# An introduction to Al

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### A definition of AI?

"The construction of computer programs that perform tasks that are, for the moment, accomplished more satisfactorily by human beings"



John McCarthy Al Pioneers with M.L Minsky

Al based on Knowledge

Al based on Data [Machine Learning] **Deep Learning** Generative Al

### 40 minutes to understand how Chat GPT works

STEP 1: Machine Learning

STEP 2 : Neurones

STEP 3: Autoencoders

STEP 4 : Sequence to Sequence Models

STEP 5: Transformers

### The different types of learning

Learn with exercises Ex. Driving license

#### **Supervised Learning**

• Learning with a **labeled** training set.

#### **Unsupervised Learning**

Discovering patterns in unlabeled data.

Learn with similitude Ex. Newton and the apple

#### Reinforcement Learning

Learning based on feedback or reward.

Learn with trial and error Ex. Ride a bike

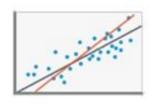
### ML to solve different types of problems



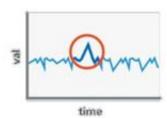
Classification (supervised – predictive)



Clustering (unsupervised – descriptive)



Regression (supervised – predictive)



Anomaly Detection (unsupervised – descriptive)

### A brief history of Deep Learning

1950

Test de Turing

1981

 Fukushima Neocognitron : lecture d'écriture manuscrite en Japonais

1988:

 Convolutional Network (CNN) de LeCun lecture d'adresse postale. 60k paramètres

2012

 Traffic Signs Challenge: Performances meilleures que les humains. AlexNet: 60 M paramètres

2016

• Alphago bat le champion du monde de go.

2024

• GPT40 : 8\*220 Milliards de paramètres

Google Gemini : 1560 Milliards de paramètres





Facebook Launches Advanced AI Effort to Find Meaning in Your Posts

A technique-called deep learning could help Facebook understand busers and their data better.



© reuters/ Kim Hong Ji

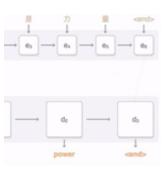
### **Solved** Applications



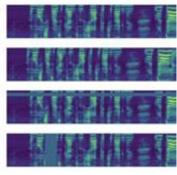
Image Classification: 92% on Image Net



**Object Detection** 



Machine Translation BLEU score 40 (34 human pro)



Speech Recognition 97% on Noisy



Sentiment analysis (amazon, twitter, ...) 96% on IMDB



Atari, Chess, Go

#### Applications still under research

BEFORE

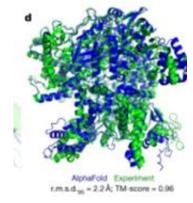


AFTER



Image - Video - 3D - World Generation Diffusion Models

Action models



**Protein Prediction** >90% AlphaFold 2



Conversation agents (LLM) outperforms humans by 30% on US Medical exam



Multi agents games: Starcraft, Diplomacy...

