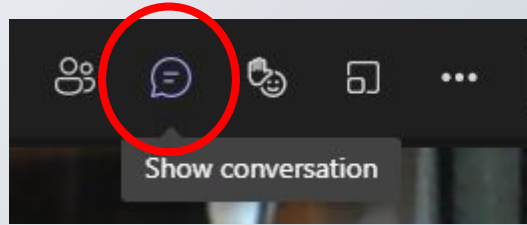


# While we wait to get started ...

Open the chat



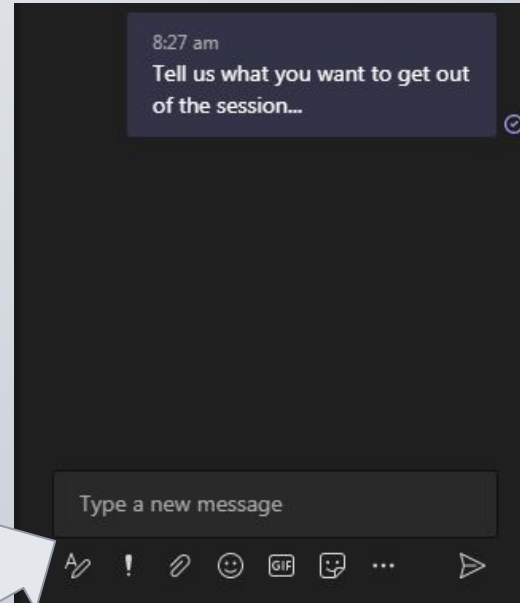
Your mic is on mute  
... and camera disabled

Tell us what you want to  
get out of the session.

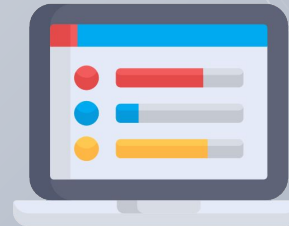
## NOTE:

your name will appear  
with your comment.

The chat won't be part  
of the recorded version.



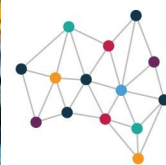
Take our poll ... you will  
find it in the chat...



DIGITAL  
TECHNOLOGIES  
HUB

# Discovering **Artificial Intelligence (AI)**

Investigating training an AI model



**DIGITAL  
TECHNOLOGIES  
HUB**

## Acknowledgement of Country



ESA acknowledges the Eastern Kulin Nation, Traditional Custodians of the land on which our head office stands, and pays our respects to Elders past and present.

We recognise the Traditional Custodians of Country across Australia and their continuing connection and contribution to lands, waters, communities and learning

# By the end of this session

Explore the **binary data** that is inputted into an AI system.

Try/observe building a Python program to:

- perform a **classification** with AI.
- perform a **regression** (line of best fit) with an AI.

Discover more about how an artificial neural network works.

# Achievement standards:

## Achievement Standard

By the end of Year 6, students explain digital system components (hardware and software) and how digital systems are connected. They explain how digital systems use data to represent a variety of data types.

Students define problems in terms of requirements and design solutions to address the problems. They incorporate repetition and user interface design to implement their digital solutions, including testing. They explain how information systems meet needs and consider sustainability in the creation and communication of ideas. They collaborate on digital projects using protocols.

## Achievement Standard

By the end of Year 8, students distinguish types of networks and defined purposes. They explain how image and audio data can be represented in digital systems.

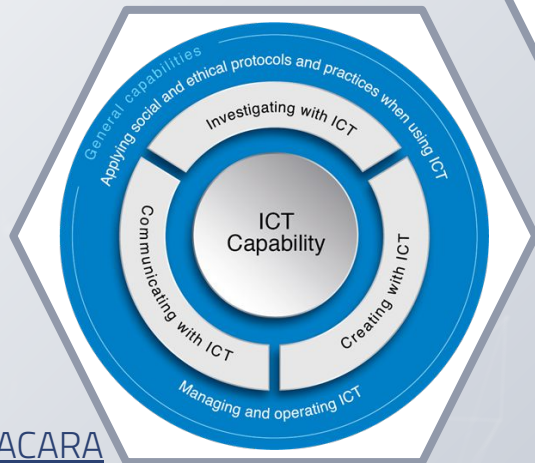
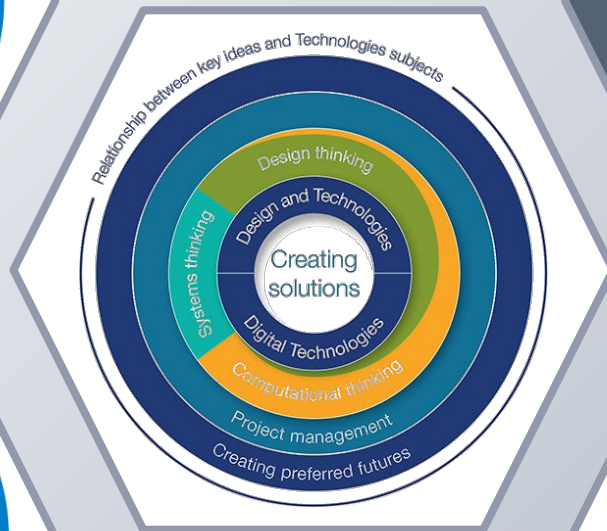
Students plan and manage digital projects. They define and decompose information. They define and decompose functional requirements and constraints. They design experiences and algorithms incorporating iterations, and test, modify and implement. They evaluate information systems against criteria of meeting needs, innovation and sustainability. They evaluate data from a range of sources. They use appropriate protocols and collaborating online.

## Achievement Standard

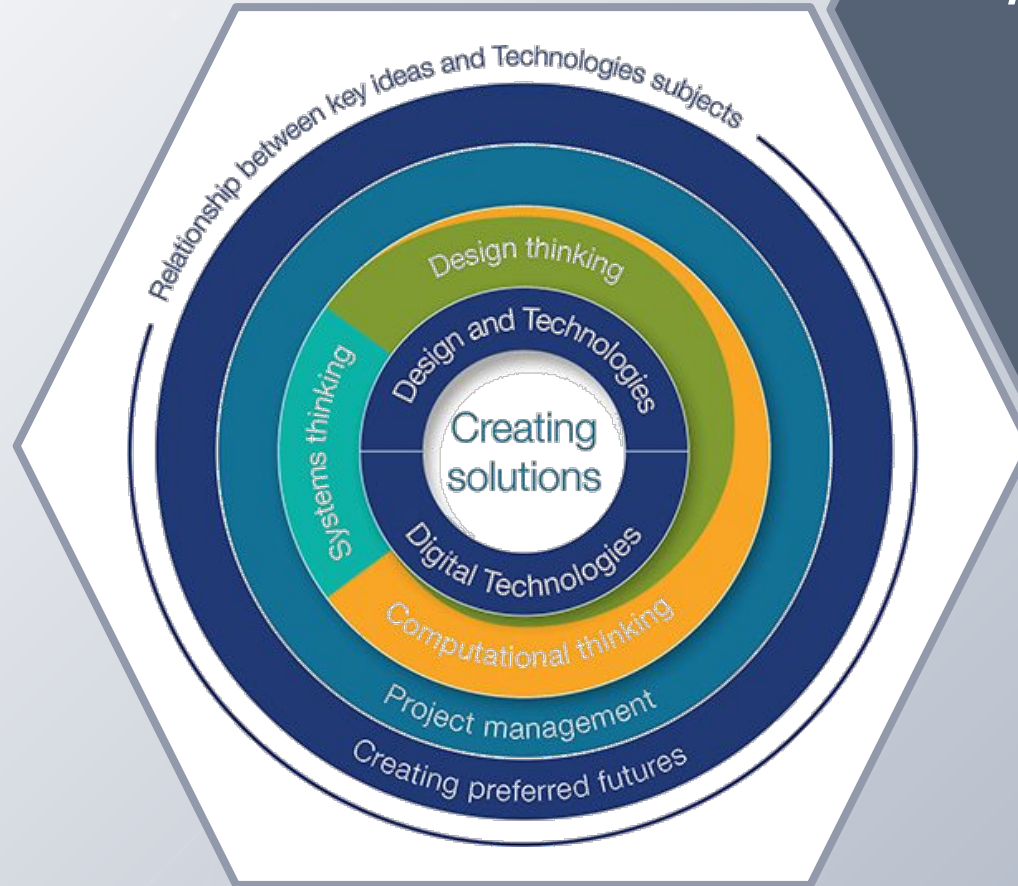
By the end of Year 10, students explain the control and management of networked digital systems and the security implications of the interaction between hardware, software and users. They explain simple data compression, and why content data are separated from presentation.

