

# Artificial Intelligence

ACTL3143 & ACTL5111 Deep Learning for Actuaries  
Patrick Laub



# Lecture Outline

- **Artificial Intelligence**
- Machine Learning
- Machine Learning Paradigms
- Neural Networks



# Different goals of AI

Artificial intelligence describes an agent which is capable of:

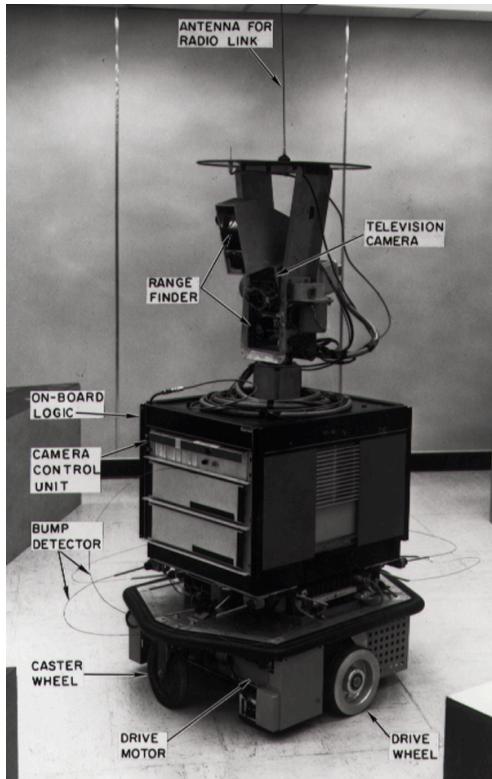
Thinking humanly	Thinking rationally
Acting humanly	Acting rationally

AI eventually become dominated by one approach, called *machine learning*, which itself is now dominated by *deep learning* (neural networks).

There are AI algorithms for simple tasks that don't use machine learning though.



# Shakey the Robot (~1966 – 1972)



Shakey the Robot

Shakey the Robot: The First Robot to Embody Artificial Intel...



# Route-finding I

At its core, a pathfinding method searches a graph by starting at one vertex and exploring adjacent nodes until the destination node is reached, generally with the intent of finding the cheapest route. Although graph searching methods such as a breadth-first search would find a route if given enough time, other methods, which “explore” the graph, would tend to reach the destination sooner. An analogy would be a person walking across a room; rather than examining every possible route in advance, the person would generally walk in the direction of the destination and only deviate from the path to avoid an obstruction, and make deviations as minor as possible. (Source: [Wikipedia](#))



A\* algorithm (1968).

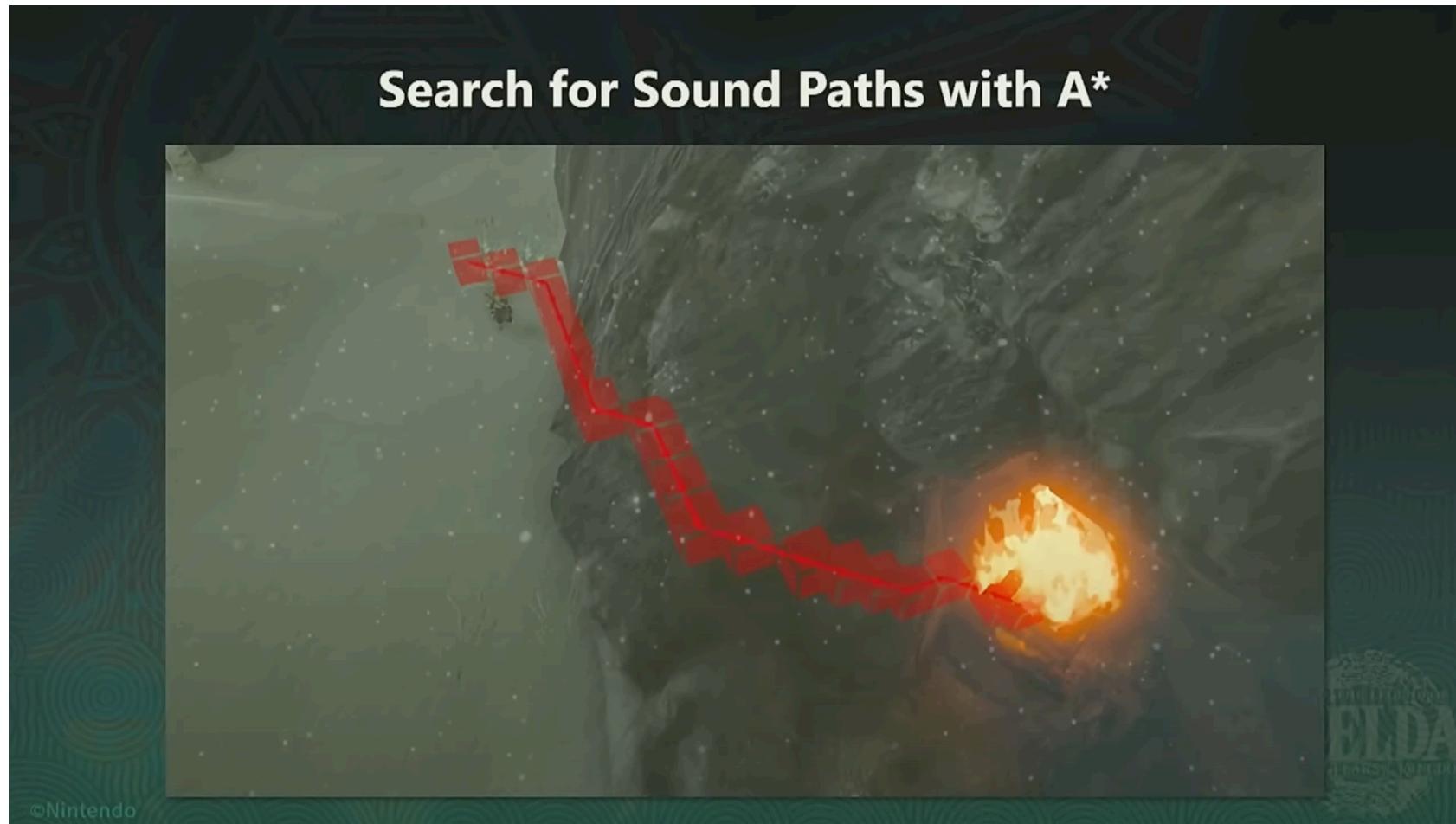


Source: Wikipedia page for [the A\\* search algorithm](#).



UNSW  
SYDNEY

# Route-finding II

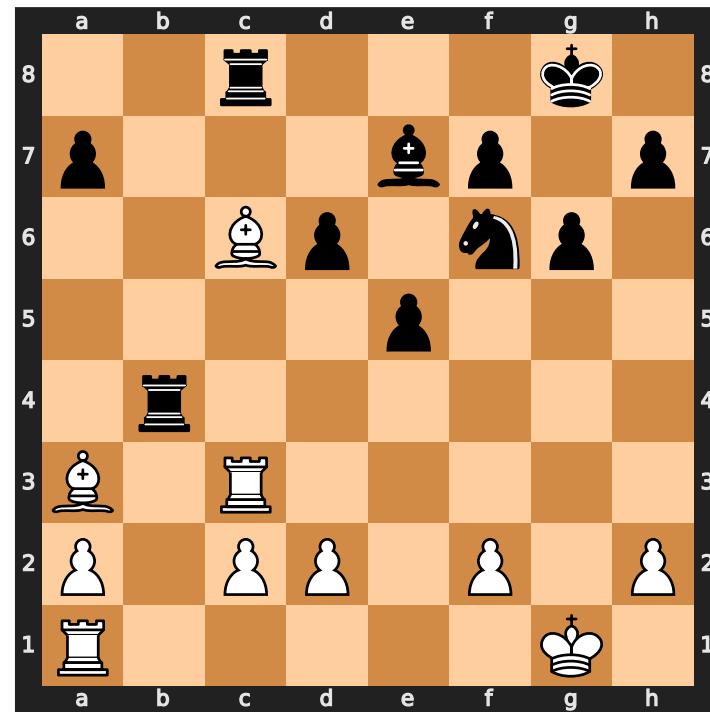


Tunes of the Kingdom: Evolving Physics and Sounds for 'The Legend of Zelda: Tears of the Kingdom', GDC 2024



# Evaluating a chess game I

Who's winning this game?

