](#)

Accelerated Cloud Computing

Scientists, artists, and engineers need access to massively parallel computational power. Google Cloud now offers virtual machines with GPUs capable of tens of teraflops of performance. Deep learning, physical simulation, and molecular modeling take hours instead of days on NVIDIA® Tesla® K80 GPUs. Regardless of the size of your workload, GCP provides the perfect GPU for your job. AMD FirePro and NVIDIA® Tesla® P100s are coming soon.



And Google ...



The image is a screenshot of the Google Cloud Platform website. At the top left is the Google Cloud Platform logo. To its right is a search bar with a magnifying glass icon and the word "Search". Further right is a "CONSOLE" link, a vertical ellipsis menu icon, and a user profile picture. Below the logo is a navigation menu with links for "Why Google", "Products", "Solutions", "Launcher", "Pricing", "Customers", "Documentation", and "Support" followed by a right-pointing chevron. The main content area features the heading "CLOUD TPU ALPHA" in large, bold, black letters. Below this heading is the sub-heading "Train and run machine learning models faster than ever before". A blue button with a white gear icon and the text "SIGN UP TO LEARN MORE" is positioned below the sub-heading. The background of this section is white with a large, faint, light-gray graphic of a stylized 'G' shape. Below the button, the heading "Accelerated Machine Learning" is displayed in a large, dark gray font. Underneath this heading is a paragraph of text: "Machine learning (ML) has the power to greatly simplify our lives. Improvements in speech recognition and language understanding help all of us interact more naturally with technology. Businesses rely on ML to strengthen network security and reduce fraud. Advances in medical imaging enabled by ML can increase the accuracy of medical diagnoses and expand access to care, ultimately saving lives." To the right of this text is a photograph of a Google Cloud TPU accelerator card, showing its green printed circuit board, various components, and multiple ports on the left and right sides.

Google Cloud Platform


Search

CONSOLE

Why Google Products Solutions Launcher Pricing Customers Documentation Support >


CLOUD TPU ALPHA

Train and run machine learning models faster than ever before

 SIGN UP TO LEARN MORE

Accelerated Machine Learning

Machine learning (ML) has the power to greatly simplify our lives. Improvements in speech recognition and language understanding help all of us interact more naturally with technology. Businesses rely on ML to strengthen network security and reduce fraud. Advances in medical imaging enabled by ML can increase the accuracy of medical diagnoses and expand access to care, ultimately saving lives.



And all major cloud platforms...

Microsoft Azure

IBM Cloud

Aliyun

Cirrascale

NIMBIX

Outscale

...