Students plan and manage digital projects using an iterative approach. They define and decompose complex problems in terms of functional and non-functional requirements. Students design and evaluate user experiences and algorithms. They design and implement modular programs, including an object-oriented program, using algorithms and data structures involving modular functions that reflect the relationships of real-world data and data entities. They take account of privacy and security requirements when selecting and validating data. Students test and predict results and implement digital solutions. They evaluate information systems and their solutions in terms of risk, sustainability and potential for innovation and enterprise. They share and collaborate online, establishing protocols for the use, transmission and maintenance of data and projects.

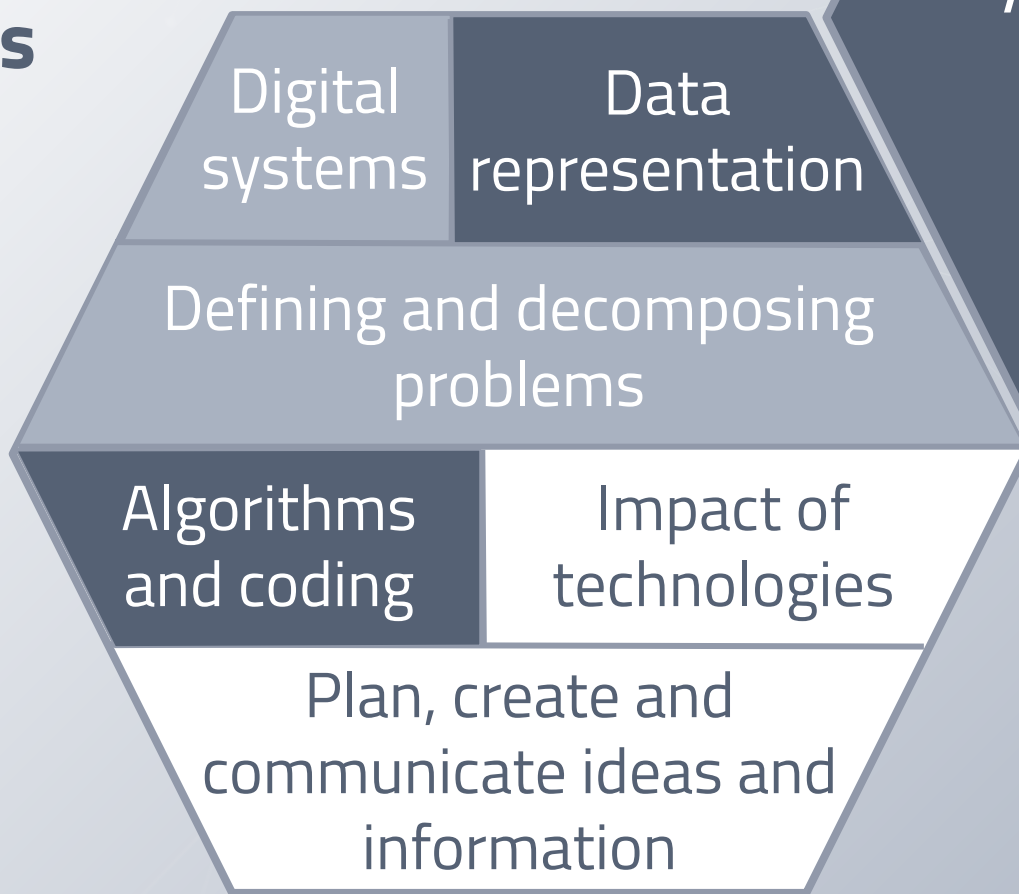
# AI topics



# AI topics



**Foci for this  
deep dive:**



AI topics

# Systems Thinking



## Design Thinking

## Computational Thinking

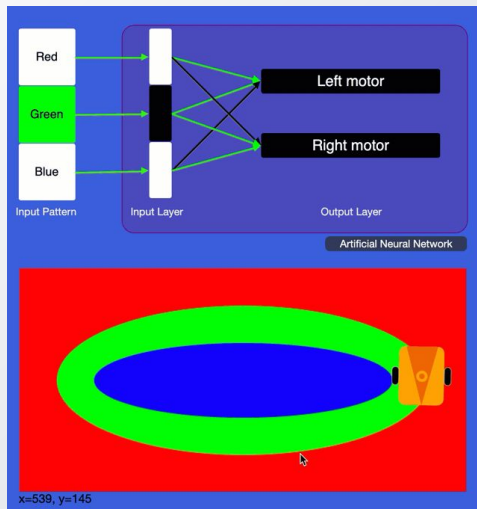




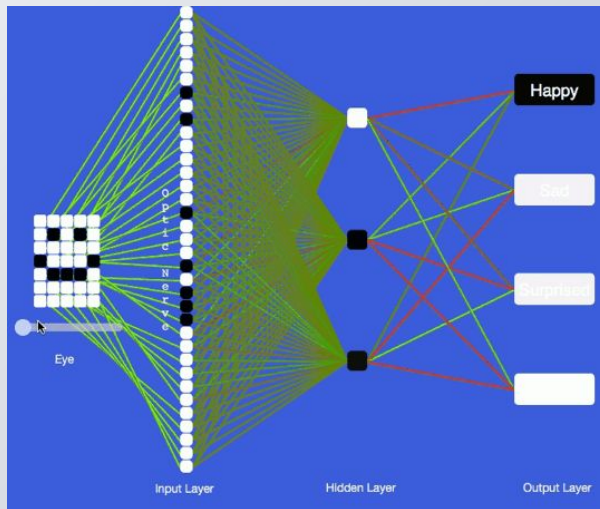
# Investigating training an **Artificial Intelligence**

A closer look  
"under the hood"

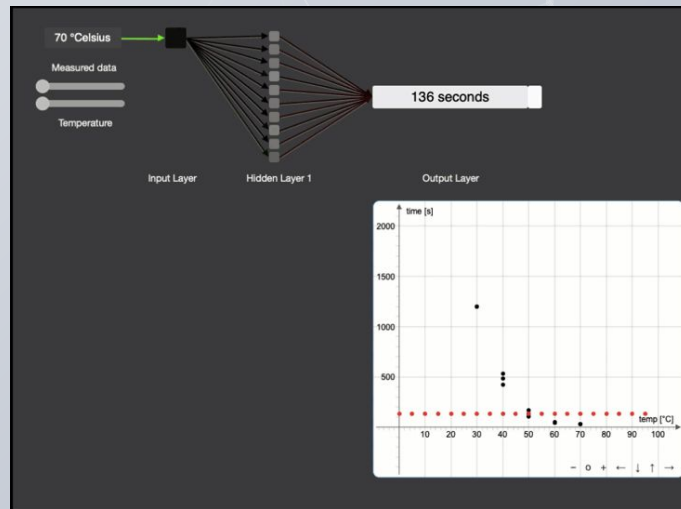
# What's so special about AI?



Driving a Robot



Emoji recognition



Prediction

AI is versatile!

In each scenario, the same underlying AI is at work.

**RESOURCE:** AI at [My Computer Brain](#)

# Can students code a neural network from scratch?

It's not recommended as a general classroom activity for DigiTech in 7-10.

Code for a "simple" neural network to work with tiny binary arrays.