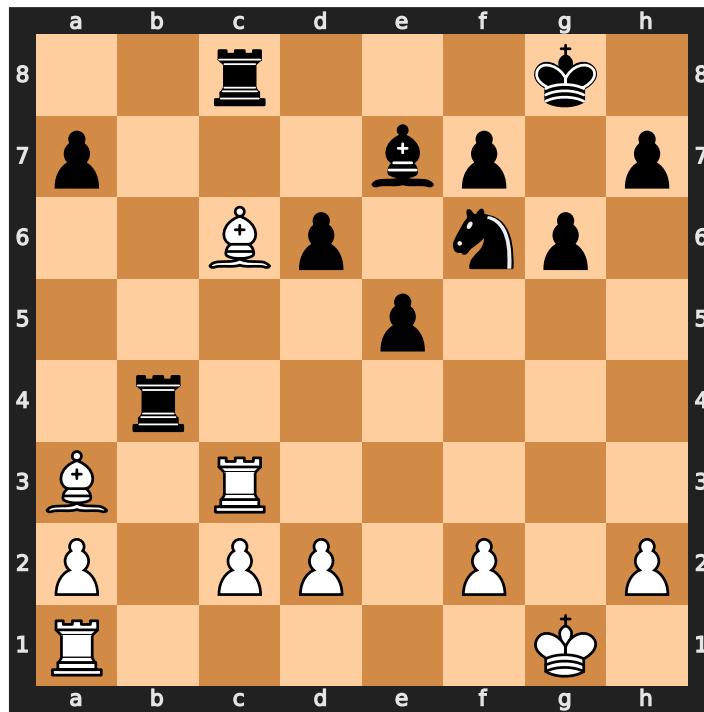


	$5 \times 1 = 5$
	$0 \times 3 = 0$
	$2 \times 3 = 6$
	$2 \times 5 = 10$
	$0 \times 9 = 0$
	$1 \times 0 = 0$

White      21

# Evaluating a chess game II

Just add up the pieces for each player.



	$6 \times 1 = 6$
	$1 \times 3 = 3$
	$1 \times 3 = 3$
	$2 \times 5 = 10$
	$0 \times 9 = 0$
	$1 \times 0 = 0$

Black      22

Overall:  $21 - 22 = -1$ .



# The minimax algorithm

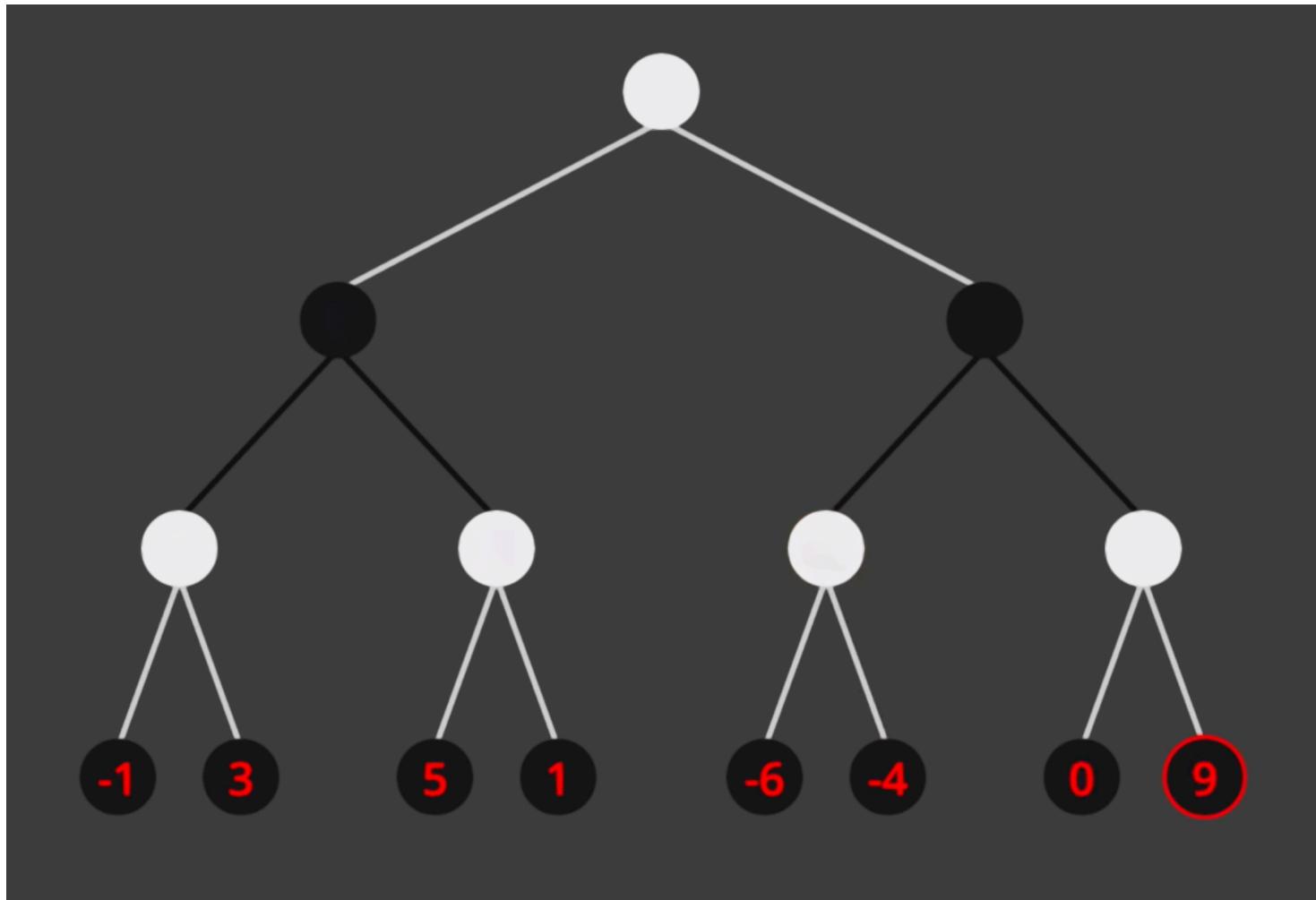


Illustration of minimax on a two-player game.



Source: Sebastian Lague (2018), [Algorithms Explained – minimax and alpha-beta pruning](#).

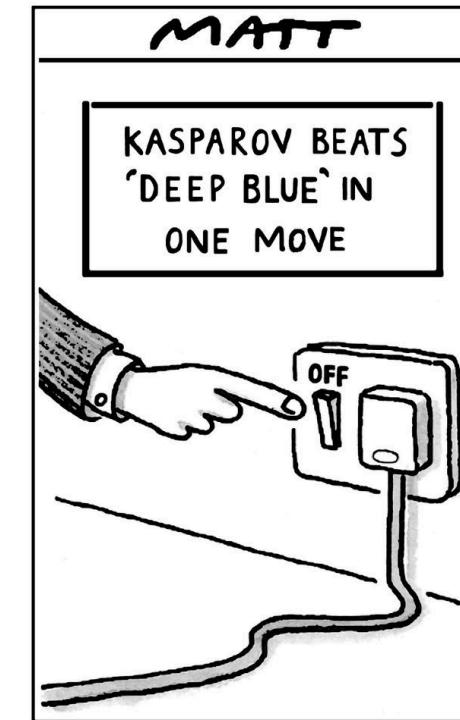


# Chess

Deep Blue (1997)



Gary Kasparov playing Deep Blue.



Cartoon of the match.



Sources: Mark Robert Anderson (2017), Twenty years on from Deep Blue vs Kasparov, The Conversation article, and Computer History Museum.



