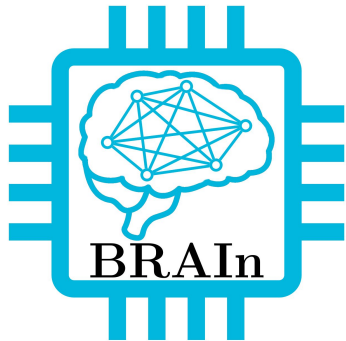


# Tutorial day: Foundation Models

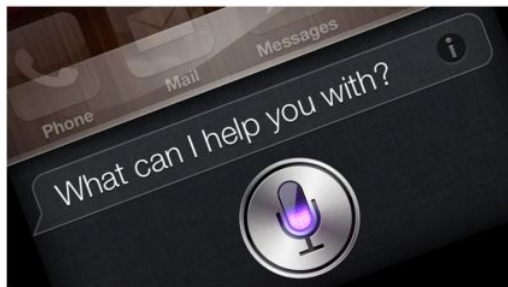
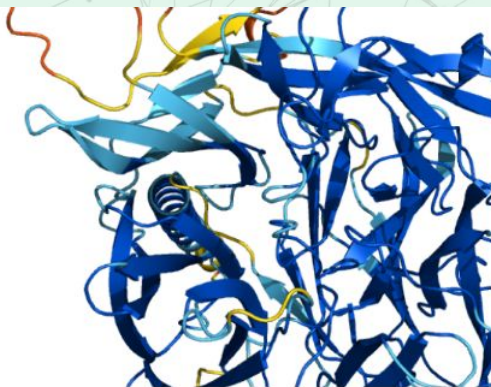
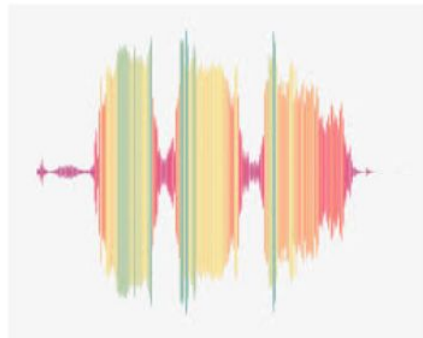
Équipe BRAIn



**IMT Atlantique**  
Bretagne-Pays de la Loire  
École Mines-Télécom



# About Deep Learning

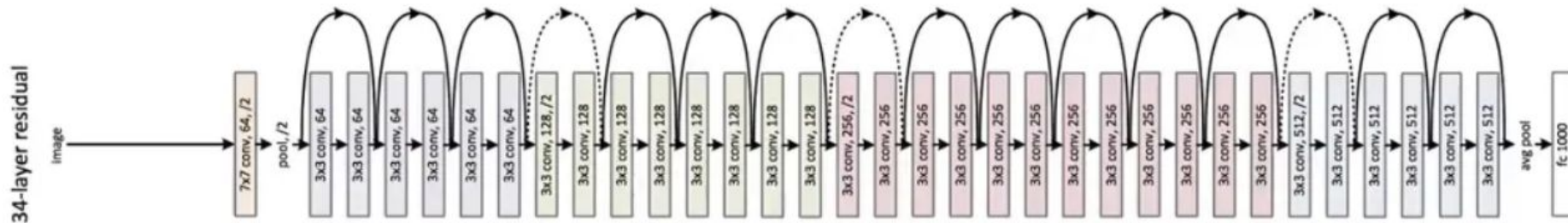


# Your AI pair programmer

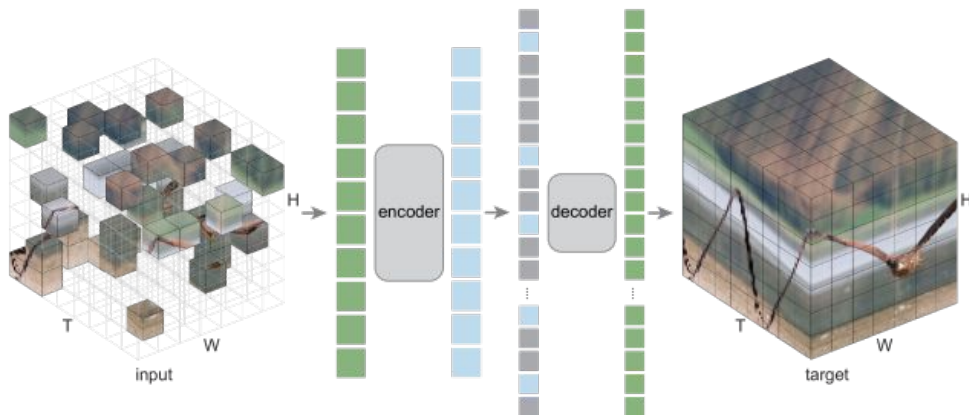
With GitHub Copilot, get suggestions for whole lines or entire functions right inside your editor.