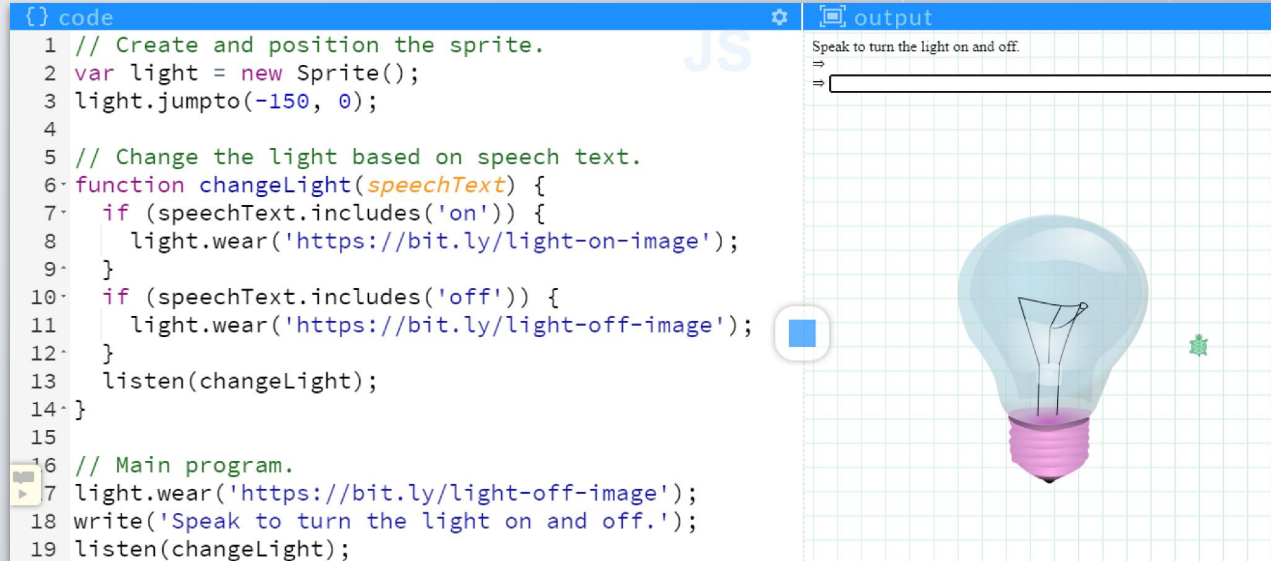
(screenshot from [Simple Neural Networks in Python](#), Aidan Wilson)

```
1 import numpy as np # helps with the math
2 import matplotlib.pyplot as plt # to plot error during training
3
4 # input data
5 inputs = np.array([[0, 1, 0],
6                    [0, 1, 1],
7                    [0, 0, 0],
8                    [1, 0, 0],
9                    [1, 1, 1],
10                   [1, 0, 1]])
11
12 # output data
13 outputs = np.array([[0], [0], [0], [1], [1], [1]])
14
15 # create NeuralNetwork class
16 class NeuralNetwork:
17
18     # initialize variables in class
19     def __init__(self, inputs, outputs):
20         self.inputs = inputs
21         self.outputs = outputs
22
23         # initialize weights as .50 for simplicity
24         self.weights = np.array([[.50], [.50], [.50]])
25         self.error_history = []
26         self.epoch_list = []
27
28     #activation function ==> S(x) = 1/(1+e^(-x))
29     def sigmoid(self, x, deriv=False):
30         if deriv == True:
31             return x * (1 - x)
32         return 1 / (1 + np.exp(-x))
```

# Can students write code to investigate and use a neural network?

YES!


Utilise an existing speech recognition system in a general purpose program (JavaScript).



The screenshot shows a code editor with two panels. The left panel, titled 'code', contains JavaScript code for a light control system. The right panel, titled 'output', shows the text 'Speak to turn the light on and off.' and a visual representation of a light bulb. The light bulb is currently off, indicated by a blue square next to it. The code defines a function 'changeLight' that takes 'speechText' as an argument. It checks if 'speechText' includes 'on' or 'off' and updates the light's image accordingly. The main program initializes the light, sets its initial image, and listens for speech input.

```
1 // Create and position the sprite.
2 var light = new Sprite();
3 light.jumpTo(-150, 0);
4
5 // Change the light based on speech text.
6 function changeLight(speechText) {
7   if (speechText.includes('on')) {
8     light.wear('https://bit.ly/light-on-image');
9   }
10  if (speechText.includes('off')) {
11    light.wear('https://bit.ly/light-off-image');
12  }
13  listen(changeLight);
14 }
15
16 // Main program.
17 light.wear('https://bit.ly/light-off-image');
18 write('Speak to turn the light on and off.');
```

Speak to turn the light on and off.



**LESSON: Home automation: General Purpose Programming** (Years 7-8)

# Can students write code to investigate and use a neural network?

YES!

Utilise an existing sentiment analysis system in a general purpose program (**Python**).

```
Alice is a hero, score: 42
Queen is a hero, score: 16
Rabbit is neutral, score: 6
Duchess is a hero, score: 18
Illustration is neutral, score: 0
Hatter is a villain, score: -5
Majesty is a villain, score: -2
Alices is a hero, score: 11
Youre is neutral, score: 1
Hearts is neutral, score: 1
```

**LESSON: Coding a sentimental chatbot** (Years 7-10)

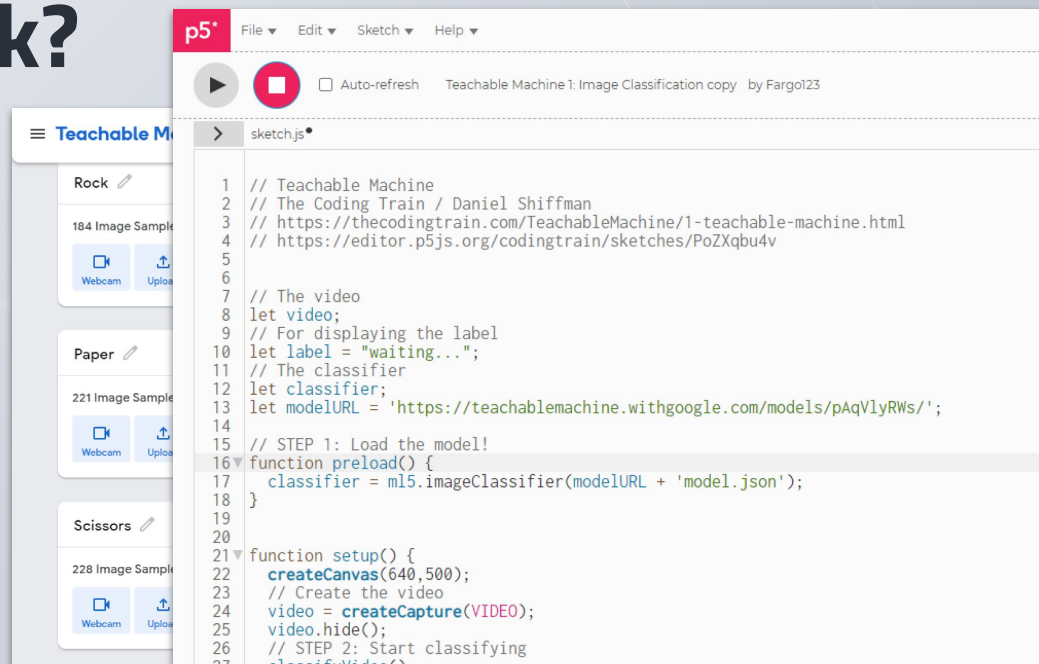
**LESSON: Book analysis with AI techniques** (Years 7-10)

# Can students write code to investigate and use a neural network?

YES!

Train their own AI model with an online tool.

Bring the model into a general purpose program (**JavaScript**) to make a decision.



**LESSON: Rock, Paper, Scissors, AI!** (Years 7-8)

# Can students write code to investigate and use a neural network?

YES!

Train *and* test an AI model with general purpose code (**Python**).

Perform **classification** and **regression**.

# First, let's talk about **data**.

## Years 7-8

Investigate how digital systems represent text, image and audio data in binary (ACTDIK024)

TEXT



T = 84

E = 69

X = 88

T = 84



01010100

01000101

01011000

01010100

# First, let's talk about **data**.

## Years 7-8

Investigate how digital systems represent text, image and audio data in binary (ACTDIK024)



# First, let's talk about **data**.

## Years 7-8

Investigate how digital systems represent text, image and audio data in binary (ACTDIK024)



R 84	R 82	R 76	R 74	R 72	R 70	R 64	R 52
G 120	G 118	G 113	G 111	G 108	G 106	G 97	G 72
B 78	B 77	B 70	B 68	B 66	B 66	B 64	B 55
R 82	R 80	R 75	R 74	R 65	R 55	R 38	R 16
G 117	G 115	G 112	G 109	G 97	G 78	G 50	G 18
B 77	B 75	B 71	B 69	B 63	B 54	B 44	B 27
R 76	R 75	R 73	R 59	R 38	R 21	R 11	R 9
G 108	G 109	G 104	G 82	G 53	G 25	G 7	G 6
B 71	B 71	B 72	B 62	B 45	B 30	B 20	B 16
R 78	R 64	R 42	R 22	R 11	R 10	R 9	R 7
G 102	G 84	G 57	G 27	G 9	G 6	G 6	G 6
B 75	B 65	B 49	B 33	B 21	B 18	B 15	B 12
R 43	R 25	R 10	R 8	R 9	R 8	R 8	R 21
G 51	G 26	G 11	G 5	G 5	G 6	G 5	G 23
B 48	B 33	B 21	B 18	B 16	B 15	B 13	B 22
R 12	R 9	R 8	R 8	R 8	R 9	R 27	R 57
G 8	G 5	G 6	G 5	G 5	G 7	G 30	G 74
B 21	B 16	B 15	B 17	B 14	B 12	B 28	B 57
R 12	R 12	R 8	R 10	R 12	R 35	R 65	R 71
G 6	G 7	G 6	G 6	G 10	G 41	G 86	G 103
B 20	B 21	B 17	B 15	B 15	B 35	B 65	B 74
R 12	R 13	R 12	R 18	R 41	R 70	R 73	R 70
G 9	G 11	G 9	G 16	G 53	G 95	G 106	G 107
B 21	B 21	B 18	B 18	B 40	B 72	B 77	B 74



84	01010100
120	01111000
78	01001110
82	01010010
118	01110110
77	01001101
76	01001100
113	01110001
70	01000110
...	...

