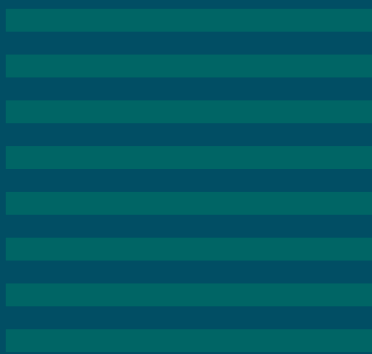




Hands On With Machine Learning

Start time

Rob Butler & Ben Sparks





Because maths matters!

Hands On With Machine Learning

1015 Start

Rob Butler & Ben Sparks

Take the Top Trumps
Cards

Sort them into two
categories

No further questions will
be taken at this time.



ROB BUTLER

Maths Education Support
Specialist

(Maths into AI Team)



BEN SPARKS

Maths Education Support
Specialist

(Maths into AI Team)

Intro

Supervised

Unsupervised

Reinforcement

Summary

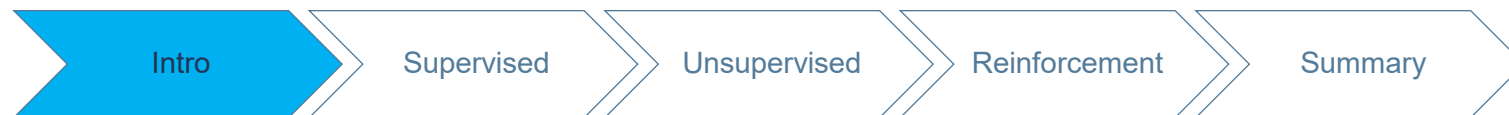
Classification

How did you group your cards?

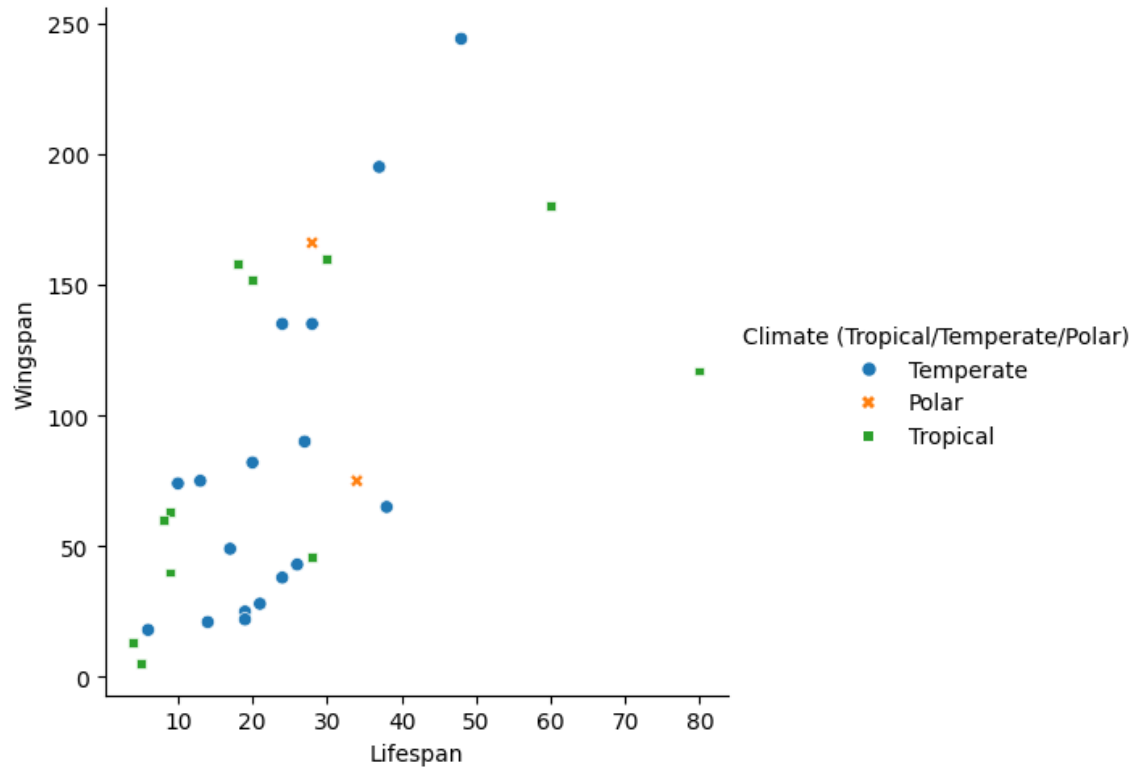
Did you give your groups names?

Which group would a bird with the following information go in?

Wingspan (cm)	Lifespan (years)	Clutch size	Independence
95	15	6	2.0



Classification as a machine learning task



Classification ideas can be applied to many different tasks

Modern AI uses machine learning to recognise patterns in data before creating models to make predictions

Types of Machine Learning

Supervised

Unsupervised

Reinforcement

Intro

Supervised

Unsupervised

Reinforcement

Summary

Supervised Machine Learning

“Tell it the answers”



Spam or Ham?

What rules would you make to detect a spam email?



Decision Tree Classification

A human built model?

“If the sender in the address book then it's not spam”

Otherwise if the subject line mentions borrowing, loans, finance, it's spam

If it mentions 'join now' in the text of the email it's spam

Etc.