# Deep Learning

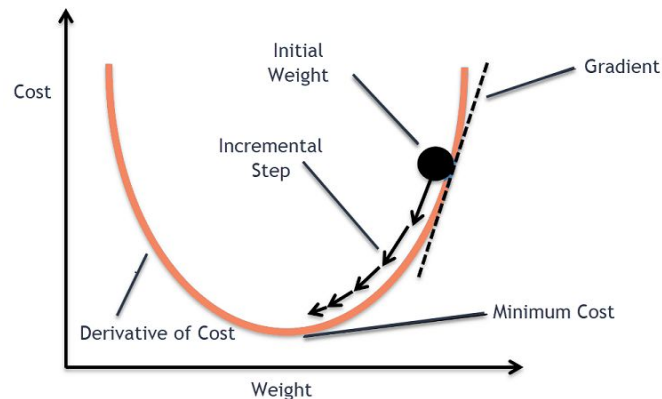
- Compositional approach:



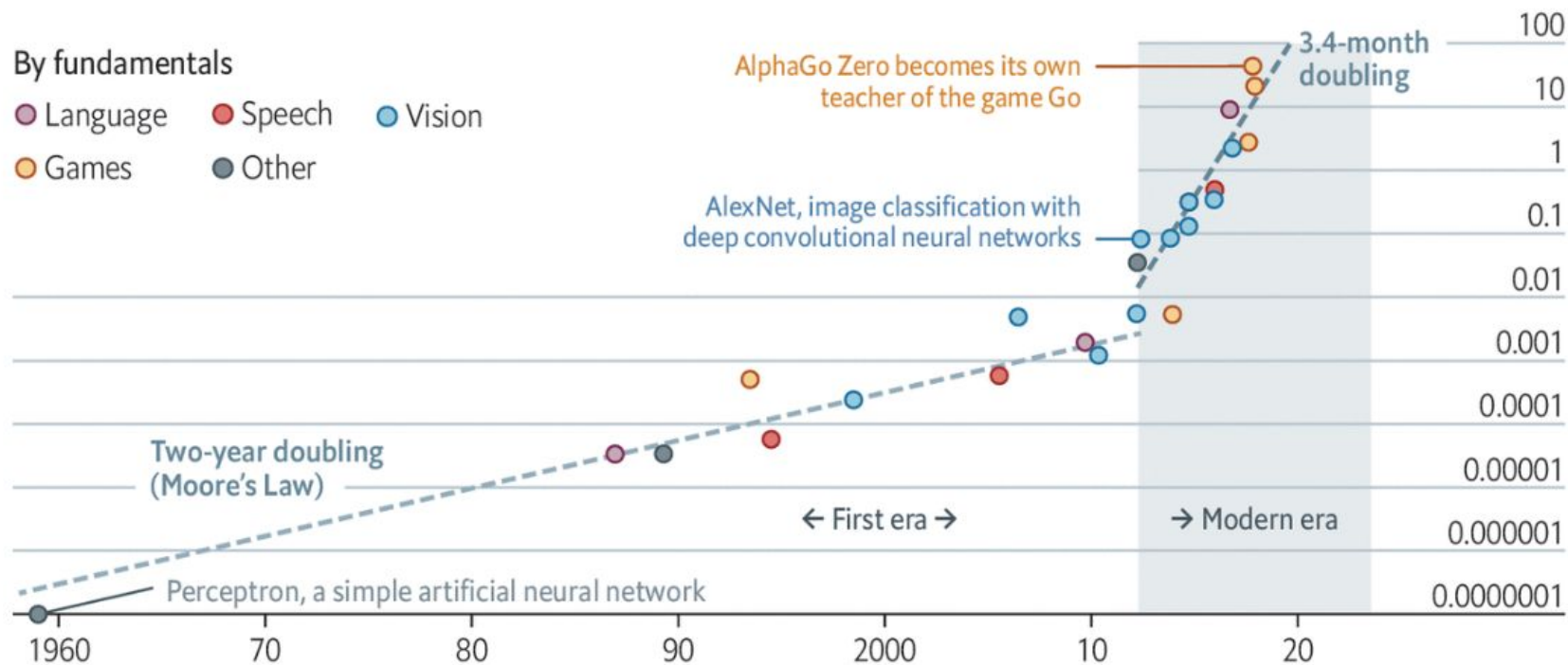
- End to end learning:



- Differential programming:



# Why it works?

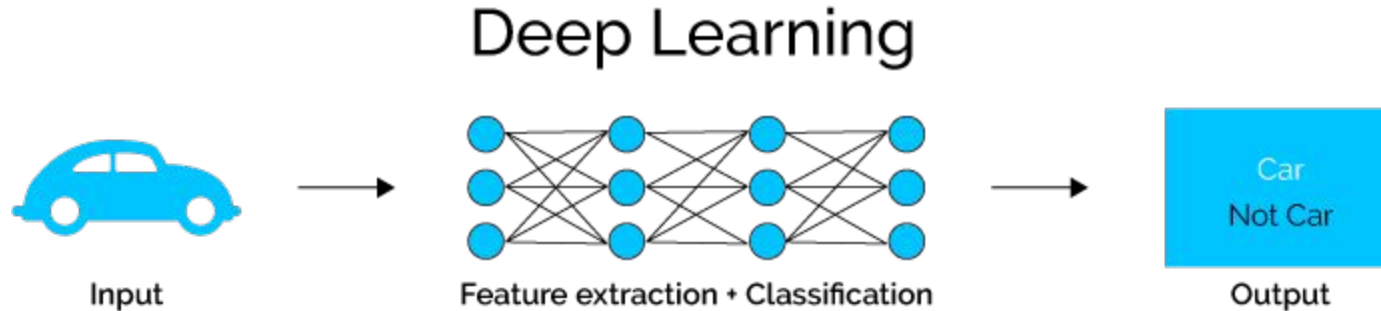
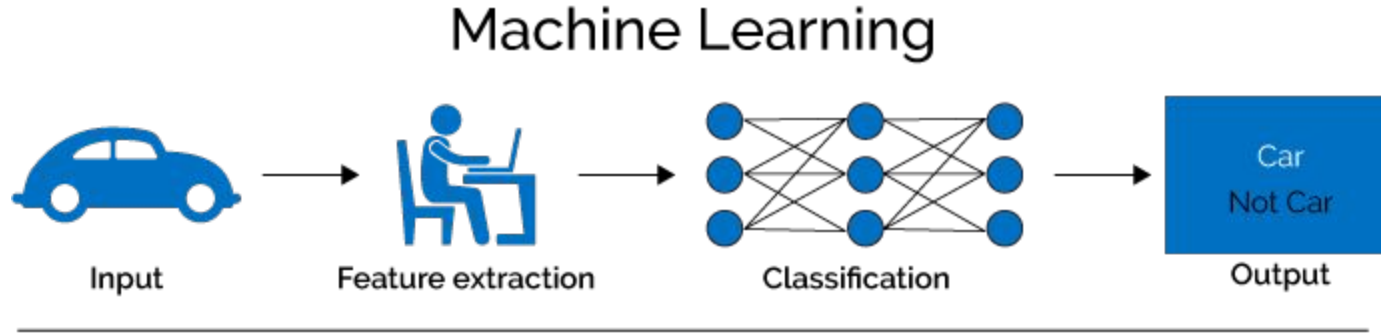


Source: OpenAI

The Economist

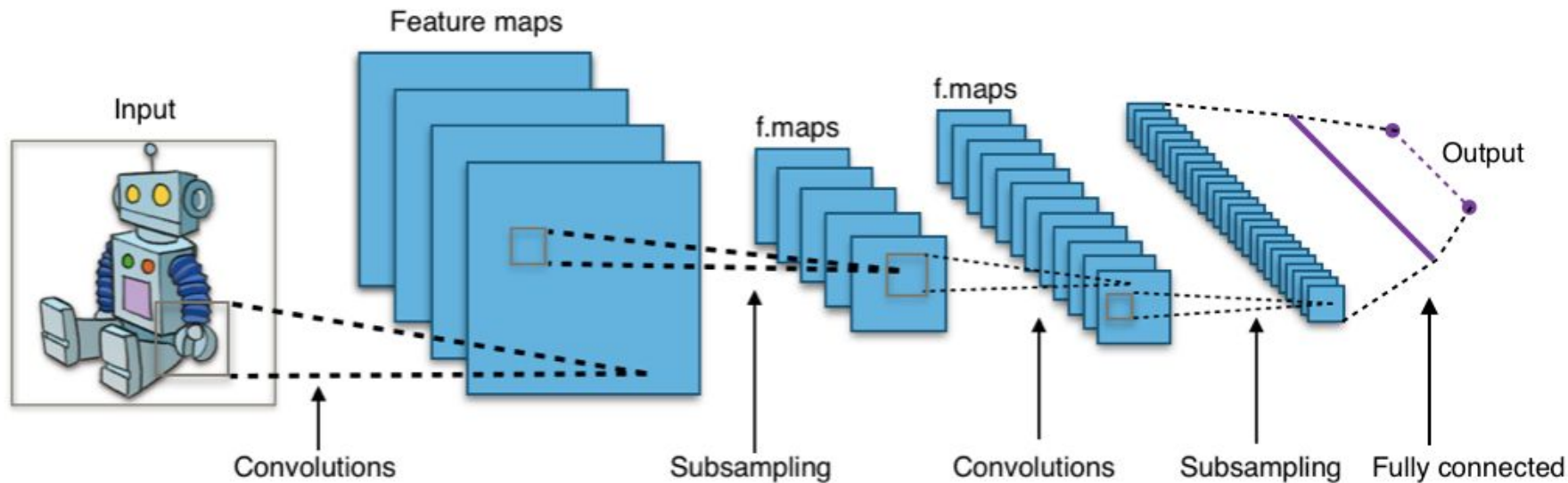
\*1 petaflop=10<sup>15</sup> calculations