# Working with pixel values

Colour Pixelation  
widget at code.org

## One bit per pixel

0 = black, 1 = white

## Two bits per pixel

00 = black, 01 = dark grey,  
10 = light grey, 11 = white

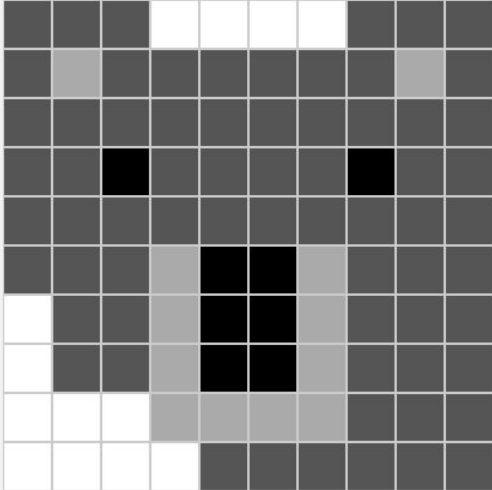
CODE Lesson 3: Color Pixelation Tutorial

**Image File Format:**

Width: 1 byte
Height: 1 byte
Bits per Pixel: 1 byte
n bits of pixel data n = Width * Height * Bits per Pixel

Image width: 10  
Image height: 10  
Bits per pixel: 2

Binary: ☒ Hexadecimal: ☐



```
0000 1010
0000 1010
0000 0010
01 01 01 11 11 11 11 01 01 01
01 10 01 01 01 01 01 01 10 01
01 01 01 01 01 01 01 01 01 01
01 01 00 01 01 01 01 00 01 01
01 01 01 01 01 01 01 01 01 01
01 01 01 10 00 00 10 01 01 01
11 01 01 10 00 00 10 01 01 01
11 01 01 10 00 00 10 01 01 01
11 11 11 10 10 10 10 01 01 01
11 11 11 11 01 01 01 01 01 01
11 11 11 11 11 01 01 01 01 01
11 11 11 11 11 11 01 01 01 01
```

Readable format Raw format Start Over

Save Image Actual size:

## Black and white

This image is a 7x5 matrix of black and white pixels

Each pixel represents a bit: **0** or **1**



# Binary representation

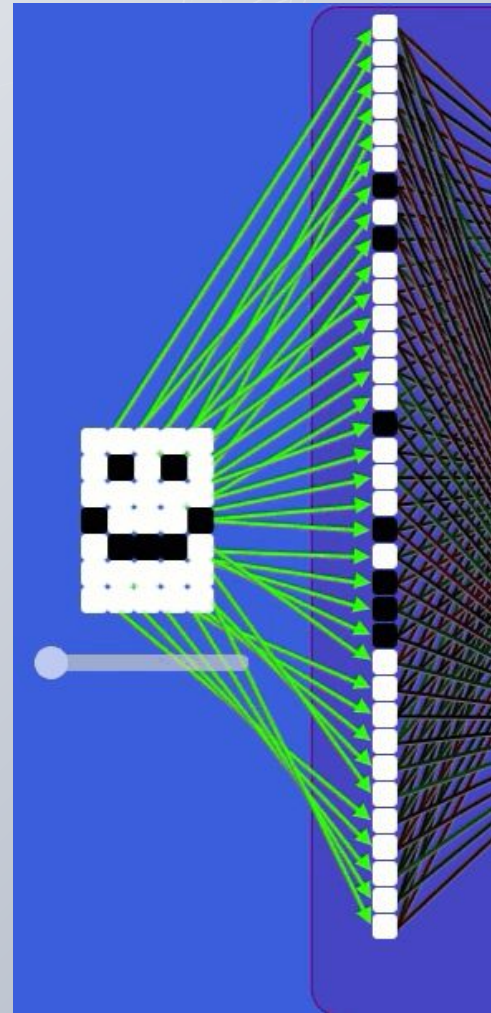
The image can be represented as binary.



00000  
01010  
00000  
10001  
01110  
00000  
00000

The rows of 5 bits can be connected  
into a 35 bit binary number

smiley = 00000 01010 00000  
10001 01110 00000 00000



# Binary representation

The binary representation is preferred over a decimal representation, as binary more closely resembles the image.

By looking at the binary number, we can actually see the picture



00000  
01010  
00000  
10001  
01110  
00000  
00000

## Decimal representation

The decimal representation is fairly disconnected from the original image.

Leading zeroes are lost.



336115712

# Binary Representation

With this approach, we can represent any B/W picture as binary numbers.



happy



sad



surprised



angry

# Binary Representation activity

What is the binary  
representations of the  
**surprised** emoji?



00000  
?

# Binary Representation activity

What is the binary  
representations of the  
**surprised** emoji?



00000  
01010  
00000  
01110  
10001  
01110  
00000

This approach is not limited to pixel graphics. We can also represent words in binary.

Let's explore home automation

Image CC-BY-SA NDB Photos ([Wikimedia Commons](#))

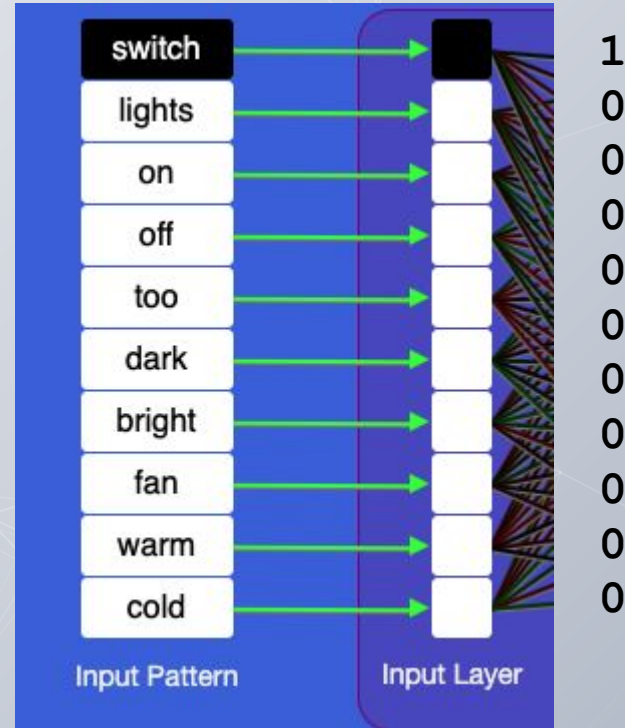


# Home automation

Entire words can be  
represented in binary

In this AI ...