Intro

Supervised

Unsupervised

Reinforcement

Summary

Recognising language

- A **vectorizer** converts strings of text into vectors
- Each dimension represents information about the words in the text, e.g. text frequency

“new files for new project”

“re evergreen deals ”

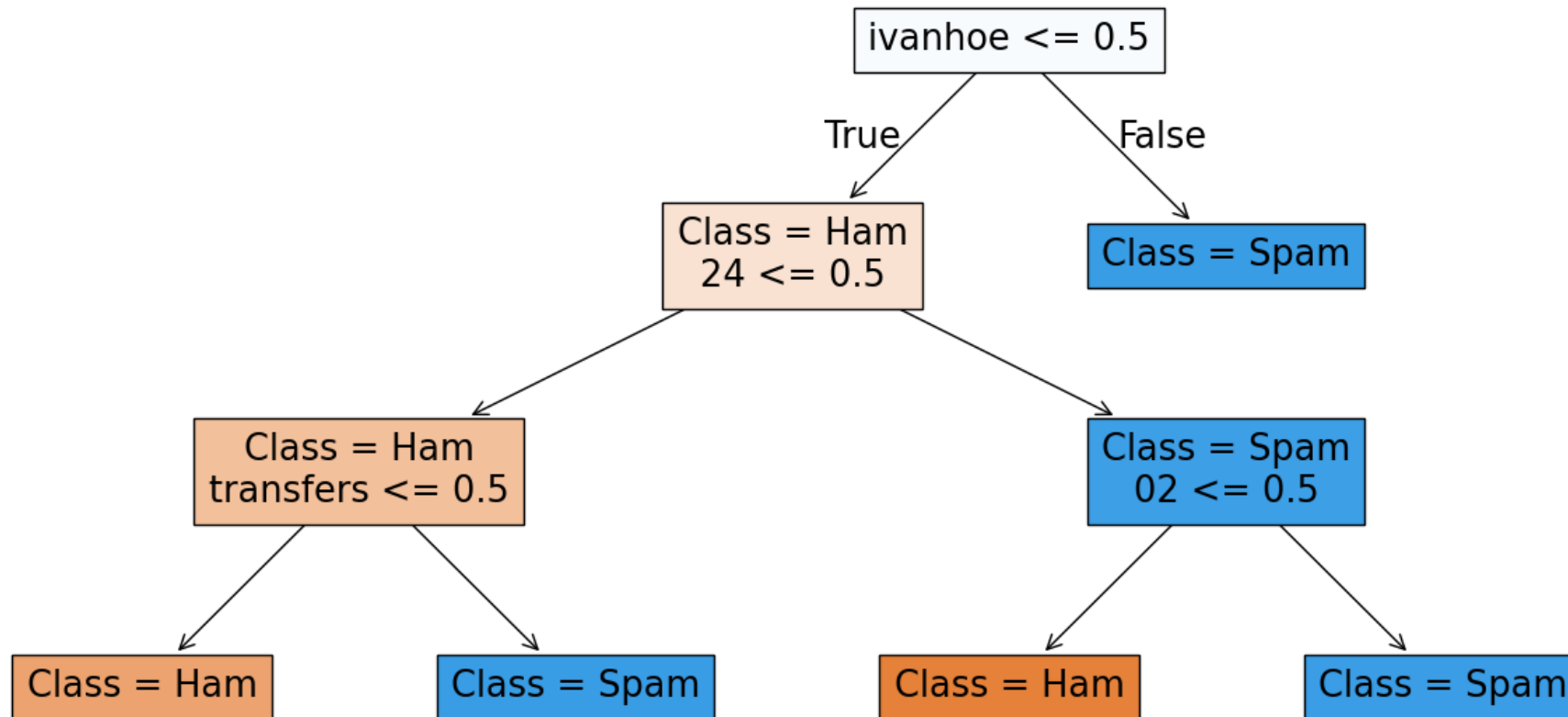
“re misc questions”

deals	evergreen	files	for	misc	new	project	re	questions
0	0	1	1	0	2	1	0	0
1	1	0	0	0	0	0	1	0
0	0	0	0	1	0	0	1	1

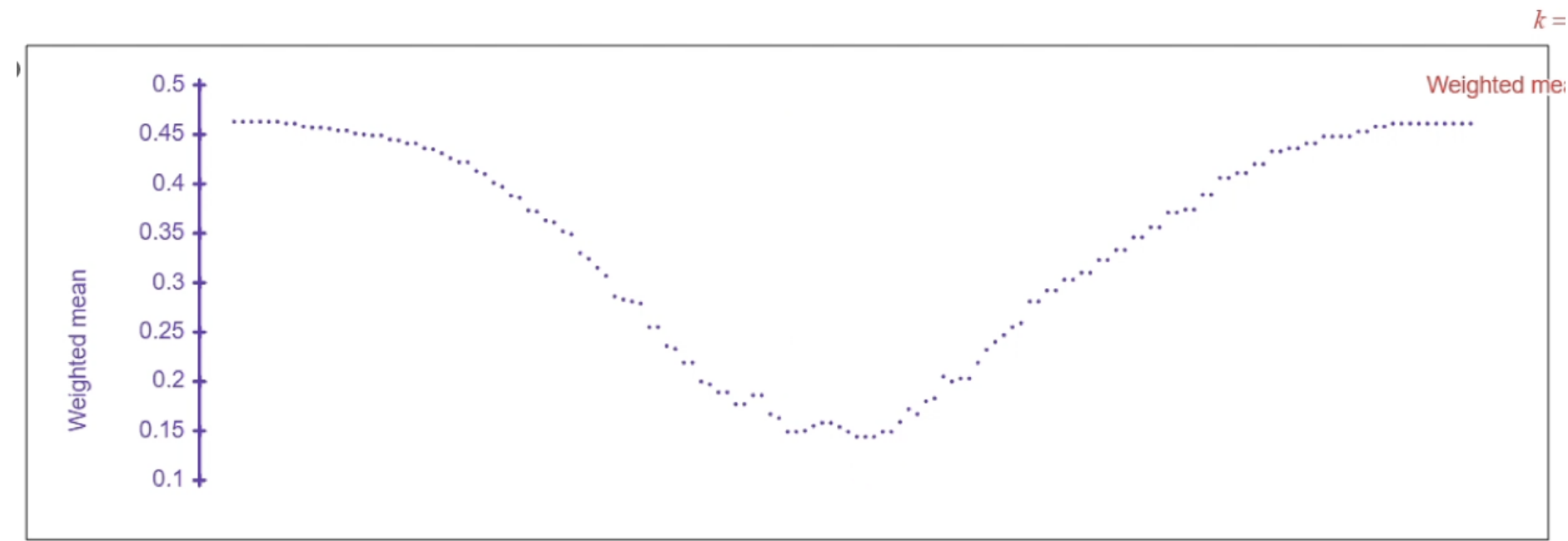
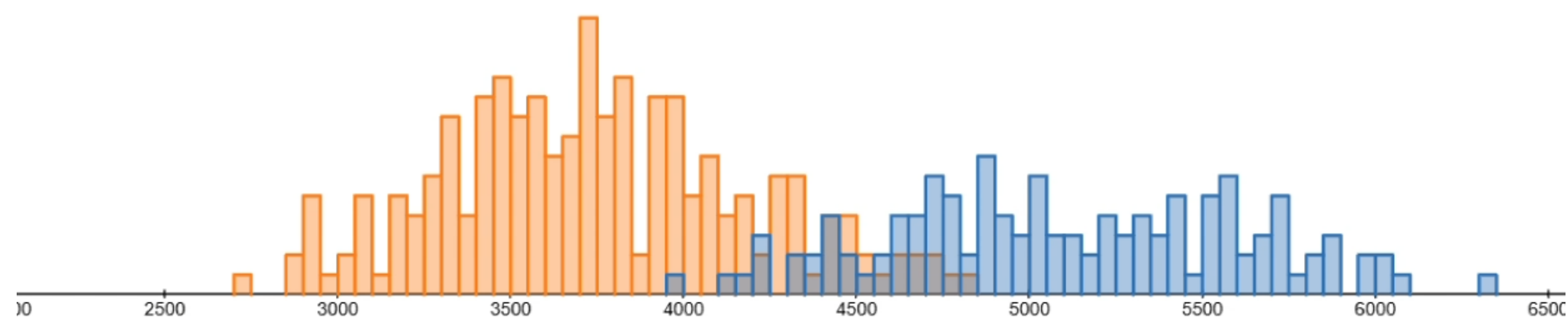
A machine built model

