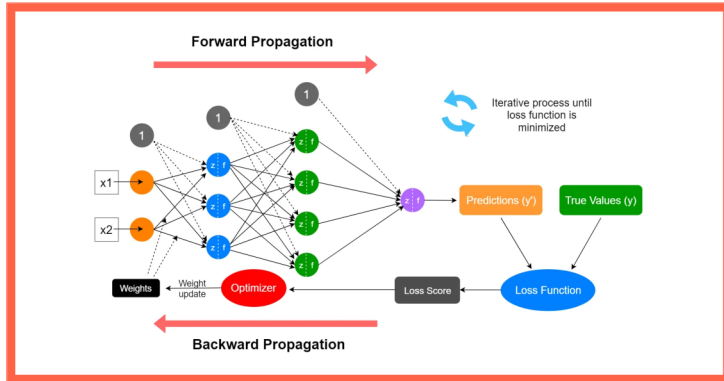


# The mathematics of AI

Or: Learning = forward, loss, backward

Accept Change what you cannot change accept



# What is machine learning?

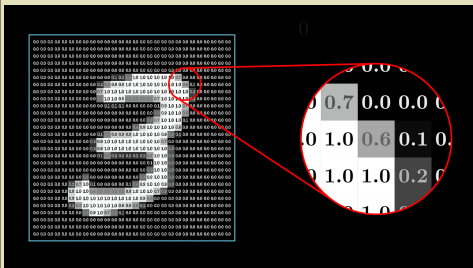
---



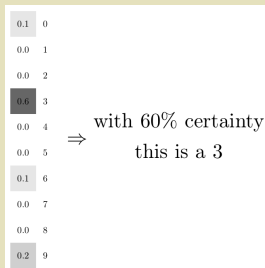
- ▶ **Task** Identify handwritten digits
- ▶ We can see this as a **function** in the following way:
  - ▶ Convert the pictures into grayscale values, e.g.  $28 \times 28$  grid of numbers
  - ▶ Flatten the result into a vector, e.g.  $28 \times 28 \mapsto$  a vector with  $28^2 = 784$  entries
  - ▶ The output is a vector with 10 entries
- ▶ We thus have a function  $\mathbb{R}^{784} \rightarrow \mathbb{R}^{10}$

# What is machine

## Input example



## Output example

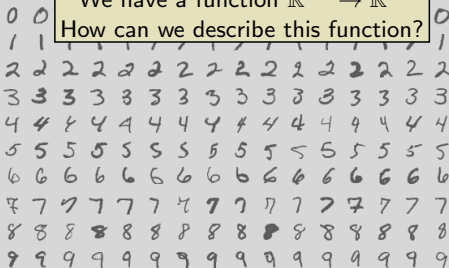


- ▶ Task Identify
- ▶ We can see this
- ▶ Convert the
- ▶ Flatten the
- ▶ The output
- ▶ We thus have a

28 grid of numbers  
with  $28^2 = 784$  entries

Task – rephrased

We have a function  $\mathbb{R}^{784} \rightarrow \mathbb{R}^{10}$   
 How can we describe this function?

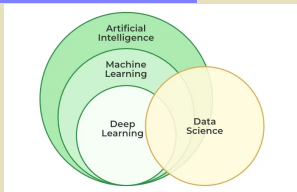


- ▶ **Task** Identify handwritten digits
- ▶ We can see this as a **function** in the following way:
  - ▶ Convert the pictures into grayscale values, e.g.  $28 \times 28$  grid of numbers
  - ▶ Flatten the result into a vector, e.g.  $28 \times 28 \mapsto$  a vector with  $28^2 = 784$  entries
  - ▶ The output is a vector with 10 entries
- ▶ We thus have a function  $\mathbb{R}^{784} \rightarrow \mathbb{R}^{10}$

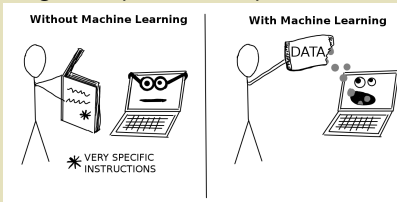
Task – rephrased

We have a function  $\mathbb{R}^{784} \rightarrow \mathbb{R}^{10}$   
How can we describe this function?