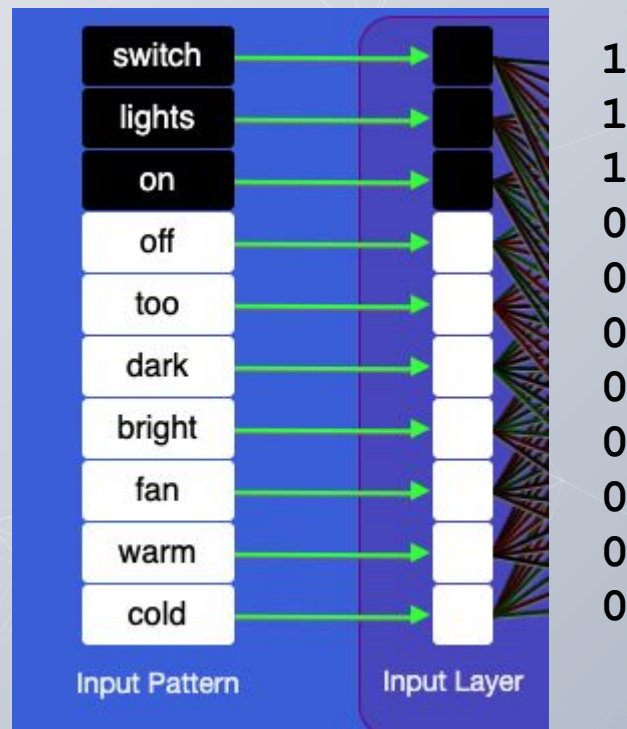
'Switch' is **1000000000**



**LESSON:** Home automation with AI (Years 5-6)

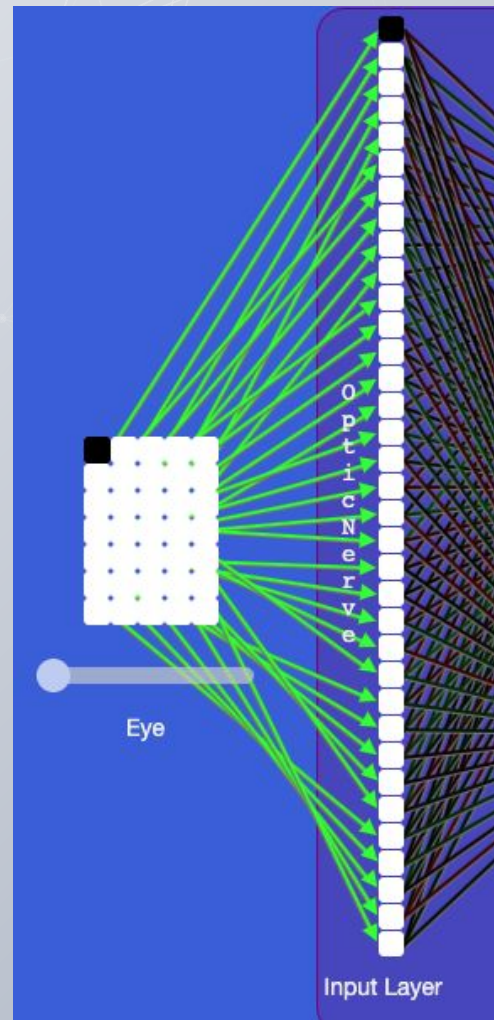
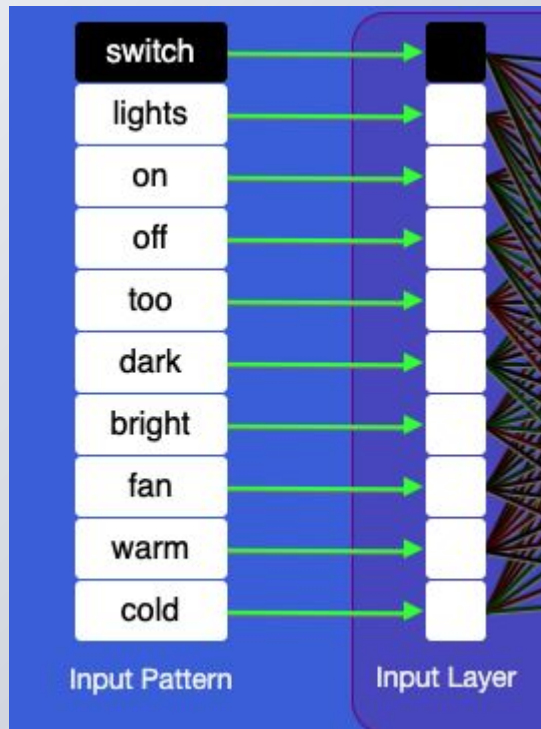
**In this AI ...**

'switch lights on'  
is **1110000000**



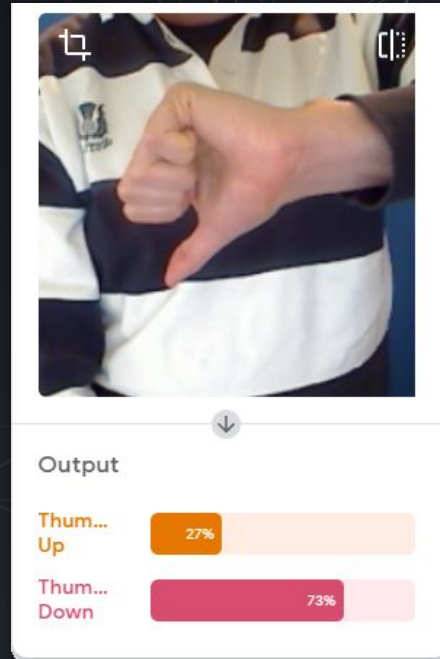
# Ultimate equality!

It does not matter for the AI whether a bit represents a word or a pixel



# Classifier

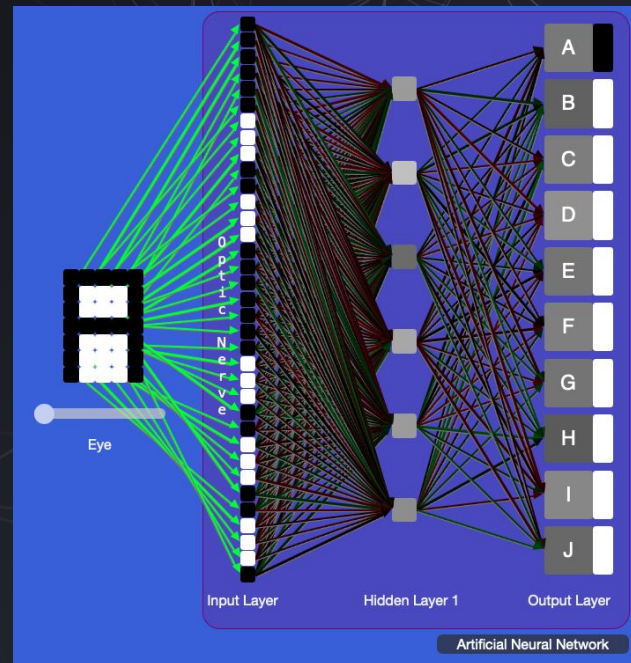
At its output, the AI tells us into which bucket (class) an object most likely belongs.



# Our hands-on example

We classify the first 10 letters of the alphabet.

visual demonstration



# Our hands-on example

Now lets run the simulation with **general purpose programming**.

Using:

- Python
- **replit.com** online environment
- **sci-kit** library

# 1. Include the sci-kit library

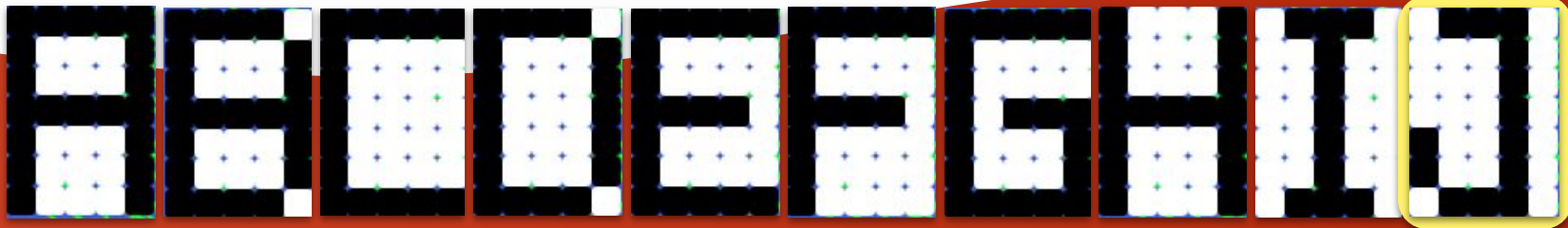
```
from sklearn.neural_network import MLPClassifier
```

