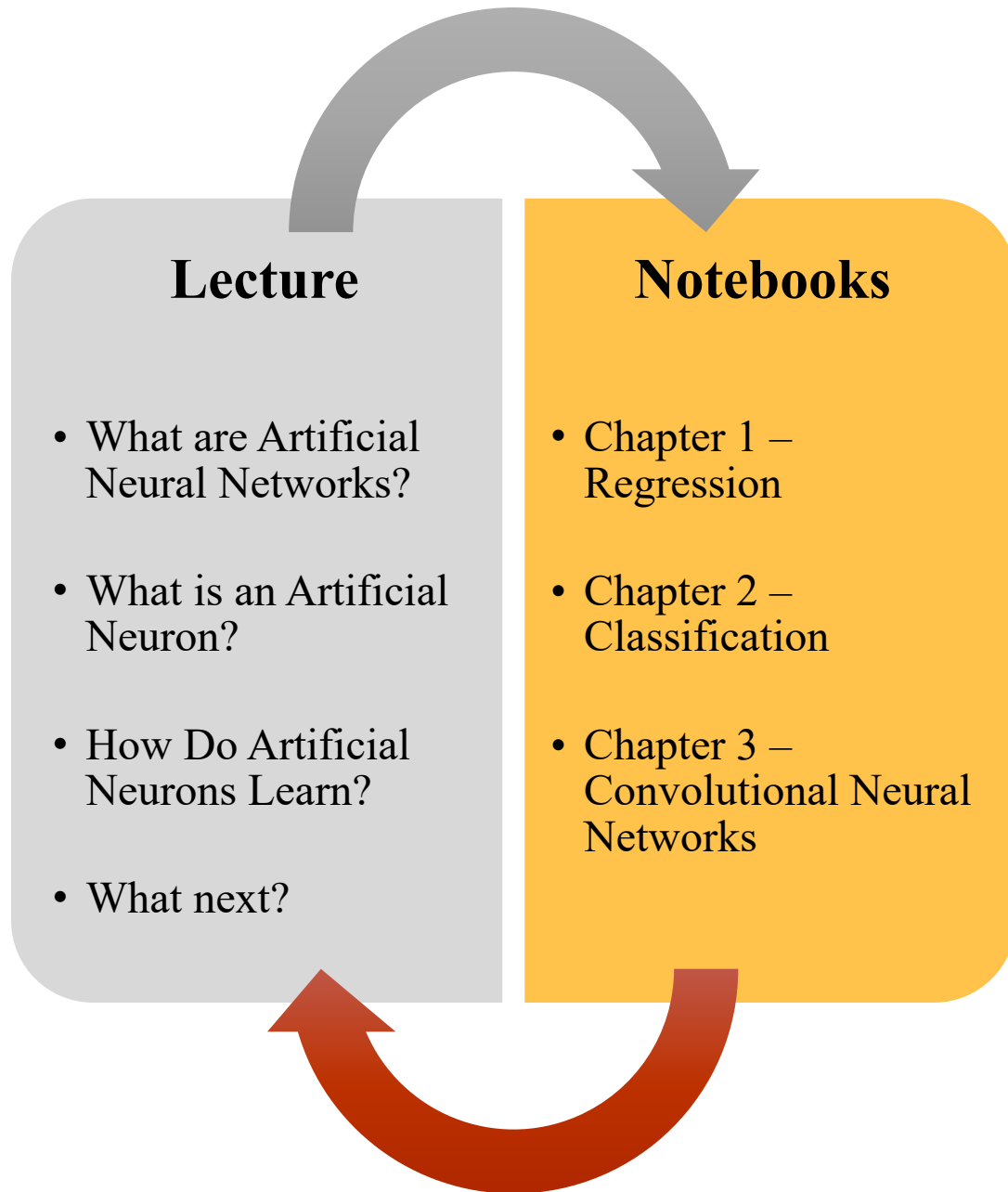


# Introduction to Neural Networks

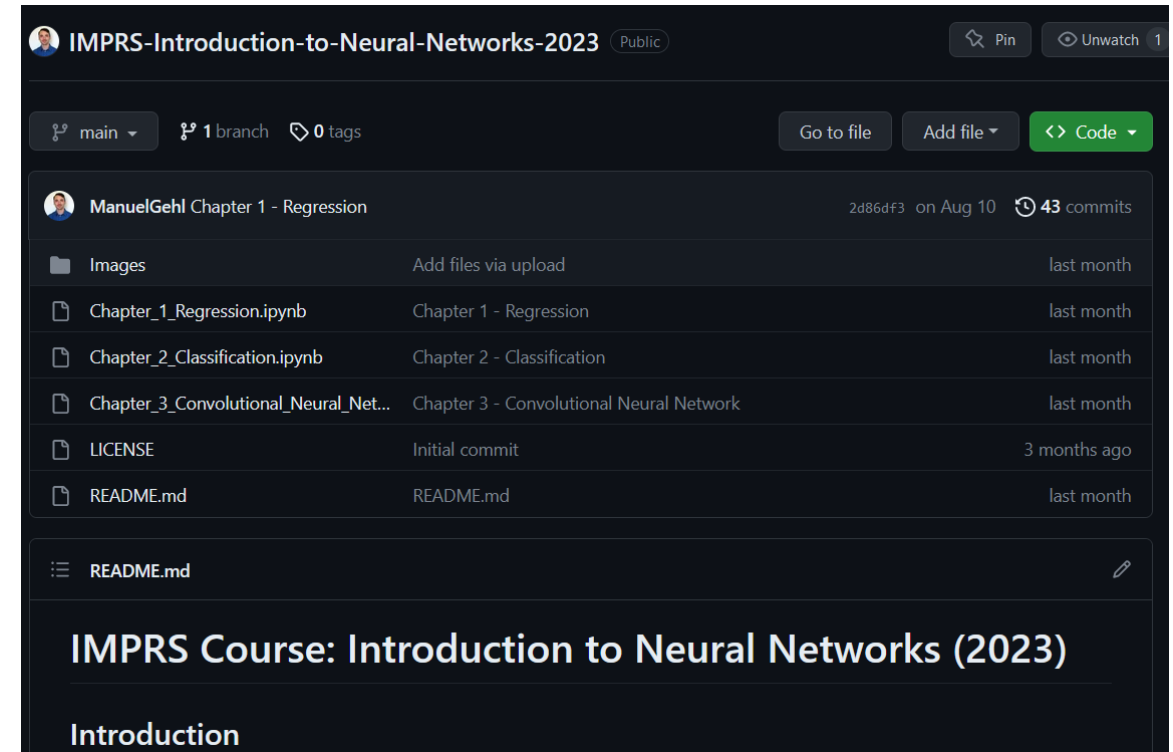
Dr. Manuel Gehl



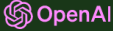
# Overview



## Course Website



# What are Artificial Neural Networks?

 OpenAI

Research ▾ Product ▾ Developers ▾ Safety Company ▾

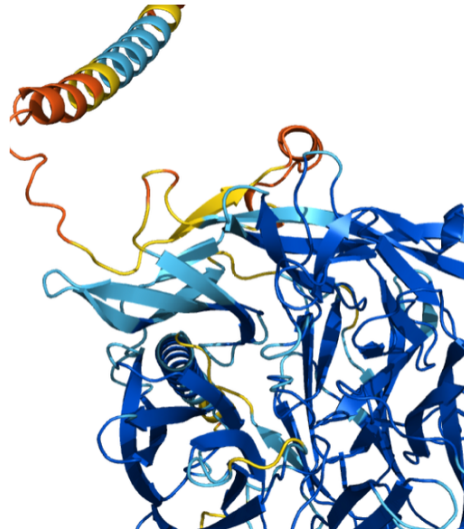
## Introducing ChatGPT

We've trained a model called ChatGPT which interacts in a conversational way. The dialogue format makes it possible for ChatGPT to answer followup questions, admit its mistakes, challenge incorrect premises, and reject inappropriate requests.

[Try ChatGPT ↗](#) [Read about ChatGPT Plus](#)

**AlphaFold** is an AI system developed by **DeepMind** that predicts a protein's 3D structure from its amino acid sequence. It regularly achieves accuracy competitive with experiment.

DeepMind and EMBL's European Bioinformatics Institute ([EMBL-EBI](#)) have partnered to create AlphaFold DB to make these predictions freely available to the scientific community. The latest database release contains over 200 million entries, providing broad coverage of [UniProt](#) (the standard repository of protein sequences and annotations). We provide individual [downloads](#) for the human proteome and for the proteomes of 47 other key organisms important in research and global health. We also provide a download for the manually curated subset of UniProt ([Swiss-Prot](#)).




Q8I3H7: May protect the malaria parasite against attack by the immune system.  
Mean pLDDT 85.57.

Input

An astronaut riding a horse in photorealistic style.

Output



[<](#) [>](#)

# What are Artificial Neural Networks?

## Deep Learning for Protein–Protein Interaction Site Prediction

[Arian R. Jamasb](#), [Ben Day](#), [Cătălina Cangea](#), [Pietro Liò](#) & [Tom L. Blundell](#) 

Protocol | [Open Access](#) | [First Online: 09 July 2021](#)

**5070** Accesses | **7** Citations | **10** Altmetric

Part of the [Methods in Molecular Biology](#) book series (MIMB, volume 2361)

## Classification and mutation prediction from non–small cell lung cancer histopathology images using deep learning

[Nicolas Coudray](#), [Paolo Santiago Ocampo](#), [Theodore Sakellaropoulos](#), [Navneet Narula](#), [Matija Snuderl](#),

[David Fenyö](#), [Andre L. Moreira](#), [Narges Razavian](#)  & [Aristotelis Tsirigos](#) 

[Nature Medicine](#) **24**, 1559–1567 (2018) | [Cite this article](#)

## Scaffolding protein functional sites using deep learning

[Jue Wang](#) , [Sidney Lisanza](#) , [David Juergens](#) , [Doug Tischer](#) , [Joseph L. Watson](#) , [Karla M. Castro](#), [Robert Ragotte](#) , [Amijai Saragovi](#) , [Lukas F. Milles](#) , [...] and [David Baker](#)  **+14 authors** [Authors Info & Affiliations](#)

*SCIENCE* • 21 Jul 2022 • Vol 377, Issue 6604 • pp. 387–394 • DOI: 10.1126/science.abn2100

## De novo design of luciferases using deep learning

[Andy Hsien-Wei Yeh](#) , [Christoffer Norn](#), [Yakov Kipnis](#), [Doug Tischer](#), [Samuel J. Pellock](#), [Declan Evans](#), [Pengchen Ma](#), [Gyu Rie Lee](#), [Jason Z. Zhang](#), [Ivan Anishchenko](#), [Brian Coventry](#), [Longxing Cao](#), [Justas Dauparas](#), [Samer Halabiya](#), [Michelle DeWitt](#), [Lauren Carter](#), [K. N. Houk](#) & [David Baker](#) 

[Nature](#) **614**, 774–780 (2023) | [Cite this article](#)

## Geometric deep learning of RNA structure

[Raphael J. L. Townshend](#) , [Stephan Eismann](#), [Andrew M. Watkins](#) , [Ramya Rangan](#) , [Masha Karelina](#) , [Rhiiju Das](#) , and [Ron O. Dror](#) 

[Authors Info & Affiliations](#)

*SCIENCE* • 26 Aug 2021 • Vol 373, Issue 6558 • pp. 1047–1051 • DOI: 10.1126/science.abe5650

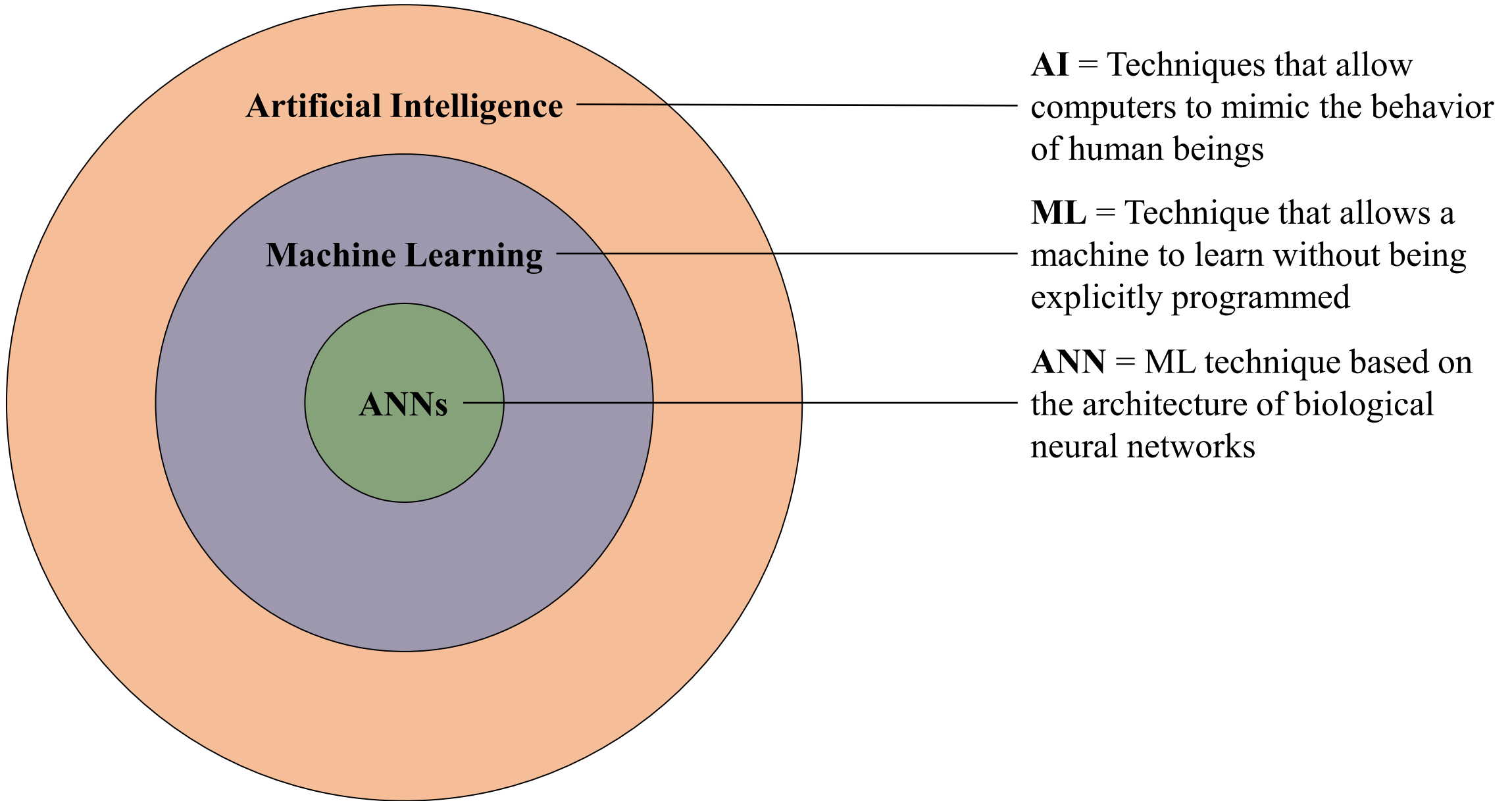
STRUCTURAL BIOLOGY

## Protein sequence design by deep learning

[Jue Wang](#) 