Machine/deep learning (today's topic)

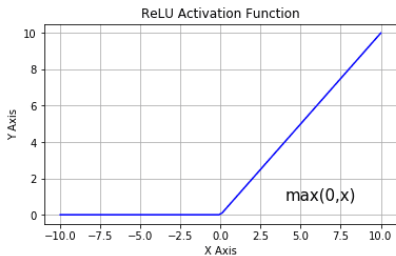
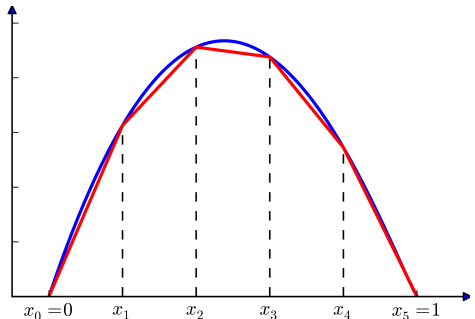


tries to answer questions of this type  
letting a computer detect patterns in data



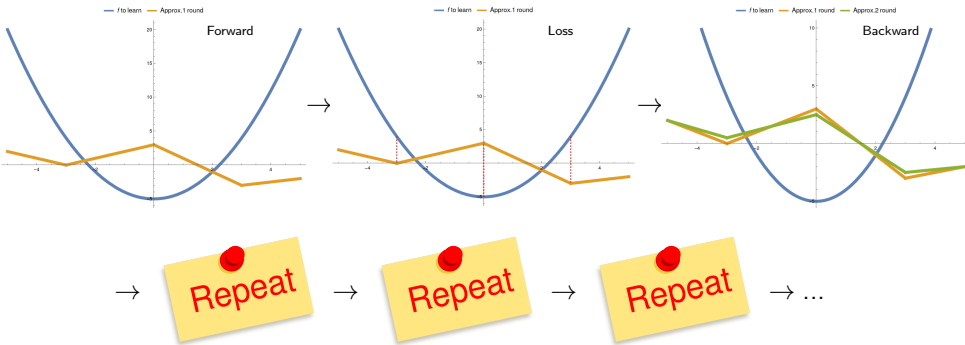
**Crucial** In ML the computer performs tasks without explicit instructions

# What is machine learning?



- ▶ **Idea** Approximate the unknown function  $\mathbb{R}^{784} \rightarrow \mathbb{R}^{10}$
- ▶ **Neural network** = a piecewise linear approximation (matrices + PL maps)
- ▶ The **matrices** = a bunch of numbers (weights) and offsets (biases)
- ▶ The **PL maps** = usually ReLU

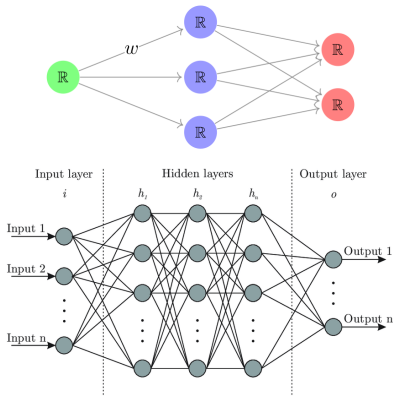
# What is machine learning?