Cogeco Peer 1

Penguin Computing

RapidSwitch

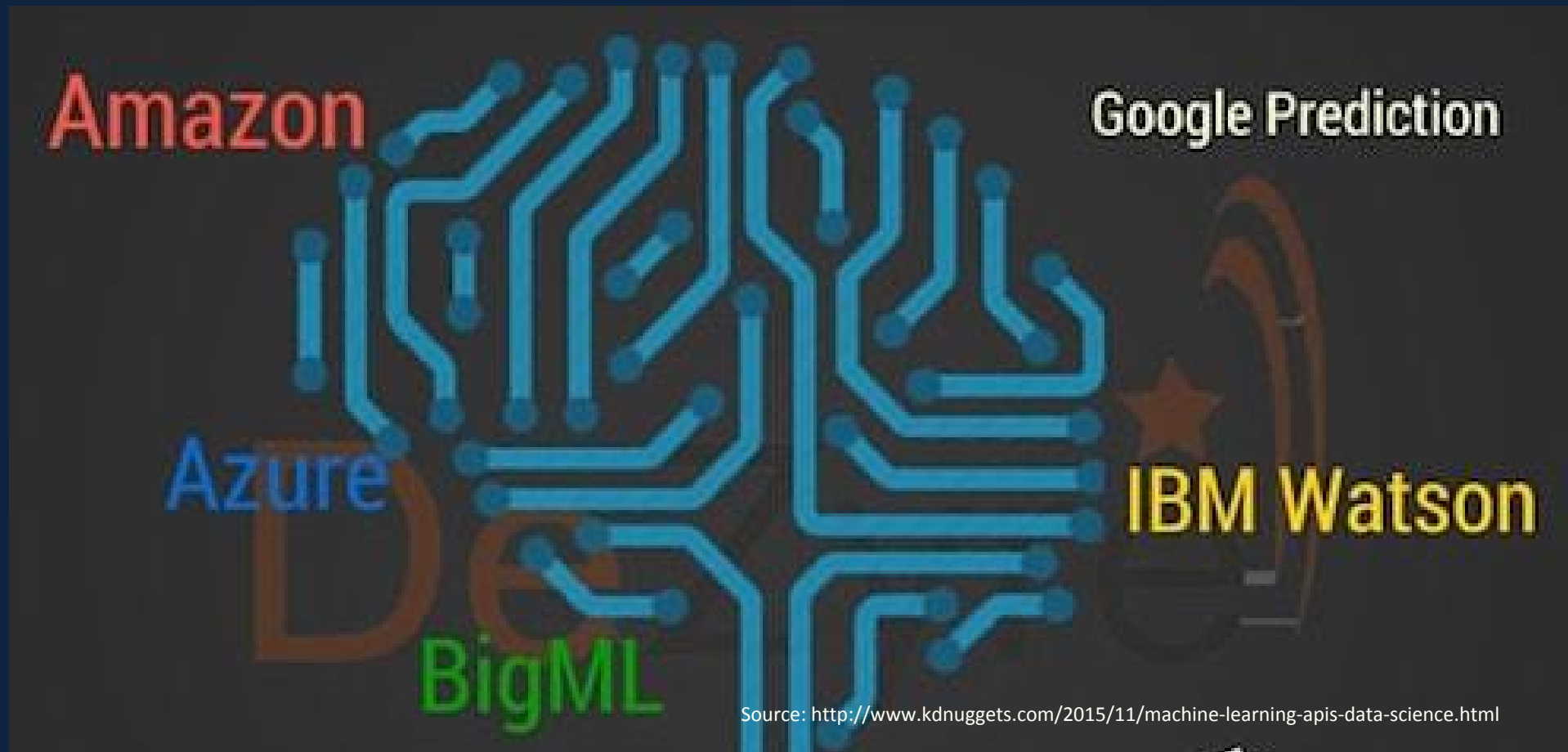
Rescale

SkyScale

SoftLayer

...

And for “less expert” people, various companies are providing a working scalable implementation of ML/AI algorithms as a Service (**AI-as-a-Service**)