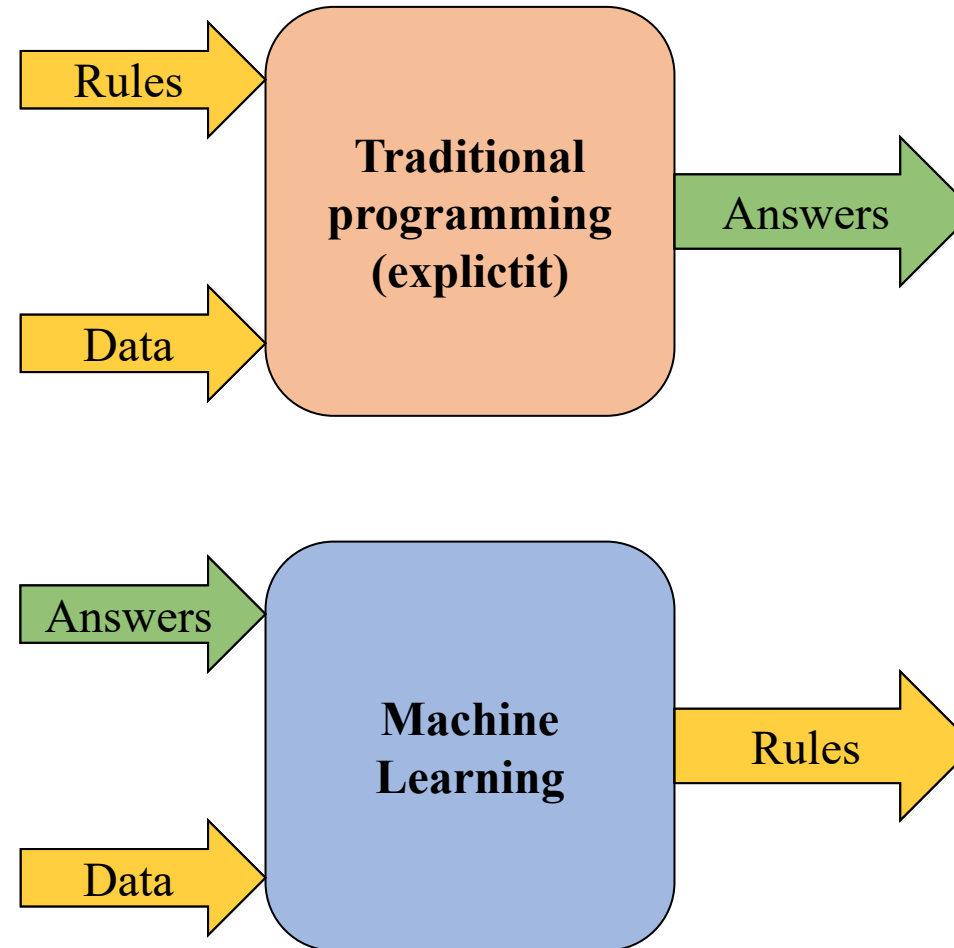
[Nature Computational Science](#) **2**, 416–417 (2022) | [Cite this article](#)

# What are Artificial Neural Networks?



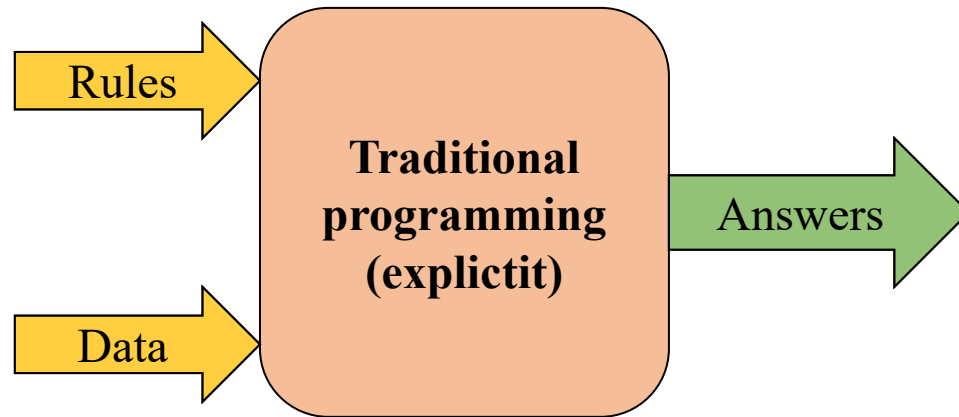
# What are Artificial Neural Networks?


**ML** = Technique that allows a machine to learn without being explicitly programmed




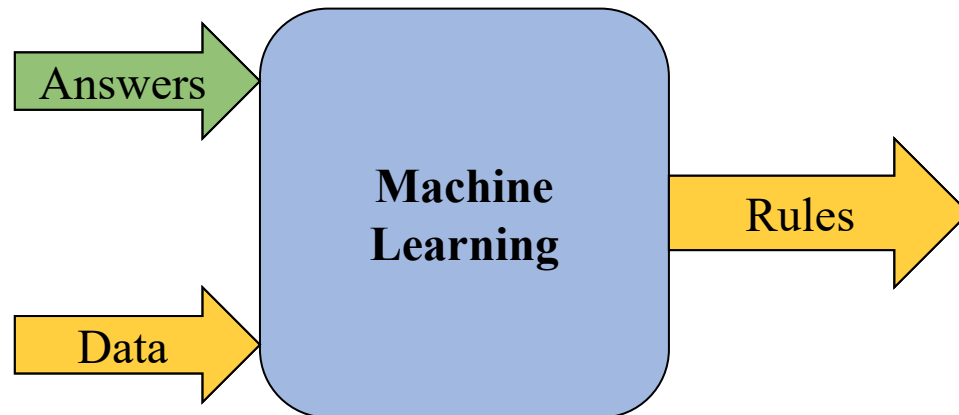
# What are Artificial Neural Networks?



**ML** = Technique that allows a machine to learn without being explicitly programmed

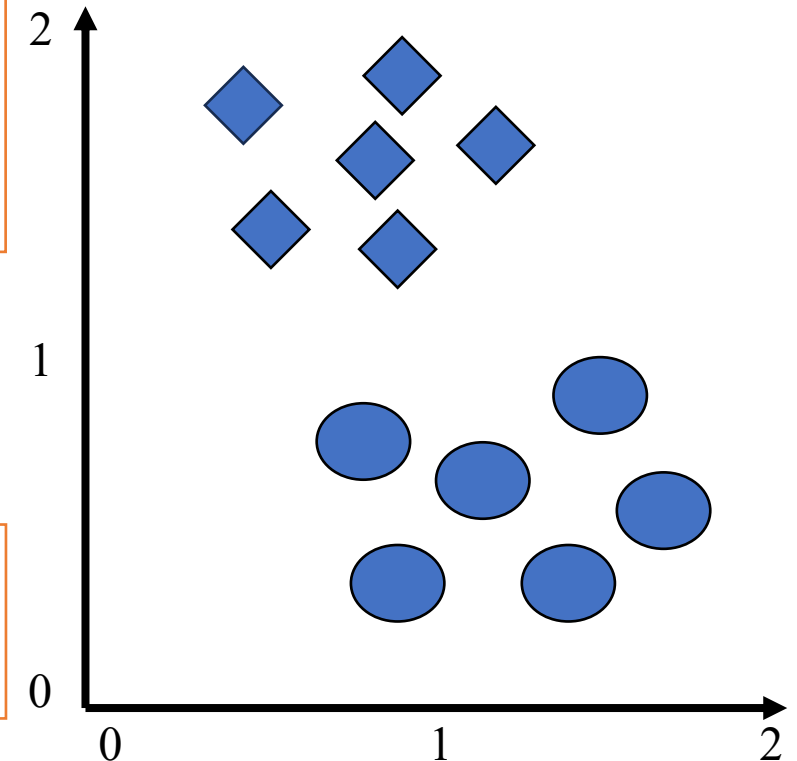


If  $x > 0.5$  and  $y > 1$ ,  
the answer is: 

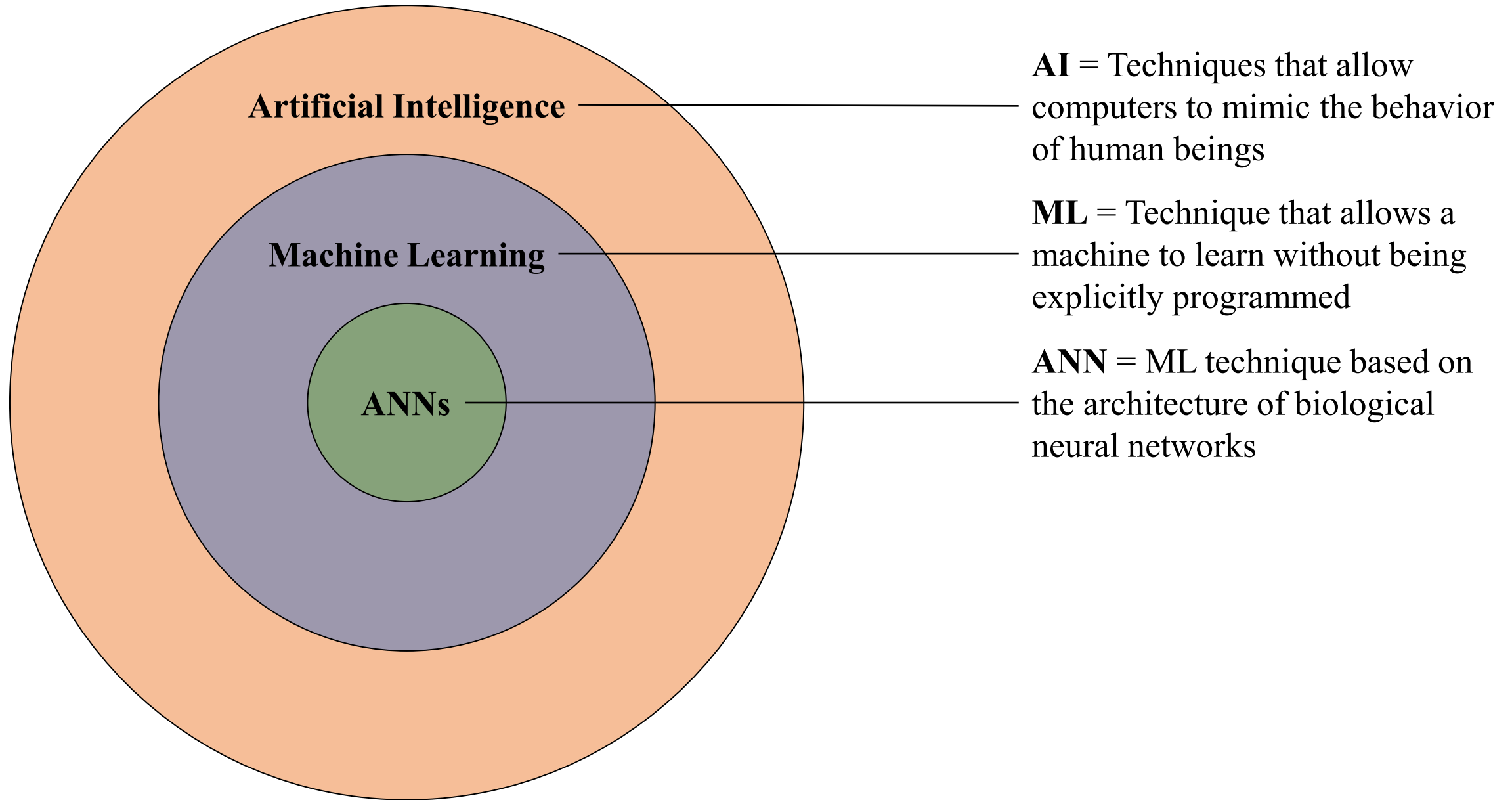
If  $x > 0.8$  and  $y < 1$ ,  
the answer is: 



 and  are two classes. Learn to distinguish them.