- starting point
- finished program

## 2. Binary representation of the letters A-J

# The data.

```
letters = [  
  [1, 1, 1, 1, 1, 1, 0, 0, 0, 1, 1, 0, 0, 0, 1, 1, 1, 1, 1, 1, 1, 0, 0, 0, 1, 1, 0, 0, 0, 1, 1, 0, 0, 0, 1],  
  [1, 1, 1, 1, 0, 1, 0, 0, 0, 1, 1, 0, 0, 0, 1, 1, 1, 1, 1, 1, 1, 0, 0, 0, 1, 1, 0, 0, 0, 1, 1, 1, 1, 1, 0],  
  [1, 1, 1, 1, 1, 1, 0, 0, 0, 0, 1, 0, 0, 0, 0, 1, 0, 0, 0, 0, 1, 0, 0, 0, 0, 1, 0, 0, 0, 0, 1, 1, 1, 1, 1],  
  [1, 1, 1, 1, 0, 1, 0, 0, 0, 1, 1, 0, 0, 0, 1, 1, 0, 0, 0, 1, 1, 0, 0, 0, 1, 1, 0, 0, 0, 1, 1, 1, 1, 1, 0],  
  [1, 1, 1, 1, 1, 1, 0, 0, 0, 0, 1, 0, 0, 0, 0, 1, 1, 1, 1, 0, 1, 0, 0, 0, 0, 1, 0, 0, 0, 0, 1, 1, 1, 1, 1],  
  [1, 1, 1, 1, 1, 1, 0, 0, 0, 0, 1, 0, 0, 0, 0, 1, 1, 1, 1, 0, 1, 0, 0, 0, 0, 1, 0, 0, 0, 0, 1, 0, 0, 0, 0],  
  [1, 1, 1, 1, 1, 1, 0, 0, 0, 0, 1, 0, 0, 0, 0, 1, 0, 1, 1, 1, 1, 1, 0, 0, 0, 1, 1, 0, 0, 0, 1, 1, 1, 1, 1],  
  [1, 0, 0, 0, 1, 1, 0, 0, 0, 1, 1, 0, 0, 0, 1, 1, 1, 1, 1, 1, 1, 1, 0, 0, 0, 1, 1, 0, 0, 0, 1, 1, 0, 0, 0, 1],  
  [0, 1, 1, 1, 0, 0, 0, 1, 0, 0, 0, 0, 1, 0, 0, 0, 0, 1, 0, 0, 0, 0, 1, 0, 0, 0, 0, 1, 0, 0, 0, 0, 1, 1, 1, 0],  
  [0, 1, 1, 1, 0, 0, 0, 0, 1, 0, 0, 0, 0, 1, 0, 0, 0, 0, 1, 0, 1, 0, 0, 0, 1, 0, 1, 0, 0, 1, 0, 0, 1, 1, 1, 0]]
```



### 3. Assign letters to classes (buckets)

```
letter_classes = ['A', 'B', 'C', 'D', 'E', 'F', 'G', 'H', 'I', 'J']
```

### 3. Configure and train the AI

```
# Set up the AI.  
AIClassifier = MLPClassifier(solver = 'lbfgs', activation = 'logistic',  
                             alpha = 1e-5, hidden_layer_sizes = (10),  
                             random_state = 1, max_iter = 200)  
  
# Train the AI.  
AIClassifier.fit(letters, letter_classes)
```

## 4. Check if the classifier works

```
# Test the AI with one of the input letters.  
prediction = AIClassifier.predict([[1, 1, 1, 1, 1,  
                                   1, 0, 0, 0, 0,  
                                   1, 0, 0, 0, 0,  
                                   1, 0, 0, 0, 0,  
                                   1, 0, 0, 0, 0,  
                                   1, 0, 0, 0, 0,  
                                   1, 1, 1, 1, 1]])  
print('The letter C is recognised as', prediction)
```

The letter C is recognised as ['C']

## 5. Classify an unknown letter

```
# Test the AI with an unknown letter.
```

```
unknown_letter = [[1, 1, 1, 1, 1,  
                   1, 0, 0, 0, 0,  
                   1, 0, 0, 0, 0,  
                   1, 1, 0, 1, 0,  
                   1, 0, 0, 0, 0,  
                   0, 0, 0, 0, 0,  
                   1, 1, 1, 1, 1]]
```

```
prediction = AIClassifier.predict(unknown_letter)
```


```
print('The unknown letter is recognised as', prediction)
```



The unknown letter is recognised as ['E']

## 6. Also get confidence level

```
confidence = AIClassifier.predict_proba(unknown_letter)
print('Confidence level is', confidence[0][4].round(2))
```



'E' is at position 4  
in the array of  
classes / buckets.

```
Confidence level is 0.97
```

# Regression

Unlike the classification models, regression models output numeric values.

They have continuous values for both dependent and independent variables.

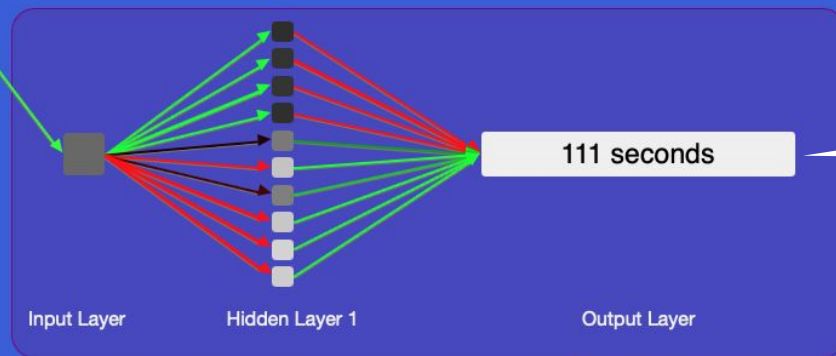
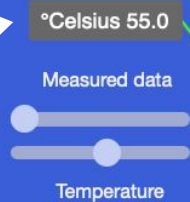
We use regression to find data points outside of a given range (**interpolation** and **extrapolation**).

# Our hands-on example

We find a **curve of best fit**  
for a chemistry experiment.  
visual demonstration



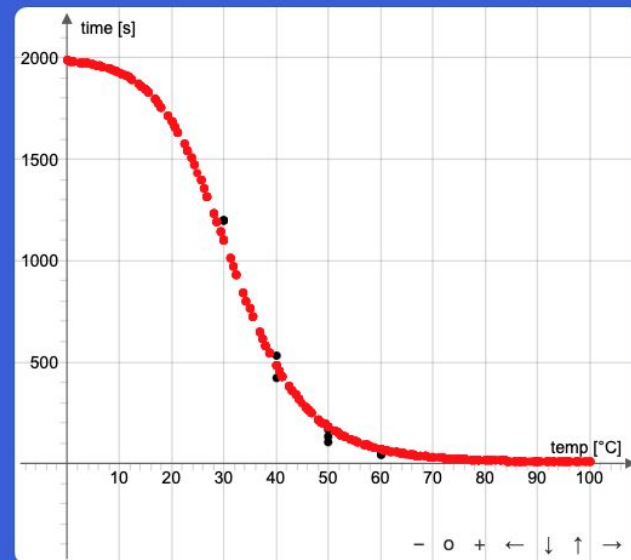
**Numeric  
input**



111 seconds

**Numeric  
output**

Artificial Neural Network



demo

# Our hands-on example

Now lets run the simulation with **general purpose programming**.

Using:

- Python
- **replit.com** online environment
- **sci-kit** library

# 1. Include the sci-kit library + others

```
from sklearn.neural_network import MLPRegressor  
import matplotlib.pyplot as plt  
import numpy as np
```

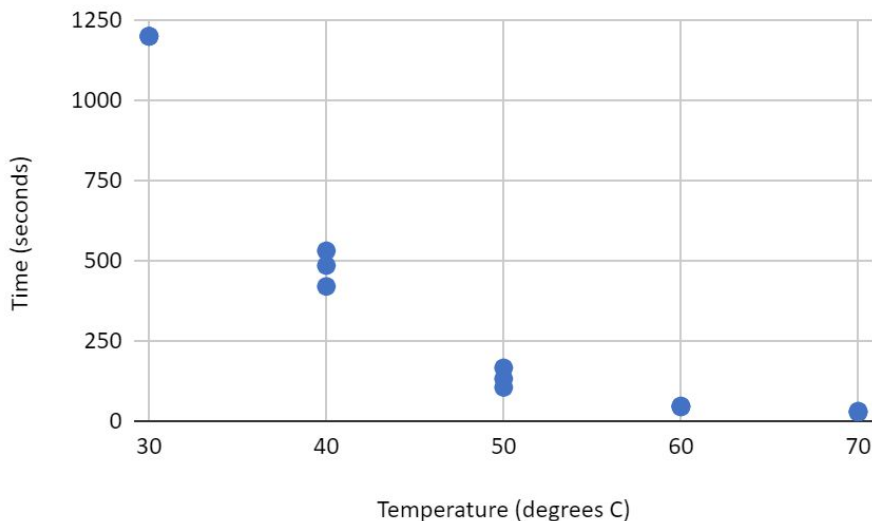
- starting point
- finished program

## 2. Raw data from the experiment

```
# The data.
```

```
temp = [[30], [30], [30], [40], [40], [40], [50], [50], [50], [60], [60], [60], [70], [70], [70]]
```

```
time = [1200, 1200, 1200, 531, 485, 420, 166, 132, 105, 47, 44, 46, 29, 27, 31]
```



### 3. Normalise (scale) the data

Als like their data in the domain and range of 0 to 1.