UNSW  
SYDNEY

# Lecture Outline

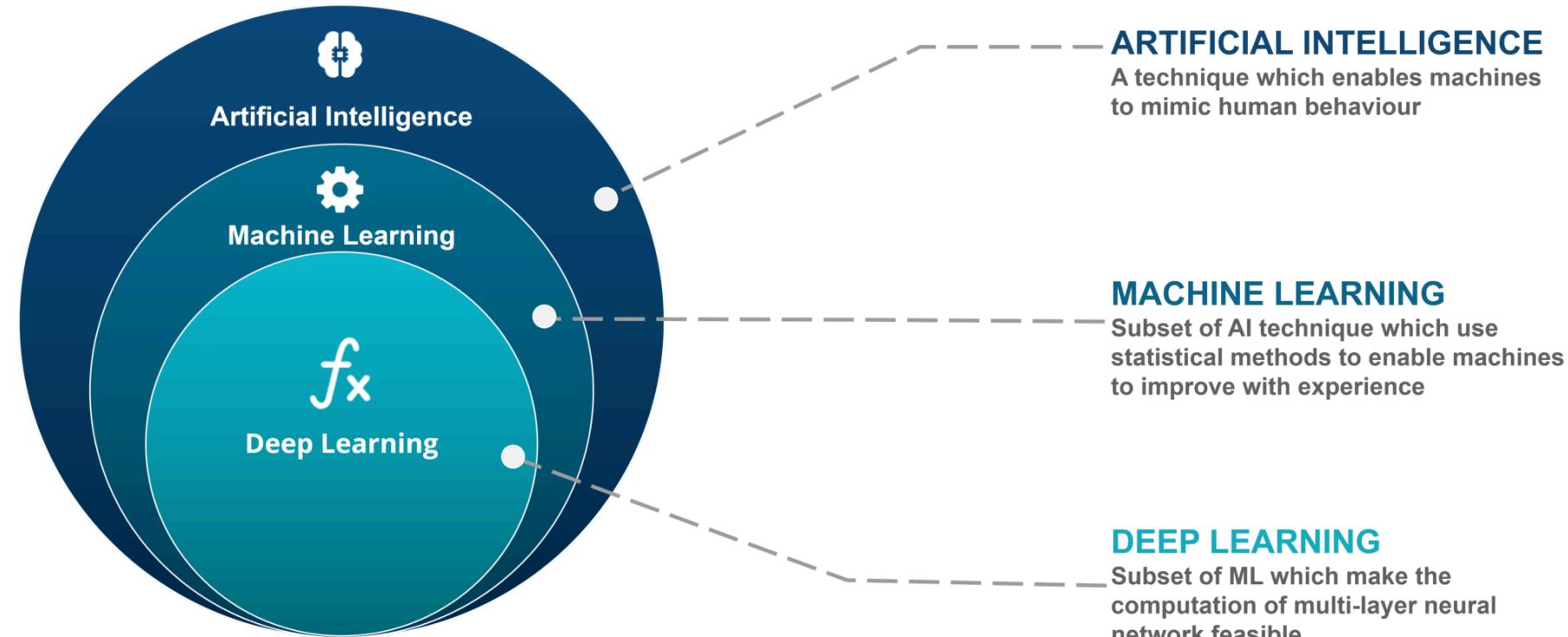
- Artificial Intelligence
- **Machine Learning**
- Machine Learning Paradigms
- Neural Networks



# Main idea

Tried *making a computer smart*, too hard!

Make a computer that can **learn** to be smart.



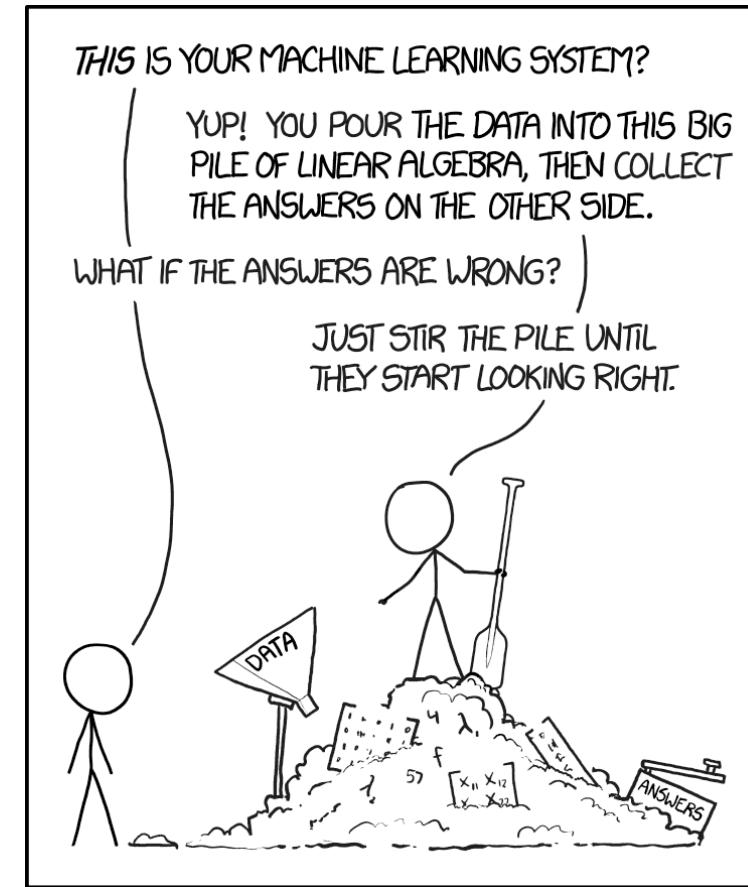
The Venn diagram of Artificial Intelligence, Machine Learning, and Deep Learning.



Source: Edureka (2020), [AI Vs Machine Learning Vs Deep Learning Edureka](#).

# Definition

“[Machine Learning is the] field of study that gives computers the ability to learn without being explicitly programmed” Arthur Samuel (1959)



# Image Classification I

What is this?



Options:

1. punching bag
2. goblet
3. red wine
4. hourglass
5. balloon



## Note

Hover over the options to see AI's prediction  
(i.e. the probability of the photo being in that category).

# Image Classification II

What is this?



Options:

1. sea urchin
2. porcupine
3. echidna
4. platypus
5. quill

# Image Classification III

What is this?

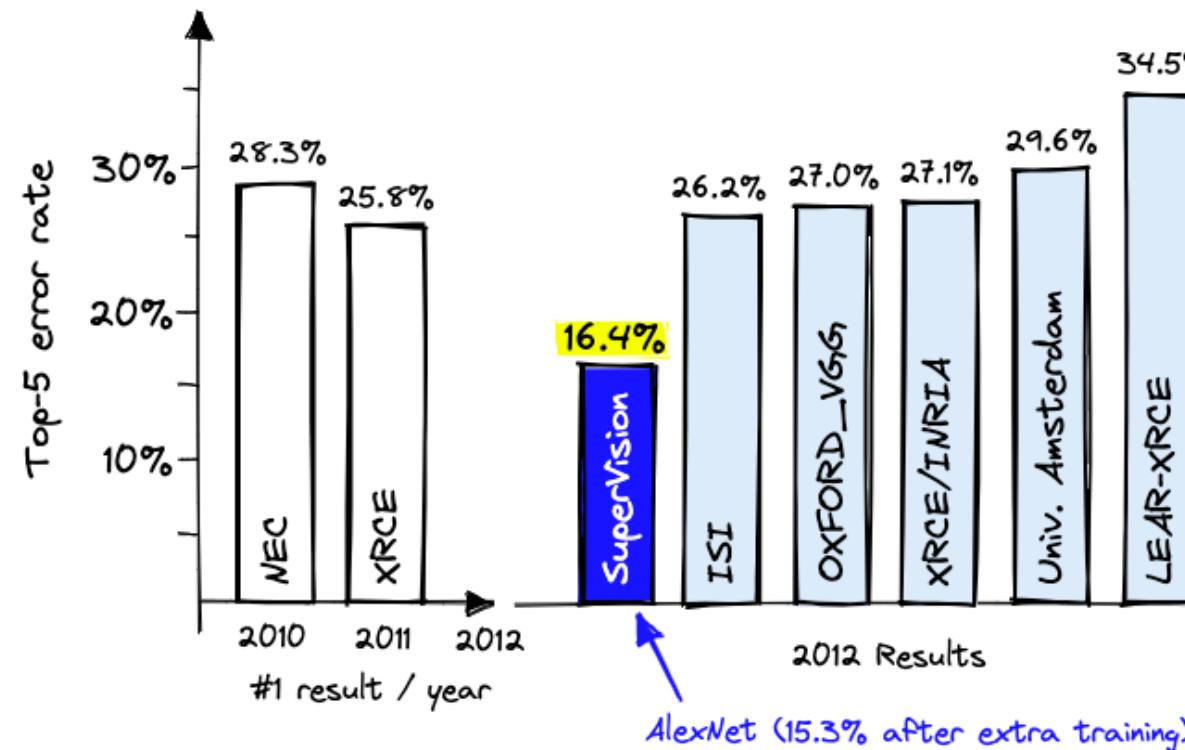


Options:

1. dingo
2. malinois
3. German shepherd
4. muzzle
5. kelpie

# ImageNet Challenge

**ImageNet** and the *ImageNet Large Scale Visual Recognition Challenge (ILSVRC)*; originally **1,000 synsets**.



AlexNet — a neural network developed by Alex Krizhevsky, Ilya Sutskever, and Geoffrey Hinton — won the ILSVRC 2012 challenge convincingly.