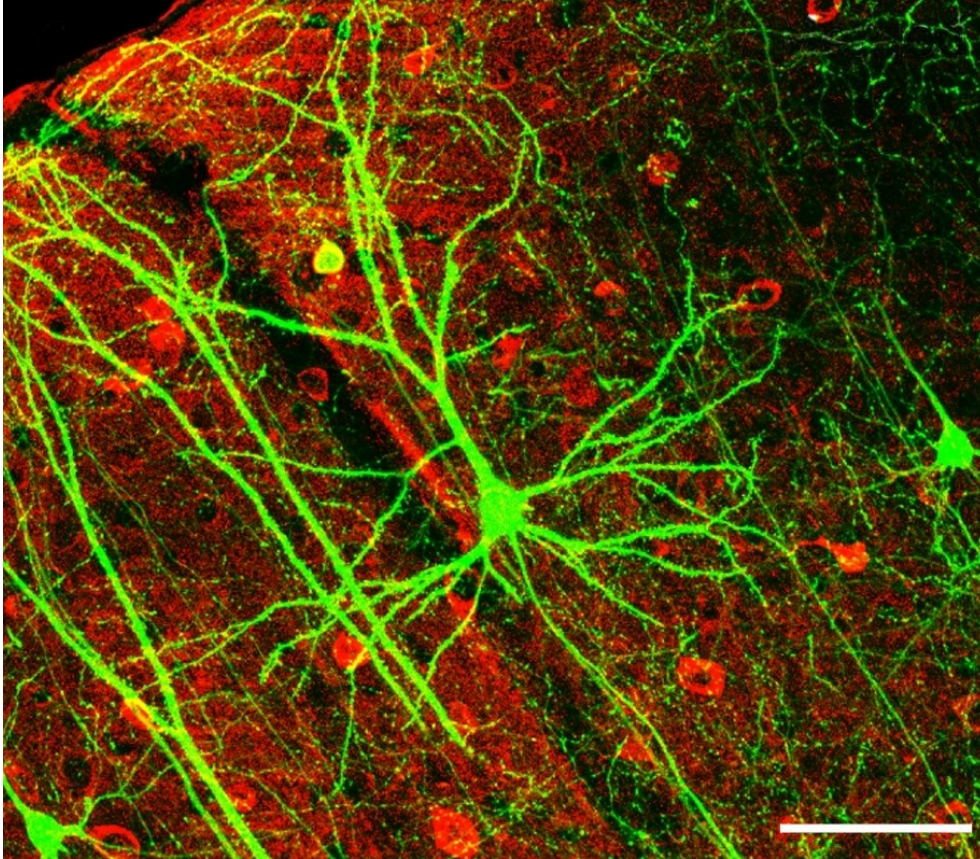


# What are Artificial Neural Networks?





# What are Artificial Neural Networks?



Wei-Chung Allen Lee, Hayden Huang, Guoping Feng, Joshua R. Sanes, Emery N. Brown, Peter T. So, Elly Nedivi, [PLoS Biol 4.e126.Fig6fNeuron](#), [CC BY 2.5](#)

1. **Neurons (units):** Basic processing unit.
2. **Connectivity:** Neurons are connected to each other.
3. **Activation:** Neurons „fire“ after exceeding threshold.
4. **Layers:** Neurons are organized in layers.
5. **Feedforward information flow:** From input to output.
6. **Feedback information flow:** From output to input.
7. **Learning and Adaptation:** Adapting connections to stimuli.
8. **Parallel processing:** Processing information simultaneously.

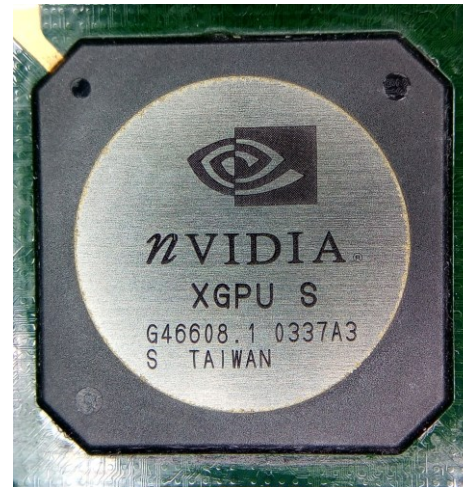
# What are Artificial Neural Networks?

## 1. Data



[Zxb](#), [Storage size comparison](#), [CC BY-SA 2.0 AT](#)

## 2. Hardware



[A7N8X](#), [Xbox GPU](#), [CC BY-SA 4.0](#)

## 3. Software



[vscode-icons](#),  
[Pytorch-svgrepo-com](#), [CC BY-SA 4.0](#)



[vscode-icons](#),  
[Tensorflow-svgrepo-com](#), [CC BY-SA 4.0](#)

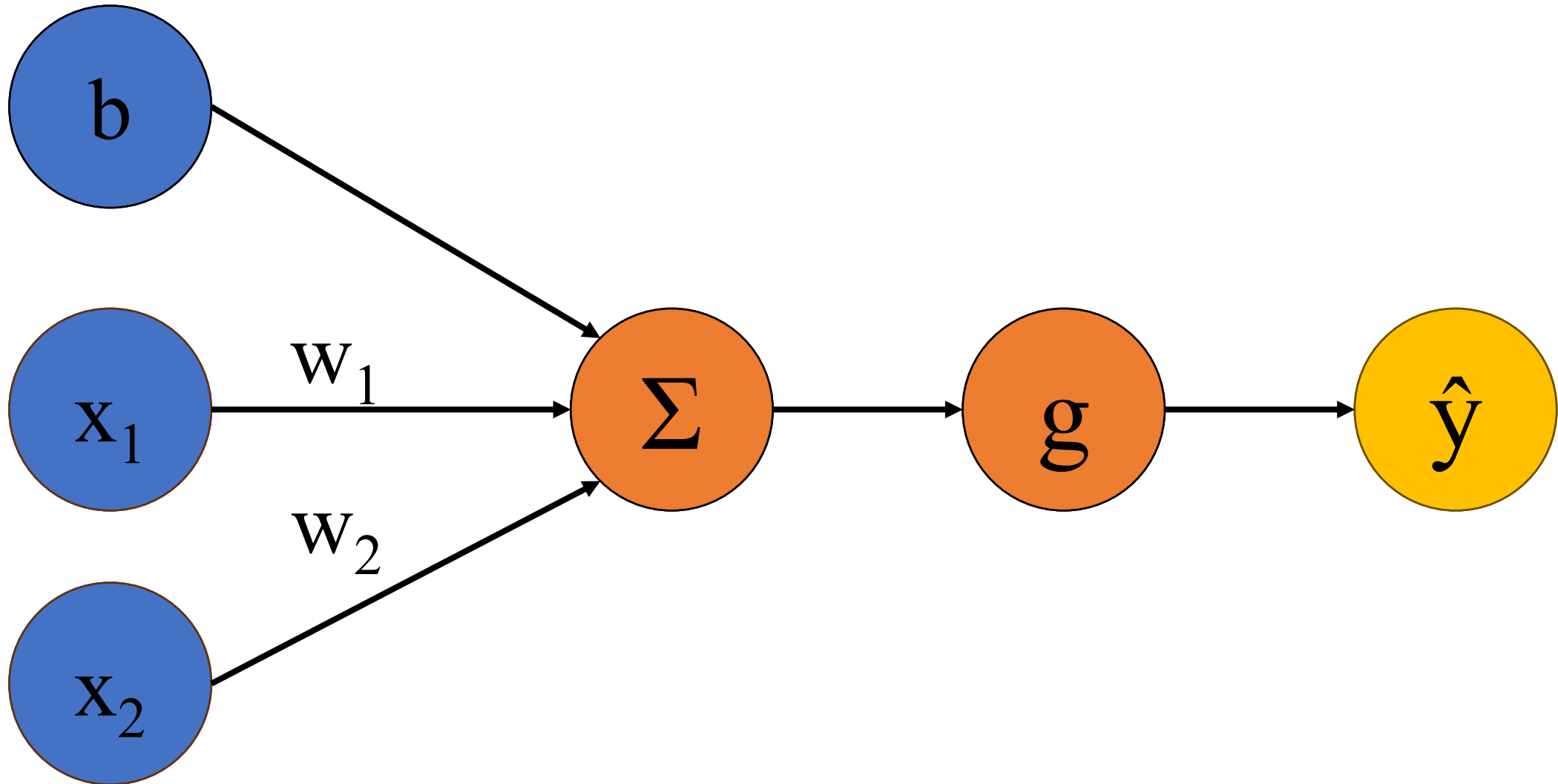
# What is an Artificial Neuron?

Inputs

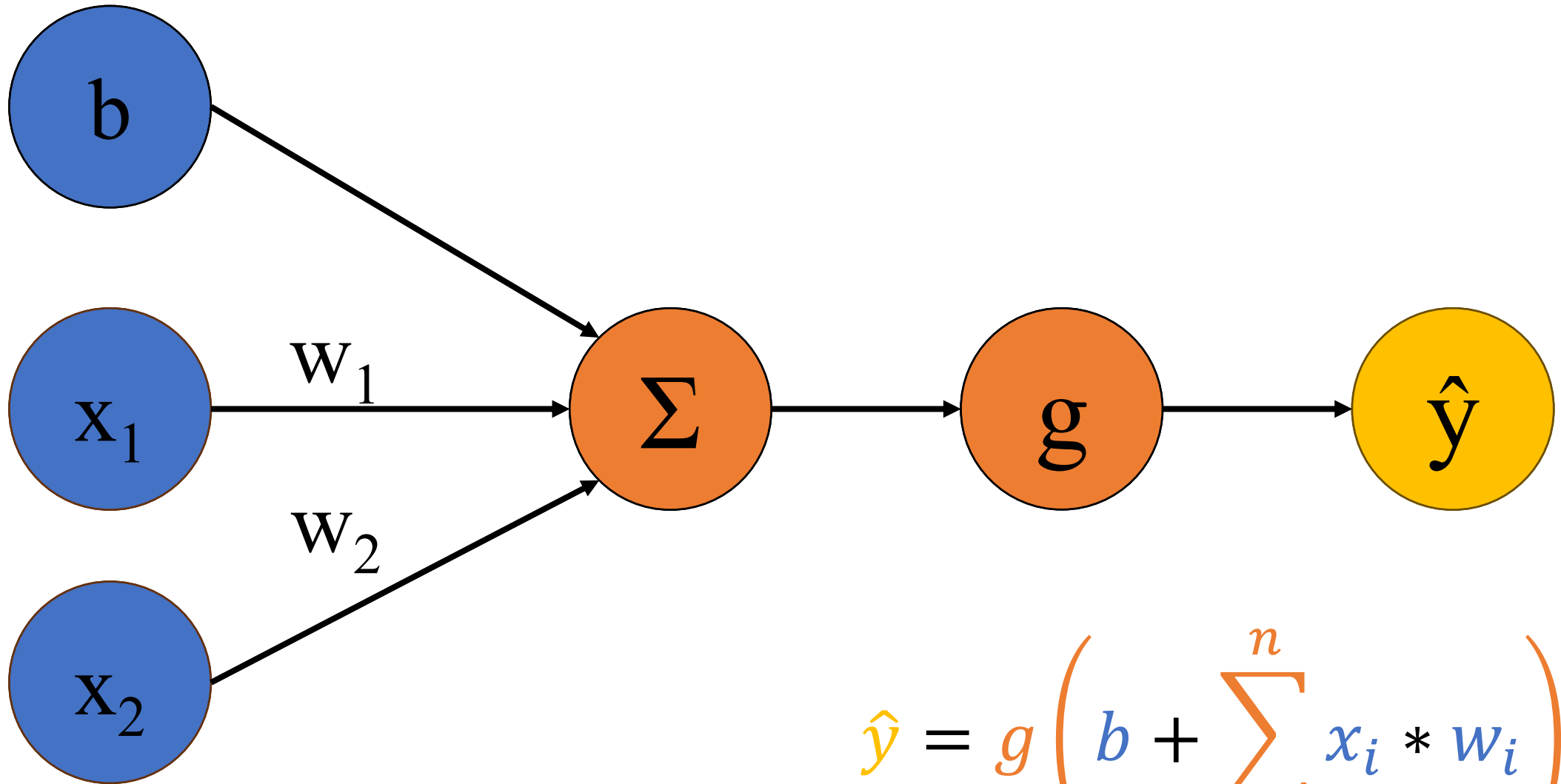
Neuron/Unit

Activation

Output

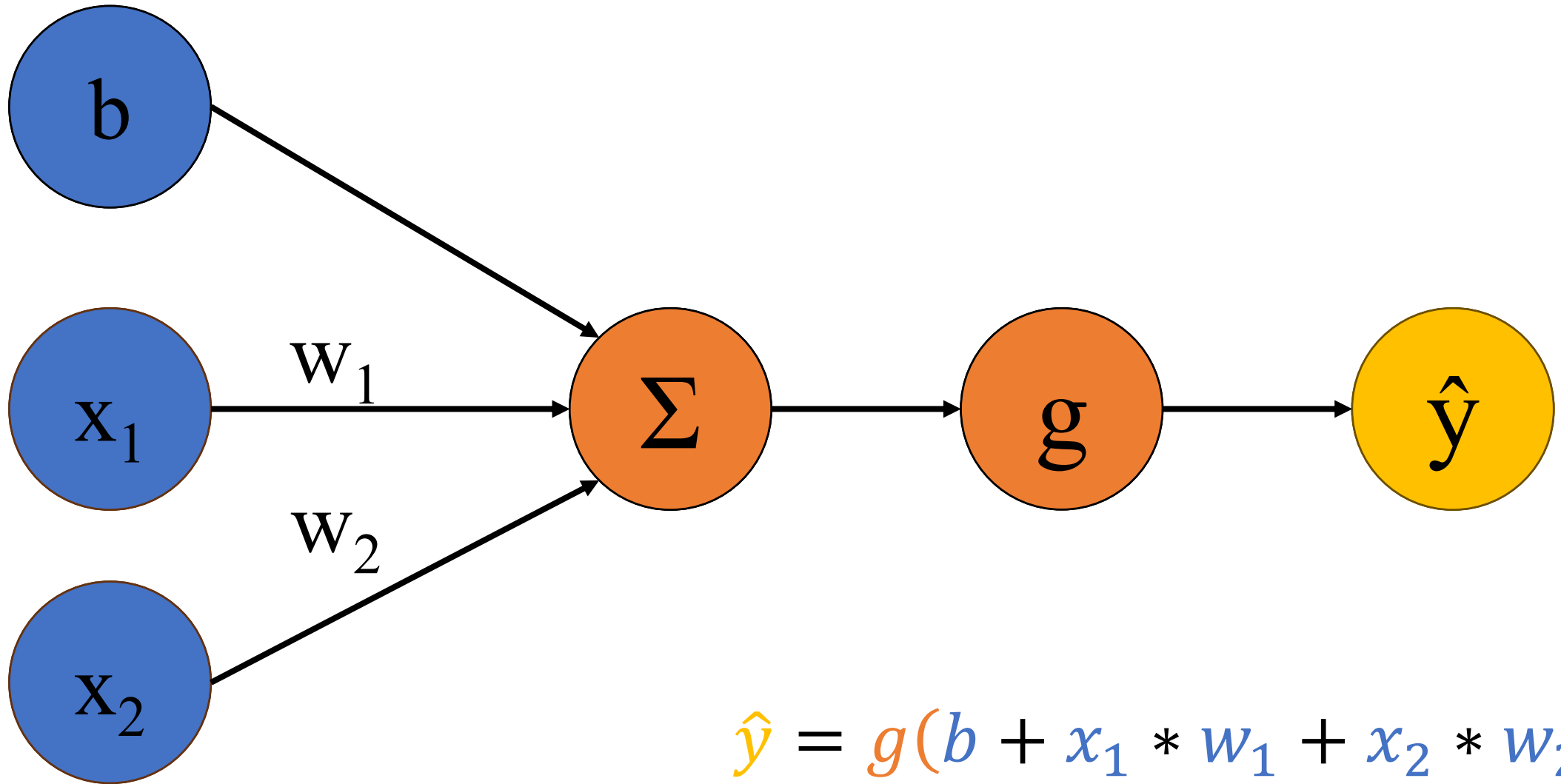


# What is an Artificial Neuron?



$$\hat{y} = g \left( b + \sum_{i=1}^n x_i * w_i \right)$$

# What is an Artificial Neuron?





# What is an Artificial Neuron?

## Example: Oktoberfest



[Heribert Pohl --- Thanks for half a million clicks!](#) from Germering bei München, Bayern, [O'zapft is! Münchens 5 Jahreszeit hat begonnen - O'zapft is! Munich 5 season, the Oktoberfest has begun \(9855483374\)](#), [CC BY-SA 2.0](#)