# From handcrafted features to handcrafted hyperparameters



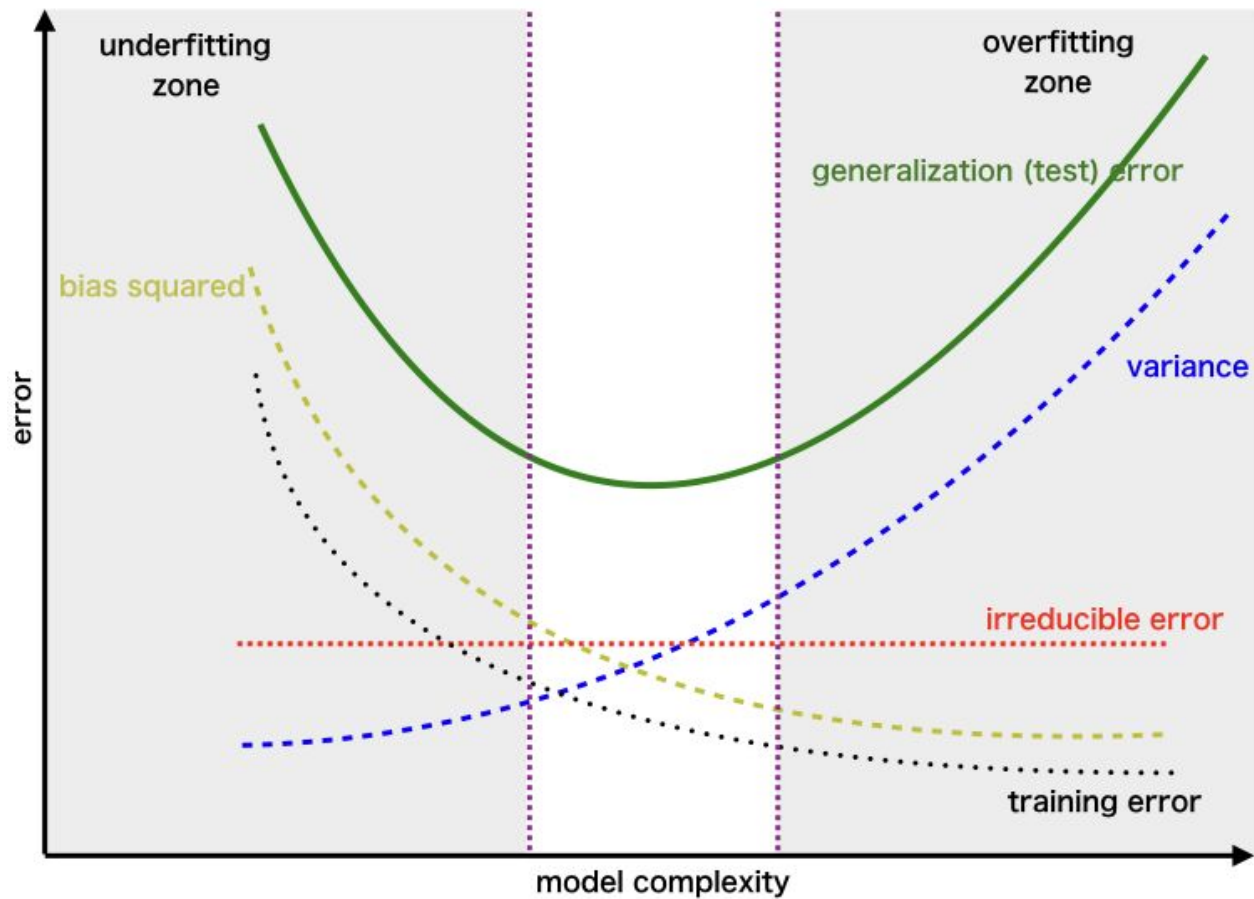
# Transformers

# About the inductive bias





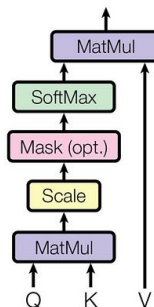
# About the inductive bias



# Transformers

- No inductive bias
- Best generalization in many domains:
  - On par with Convolutions for image:
  - SOTA for natural language processing
- Few hyperparameters
- Require large datasets to train

Scaled Dot-Product Attention



Multi-Head Attention

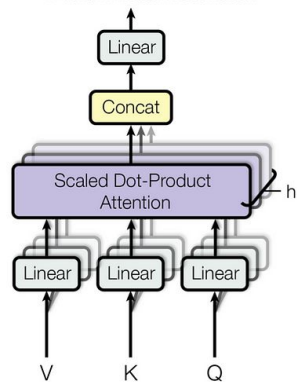
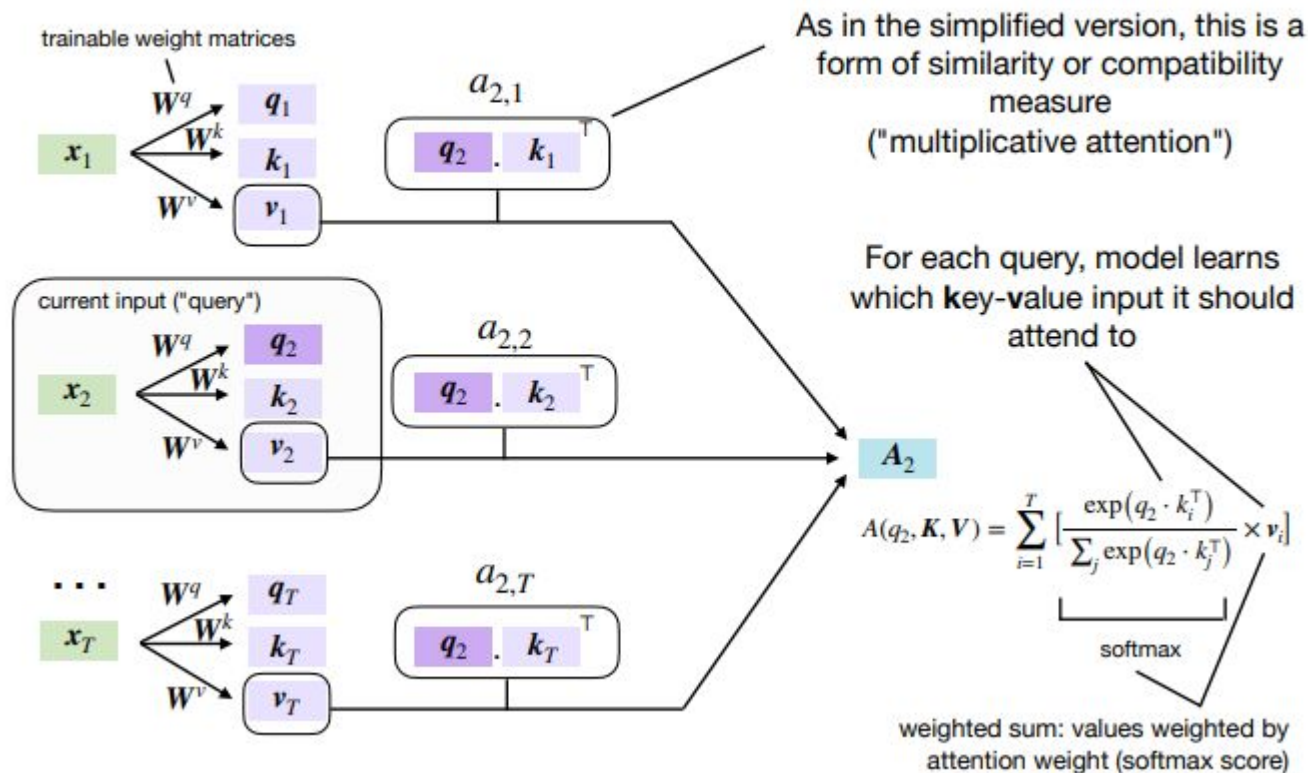


Figure 2: (left) Scaled Dot-Product Attention. (right) Multi-Head Attention consists of several attention layers running in parallel.