## ▶ Machine learning mantra

- ▶ **Forward** = calculate an approximation (start with random inputs)
- ▶ **Loss** = compare to real data
- ▶ **Backward** = adjust the approximation

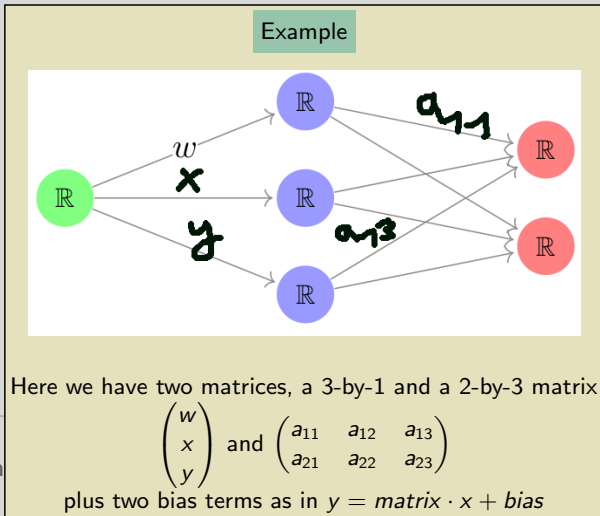
# What is a neural network (nn)?



- ▶ **NN** = a directed graph as above
- ▶ The task of a nn is to **approximate** an unknown function
- ▶ It **consist of** neurons = entries of vectors, and weights = entries of matrices



# What is a neural network (nn)?



► NN = a

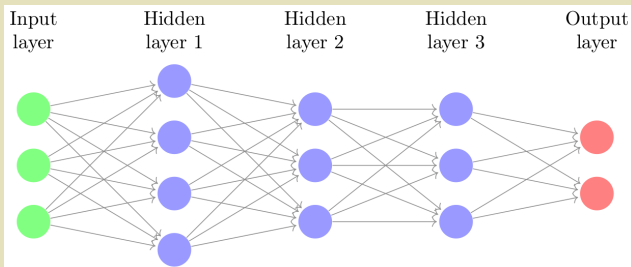
► The task of a nn is to approximate an unknown function

► It consist of neurons = entries of vectors, and weights = entries of matrices

# What is a neural network (nn)?



## Example



Here we have four matrices (plus four biases), whose composition gives a map

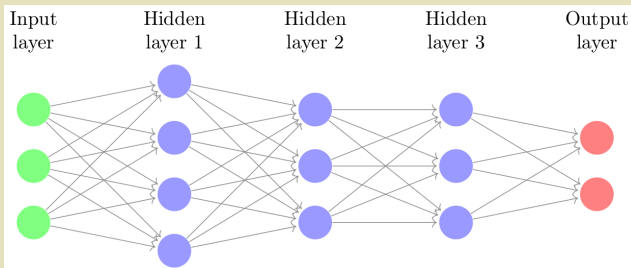
$$\mathbb{R}^3 \rightarrow \mathbb{R}^4 \rightarrow \mathbb{R}^3 \rightarrow \mathbb{R}^3 \rightarrow \mathbb{R}^2$$

- ▶ The task of a nn is to approximate an unknown function
- ▶ It consist of neurons = entries of vectors, and weights = entries of matrices

# What is a neural network (nn)?

Actually...

we need nonlinear maps as well, say ReLU applied componentwise



Here we have four matrices, whose composition gives a map



▶ NN = a

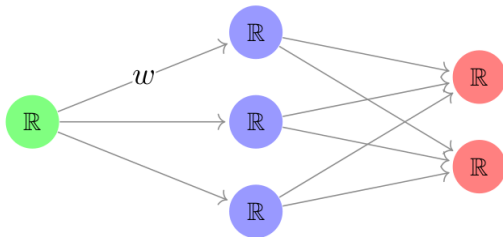
But ignore that for now

▶ The task ReLU doesn't learn anything, its just brings in nonlinearity

▶ It consist of neurons = entries of vectors, and weights = entries of matrices

# What is a neural network (nn)?

---



$$\left( \begin{array}{c|c} a_{11}^1 & b_1^1 \\ a_{12}^1 & b_2^1 \\ a_{13}^1 & b_3^1 \end{array} \right), \quad \left( \begin{array}{ccc|c} a_{11}^2 & a_{12}^2 & a_{13}^2 & b_1^2 \\ a_{21}^2 & a_{22}^2 & a_{23}^2 & b_2^2 \end{array} \right)$$