**Prediction:** Are you going to the Oktoberfest this Saturday?

**Criteria (weights):**

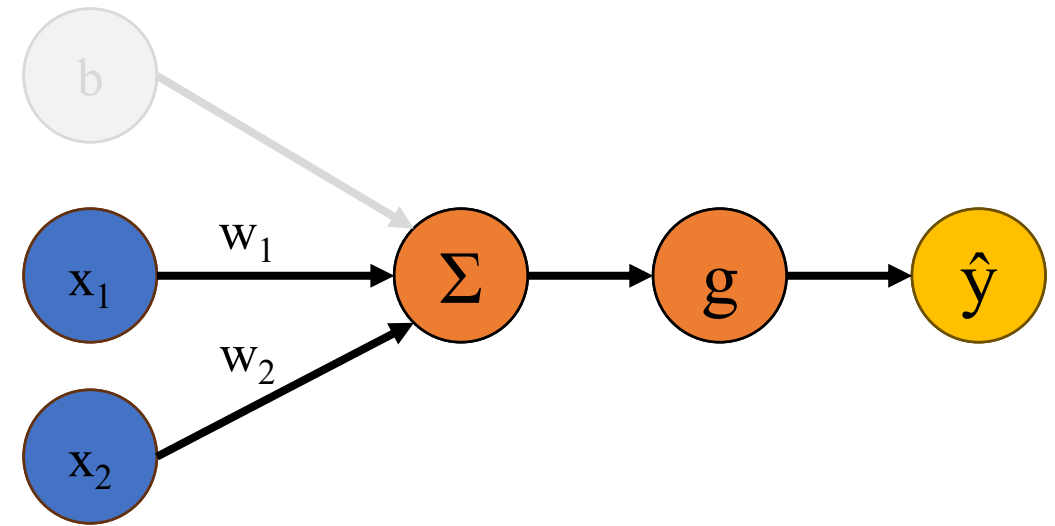
1. Do you like beer?
  - Yes (1) – No (0)
2. How important is the weather to you?
  - Very (1) – Not at all (0)

# What is an Artificial Neuron?

**Prediction:** Are you going to the Oktoberfest this Saturday?

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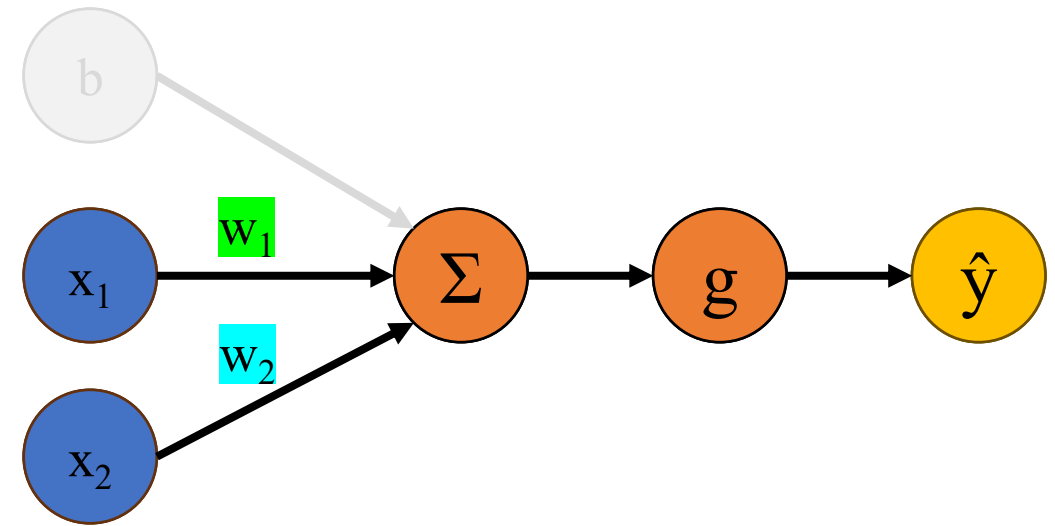
$$\hat{y} = g(x_1 * w_1 + x_2 * w_2)$$

# What is an Artificial Neuron?

**Prediction:** Are you going to the Oktoberfest this Saturday?

**Criteria:**

1. Do you like beer?
  - Yes (1) – No (0)
2. How important is the weather to you?
  - Very (1) – Not at all (0)



$$\hat{y} = g(x_1 * w_1 + x_2 * w_2)$$

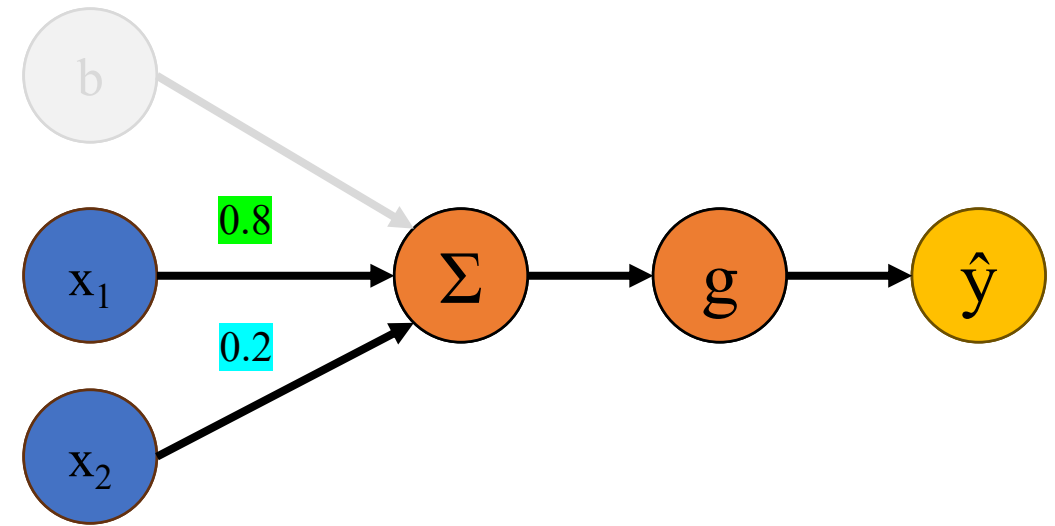


# What is an Artificial Neuron?

**Prediction:** Are you going to the Oktoberfest this Saturday?

**Person 1:**

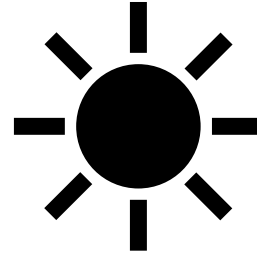
1. **Do you like beer?**
  - „Pretty much“ (0.8)
2. **How important is the weather to you?**
  - “There is no bad weather, only the wrong clothes” (0.2)



$$\hat{y} = g(x_1 * \mathbf{0.8} + x_2 * \mathbf{0.2})$$

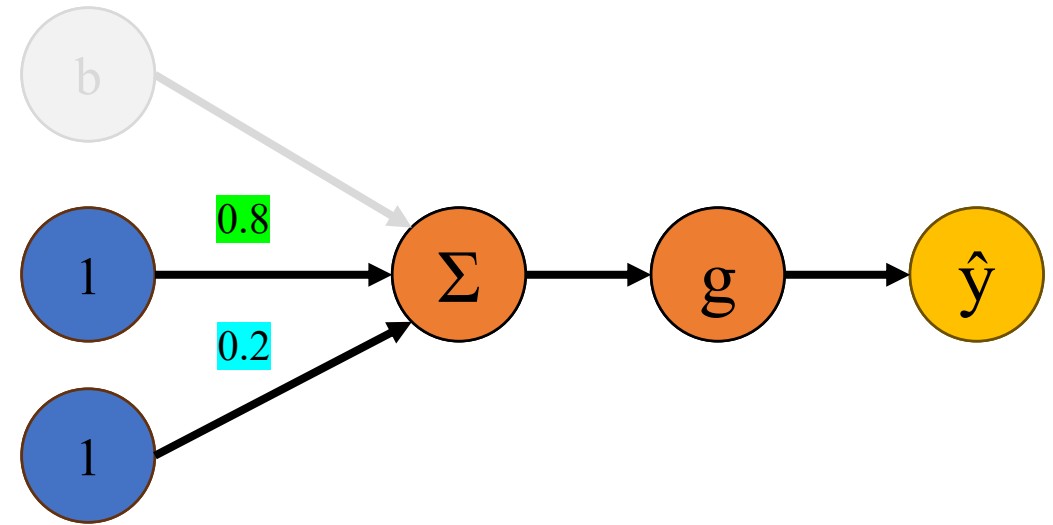
# What is an Artificial Neuron?

**Prediction:** Are you going to the Oktoberfest this Saturday?



Person 1:

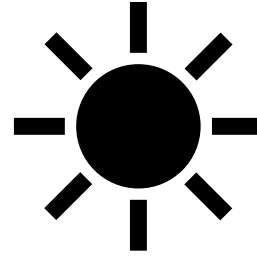
1. Do you like beer?
  - „Pretty much“ (0.8)
2. How important is the weather to you?
  - “There is no bad weather, only the wrong clothes” (0.2)



$$\hat{y} = g(1 * 0.8 + 1 * 0.2)$$
$$\hat{y} = g(1)$$

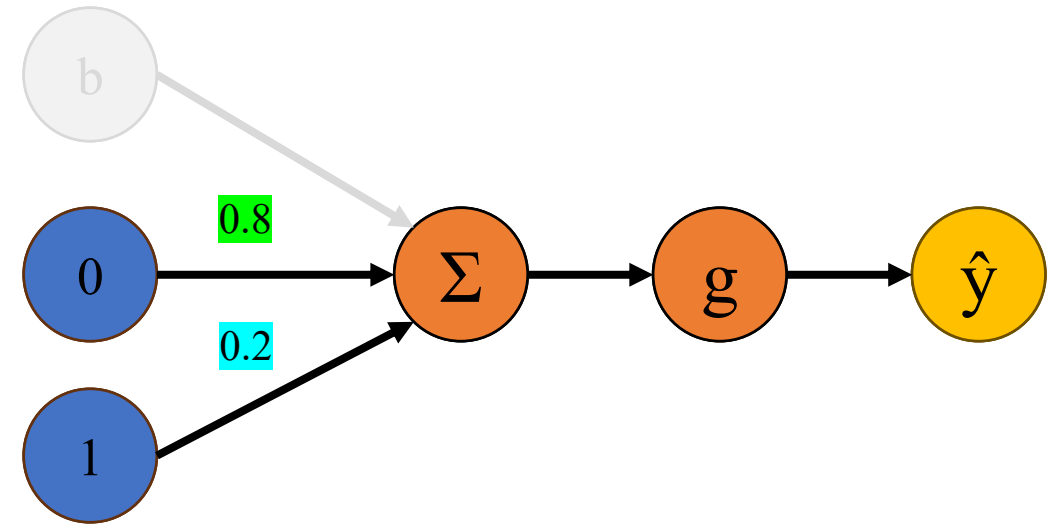
# What is an Artificial Neuron?

**Prediction:** Are you going to the Oktoberfest this Saturday?



**Person 1:**

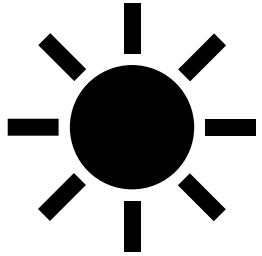
1. **Do you like beer?**
  - „Pretty much“ (0.8)
2. **How important is the weather to you?**
  - “There is no bad weather, only the wrong clothes” (0.2)



$$\hat{y} = g(0 * 0.8 + 1 * 0.2)$$
$$\hat{y} = g(0.2)$$

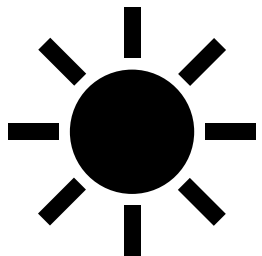
# What is an Artificial Neuron?