### **STEP 1: Machine Learning**

STEP 2 : Neurones

STEP 3: Autoencoders

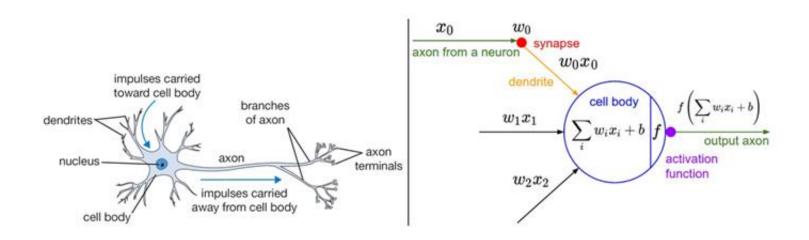
STEP 4 : Sequence to Sequence Models

STEP 5: Transformers

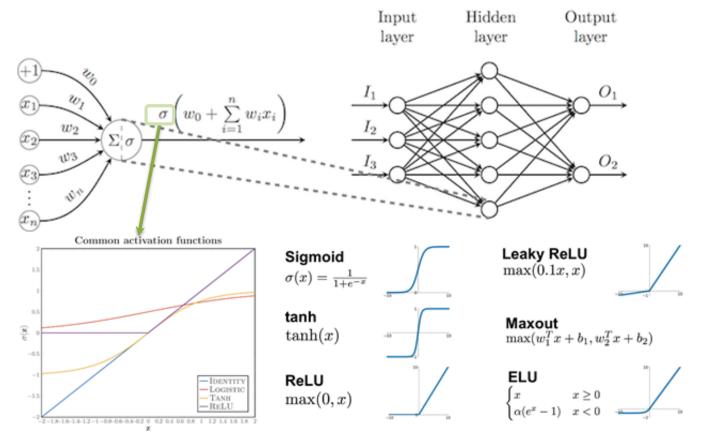


### **Neurons**

Neurons are trained to filter and detect **features** such as edges, shapes, textures, by receiving weighted inputs from the previous neurons, transforming it with an activation function and passing it to the outgoing connections.



### **Activation Functions**



STEP 1 : Machine Learning

**STEP 2 : Neurones** 

STEP 3: Autoencoders

STEP 4 : Sequence to Sequence Models

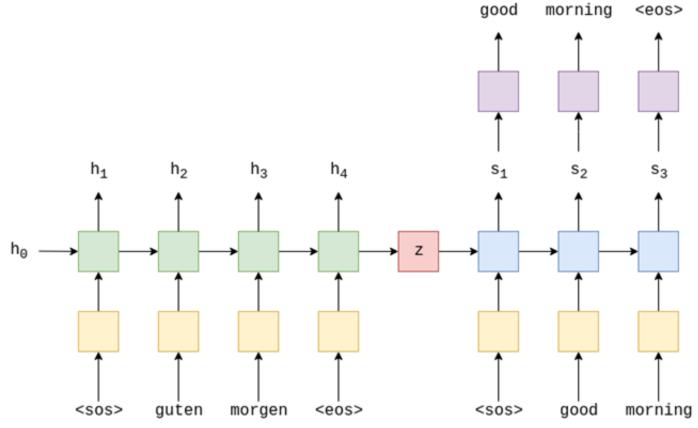
STEP 5: Transformers

playground.tensorflow.org/

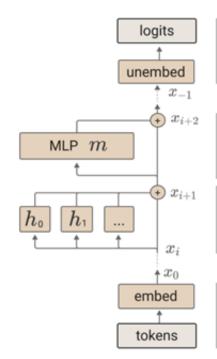
#### STEP 3: Autoencoders Discriminator Generator **GAN:** Adversarial $\mathbf{x}'$ (0/1 $\mathbf{x}'$ $D(\mathbf{x})$ $G(\mathbf{z})$ training Encoder Decoder VAE: maximize $\mathbf{x}$ $q_{\phi}(\mathbf{z}|\mathbf{x})$ $p_{\theta}(\mathbf{x}|\mathbf{z})$ variational lower bound Inverse Flow $\mathbf{x}'$ Flow-based models: $\mathbf{x}$ $f^{-1}(\mathbf{z})$ $f(\mathbf{x})$ Invertible transform of distributions Diffusion models: $\rightarrow$ $\mathbf{x}_1$ Gradually add Gaussian noise and then reverse

STEP 4 : Sequence to

Sequence Models



#### STEP 5: Transformers



The final logits are produced by applying the unembedding.

$$T(t) = W_U x_{-1}$$

An MLP layer, m, is run and added to the residual stream.

$$x_{i+2} = x_{i+1} + m(x_{i+1})$$

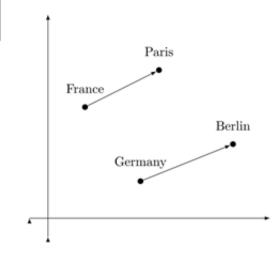
Each attention head, h, is run and added to the residual stream.

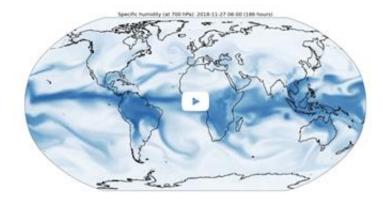
$$x_{i+1} \ = \ x_i \ + \ \sum
_{h \in H_i} h(x_i)$$

Token embedding.