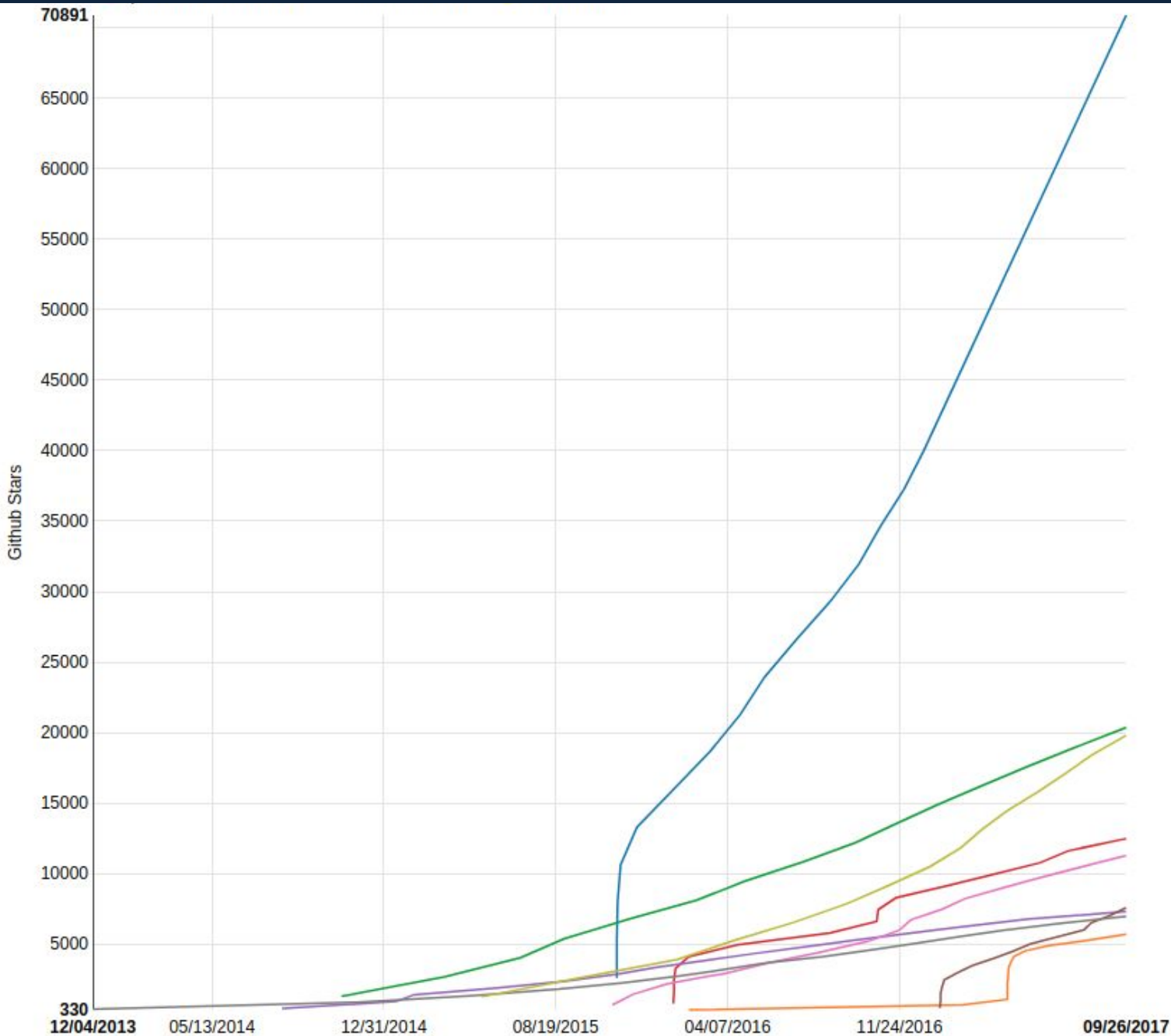
**An open-source world for the
Deep Learning community**