**Prediction:** Are you going to the Oktoberfest this Saturday?



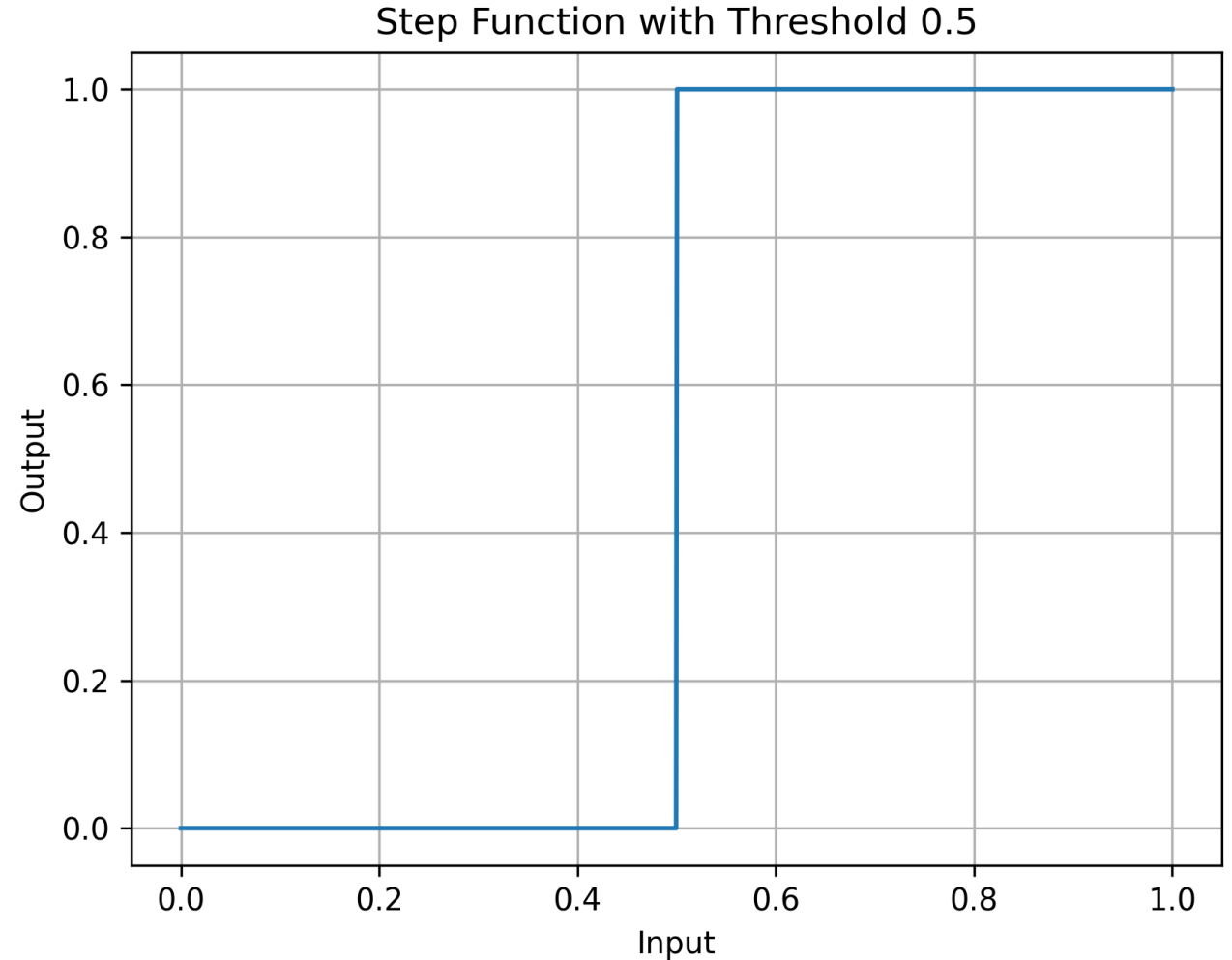
$$\hat{y} = g(1 * 0.8 + 1 * 0.2)$$

$$\hat{y} = g(1)$$



$$\hat{y} = g(0 * 0.8 + 1 * 0.2)$$

$$\hat{y} = g(0.2)$$



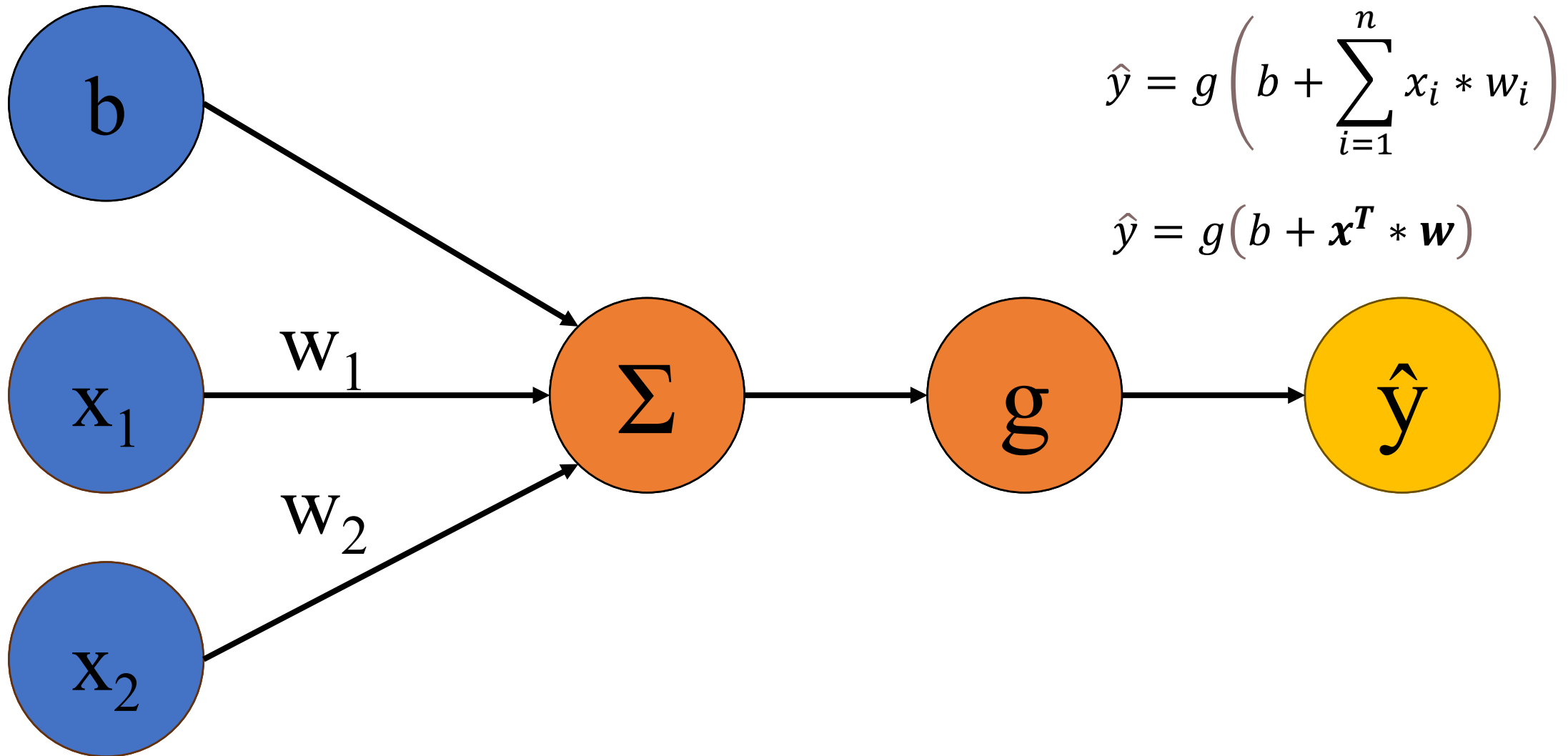
# What is an Artificial Neuron?

Inputs

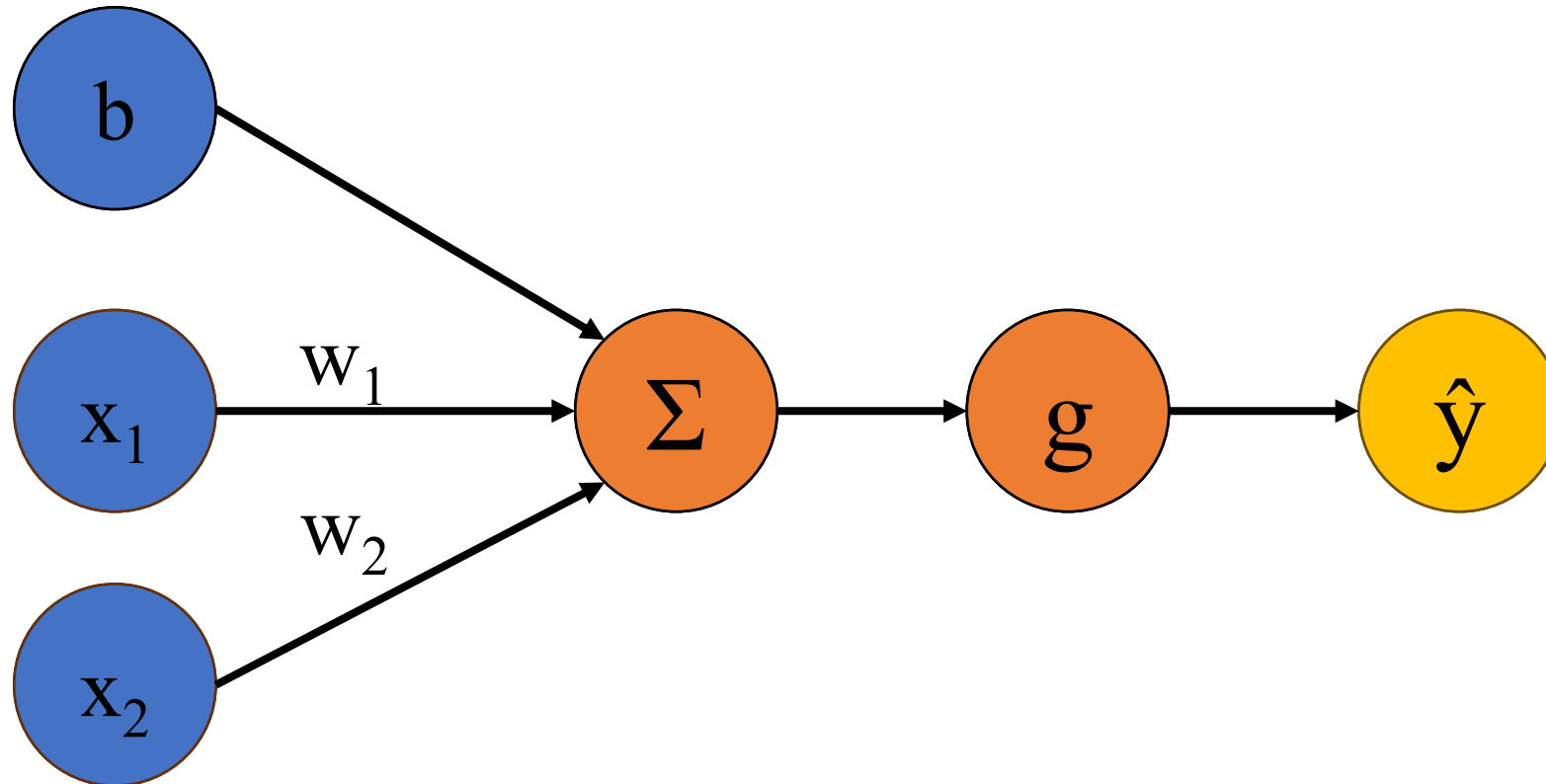
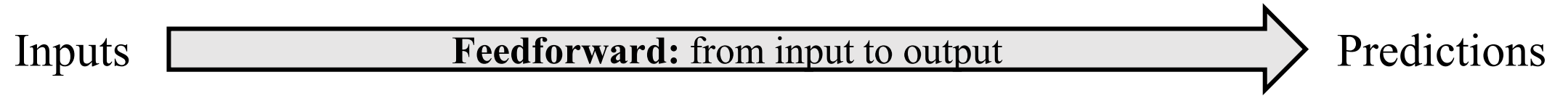
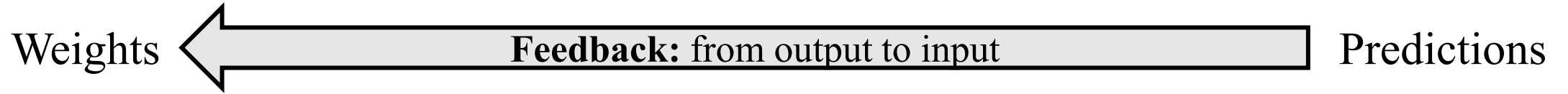
Neuron/Unit

Activation

Output



# How Do Artificial Neurons Learn?



# How Do Artificial Neurons Learn?

**Loss functions** measure difference between true values ( $y$ ) and predicted values ( $\hat{y}$ ):

- Mean Absolute Error (MAE)
- Mean Squared Error (MSE)
- Binary Crossentropy
- Categorical Crossentropy
- ...

$$\sum_{i=1}^n \frac{|\hat{y}_i - y_i|}{n}$$

$$\sum_{i=1}^n \frac{y_i * \log(p(y_i)) + (1 - y_i) * (\log(1 - p(y_i)))}{n}$$

$p(y_i)$  = Probability the model assigns the value to be  $\hat{y}$

# How Do Artificial Neurons Learn?

**Loss functions** measure difference between true values ( $y$ ) and predicted values ( $\hat{y}$ ):

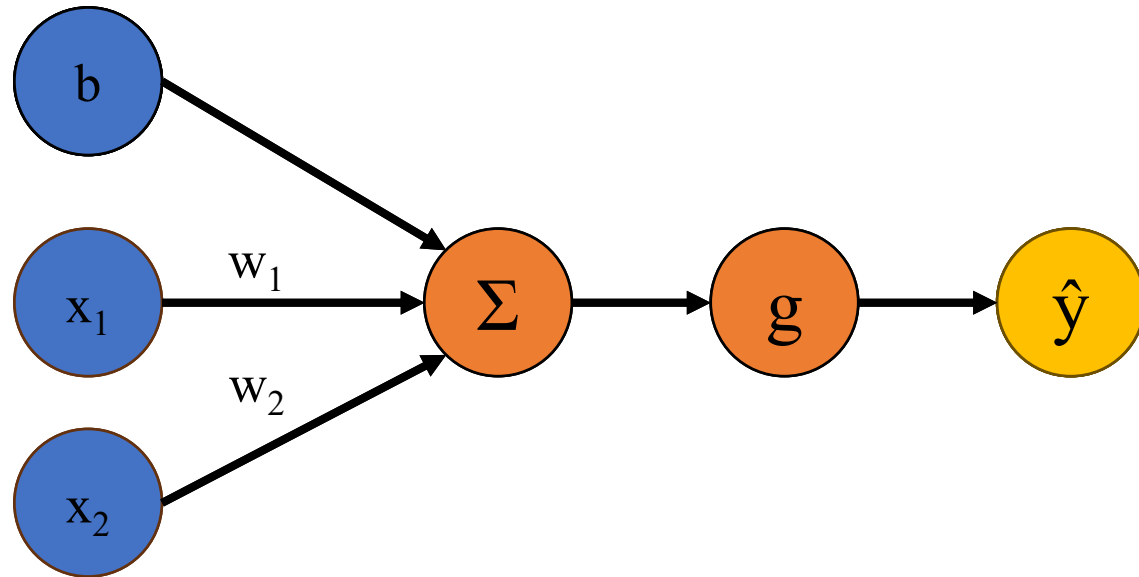
- Mean Absolute Error (MAE)
- Mean Squared Error (MSE)
- Binary Crossentropy
- Categorical Crossentropy
- ...