# Needed a graphics card

A graphics processing unit (GPU)

**“4.2. Training on multiple GPUs** A single GTX 580 GPU has only 3GB of memory, which limits the maximum size of the networks that can be trained on it. It turns out that 1.2 million training examples are enough to train networks which are too big to fit on one GPU. Therefore we spread the net across two GPUs.”



# Lee Sedol plays AlphaGo (2016)

AlphaGo - The Movie | Full award-winning documentary



Deep Blue was a win for AI, AlphaGo a win for ML.



# GANs



# DALL-E



# GPT



# Code generation

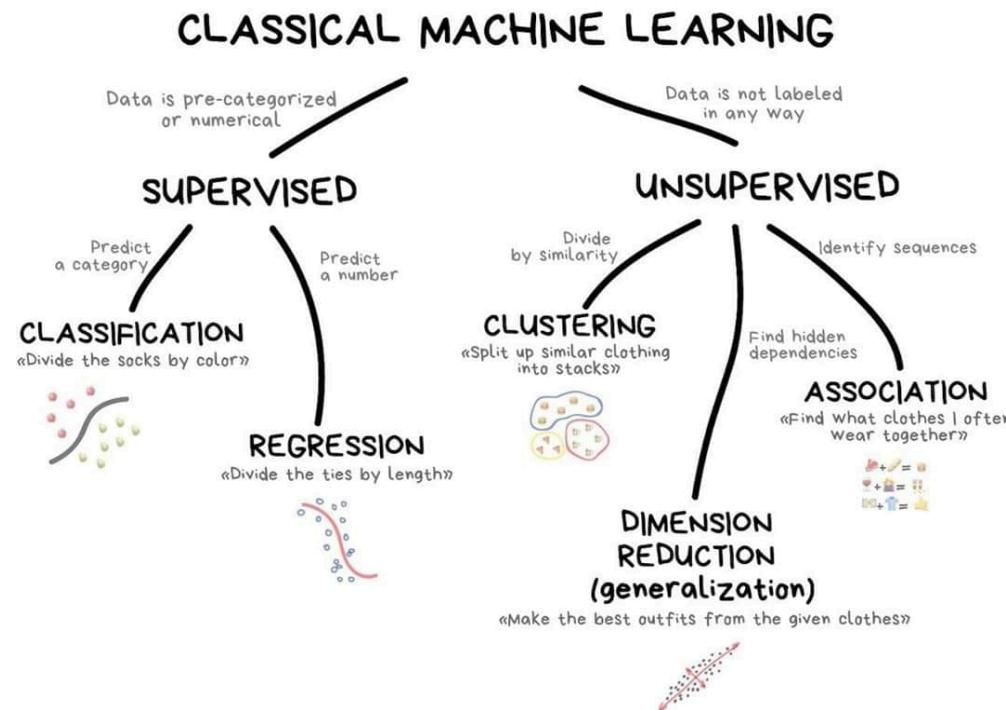


# Lecture Outline

- Artificial Intelligence
- Machine Learning
- **Machine Learning Paradigms**
- Neural Networks



# A taxonomy of problems



Machine learning categories in ACTL3142.

New ones:

- Reinforcement learning
- Semi-supervised learning
- Active learning



Source: Kaggle, [Getting Started](#).



UNSW  
SYDNEY

# Supervised learning

The main focus of this course.

## Regression

- Given policy  $\hookrightarrow$  predict the rate of claims.
- Given policy  $\hookrightarrow$  predict claim severity.
- Given a reserving triangle  $\hookrightarrow$  predict future claims.

## Classification

- Given a claim  $\hookrightarrow$  classify as fraudulent or not.
- Given a customer  $\hookrightarrow$  predict customer retention patterns.



# Supervised learning: mathematically

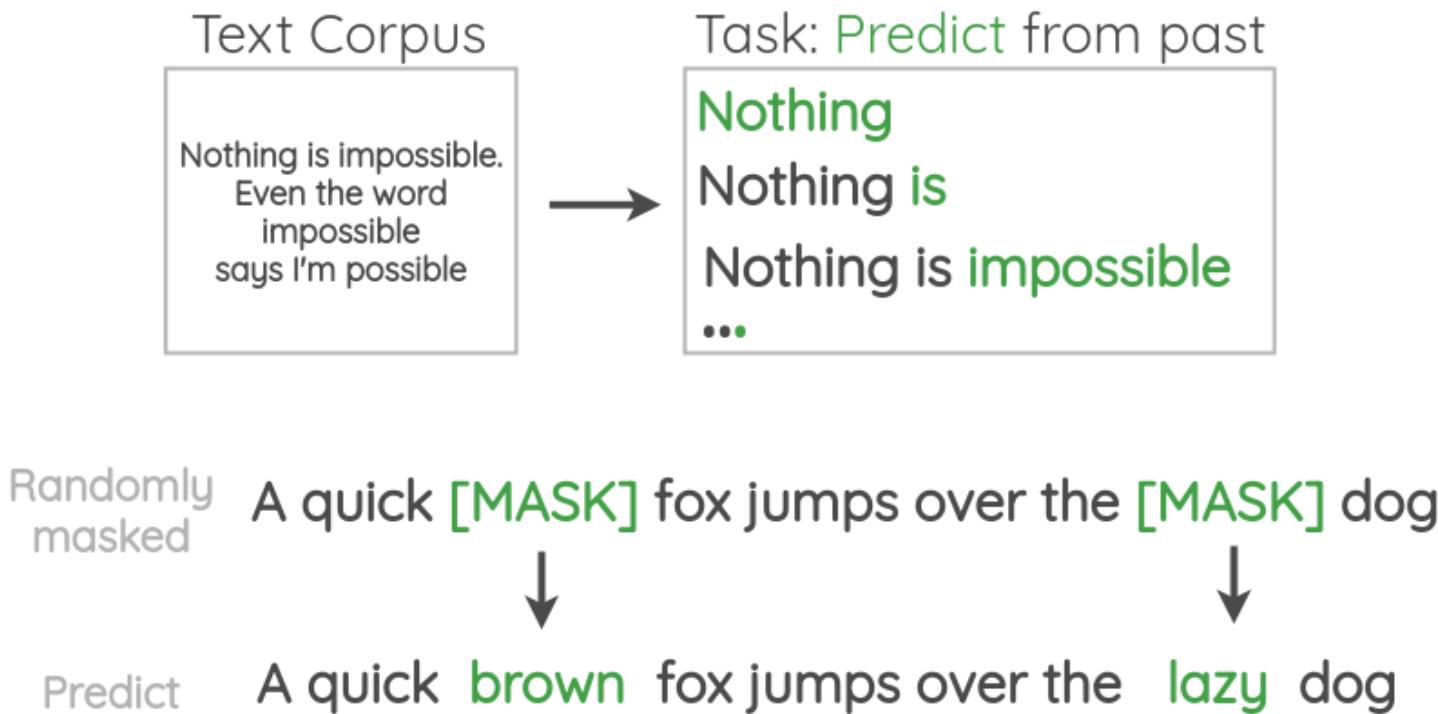
Background	A Recipe for Machine Learning
1. Given training data: $\{\mathbf{x}_i, \mathbf{y}_i\}_{i=1}^N$	3. Define goal: $\boldsymbol{\theta}^* = \arg \min_{\boldsymbol{\theta}} \sum_{i=1}^N \ell(f_{\boldsymbol{\theta}}(\mathbf{x}_i), \mathbf{y}_i)$
2. Choose each of these: <ul style="list-style-type: none"> <li>– Decision function <math>\hat{\mathbf{y}} = f_{\boldsymbol{\theta}}(\mathbf{x}_i)</math></li> <li>– Loss function <math>\ell(\hat{\mathbf{y}}, \mathbf{y}_i) \in \mathbb{R}</math></li> </ul>	4. Train with SGD: (take small steps opposite the gradient) $\boldsymbol{\theta}^{(t+1)} = \boldsymbol{\theta}^{(t)} - \eta_t \nabla \ell(f_{\boldsymbol{\theta}}(\mathbf{x}_i), \mathbf{y}_i)$

A recipe for supervised learning.