This is very different to how a human would make the decision.

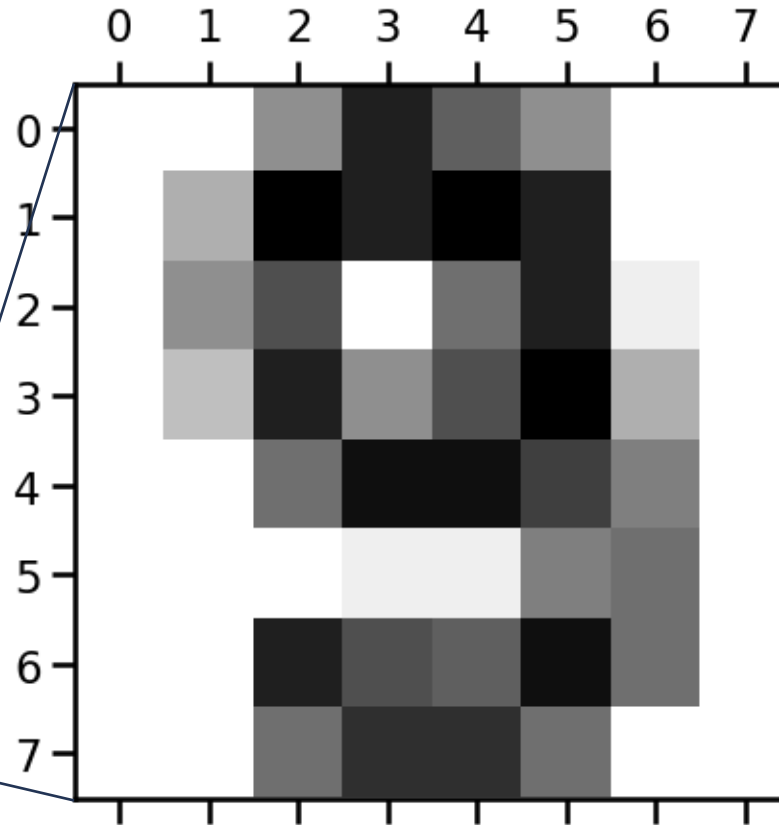
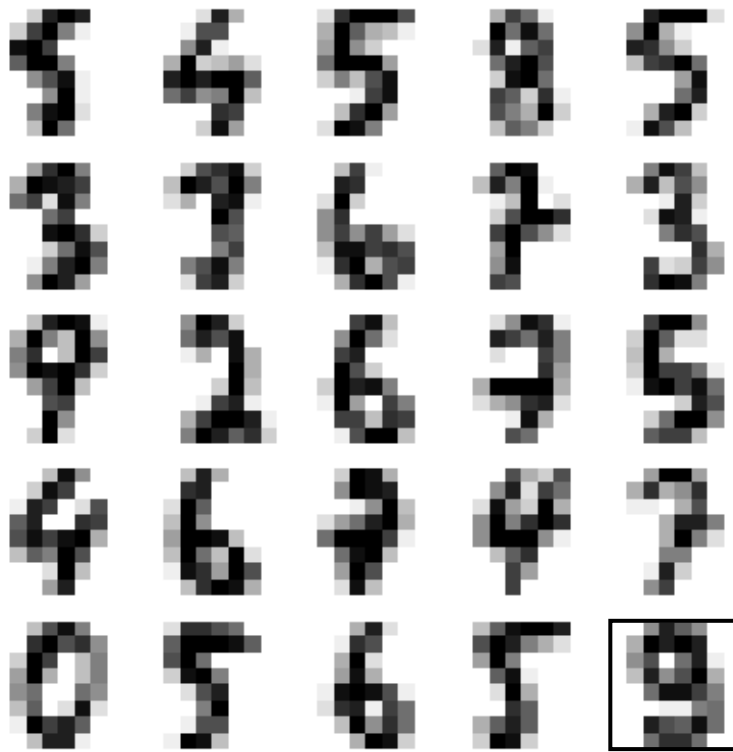
86.4% accurate (on this data set)



HOW DOES IT WORK?