**Training** an artificial neuron (or network; learning) refers to the **minimization of the loss function**

The error is a function of all neurons, which in turn are a function of all weights. The error is **backpropagated** through the network and the weights are adjusted to minimize the loss function.



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The error is a function of all neurons, which in turn are a function of all weights. The error is **backpropagated** through the network and the weights are adjusted to minimize the loss function.

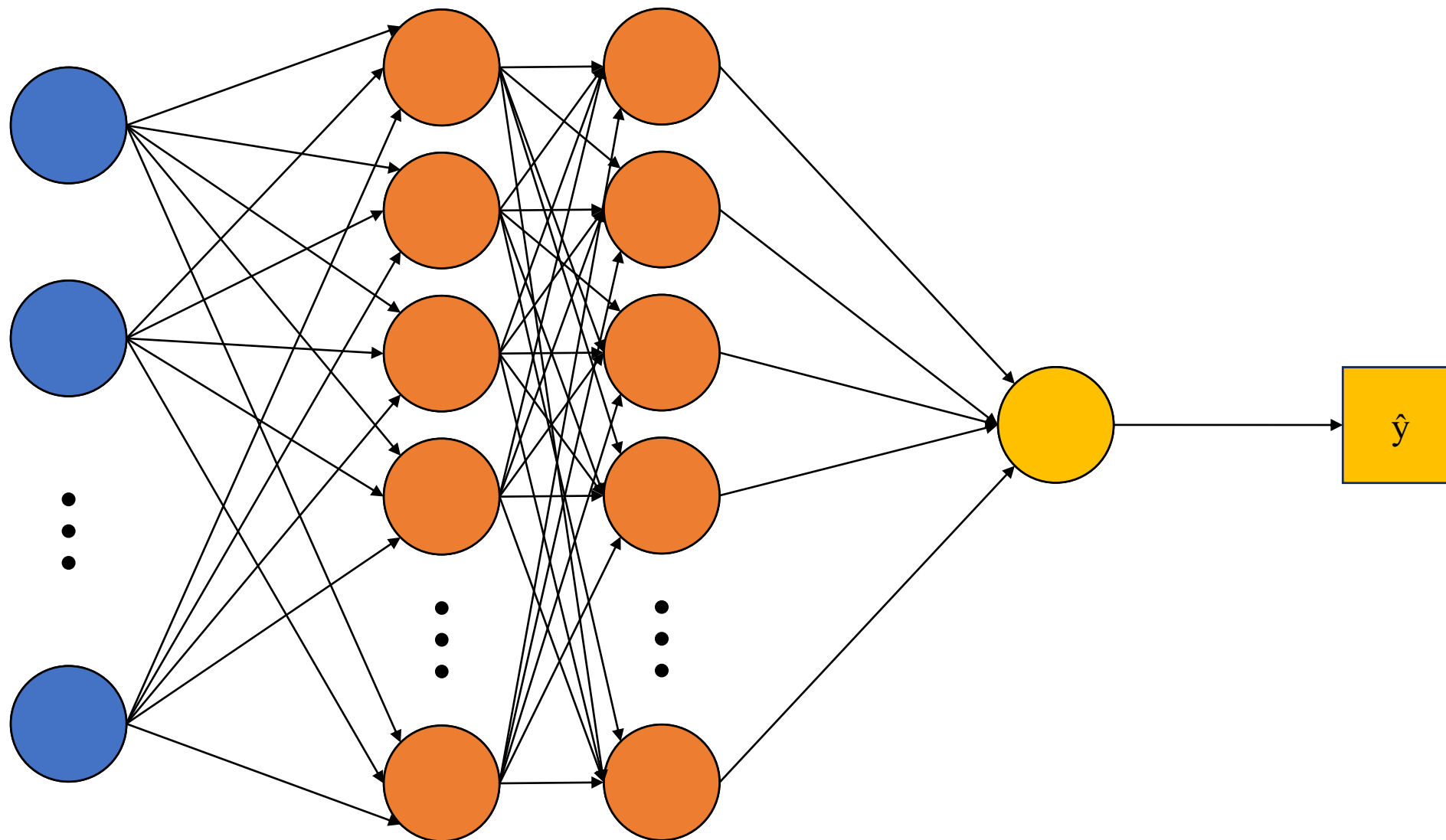
# What is an Artificial Neural Network?

Input layer

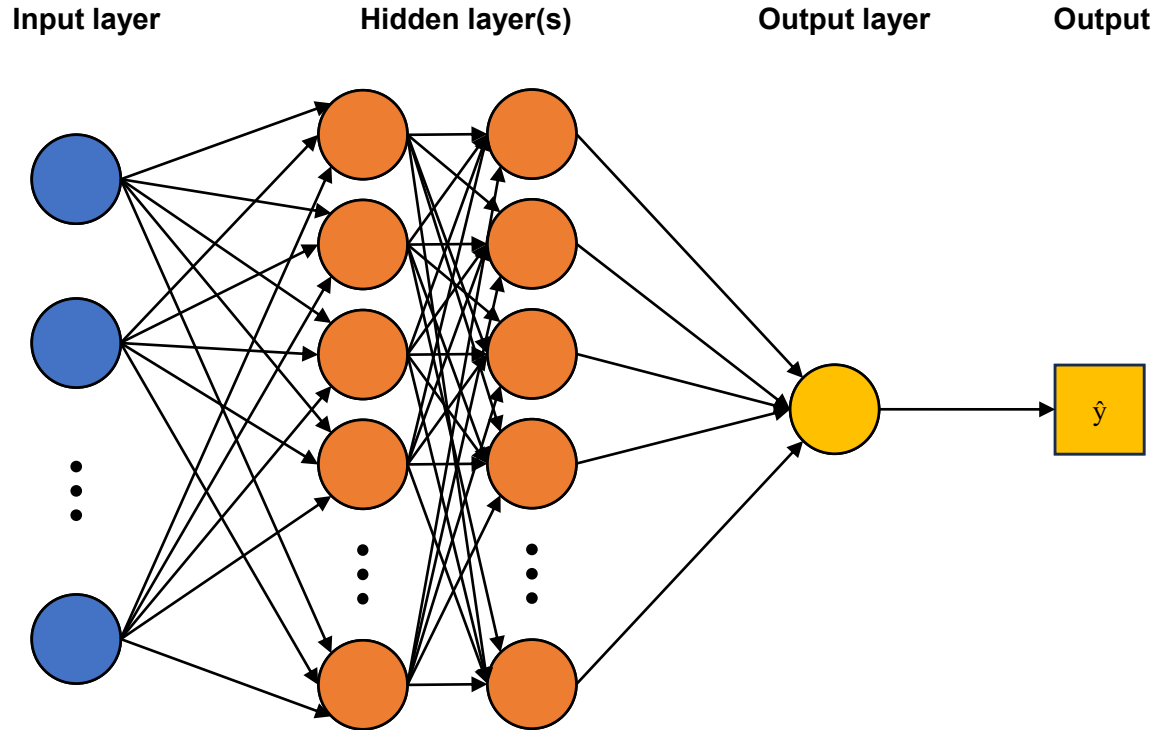
Hidden layer(s)

Output layer

Output



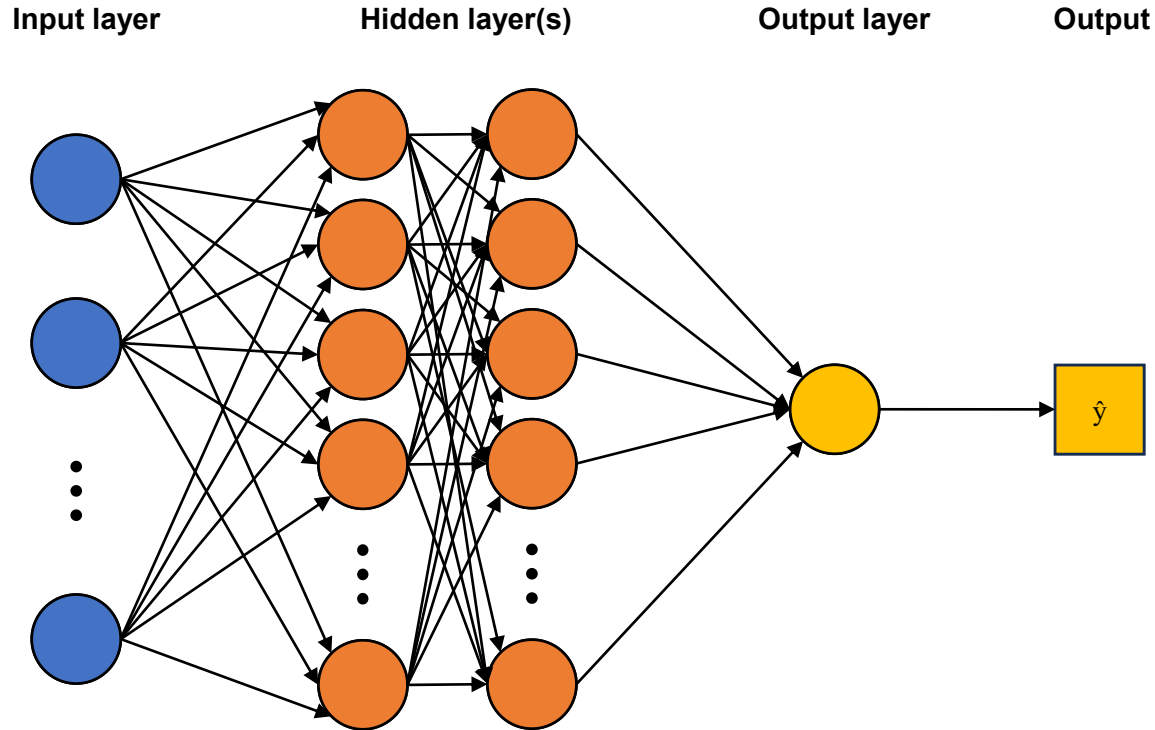
# What is an Artificial Neural Network?



**Universal approximation theorem:**  
An ANN with enough neurons in the hidden layer can approximate any continuous function with arbitrary precision.

**Deep Learning:** Multiple hidden layers.

# What is an Artificial Neural Network?



- Pattern Detector
- High-dimensional feature extractor
- Function approximator
- Computational model

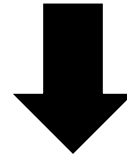
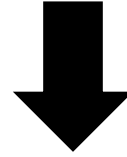
# Chapter 1 - Regression

---

- Regression model for Bradford assay
- Model for growth curve during diauxic growth
- Model for a circadian pattern of enzyme activity
- Human vs. Machine Learning
- The curse of small data sets
- Overfitting, Underfitting, and Holdout Validation