# Self-supervised learning

Data which ‘labels itself’. Example: language model.



‘Autoregressive’ (e.g. GPT) versus ‘masked’ model (e.g. BERT).



Source: Amit Chaudhary (2020), [Self Supervised Representation Learning in NLP](#).

# Example: image inpainting



Original image



Randomly remove a part



Try to fill it in from context

Other examples: image super-resolution, denoising images.



Cf. Liu et al. (2018), [Image Inpainting for Irregular Holes using Partial Convolutions](#).



UNSW  
SYDNEY

# Example: Deoldify images #1



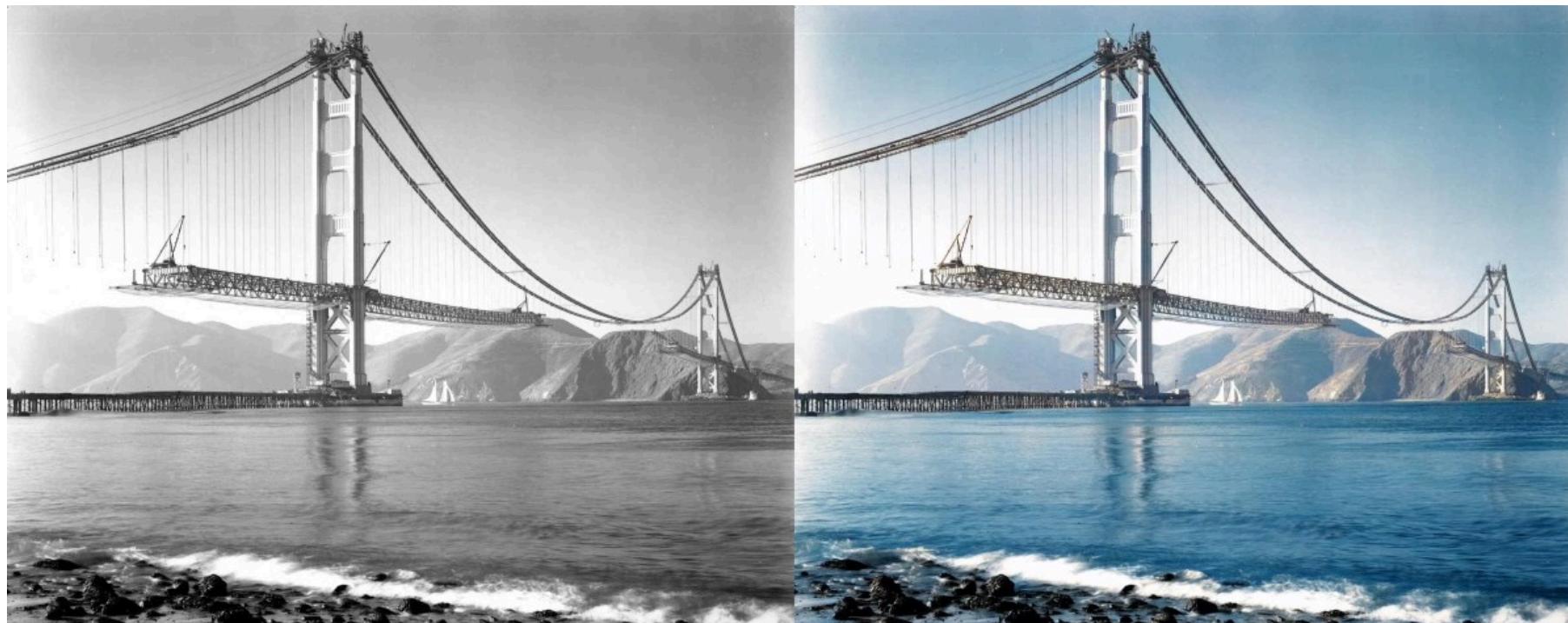
A deoldified version of the famous “Migrant Mother” photograph.

Source: [Deoldify package](#).



UNSW  
SYDNEY

# Example: Deoldify images #2



A deoldified Golden Gate Bridge under construction.



Source: [Deoldify package](#).



UNSW  
SYDNEY

# Lecture Outline

- Artificial Intelligence
- Machine Learning
- Machine Learning Paradigms
- **Neural Networks**