Many **open-source DL software**
have greased the innovation process



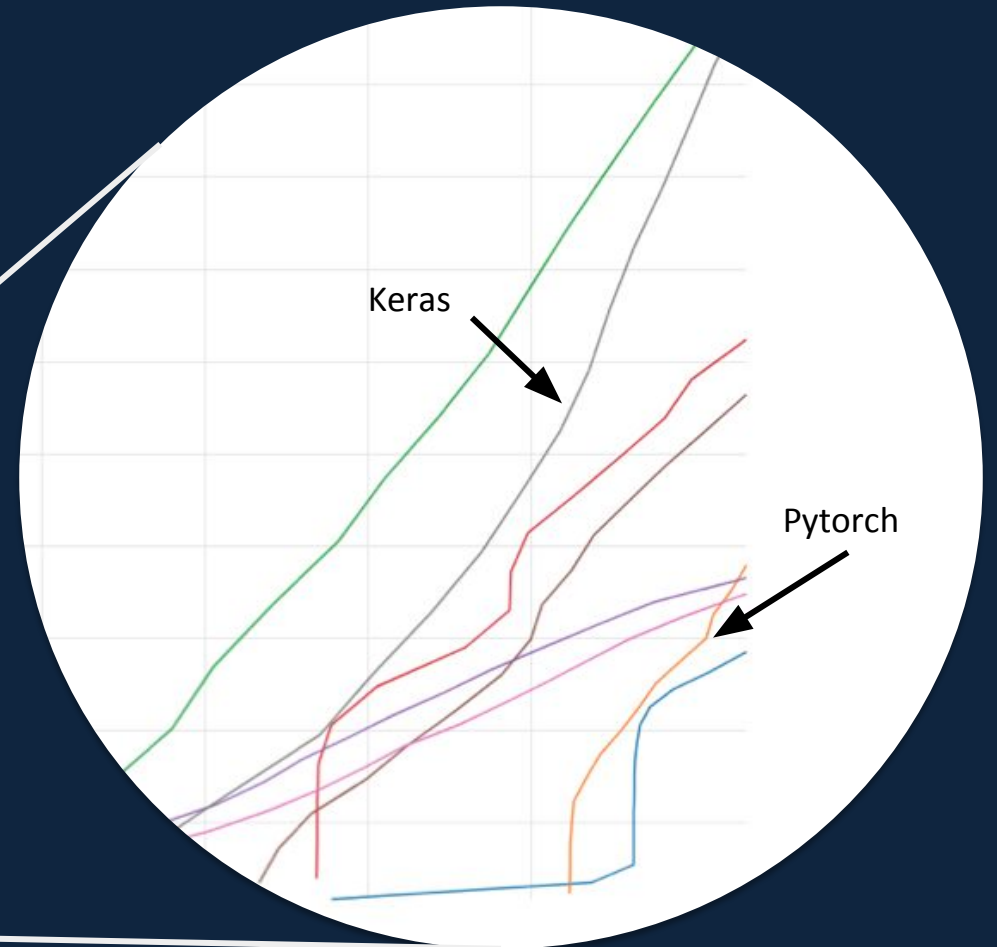
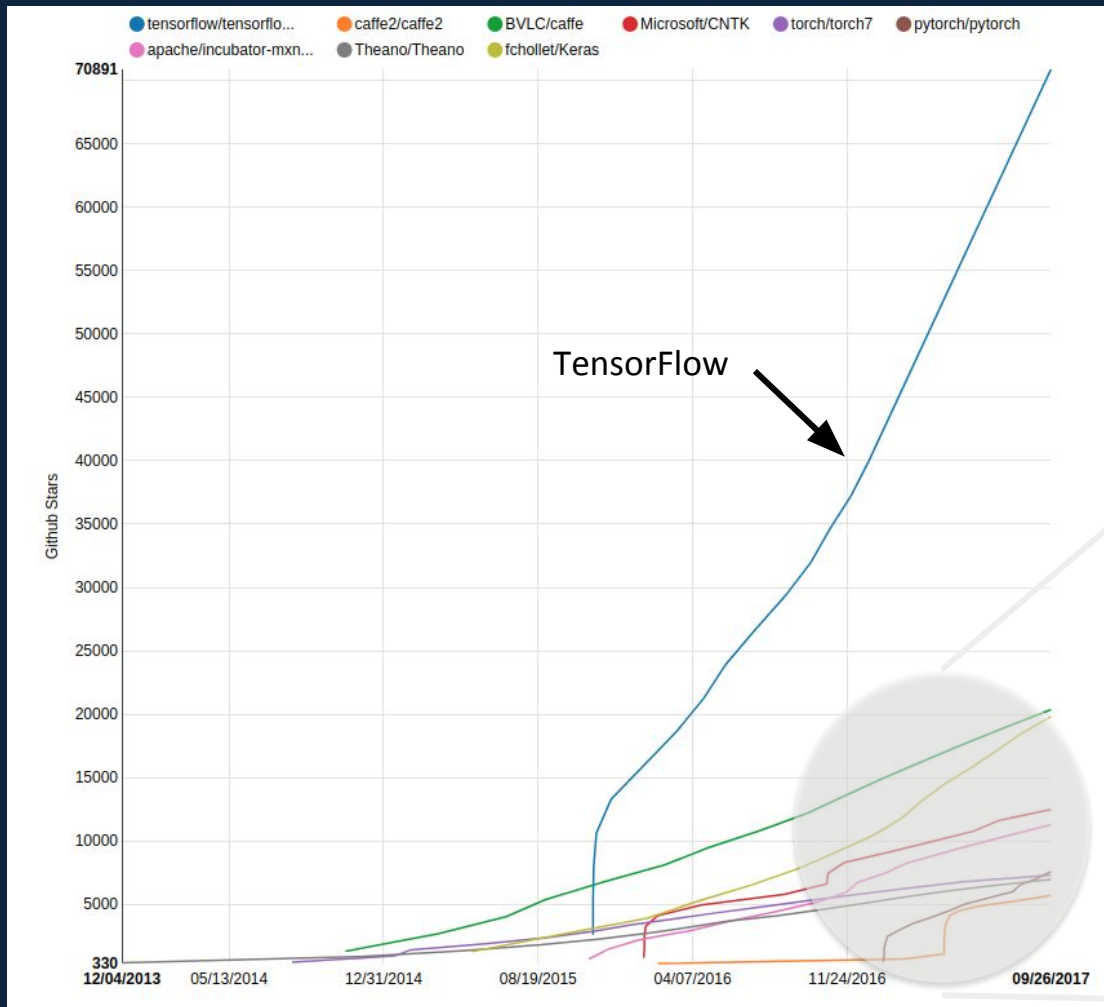
Github Stars