# Holdout validation

---

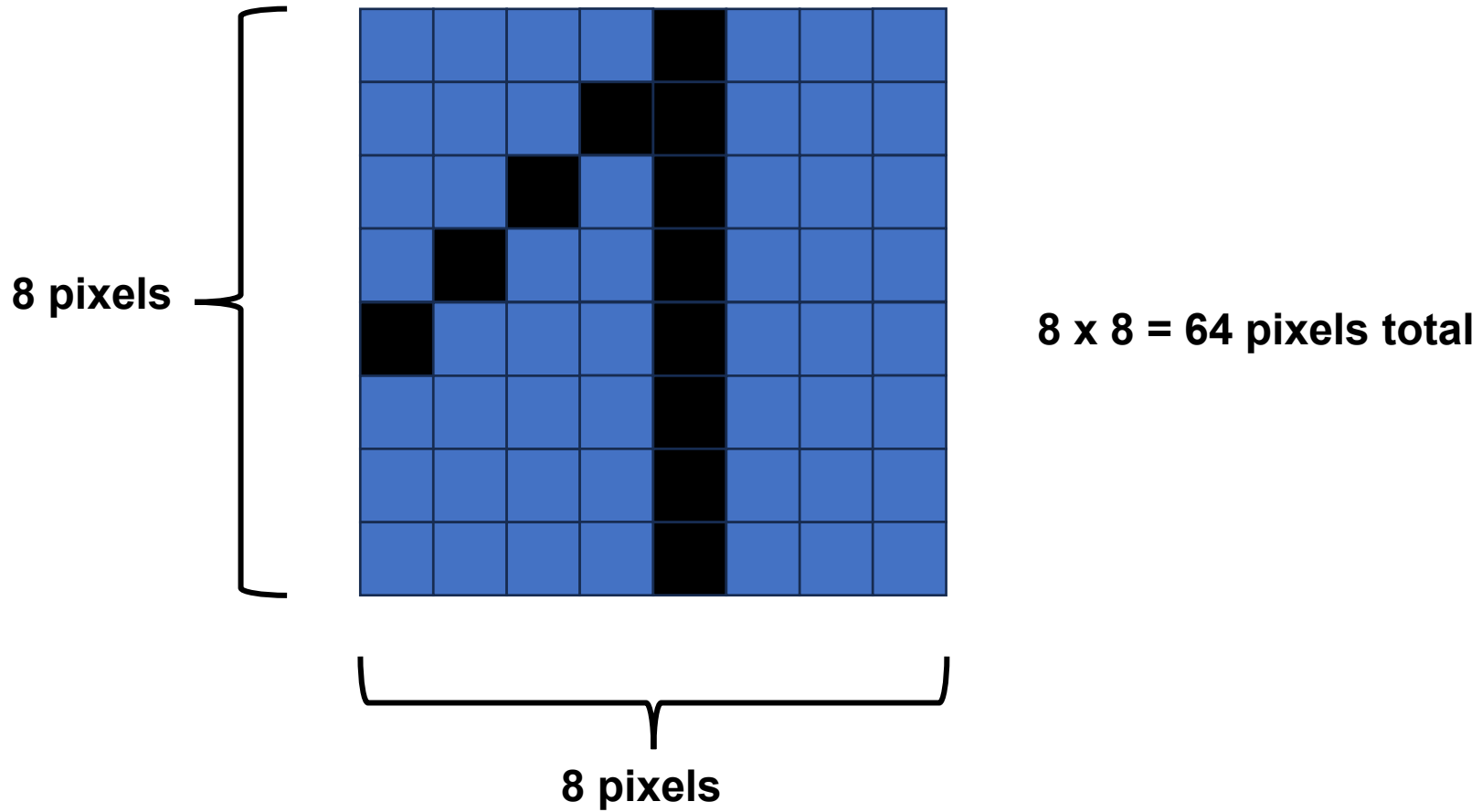


# Chapter 2 - Classification

---

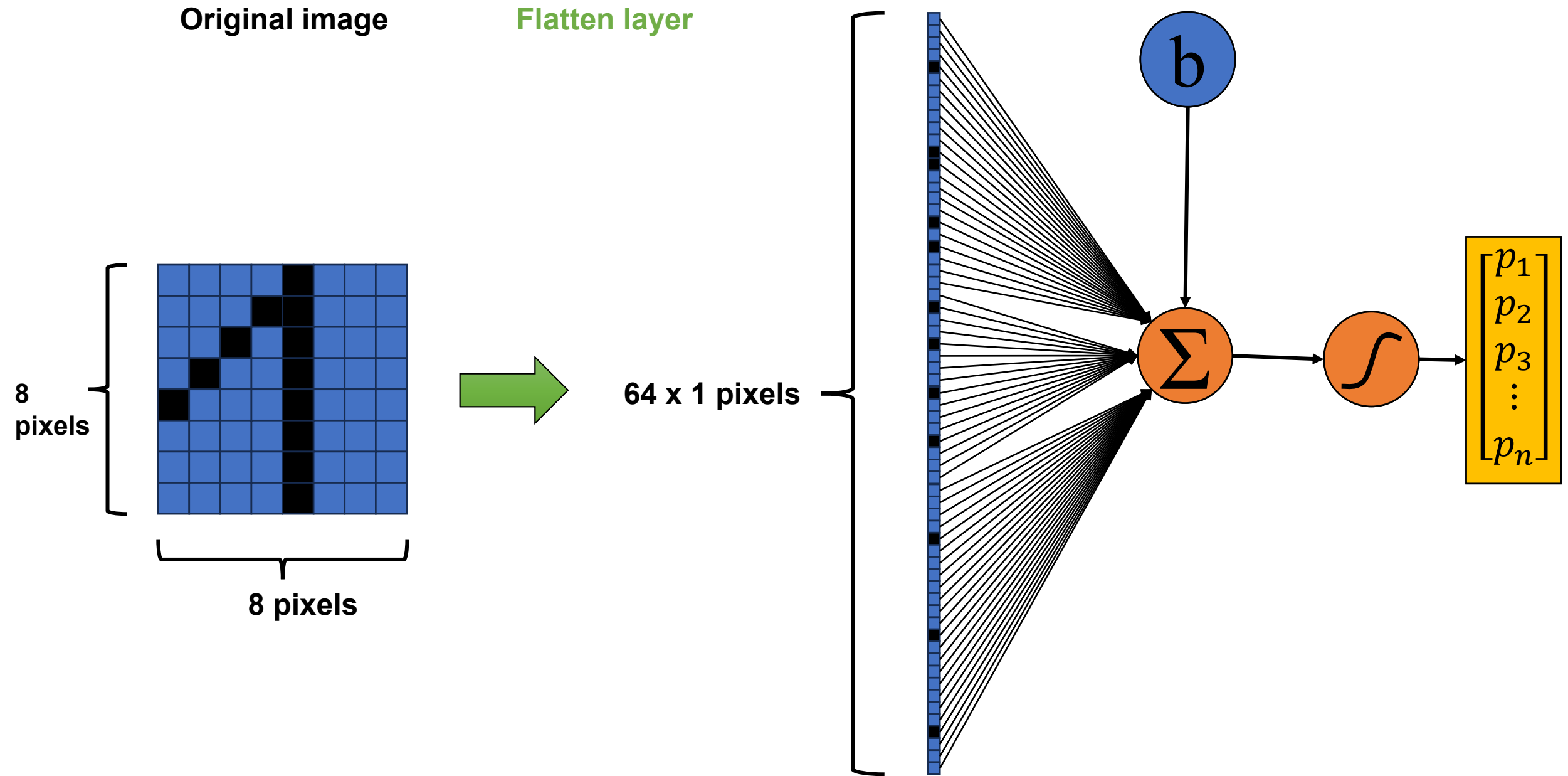
- Binary classification to distinguish microbial cells
- Multi-class classification to distinguish microbial cells
- Simple image classification using the MNIST dataset
- Precision and Recall, Confusion matrix and decision boundaries
- Black box problem

# Multi-class classification network topology for images





# Multi-class classification network topology for images

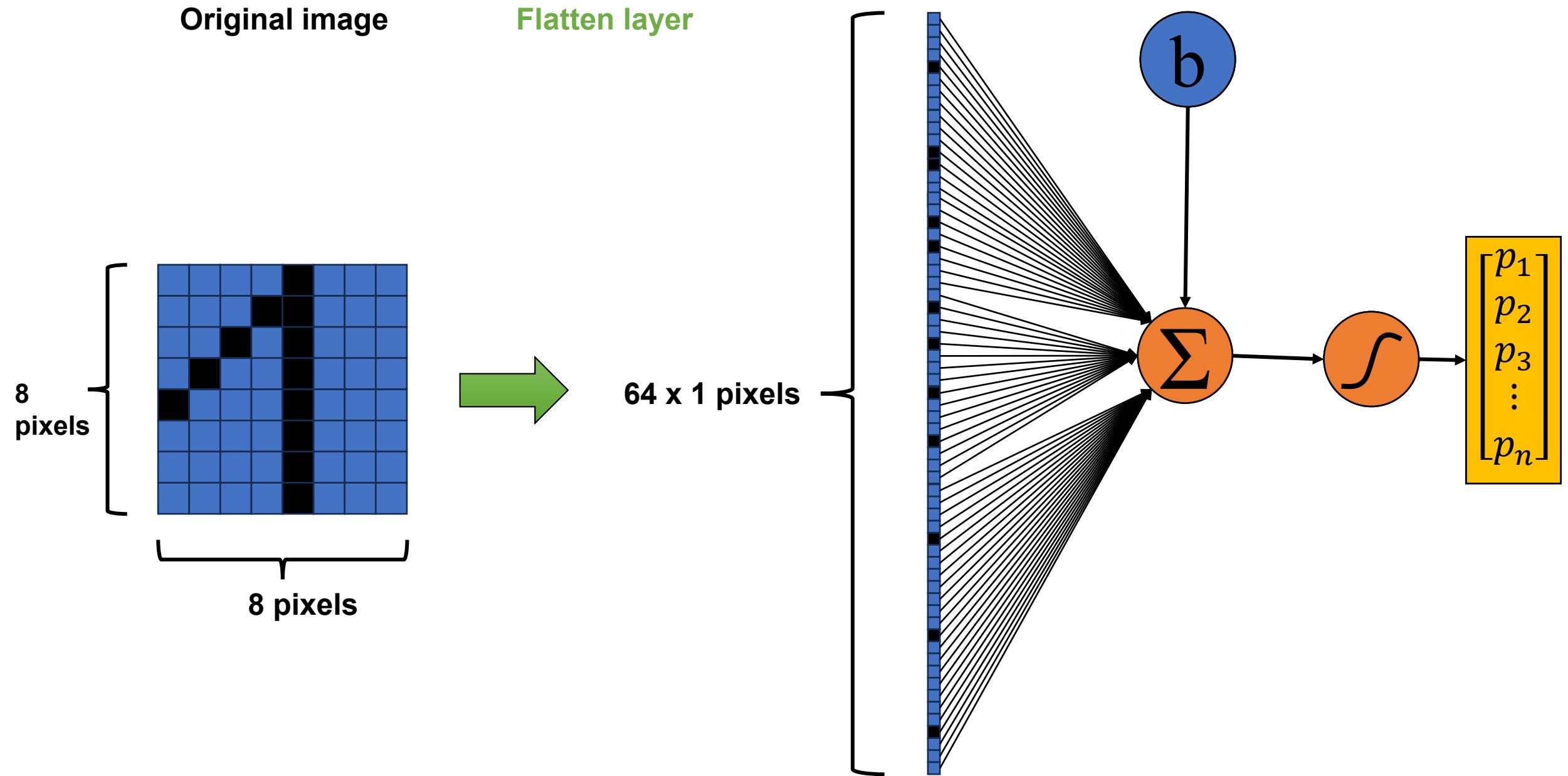


# Chapter 3 – Convolutional Neural Networks

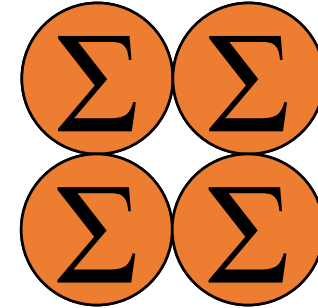
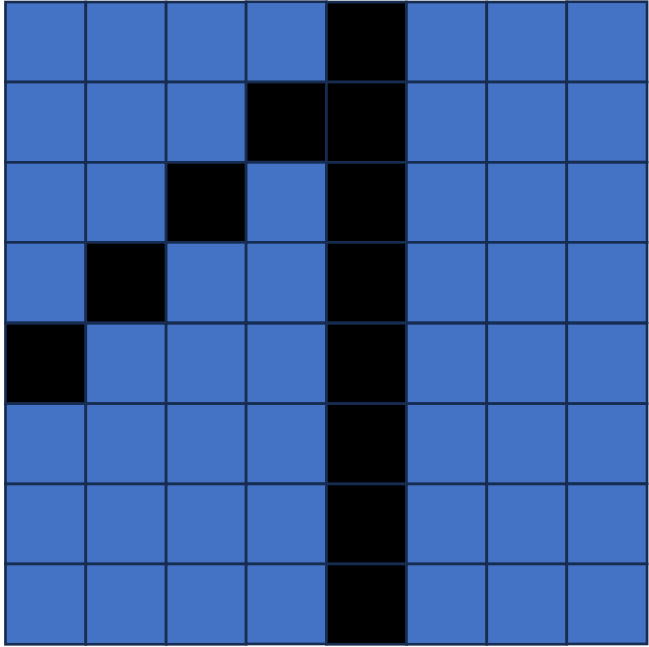
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- CNN for healthy and infected cell classification
- Convolution layer and pooling layer
- Transfer learning

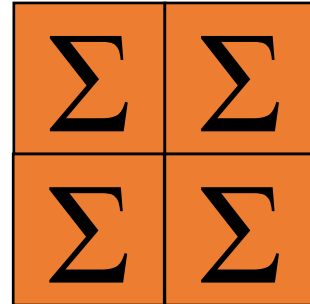
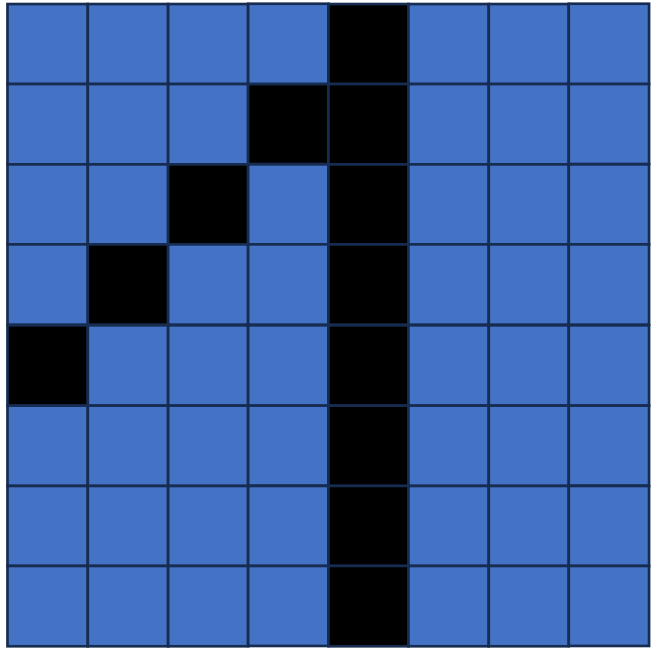
# Multi-class classification network topology for images



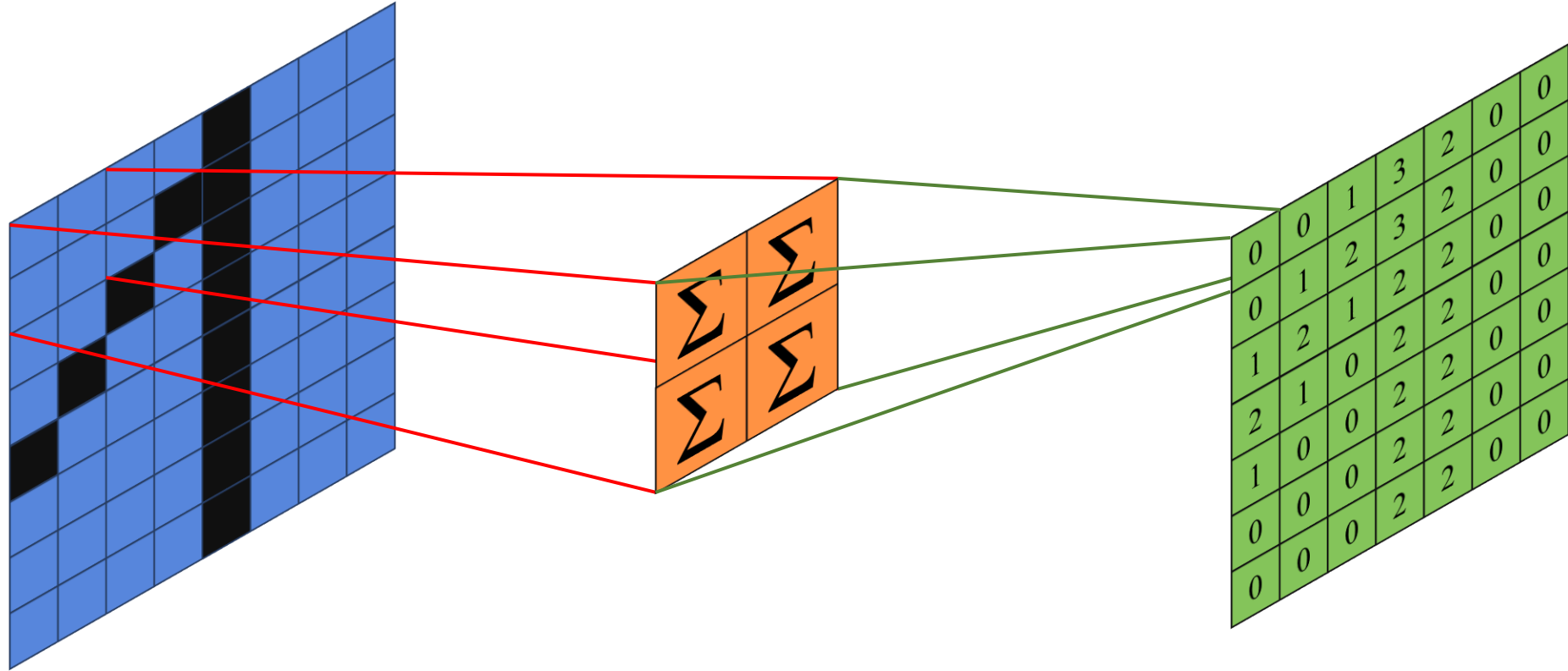
# What is a Convolutional Layer?