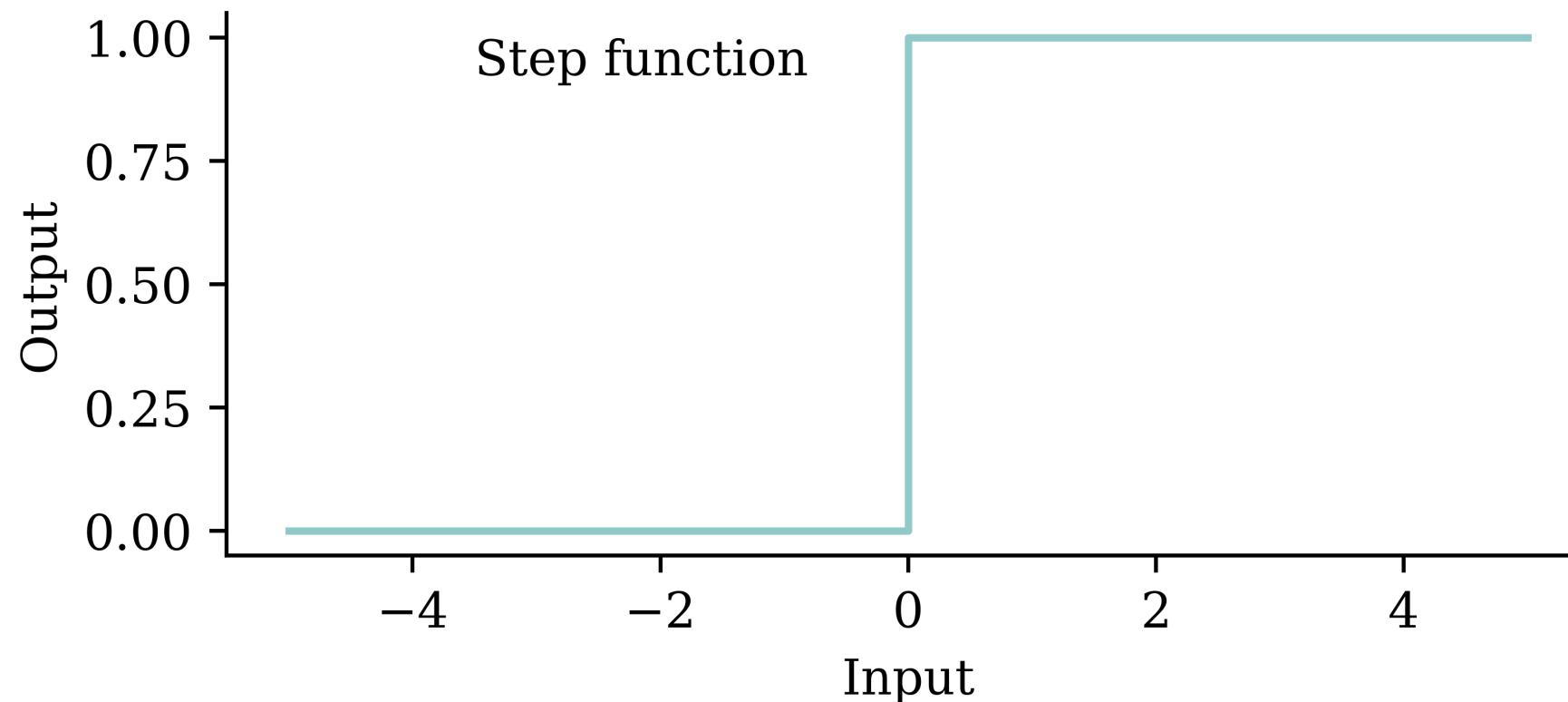


# How do real neurons work?

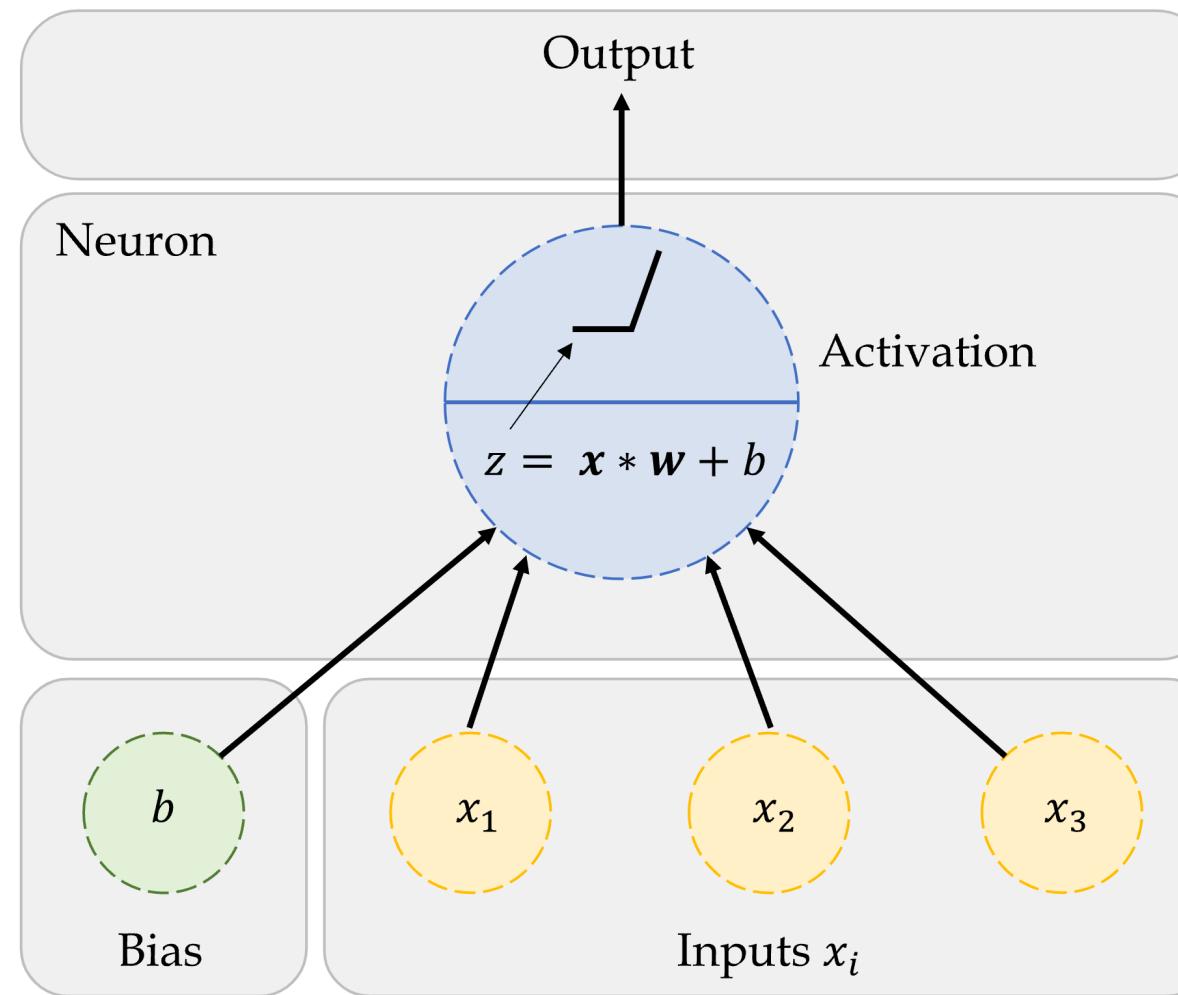
2-Minute Neuroscience: The Neuron



# A neuron ‘firing’



# An artificial neuron



A neuron in a neural network with a ReLU activation.



Source: Marcus Lautier (2022).

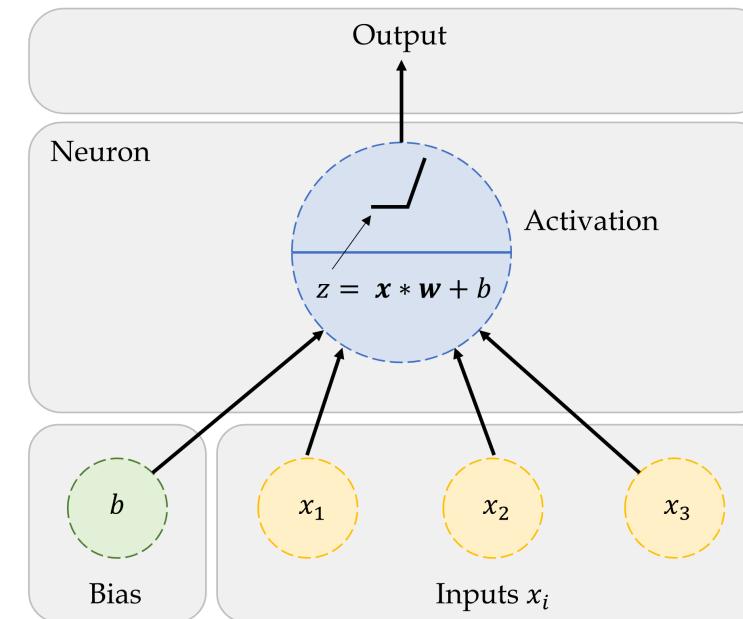
# One neuron

$$\begin{aligned} z &= x_1 \times w_1 + \\ &\quad x_2 \times w_2 + \\ &\quad x_3 \times w_3. \end{aligned}$$

$$a = \begin{cases} z & \text{if } z > 0 \\ 0 & \text{if } z \leq 0 \end{cases}$$

Here,  $x_1, x_2, x_3$  is just some fixed data.

The weights  $w_1, w_2, w_3$  should be ‘learned’.



A neuron in a neural network with a ReLU activation.



Source: Marcus Lautier (2022).