Recognising handwritten digits


$$\begin{pmatrix} 0 & 0 & 7 & 14 & 10 & 7 & 0 & 0 \\ 0 & 5 & 16 & 14 & 16 & 14 & 0 & 0 \\ 0 & 7 & 11 & 0 & 9 & 14 & 1 & 0 \\ 0 & 4 & 14 & 7 & 11 & 16 & 5 & 0 \\ 0 & 0 & 9 & 15 & 15 & 12 & 8 & 0 \\ 0 & 0 & 0 & 1 & 1 & 8 & 9 & 0 \\ 0 & 0 & 14 & 11 & 10 & 15 & 9 & 0 \\ 0 & 0 & 9 & 13 & 13 & 9 & 0 & 0 \end{pmatrix}$$

Intro

Supervised

Unsupervised

Reinforcement

Summary

Recognising handwritten digits


$$\begin{pmatrix} 0 & 0 & 7 & 14 & 10 & 7 & 0 & 0 \\ 0 & 5 & 16 & 14 & 16 & 14 & 0 & 0 \\ 0 & 7 & 11 & 0 & 9 & 14 & 1 & 0 \\ 0 & 4 & 14 & 7 & 11 & 16 & 5 & 0 \\ 0 & 0 & 9 & 15 & 15 & 12 & 8 & 0 \\ 0 & 0 & 0 & 1 & 1 & 8 & 9 & 0 \\ 0 & 0 & 14 & 11 & 10 & 15 & 9 & 0 \\ 0 & 0 & 9 & 13 & 13 & 9 & 0 & 0 \end{pmatrix}$$

64 pixels become 64 features



{0, 0, 7, 14, 10, 7, 0, 0, 0, 5, 16, 14, 16, 14, 0, 0, 0, 7, 1

What happens when a decision tree is trained on this data?

Intro

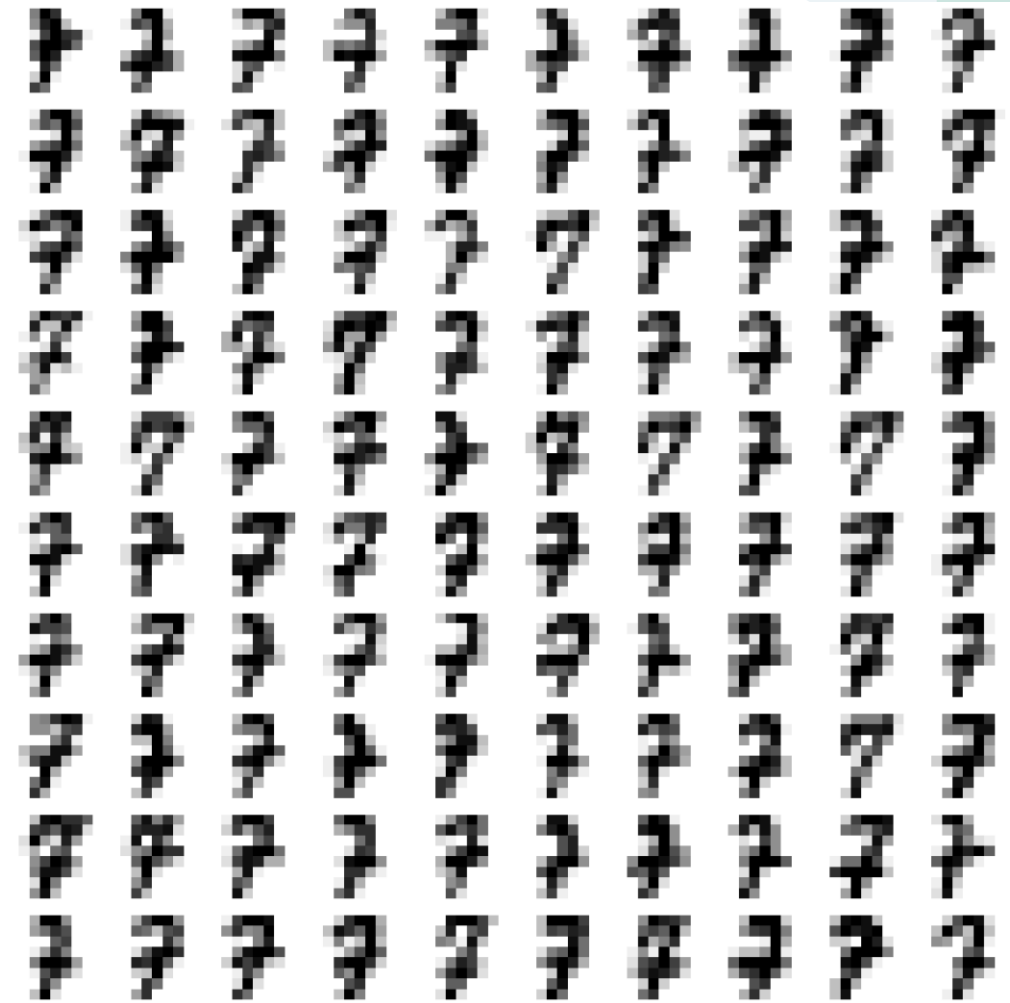
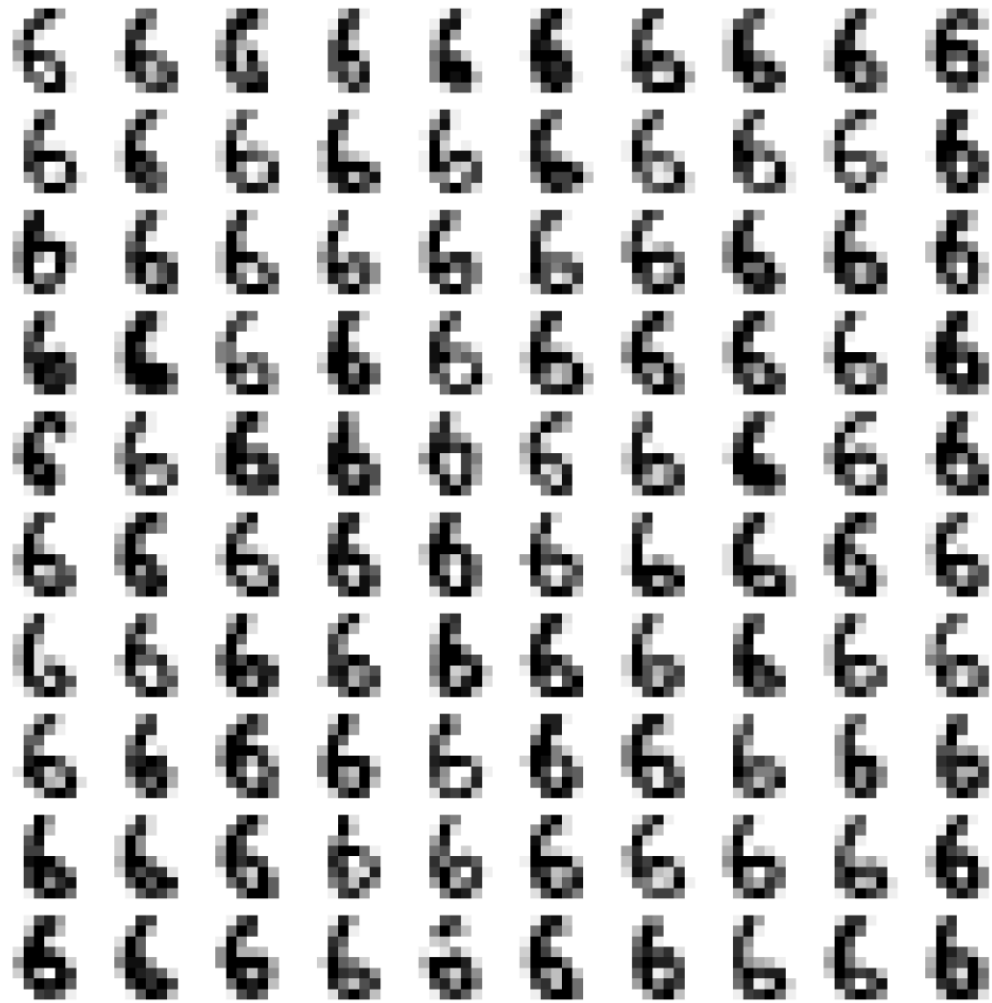
Supervised