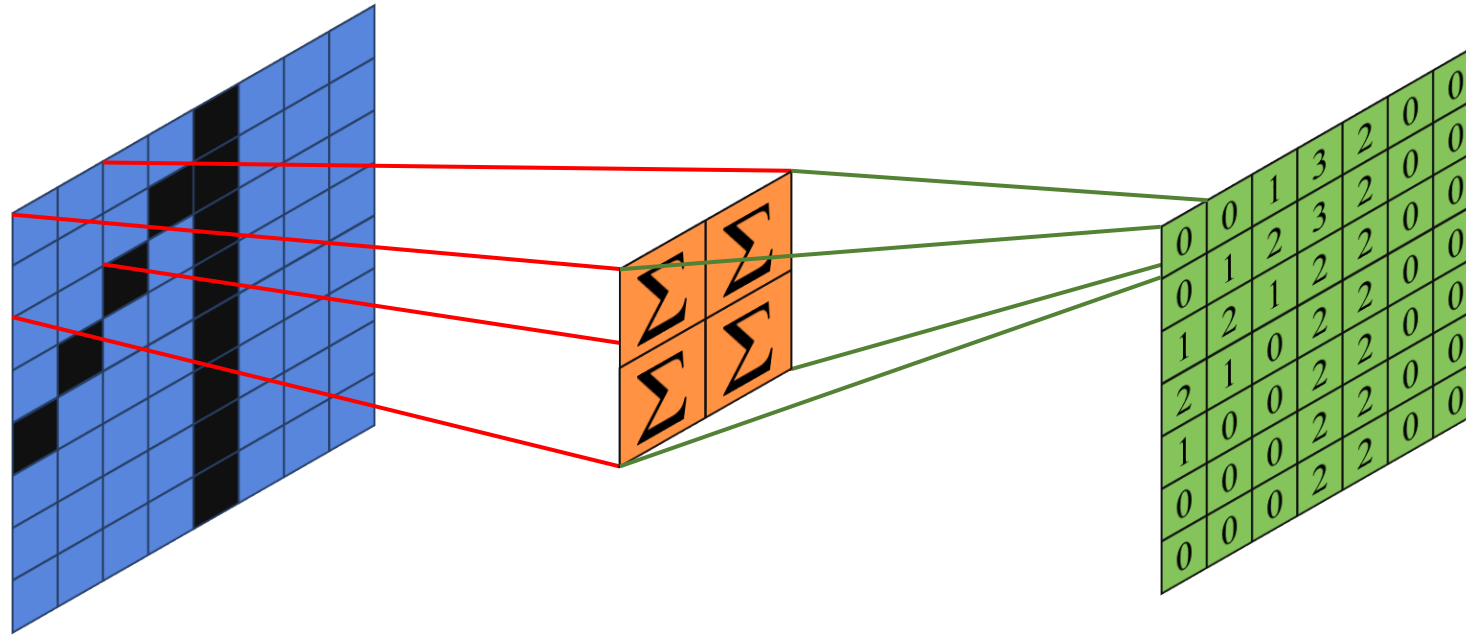
# What is a Convolutional Layer?



# What is a Convolutional Layer?



# What is a Convolutional Layer?

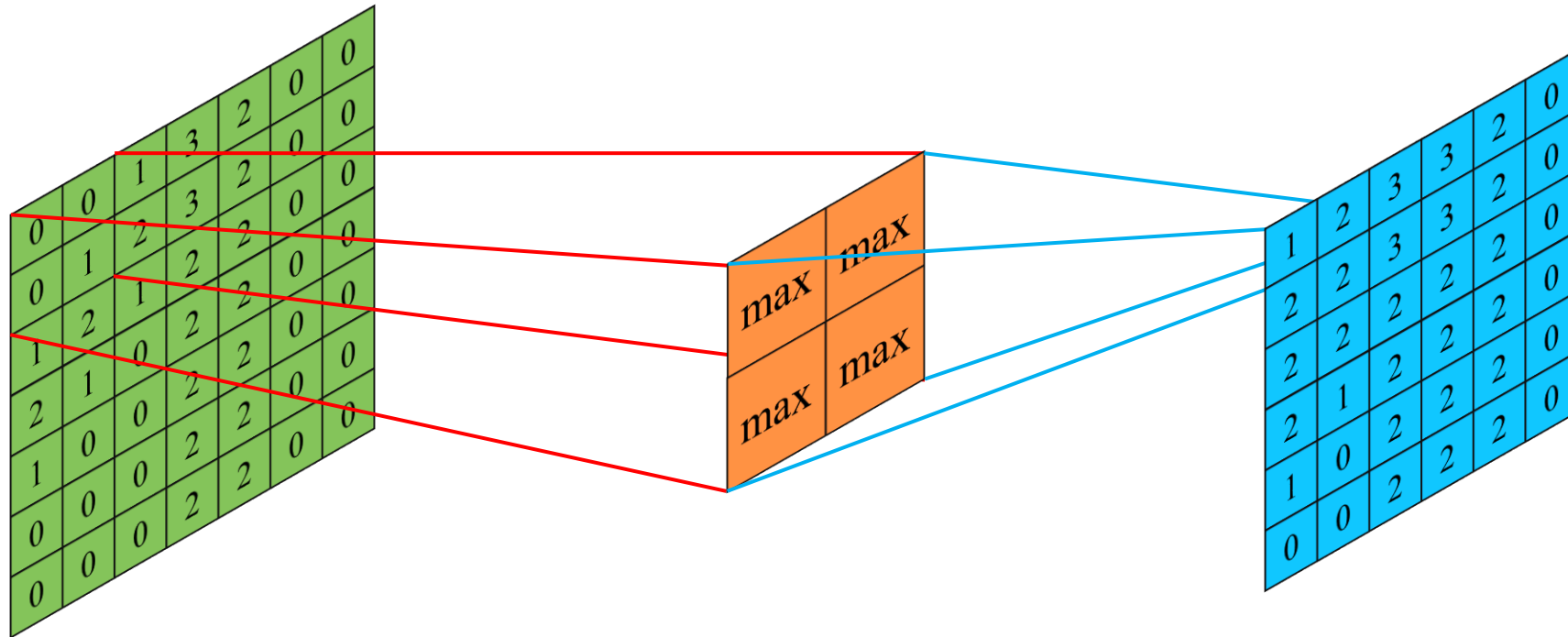


Convolutional layers preserve the intrinsic structure of the data. The output is called a **feature map**. As a result, they perform very well on **structured data**.

Examples include:

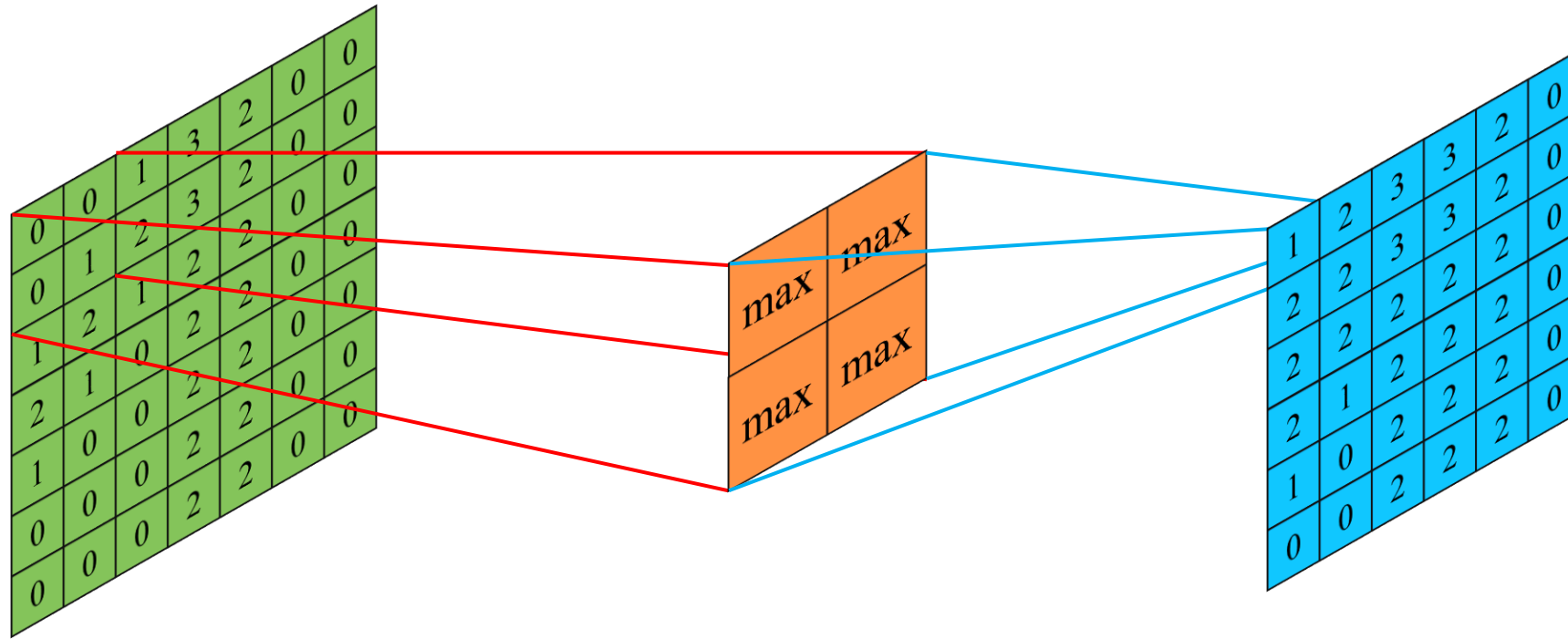
- Computer Vision
- Natural Language Processing
- Time series forecasting and classification
- Molecular structure prediction

# What is a Pooling Layer?





# What is a Pooling Layer?

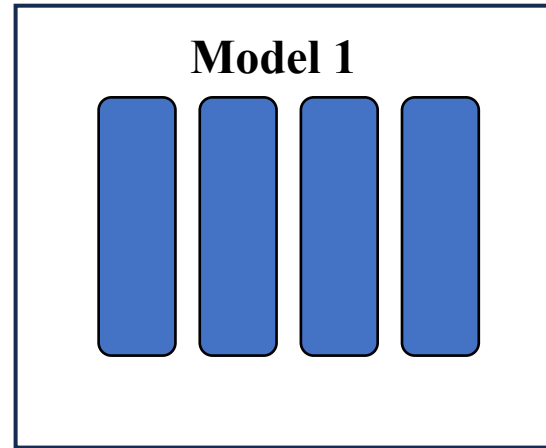


Pooling layers reduce the size of feature maps while preserving **meaningful features**.

# What is Transfer Learning?



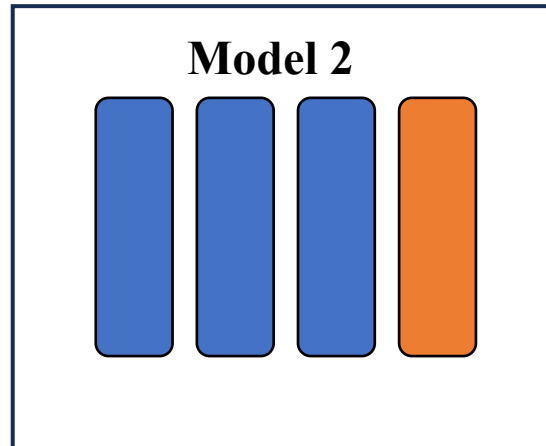
[Alvesgaspar](#), [Cat](#)  
November 2010-1a, [CC](#)  
[BY-SA 3.0](#)



**Prediction 1**

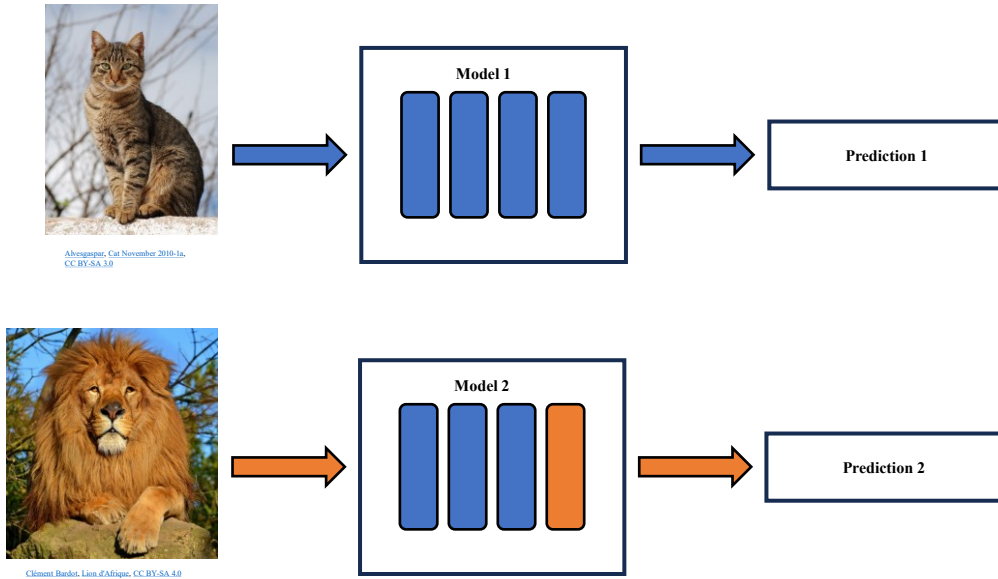


[Clément Bardot](#), [Lion](#)  
[d'Afrique](#), [CC BY-SA 4.0](#)



**Prediction 2**

# What is Transfer Learning?



Transfer learning is a technique that uses a **pre-trained model** for a related task.

Benefits:

- **Faster training**
- **Improved performance** with small amount of data
- **Generalization:** Pre-trained models have typically learned general features

# What next?