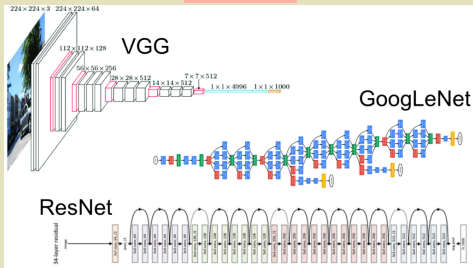
- ▶ The  $a_{ij}^k$  and  $b_i^k$  are the **parameters** of our nn
- ▶  $k$  = number of the **layer**
- ▶ Deep = **many layers** = better approximation

# What is a neural network (nn)?

## The point

Many layers  $\rightarrow$  many parameters  
These are good for approximating real world problems

## Examples



ResNet-152 with 152. layers (used in transformer models such as ChatGPT)  
VGG-19 with 19 layers (used in image classification)  
GoogLeNet with 22 layers (used in face detection)

▶  $k$  = number of the layer

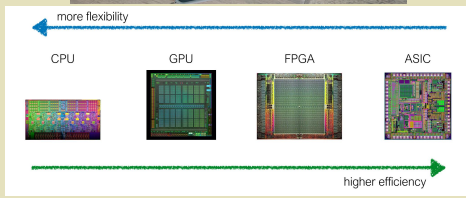
▶ Deep = many layers = better approximation

# What is a neural network (nn)?

Side fact

Gaming has improved AI!?

A GPU can do e.g. matrix multiplications faster than a CPU and lots of nn run on GPUs



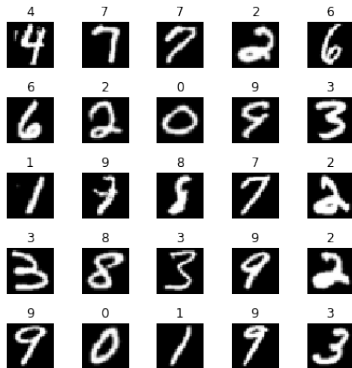
▶ The  $a_{ij}^k$  and  $b_{ij}^k$

▶  $k$  = number of

▶ Deep = many layers = better approximation

## How learning works

---



- 
- ▶ **Supervised learning** Create a dataset with answers, e.g. pictures of handwritten digits plus their label
  - ▶ There are other forms of learning e.g. unsupervised, which I skip
  - ▶ **Split** the data into  $\approx 80\%$  training and  $\approx 20\%$  testing data

Idea to keep in mind

How to train students?

There are lectures, exercises etc., the training data

There is a final exam, the testing data

Upon performance, we let them out into the wild

## Human Learning



Talent

Learning  
Materials

Skills  
learned

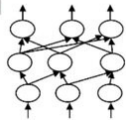


## Machine Learning

Models

Data

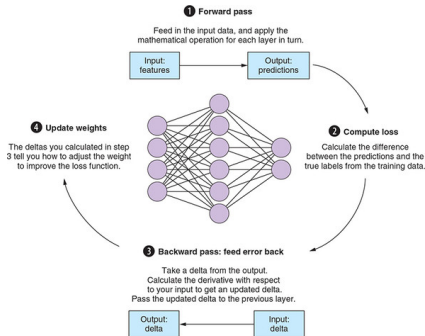
Use in  
real world



Split the data into  $\approx 80\%$  training and  $\approx 20\%$  testing data



# How learning works



- ▶ **Forward** Run the  $nn =$  function on the training data
- ▶ **Loss** Calculate the difference “results - answers” ( $\Rightarrow$  loss function)
- ▶ **Backward** Change the parameters trying to minimize the loss function
- ▶ **Repeat**

# How learning works

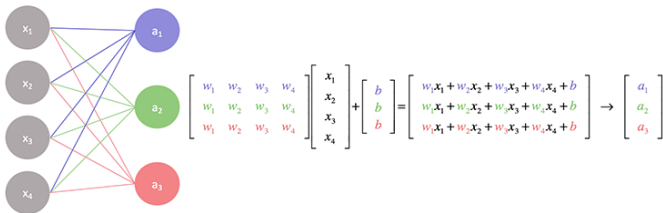
## 1 Forward pass

Feed in the input data, and apply the mathematical operation for each layer in turn.