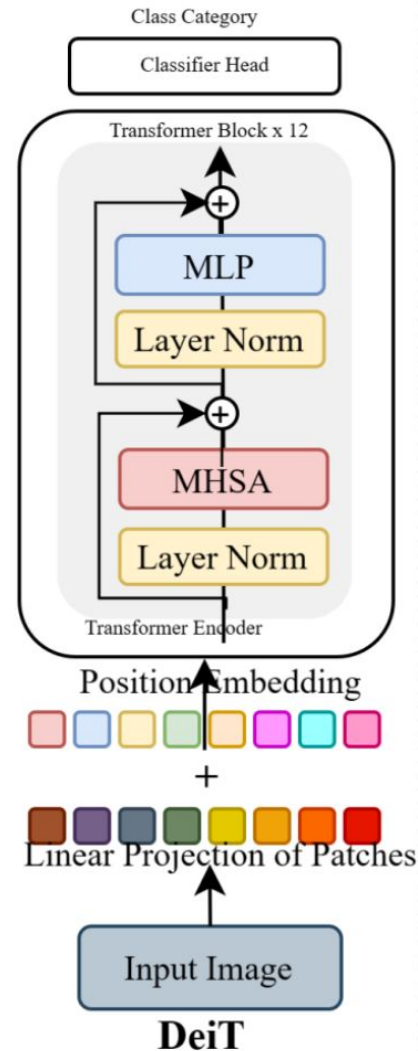
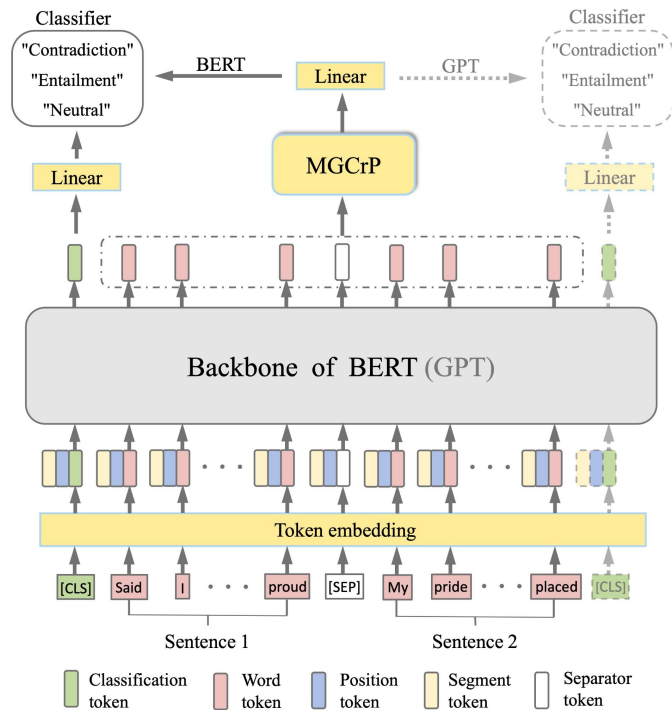
# Self-attention

## Self-Attention Mechanism



# Transformer architectures

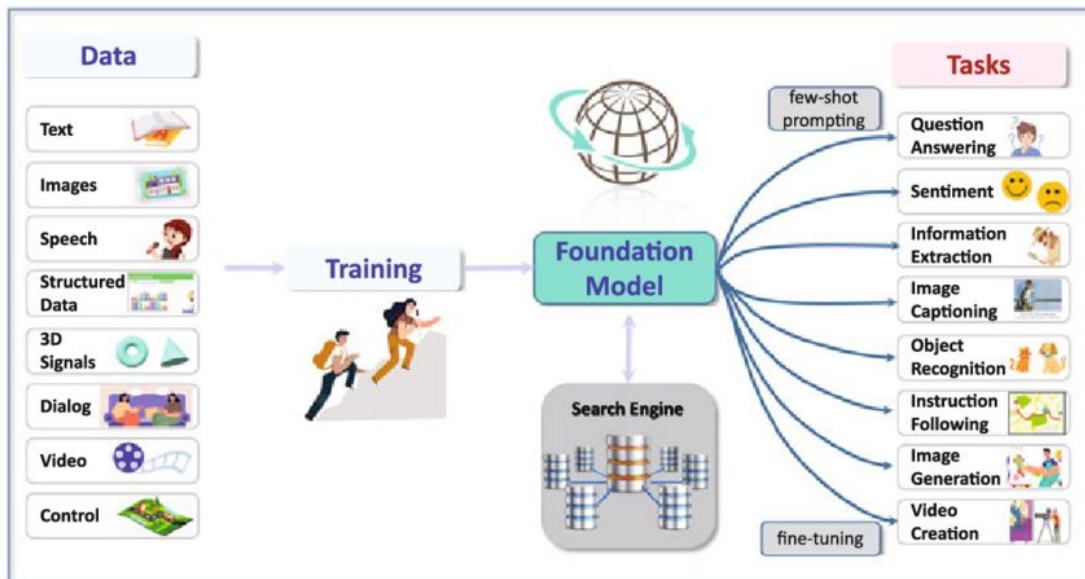
- Hyperparameters:
  - Inner dim
  - Number of heads
  - Number of Transformer blocks
- Position embedding helps incorporate a notion of structure
- Inputs are fixed in size