Unsupervised

Reinforcement

Summary

Recognising handwritten digits



Intro

Supervised

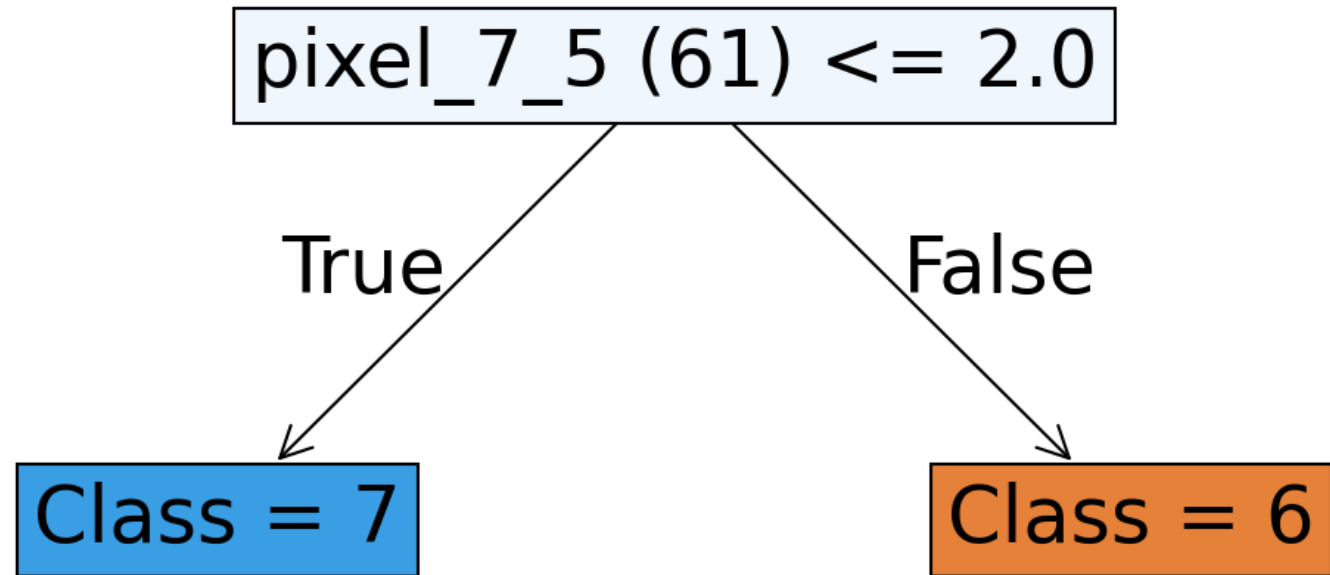
Unsupervised

Reinforcement

Summary

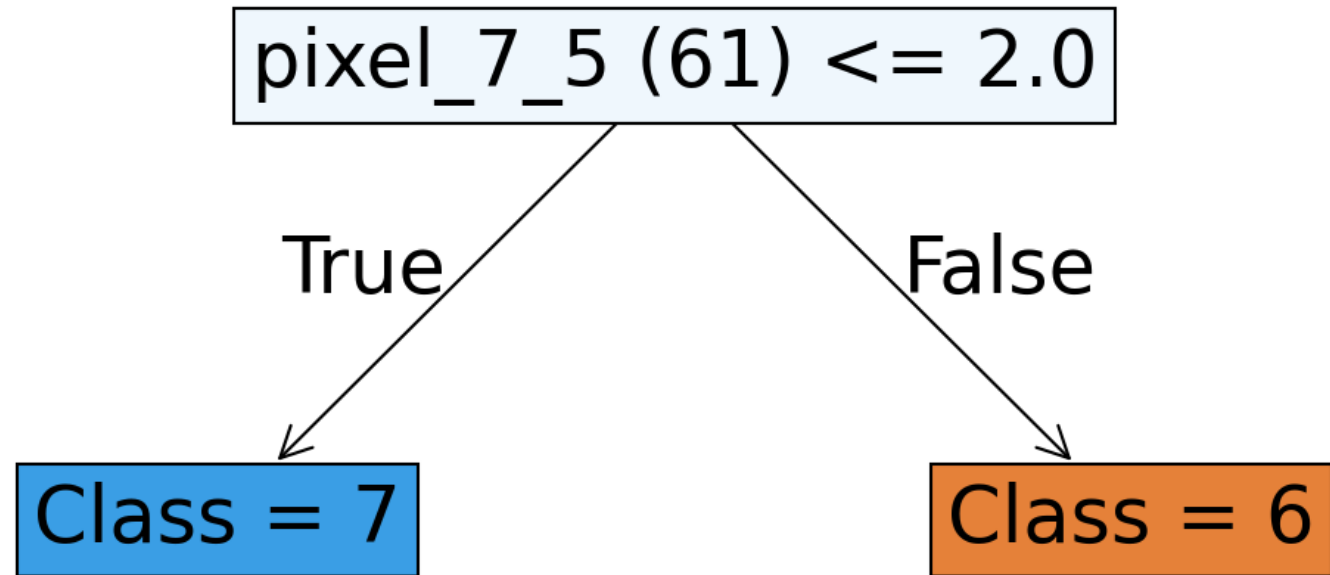
Recognising handwritten digits

	0	1	2	3	4	5	6	7
0	0	1	2	3	4	5	6	7
1	8	9	10	11	12	13	14	15
2	16	17	18	19	20	21	22	23
3	24	25	26	27	28	29	30	31
4	32	33	34	35	36	37	38	39
5	40	41	42	43	44	45	46	47
6	48	49	50	51	52	53	54	55
7	56	57	58	59	60	61	62	63



Recognising handwritten digits

	0	1	2	3	4	5	6	7
0	0	1	2	3	4	5	6	7
1	8	9	10	11	12	13	14	15
2	16	17	18	19	20	21	22	23
3	24	25	26	27	28	29	30	31
4	32	33	34	35	36	37	38	39
5	40	41	42	43	44	45	46	47
6	48	49	50	51	52	53	54	55
7	56	57	58	59	60	61	62	63