Input: → Output:

### Forward

Boils down to a bunch of matrix multiplications



▶ Forward

▶ Loss

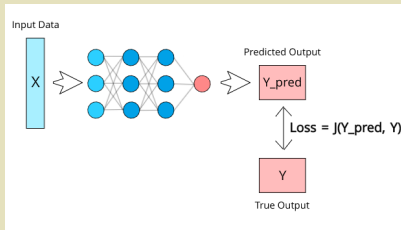
followed by the nonlinear activation e.g. ReLU

▶ Backward Change the parameters trying to minimize the loss function

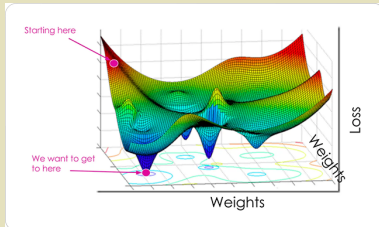
▶ Repeat

# Loss

The difference between real values and predictions



Task Minimize loss function



▶ Forward Run

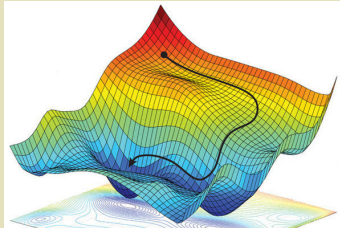
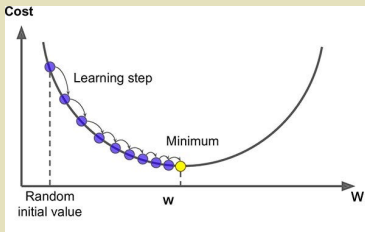
▶ Loss Calculation

▶ Backward Pass

▶ Repeat

Backward

This is running gradient descent on the loss function



Slogan Adjust parameters following the steepest descent

▶ Forward

▶ Loss Ca

▶ Backward

▶ Repeat

# How learning works

1 Forward pass

Feed in the input data, and apply the

And what makes it even better:

You can try it yourself

My favorite tool is PyTorch but there are also other methods



Let us see how!

▶ Forward

▶ Loss C

▶ Backward Change the parameters trying to minimize the loss function

▶ Repeat

### What is machine learning?

- Task** Identify handwritten digits
- We can see this as a **function** in the following way:
  - Convert the pictures into grayscale values, e.g.  $28 \times 28$  grid of numbers
  - Flatten the result into a vector, e.g.  $28 \times 28 = 28 \rightarrow$  a vector with  $28^2 = 784$  entries
  - The output is a vector with 10 entries
- We thus have a function  $\mathbb{R}^{784} \rightarrow \mathbb{R}^{10}$

The mathematics of ML | ML Learning | Research, Alex, Bealbert | April 2024 | 1 / 5

### What is machine learning?

- Task** Identify handwritten digits
- We can see this as a **function** in the following way:
  - Convert the pictures into grayscale values, e.g.  $28 \times 28$  grid of numbers
  - Flatten the result into a vector, e.g.  $28 \times 28 = 28 \rightarrow$  a vector with  $28^2 = 784$  entries
  - The output is a vector with 10 entries
- We thus have a function  $\mathbb{R}^{784} \rightarrow \mathbb{R}^{10}$

The mathematics of ML | ML Learning | Research, Alex, Bealbert | April 2024 | 1 / 5

### What is machine learning?

- Task** Approximate the unknown function  $\mathbb{R}^2 \rightarrow \mathbb{R}^2$
- Neural network** = a piecewise linear approximation (matrices + PL maps)
- The **matrices** = a bunch of numbers (weights) and offsets (biases)
- The **PL maps** = usually ReLU