Can be any real number between 0 and 1.

$$x \in (0,1)$$

### 3. Normalise (scale) the data

```
# Normalise temperature by 100.
temp_normalised = [x[0] / 100 for x in temp]
temp_normalised = np.array(temp_normalised)
temp_normalised = temp_normalised.reshape(len(temp_normalised), -1)
#temp_normalised = [[0.3], [0.30], [0.30], [0.40], [0.40], [0.40], [0.50],
                    [0.50], [0.50], [0.60], [0.60], [0.60], [0.70], [0.70], [0.70]]

# Normalise time by 2000.
time_normalised = [x / 2000 for x in time]
#time_normalised = [0.60000, 0.60000, 0.60000, 0.26550, 0.24250, 0.21000, 0.08300,
                   0.06600, 0.05250, 0.02350, 0.02200, 0.02300, 0.01450, 0.01350, 0.01550]
```

## 4. Configure and train the AI

```
# Set up the AI.  
AI_regressor = MLPRegressor(solver = 'lbfgs', activation = 'logistic',  
                             alpha = 1e-4, hidden_layer_sizes = (10, 1),  
                             random_state = 1, max_iter = 1000000)  
  
# Train the AI.  
AI_regressor.fit(temp_normalised, time_normalised)
```

## 5. Make a single prediction

```
# Make a single prediction for temp = 35 degrees (x = 0.35).  
prediction = AI_regressor.predict([[0.35]])  
print('For 35 degrees, time is', prediction * 2000, 'milliseconds.')
```



Denormalise the  
time result.

```
For 35 degrees, time is [799.02094774] milliseconds.
```

## 6. Plot original data and AI line of best fit

```
# Plot experiment data.
plt.scatter(temp, time, color = 'orange')

# Plot AI data. Use AI to predict y values for a series of x values.
x = np.arange(0, 1, 0.01) # start, stop, step
y = AI_regressor.predict(x.reshape(len(x), -1)) # put each x-value in its own array,
# suitable for predict()

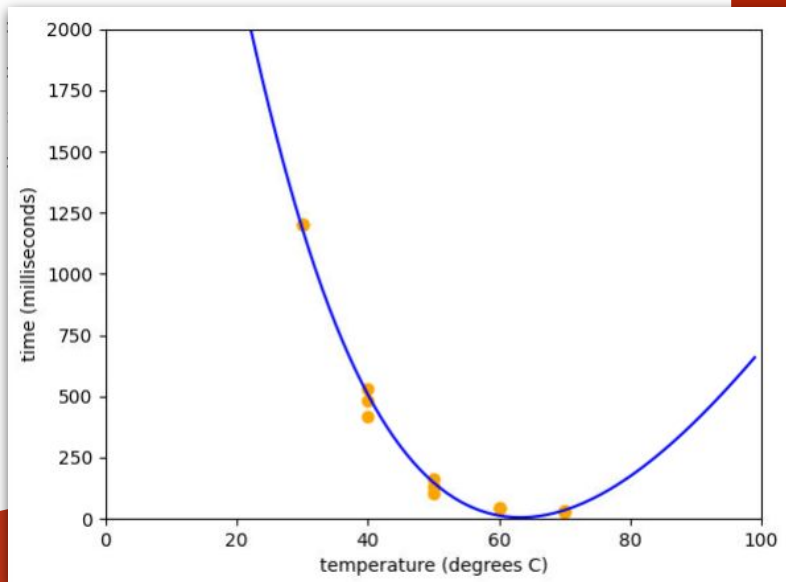
plt.plot(x * 100, y * 2000, color = 'blue') # denormalise x and y for plot

# Limit plot to an appropriate range and domain.
plt.xlim(0, 100)
plt.ylim(0, 2000)
plt.xlabel('temperature (degrees C)')
plt.ylabel('time (milliseconds)')
plt.show()
```

Denormalise the  
time result.

## 6. Plot original data and AI line of best fit

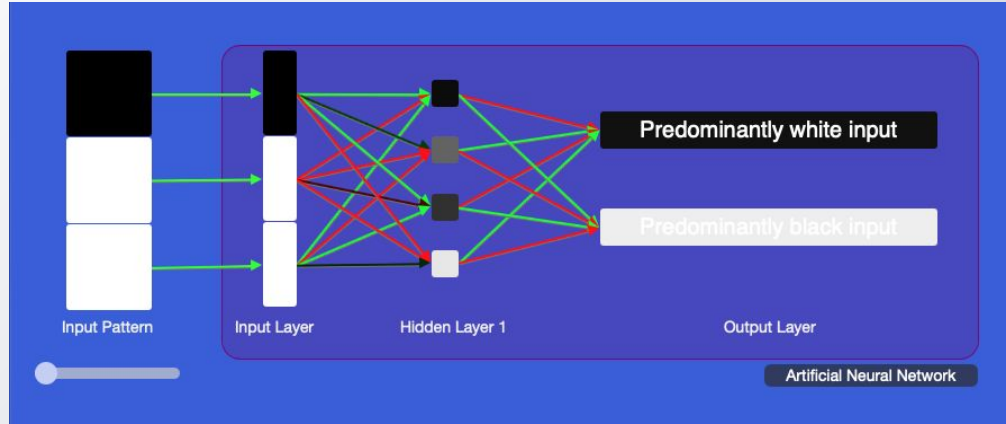
```
# Plot experiment data.  
plt.scatter(temp, time, color = 'orange')  
  
# Plot AI data. Use AI to predict y values for a series of x values.  
x = np.arange(0, 1, 0.01)  
y = AI_regressor.predict(x.reshape(len(x), -1))  
  
plt.plot(x * 100, y * 2000, color = 'blue')  
  
# Limit plot to an appropriate range and domain.  
plt.xlim(0, 100)  
plt.ylim(0, 2000)  
plt.xlabel('temperature (degrees C)')  
plt.ylabel('time (milliseconds)')  
plt.show()
```



# Science Alert



# Inside a neural network



**RESOURCE:** [My Computer Brain](#)

**LESSON COMING SOON:** Inner Workings of an AI  
(Years 9-10)

**EXPLAINER VIDEO:** [Introduction to AI and machine learning](#)

Hidden Layer 1					Output Layer				
Input	Weight	Product	Sum	Output					
1.00	3.43	3.43							
0.00	-1.37	0.00	3.43	0.9686					
0.00	3.73	0.00							

input fields are mainly white									
Input	Weight	Product	Sum	Output					
0.97	-4.23	-4.10							
0.57	13.61	7.72							
0.78	-2.54	-1.98	2.76161	0.9405658					
0.08	13.46	1.12							

input fields are mainly black									
Input	Weight	Product	Sum	Output					
0.97	4.15	4.02							
0.57	-13.59	-7.71							
0.78	2.62	2.04	-2.7663	0.0591743					
0.08	-13.47	-1.12							

input fields are mainly white									
Input	Weight	Product	Sum	Output					
1.00	0.27	0.27							
0.00	-1.81	0.00	0.27	0.5671					
0.00	-1.81	0.00							