Recognising handwritten digits

	0	1	2	3	4	5	6	7
0	0	1	2	3	4	5	6	7
1	8	9	10	11	12	13	14	15
2	16	17	18	19	20	21	22	23
3	24	25	26	27	28	29	30	31
4	32	33	34	35	36	37	38	39
5	40	41	42	43	44	45	46	47
6	48	49	50	51	52	53	54	55
7	56	57	58	59	60	61	62	63

pixel_7_5 (61) ≤ 2.0

True

False

Class = 7

Class = 6

Intro

Supervised

Unsupervised

Reinforcement

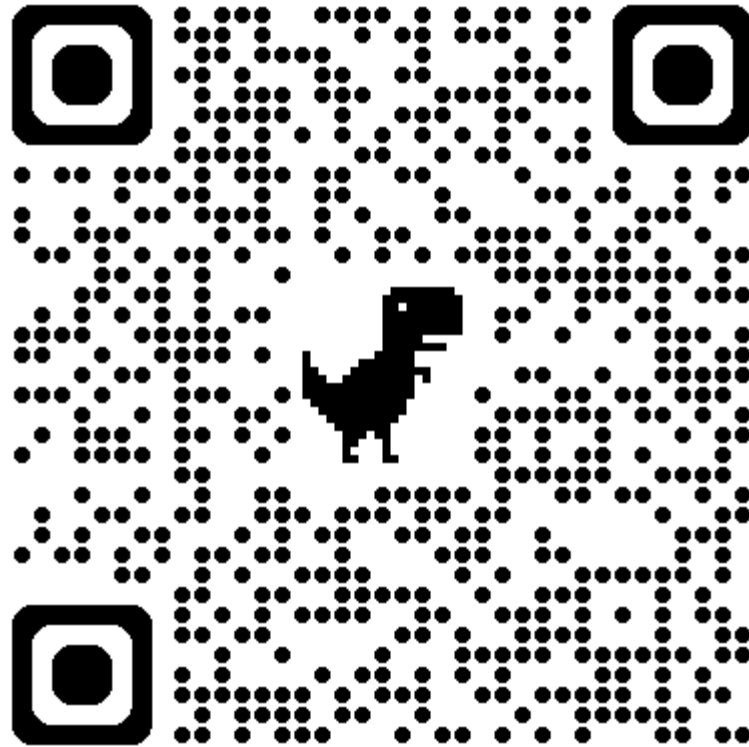
Summary

Unsupervised Machine Learning

“Don’t tell it the answers”



Controversial Opinions

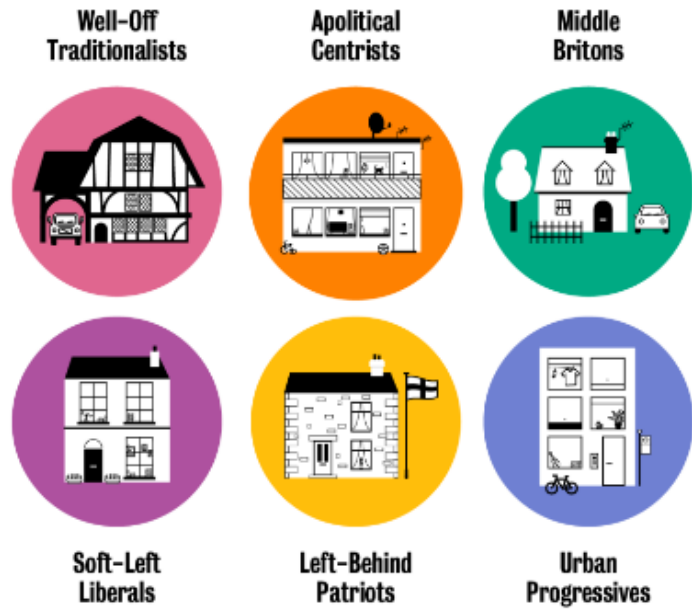


<https://forms.gle/C1pyc5uQxA37wHdy9>

Clustering

Explore our six voter types for a more nuanced understanding of the dividing lines in British politics.