# One neuron with bias

$$z = x_1 \times w_1 +$$

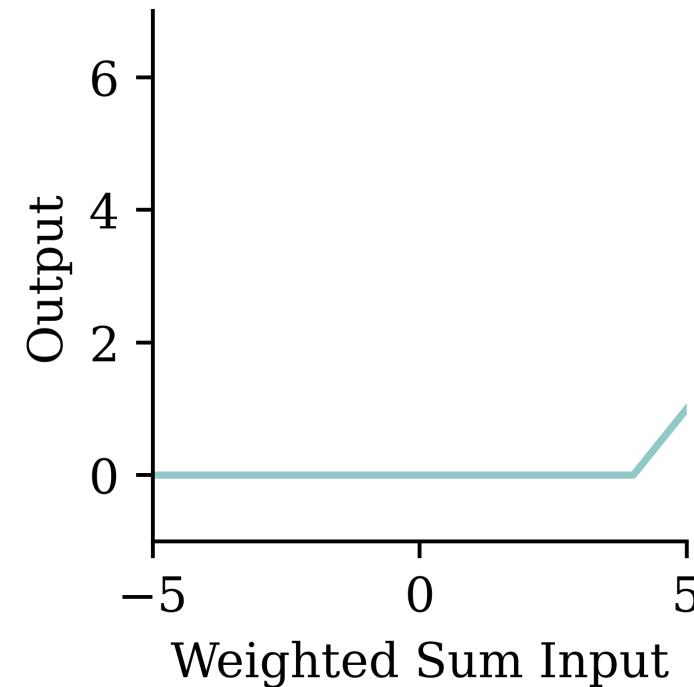
$$x_2 \times w_2 +$$

$$x_3 \times w_3 + b.$$

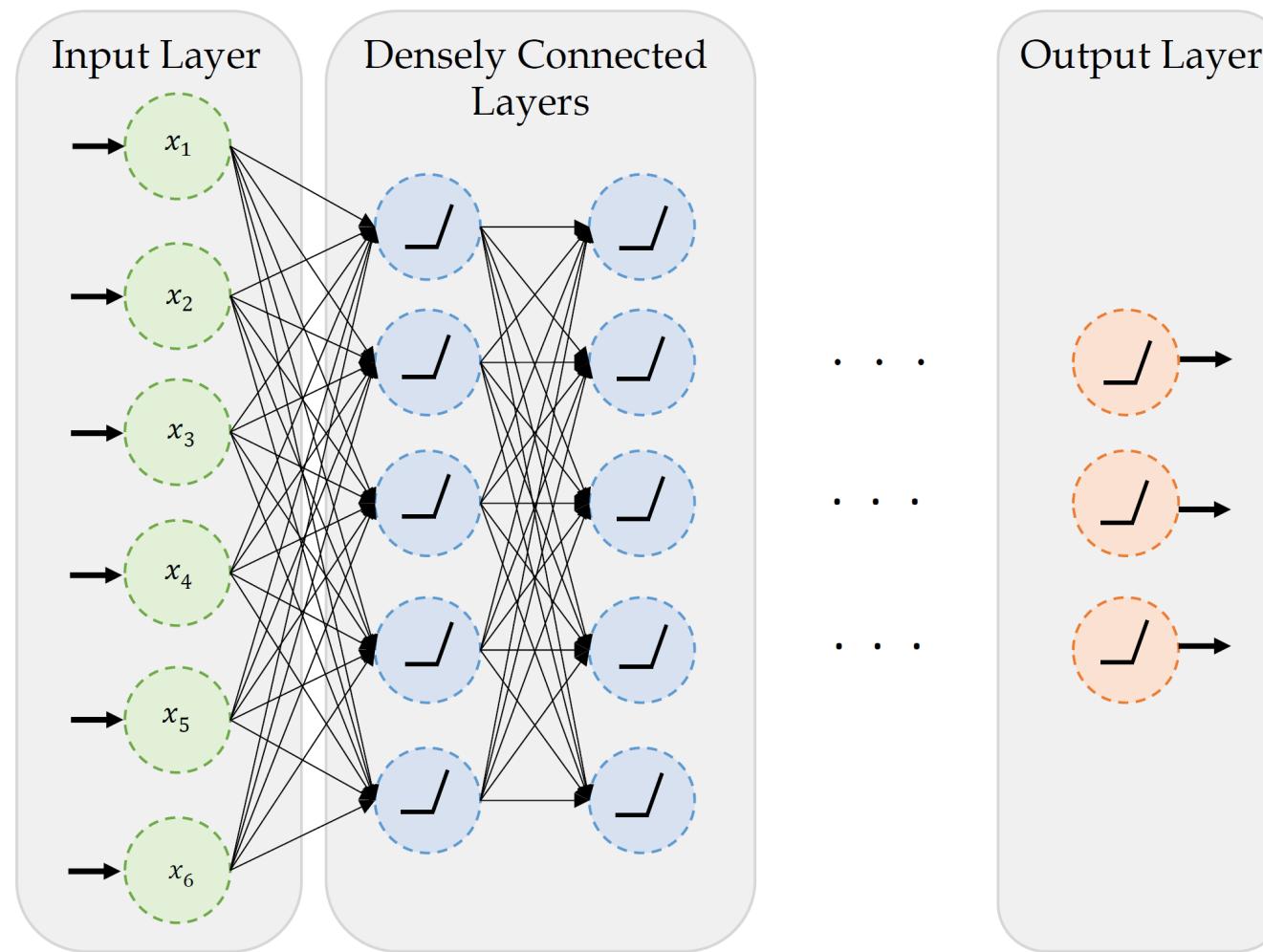
$$a = \begin{cases} z & \text{if } z > 0 \\ 0 & \text{if } z \leq 0 \end{cases}$$

The weights  $w_1, w_2, w_3$  and bias  $b$  should be ‘learned’.

Bias = -4	0	4
-----------	---	---



# A basic neural network



A basic fully-connected/dense network.



Source: Marcus Lautier (2022).

# Step-function activation

## Perceptrons

Brains and computers are binary, so make a perceptron with binary data. Seemed reasonable, impossible to train.

## Modern neural network

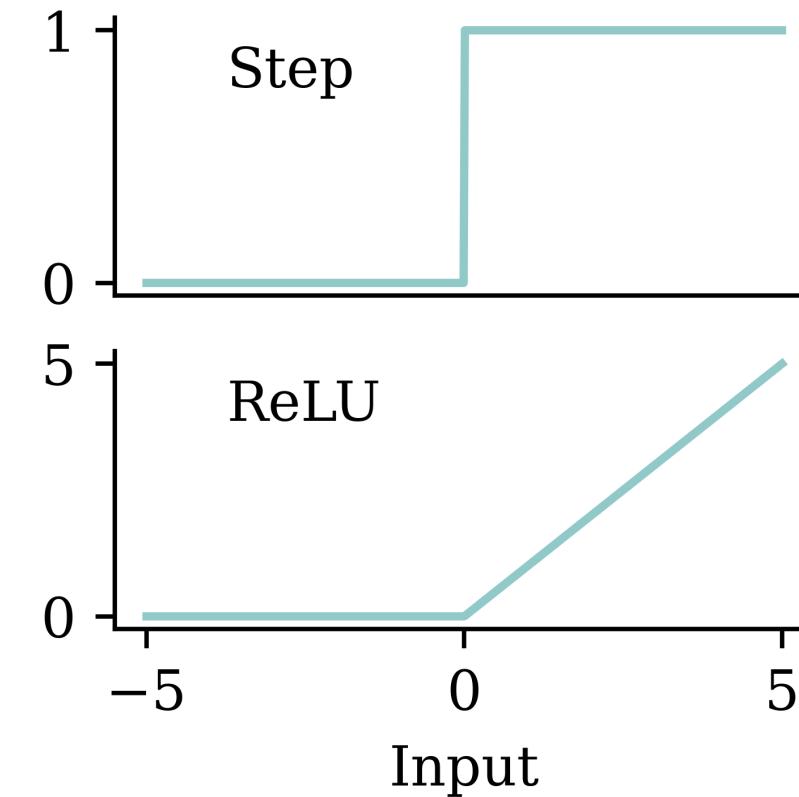
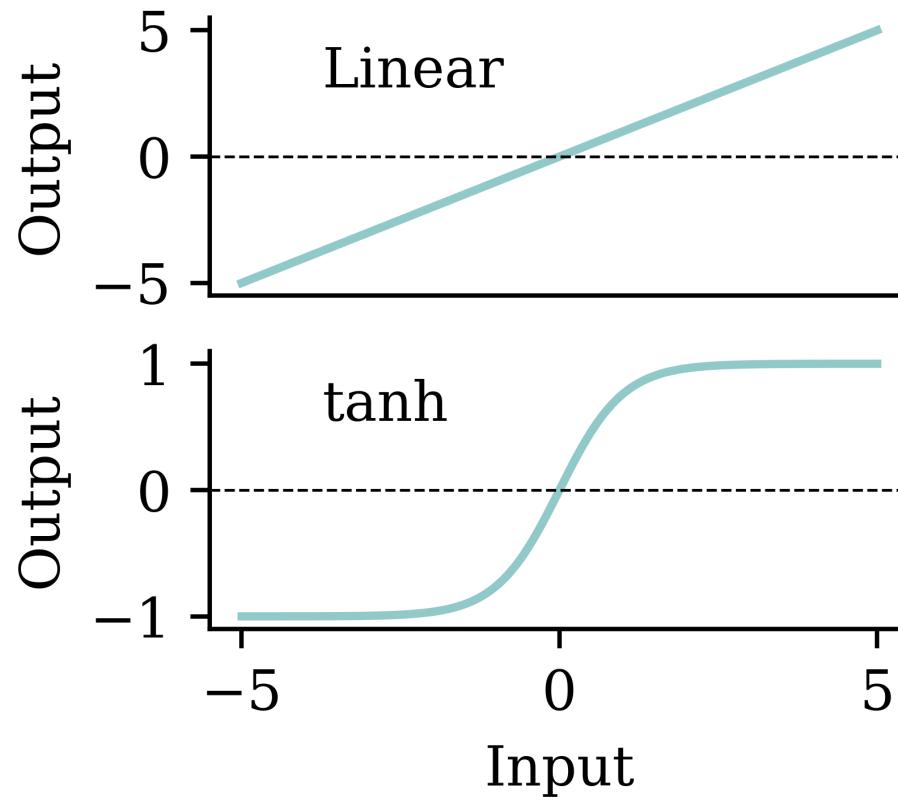
Replace binary state with continuous state. Still rather slow to train.



### Note

It's a **neural network** made of **neurons**, not a “neuron network”.