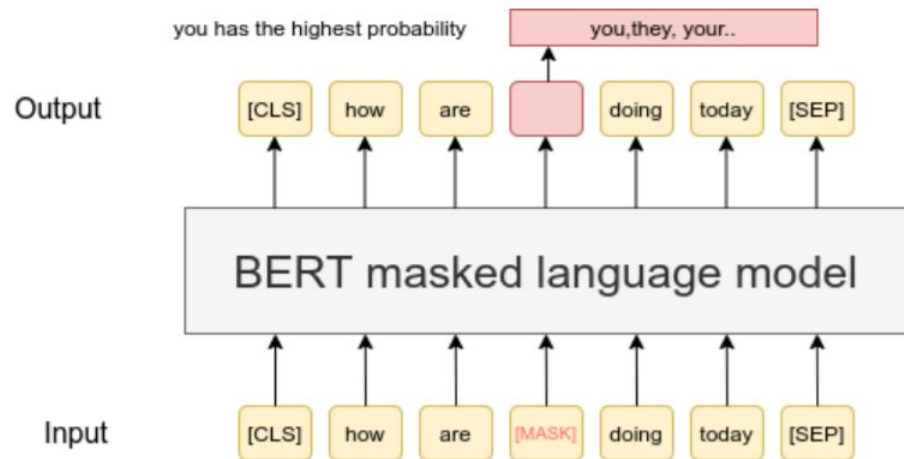
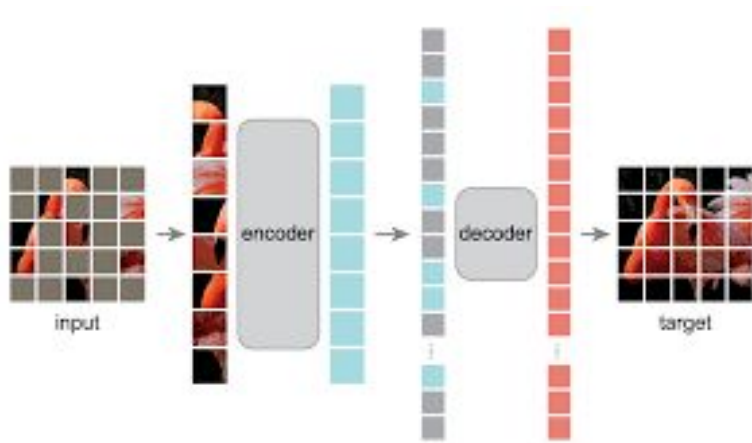
# From generalization to particularization: Foundation Models