 National Centre
for Social Research



Dividing Lines: A Classification of British Voters

<https://natcen.ac.uk/dividing-lines>

SEVEN SEGMENTS OF BRITAIN

<https://www.moreincommon.org.uk/seven-segments/>



Callback to Top Trumps

Cluster 1: Large wingspan, long lived, small clutch, independent, ugly birds

BIG UGLY BIRDS

Cluster 2: Quite small wingspan, short lived, small clutch, cute

SMALL CUTE BIRDS

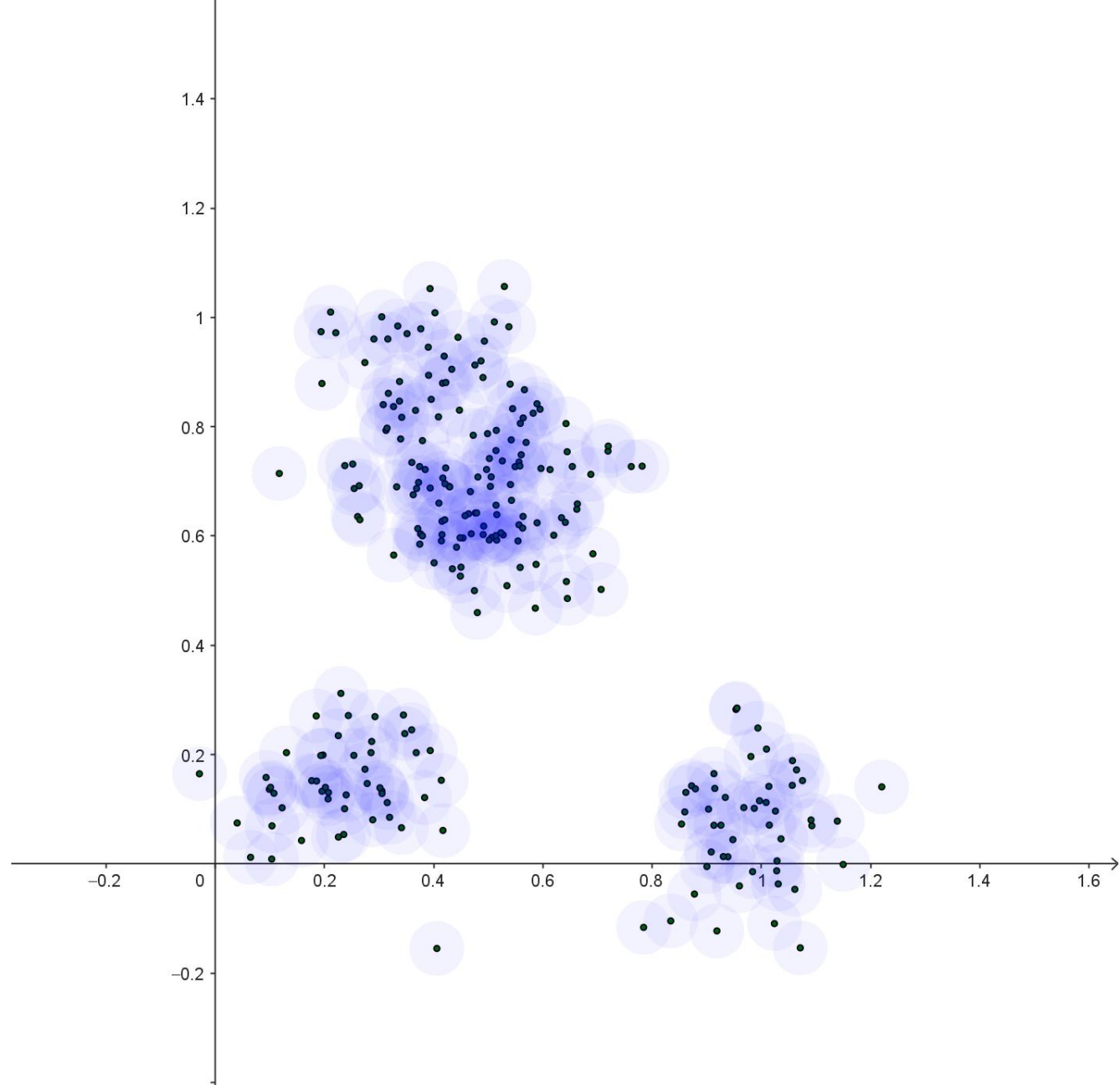
Cluster 3: small wingspan, fairly short-lived, medium clutch size, dependent, not very cute

SMALL UGLY BIRDS

Cluster 4: Medium wingspan, medium length life, large clutch size, pretty cute but not very

MEH BIRDS

Clustering



Intro

Supervised

Unsupervised

Reinforcement

Summary

Controversial Opinions

Your data?

Reinforcement Learning

^{bot}
“~~Good dog~~” and “~~Bad dog~~”^{bot}



Reinforcement Learning



Reinforcement learning: Training a simple game bot

Higher or Lower, with a single die

**Will the next roll will be
Higher or Lower?**

How might you train a 'machine' to play well?



Higher or Lower

What happens next?

(& what questions do you have?)



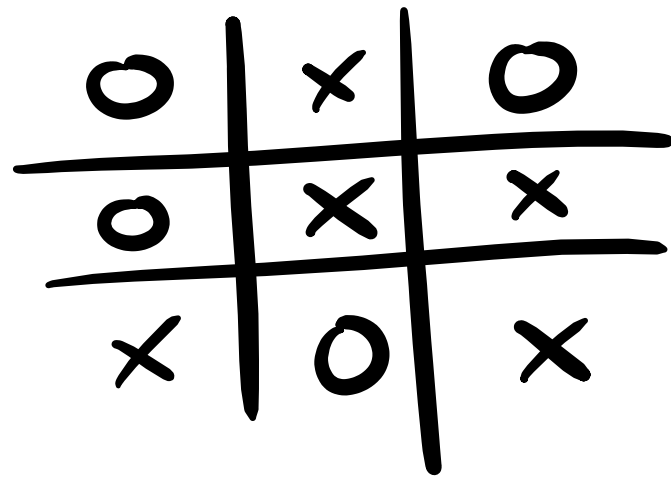
What is reinforcement learning?