The mathematics of ML | ML Learning | Research, Alex, Bealbert | April 2024 | 1 / 5

### What is machine learning?

- Machine learning matrix**
  - Forward** = calculate an approximation (start with random inputs)
  - Loss** = compare to real data
  - Backward** = adjust the approximation

The mathematics of ML | ML Learning | Research, Alex, Bealbert | April 2024 | 1 / 5

### What is a neural network (nn)?

How we have two matrices, a 3-by-1 and a 2-by-3 matrix

$$\begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix} \text{ and } \begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \end{pmatrix}$$

The task is to find the approximation to the unknown function

The **weights** = entries of vectors, and weights = entries of matrices

The mathematics of ML | ML Learning | Research, Alex, Bealbert | April 2024 | 1 / 5

### How learning works

- Supervised learning** = Create a dataset with answers, e.g. pictures of handwritten digits plus their label
- There are other forms of learning e.g. unsupervised, which I skip
- Split** the data into ~80% training and ~20% testing data

The mathematics of ML | ML Learning | Research, Alex, Bealbert | April 2024 | 1 / 5

### How learning works

**Forward**

Balls down to a bunch of matrix multiplications

$$\begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix} \begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \end{pmatrix} = \begin{pmatrix} 1 \cdot 1 + 2 \cdot 4 + 3 \cdot 1 & 1 \cdot 2 + 2 \cdot 5 + 3 \cdot 1 & 1 \cdot 3 + 2 \cdot 6 + 3 \cdot 1 \\ 1 \cdot 2 + 2 \cdot 4 + 3 \cdot 1 & 1 \cdot 3 + 2 \cdot 6 + 3 \cdot 1 & 1 \cdot 4 + 2 \cdot 7 + 3 \cdot 1 \\ 1 \cdot 3 + 2 \cdot 6 + 3 \cdot 1 & 1 \cdot 4 + 2 \cdot 7 + 3 \cdot 1 & 1 \cdot 5 + 2 \cdot 8 + 3 \cdot 1 \end{pmatrix} = \begin{pmatrix} 14 & 15 & 16 \\ 15 & 16 & 17 \\ 16 & 17 & 18 \end{pmatrix}$$

followed by the nonlinear activation e.g. ReLU

- Forward** = calculate an approximation (start with random inputs)
- Loss** = compare to real data
- Backward** = Change the parameters trying to minimize the loss function
- Repeat**

The mathematics of ML | ML Learning | Research, Alex, Bealbert | April 2024 | 1 / 5

### How learning works

**Loss**

The difference between real values and predictions

**Task** Minimize loss function

- Forward** = calculate an approximation (start with random inputs)
- Loss** = Compare to real data
- Backward** = Change the parameters trying to minimize the loss function
- Repeat**

The mathematics of ML | ML Learning | Research, Alex, Bealbert | April 2024 | 1 / 5

### How learning works

**Backward**

This is running gradient descent on the loss function

- Forward** = calculate an approximation (start with random inputs)
- Loss** = Compare to real data
- Backward** = Change the parameters trying to minimize the loss function
- Repeat**
- Steps** = Adjust parameters following the steepest descent

The mathematics of ML | ML Learning | Research, Alex, Bealbert | April 2024 | 1 / 5

There is still much to do...

### What is machine learning?

- Task** Identify handwritten digits
- We can see this as a **function** in the following way:
  - Convert the pictures into grayscale values, e.g.  $28 \times 28$  grid of numbers
  - Flatten the result into a vector, e.g.  $28 \times 28 = 28 \rightarrow$  a vector with  $28^2 = 784$  entries
  - The output is a vector with 10 entries
- We thus have a function  $g^{ML} = \mathbb{R}^{784} \rightarrow \mathbb{R}^{10}$

The mathematics of ML | ML Learning | Research, Alex, Bealbert | April 2024 | 1 / 5