# Try different activation functions



# Flexible

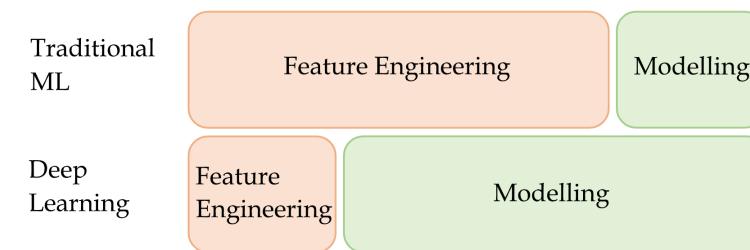
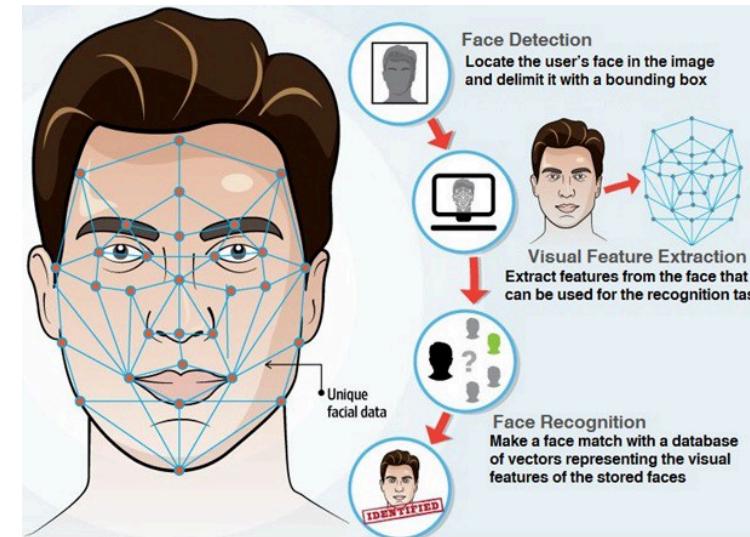
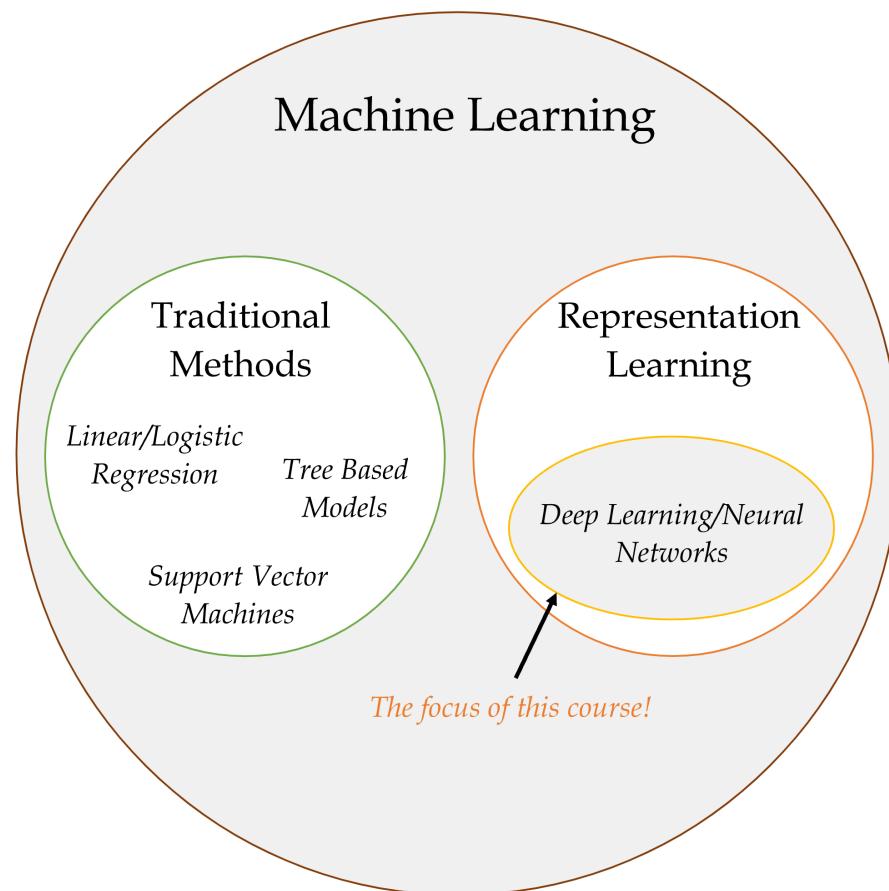
One can show that an MLP is a **universal approximator**, meaning it can model any suitably smooth function, given enough hidden units, to any desired level of accuracy (Hornik 1991). One can either make the model be “wide” or “deep”; the latter has some advantages...



Source: Murphy (2012), Machine Learning: A Probabilistic Perspective, 1st Ed, p. 566.



# Feature engineering



Sources: Marcus Lautier (2022) & Fenjiro (2019), *Face Id: Deep Learning for Face Recognition*, Medium.

# The deep learning hammer

Deep learning is not always the answer!



The map of data science.



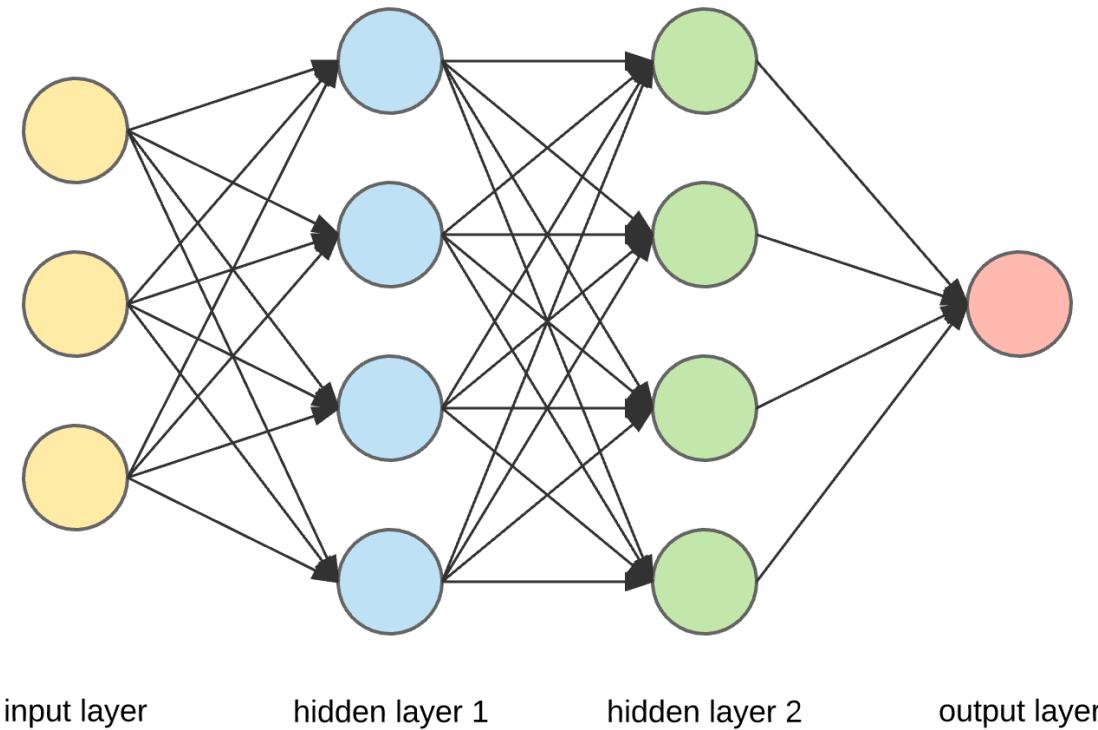
Source: Serge Masis (2022), [LinkedIn post](#).

# Quiz

In this ANN, how many of the following are there:

- features,
- targets,
- weights,
- biases, and
- parameters?

What is the depth?



An artificial neural network.



Source: Dertat (2017), *Applied Deep Learning - Part 1: Artificial Neural Networks*, Medium.



# Package Versions

```
1 from watermark import watermark  
2 print(watermark(python=True, packages="keras,matplotlib,numpy,pandas,seaborn,scipy,torch"))
```

Python implementation: CPython

Python version : 3.11.9

IPython version : 8.24.0

keras : 3.3.3

matplotlib: 3.9.0

numpy : 1.26.4

pandas : 2.2.2

seaborn : 0.13.2

scipy : 1.11.0

torch : 2.3.0

tensorflow: 2.16.1

tf\_keras : 2.16.0



# Glossary

- activations, activation function
- artificial neural network
- biases (in neurons)
- classification problem
- deep network, network depth
- dense or fully-connected layer
- feed-forward neural network
- labelled/unlabelled data
- machine learning
- neural network architecture
- perceptron
- ReLU
- representation learning
- sigmoid activation function
- targets
- training/test split
- weights (in a neuron)