- tensorflow/tensorflo...
- apache/incubator-mxn...
- caffe2/caffe2
- Theano/Theano
- Microsoft/CNTK
- pytorch/pytorch
- torch/torch7



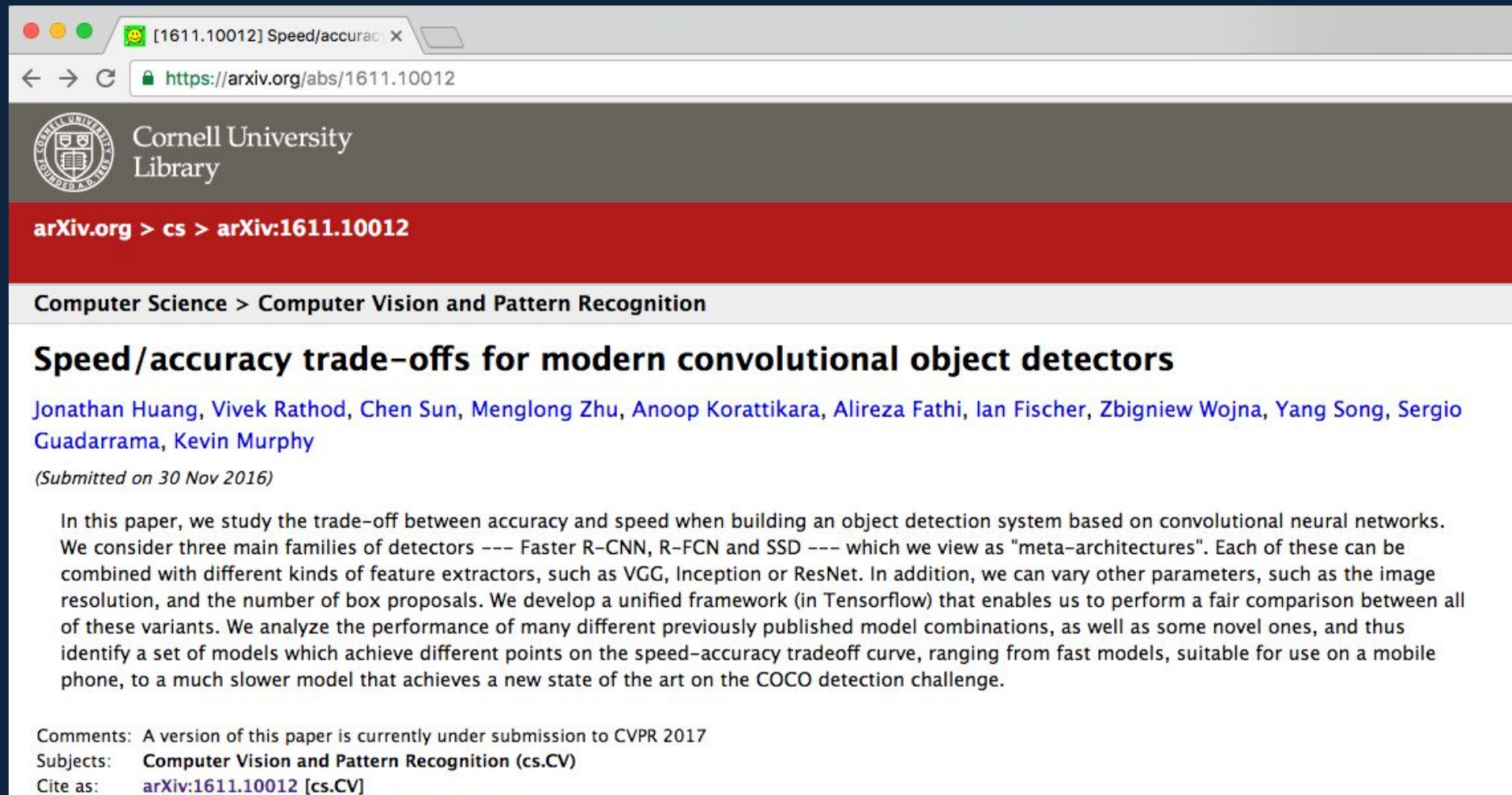
source: Francesc Sastre

In this course: we will consider Keras



frameworks with more slope


and no less important, **an open-publication ethic**, whereby many researchers publish their results immediately on a database without awaiting peer-review approval.



The image is a screenshot of a web browser window. The address bar shows the URL <https://arxiv.org/abs/1611.10012>. The page header includes the Cornell University Library logo and the text "Cornell University Library". Below the header, a red navigation bar contains the text "arXiv.org > cs > arXiv:1611.10012". A grey breadcrumb trail shows "Computer Science > Computer Vision and Pattern Recognition". The main title of the paper is "Speed/accuracy trade-offs for modern convolutional object detectors". The authors listed are Jonathan Huang, Vivek Rathod, Chen Sun, Menglong Zhu, Anoop Korattikara, Alireza Fathi, Ian Fischer, Zbigniew Wojna, Yang Song, Sergio Guadarrama, and Kevin Murphy. The submission date is noted as "(Submitted on 30 Nov 2016)". The abstract text describes the study of trade-offs between accuracy and speed in object detection systems based on convolutional neural networks, mentioning families of detectors like Faster R-CNN, R-FCN, and SSD, and feature extractors like VGG, Inception, and ResNet. At the bottom, there are comments, subjects, and citation information.

[1611.10012] Speed/accurac X

← → ↻ <https://arxiv.org/abs/1611.10012>

 Cornell University
Library

arXiv.org > cs > arXiv:1611.10012

Computer Science > Computer Vision and Pattern Recognition

Speed/accuracy trade-offs for modern convolutional object detectors

Jonathan Huang, Vivek Rathod, Chen Sun, Menglong Zhu, Anoop Korattikara, Alireza Fathi, Ian Fischer, Zbigniew Wojna, Yang Song, Sergio Guadarrama, Kevin Murphy