### What is machine learning?

- Task** Identify handwritten digits
- We can see this as a **function** in the following way:
  - Convert the pictures into grayscale values, e.g.  $28 \times 28$  grid of numbers
  - Flatten the result into a vector, e.g.  $28 \times 28 = 28 \rightarrow$  a vector with  $28^2 = 784$  entries
  - The output is a vector with 10 entries
- We thus have a function  $g^{ML} = \mathbb{R}^{784} \rightarrow \mathbb{R}^{10}$

The mathematics of ML | ML Learning | Research, Alex, Bealbert | April 2024 | 1 / 5

### What is machine learning?

- Task** Approximate the unknown function  $\mathbb{R}^{2M} \rightarrow \mathbb{R}^{2M}$
- Neural network** = a piecewise linear approximation (matrices + PL maps)
- The **matrices** = a bunch of numbers (weights) and offsets (biases)
- The **PL maps** = usually ReLU

The mathematics of ML | ML Learning | Research, Alex, Bealbert | April 2024 | 1 / 5

### What is machine learning?

- Machine learning matrix**
  - Forward** = calculate an approximation (start with random inputs)
  - Loss** = compare to real data
  - Backward** = adjust the approximation
- Repeat** (multiple times)

The mathematics of ML | ML Learning | Research, Alex, Bealbert | April 2024 | 1 / 5

### What is a neural network (nn)?

How we have two matrices, a 3-by-1 and a 2-by-3 matrix

$$\begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix} \text{ and } \begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \end{pmatrix}$$

**Task** plus two lists terms as  $2 \times 3 = \text{matrix} \times \text{list}$

The **task** for a NN is to approximate an unknown function

The **input** of neurons = entries of vectors, and weights = entries of matrices

The mathematics of ML | ML Learning | Research, Alex, Bealbert | April 2024 | 1 / 5

### How learning works

- Supervised learning** Create a dataset with answers, e.g. pictures of handwritten digits plus their label
- There are other forms of learning e.g. unsupervised, which I skip
- Split** the data into ~80% training and ~20% testing data

The mathematics of ML | ML Learning | Research, Alex, Bealbert | April 2024 | 1 / 5

### How learning works

**Forward**

Balls down to a bunch of matrix multiplications

$$\begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix} \times \begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \end{pmatrix} = \begin{pmatrix} 1 \cdot 1 + 2 \cdot 4 + 3 \cdot 7 & 1 \cdot 2 + 2 \cdot 5 + 3 \cdot 8 & 1 \cdot 3 + 2 \cdot 6 + 3 \cdot 9 \\ 2 \cdot 1 + 3 \cdot 4 + 4 \cdot 7 & 2 \cdot 2 + 3 \cdot 5 + 4 \cdot 8 & 2 \cdot 3 + 3 \cdot 6 + 4 \cdot 9 \\ 3 \cdot 1 + 4 \cdot 4 + 5 \cdot 7 & 3 \cdot 2 + 4 \cdot 5 + 5 \cdot 8 & 3 \cdot 3 + 4 \cdot 6 + 5 \cdot 9 \end{pmatrix}$$

**Task** followed by the nonlinear activation e.g. ReLU

- Forward** Calculate
- Loss** Calculate
- Backward** Change the parameters trying to minimize the loss function
- Repeat** (multiple times)

The mathematics of ML | ML Learning | Research, Alex, Bealbert | April 2024 | 1 / 5

### How learning works

**Loss**

The difference between real values and predictions

**Task** Minimize loss function

- Forward** Calculate
- Loss** Calculate
- Backward** Change the parameters trying to minimize the loss function
- Repeat** (multiple times)

The mathematics of ML | ML Learning | Research, Alex, Bealbert | April 2024 | 1 / 5

### How learning works

**Backward**

This is running gradient descent on the loss function

- Forward** Calculate
- Loss** Calculate
- Backward** Change the parameters trying to minimize the loss function
- Repeat** (multiple times)
- Steps** Adjust parameters following the steepest descent

The mathematics of ML | ML Learning | Research, Alex, Bealbert | April 2024 | 1 / 5

Thanks for your attention!