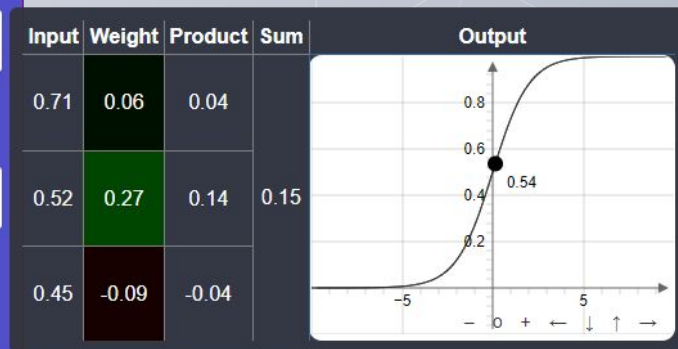
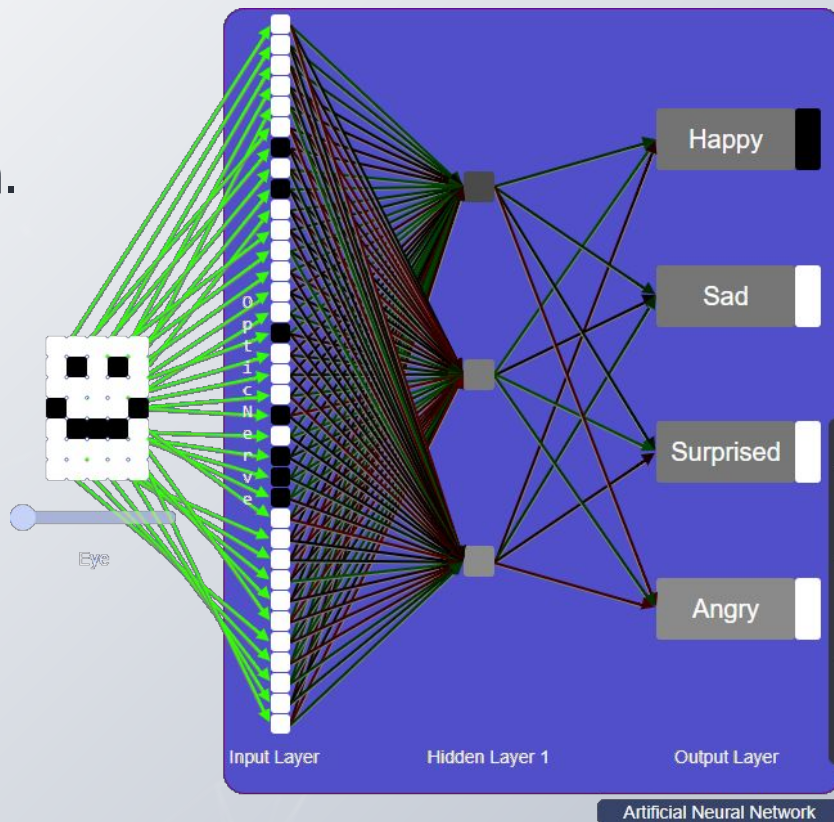
input fields are mainly black									
Input	Weight	Product	Sum	Output					
1.00	1.26	1.26							
0.00	-0.34	0.00	1.26	0.779					
0.00	2.51	0.00							

input fields are mainly white									
Input	Weight	Product	Sum	Output					
1.00	-2.40	-2.40							
0.00	-1.43	0.00	-2.4	0.0832					
0.00	0.18	0.00							

# Investigating a little further

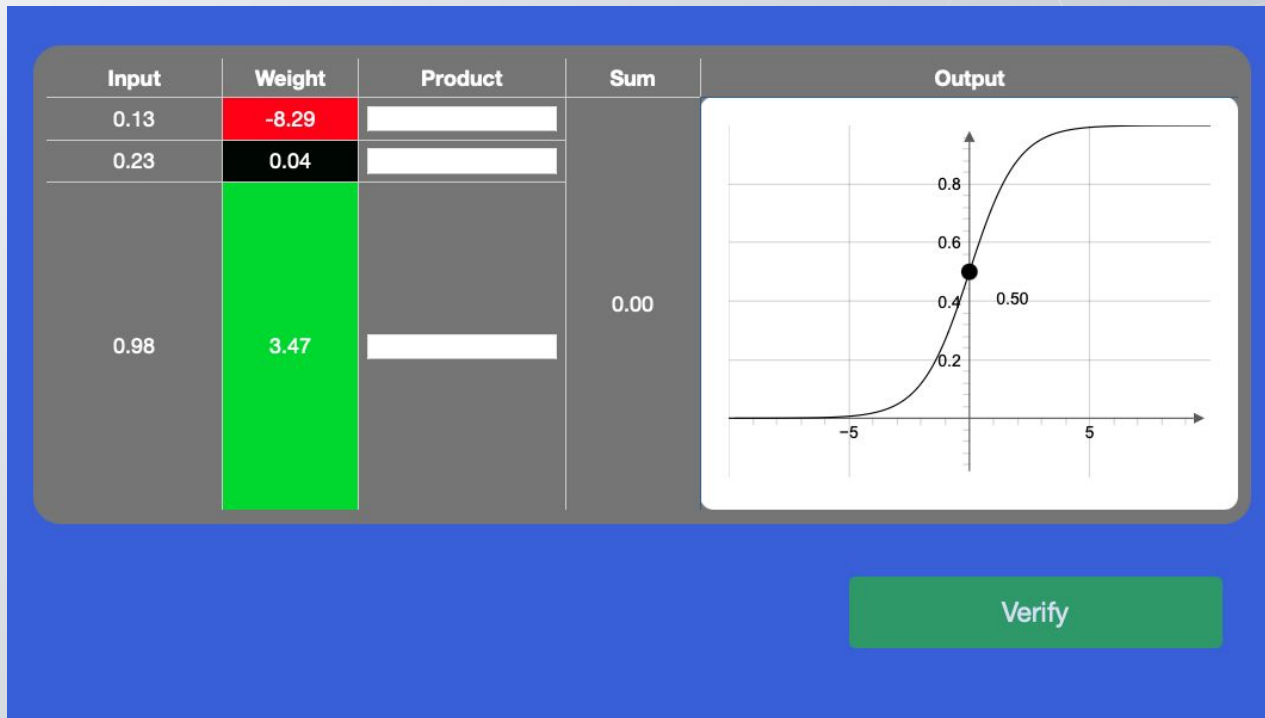
Inside the  
perceptron.



**Demo**

# Inside the Perceptron

Inside the perceptron.



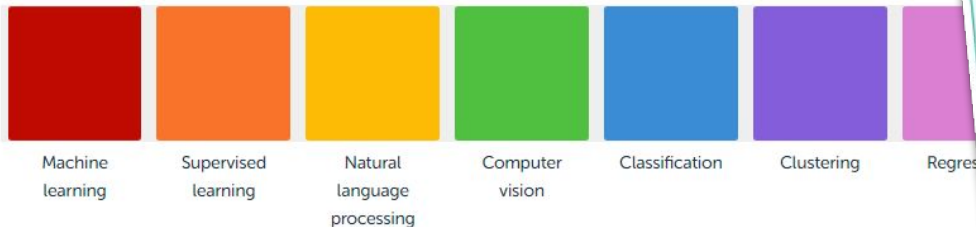
Demo

## Artificial Intelligence lesson plans

Humans display natural intelligence in contrast to machines that demonstrate artificial intelligence (AI).

AI has various definitions however for our purposes we are using the definition 'any device that perceives its environment and takes actions that maximize its chance of successfully achieving its goals' [1]. [Read more...](#)

The following lesson ideas cover a range of specialisations and subsets as indicated by the colour coding. Click on the coloured squares to learn more about each definition.



# Lesson plans

Artificial Intelligence

[Access DT Hub AI lesson plans](#)



# A chance to ask questions ...



## Use the chat...

How can you incorporate these teaching ideas?

What do you feel more confident about?

What do you still need to know?

# Next steps

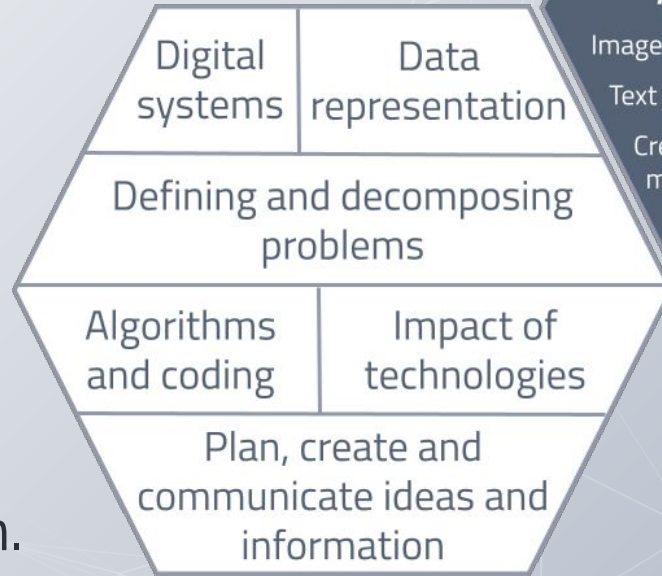
Making a commitment to implementing AI in your classroom

Use the chat to **write your idea** of where you will include AI as part of your teaching and learning program.

Connecting and sharing with the group.

email:

**[digitaltechnologieshub@esa.edu.au](mailto:digitaltechnologieshub@esa.edu.au)**



## AI topics

Image recognition

Text & speech recognition

Creating & using AI models (machine learning)

Bias and ethical issues



**DIGITAL  
TECHNOLOGIES  
HUB**

# Other Deep Dives

Deep dive 3: Natural language processing for large text analysis

Tues 24th August 2021

Deep dive 4: AI, ethics and systems thinking

Tues 7th Sept 2021