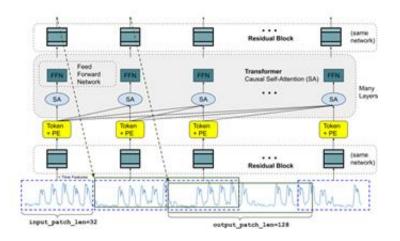
$$x_0 \; = \; W_E t$$

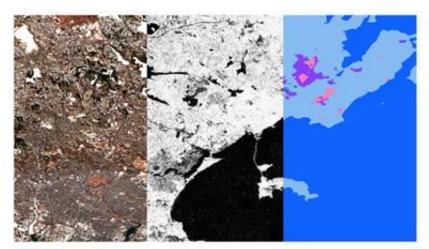




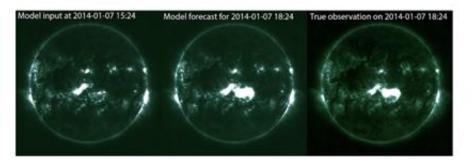


A selection of GraphCast's predictions rolling across 10 days showing specific humidity at 700 hectopascals (about 3 km above surface), surface temperature, and surface wind speed.



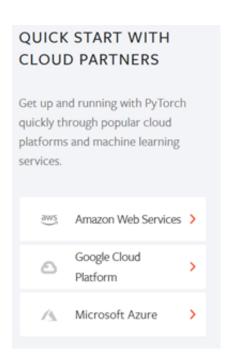


TerraMind's any-to-any generative capabilities demonstrated on a scene over Boston. From left to right: (1) optical input. (2) synthetic radar generated from optical imagery, and (3) generated land use classification.



These images compare the ground-truth data (right) with model output (center) for solar flares, which are the events behind most space weather. Surya's prediction is very close to what happened in reality (right). These preliminary results suggest that Surya has learned enough solar physics to predict the structure and evolution of a solar flare by looking at its beginning phase. NASA/SDO/ODSI IMPACT AI Team

## It may cost some money to train and deploy



https://pytorch.org/

Cloud is the easiest option (by far...)
Optimized stacks for training and inference

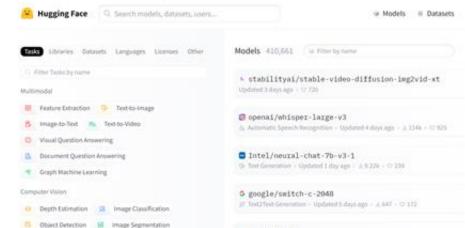
But ...
Exploding costs depending on the number of parameters

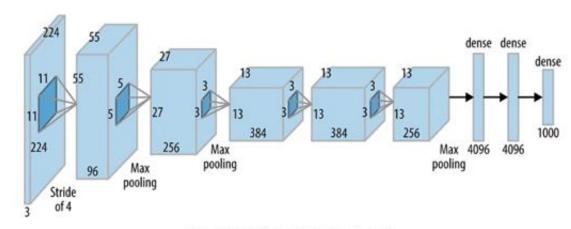
	Training cost
DeepMind AlphaGO	35 Million \$
GPT3	12 Million \$
CoAtNet (top 1 ImageNet)	250 000 \$
BERT	7000\$
Yolo V5	100\$
ResNet 50	10\$

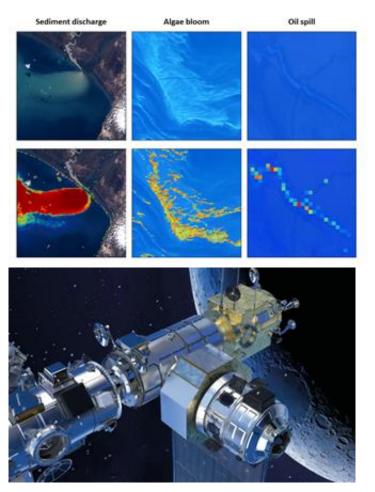
# Start from a model already trained

Hugging Face/Pytorchhub: state of the art models with weight already tuned => we add images, continue training and voilà!

Articles and blogs describe architectures (how many layers, which types), which are known to work well on a given problem









### IA@IRAP 2025