# What is a foundation model?

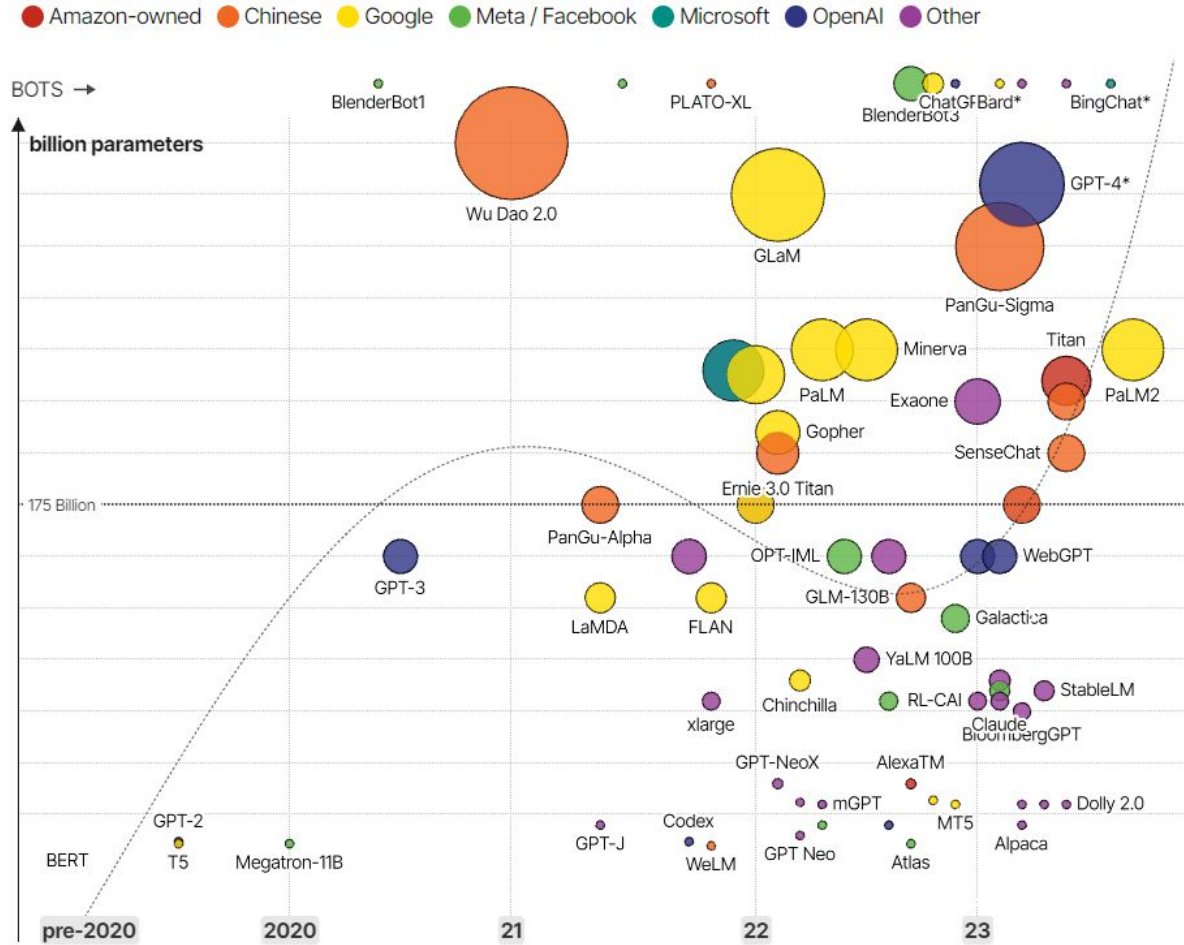
- Model trained on an internet-scale dataset
- Training task is typically not straightforward:
  - Self-supervision (MAE, sequence prediction...)
  - Pretext tasks
- They are multipurpose and generic
- Particularization vs. generalization



# Self supervision to leverage huge datasets



# Foundation models are rising (very) fast



David McCandless, Tom Evans, Paul Barton

Information is Beautiful // May 2023

source: news reports, [LifeArchitect.ai](https://lifeaiarchitect.ai)

\* = parameters undisclosed // see [the data](#)



Quick notes before diving into the details

# 101 vocabulary

- ***Hyperparameters***: values to be chosen before setting up an architecture and training it
- ***Token***: unit vector to describe data
- ***Freezing weights***: fixing some parameter values during training, so that they cannot be updated
- ***Fine-tuning***: training slowly to find a way to adapt without overfitting
- ***Adapter***: small complement architecture to be trained to adapt a large model with few parameters
- ***Latent space***: representation of the input data deep into the considered architecture

# The price of foundation models

- **Can we train foundation models?**
  - Short answer: no
- **Can we use foundation models?**
  - Most of them on standard desktop machines (with a strong GPU inside)
  - Some of them using only CPU (but slow)
  - Some of them using expensive hardware (e.g. A100)
- **Can we download training datasets?**
  - Often, but not always

# Use cases and limitations

- **Use cases:**

- Process and automatically organize collections of data
- Use as components in systems
- Quick prototyping

- **Limitations:**

- Black box
- Biases
- May require heavy computations

# Program of the day

## **Now**

10h: Global introduction

11h: Vision with DINOv2 + SAM

## **12h: Lunch break**

14h: Natural Language Processing with Llama2

15h: MultiModal vision/text with CLIP

## **16h: Coffee break**

16h30: MultiModal audio/text with CLAP