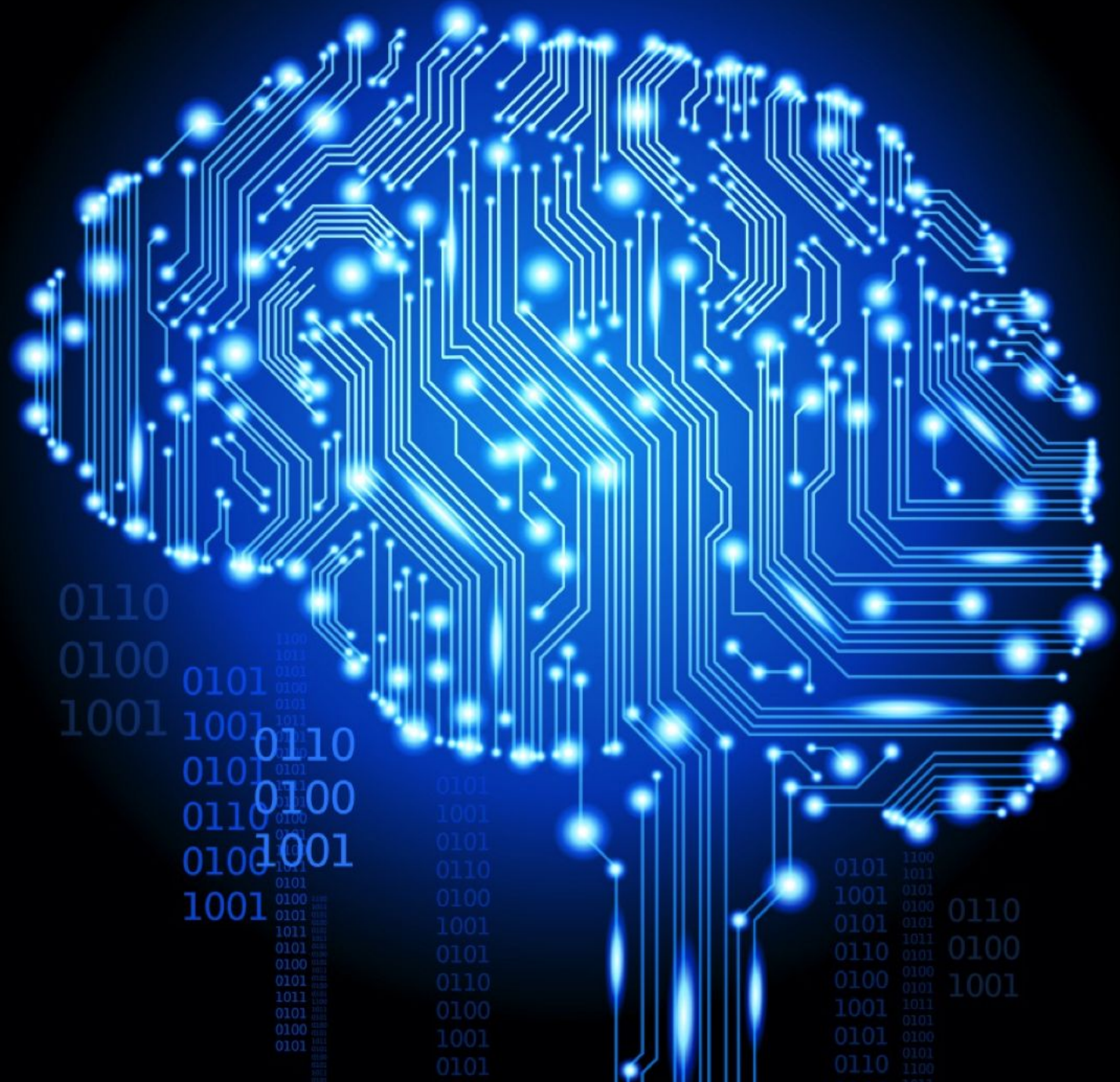
(Submitted on 30 Nov 2016)

In this paper, we study the trade-off between accuracy and speed when building an object detection system based on convolutional neural networks. We consider three main families of detectors --- Faster R-CNN, R-FCN and SSD --- which we view as "meta-architectures". Each of these can be combined with different kinds of feature extractors, such as VGG, Inception or ResNet. In addition, we can vary other parameters, such as the image resolution, and the number of box proposals. We develop a unified framework (in Tensorflow) that enables us to perform a fair comparison between all of these variants. We analyze the performance of many different previously published model combinations, as well as some novel ones, and thus identify a set of models which achieve different points on the speed-accuracy tradeoff curve, ranging from fast models, suitable for use on a mobile phone, to a much slower model that achieves a new state of the art on the COCO detection challenge.

Comments: A version of this paper is currently under submission to CVPR 2017

Subjects: **Computer Vision and Pattern Recognition (cs.CV)**

Cite as: **arXiv:1611.10012 [cs.CV]**



JORDI TORRES | FRANCESC SASTRE