Reinforcement learning:

Evolving a strategy through reward/punishment

Examples:

Noughts & crosses, chess, Go



HOLDIR: (Higher Or Lower Dice Robot)

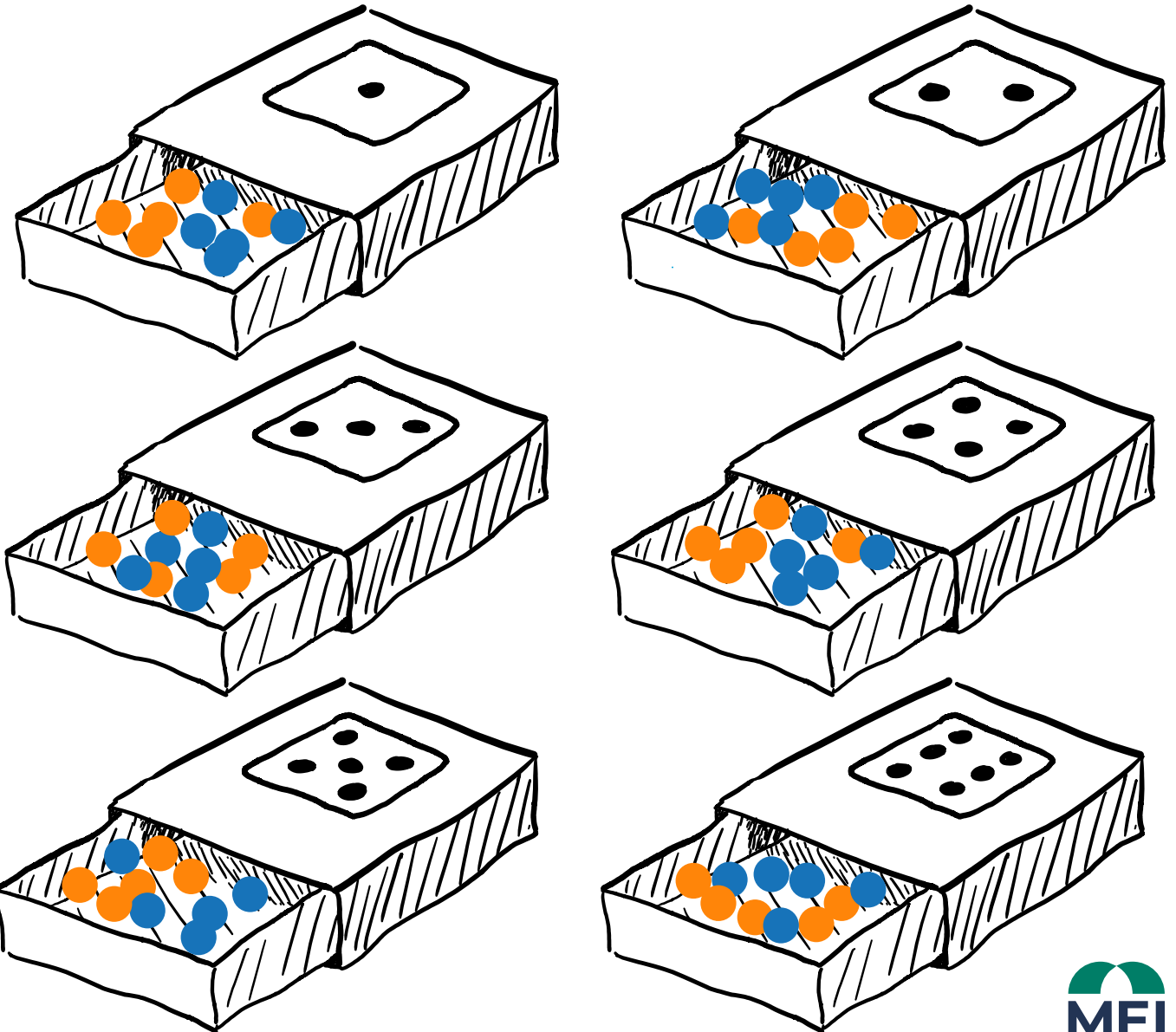
This game has 6 states:
(previous roll was 1, 2, 3, 4, 5, or 6)

Use 6 'matchboxes', labelled appropriately

Start with equal amounts of
blue and orange beads in each box

- Blue = Guess Higher
- Orange = Guess Lower

The 'machine' decides how to play by picking
a bead at random from the appropriate box.



Reward and/or punish

If the guess is incorrect:

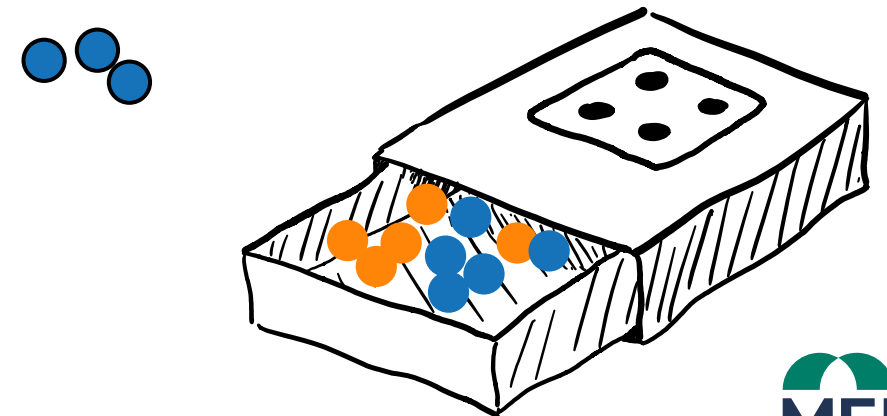
‘Punish’ by not replacing the chosen bead

If the guess is correct:

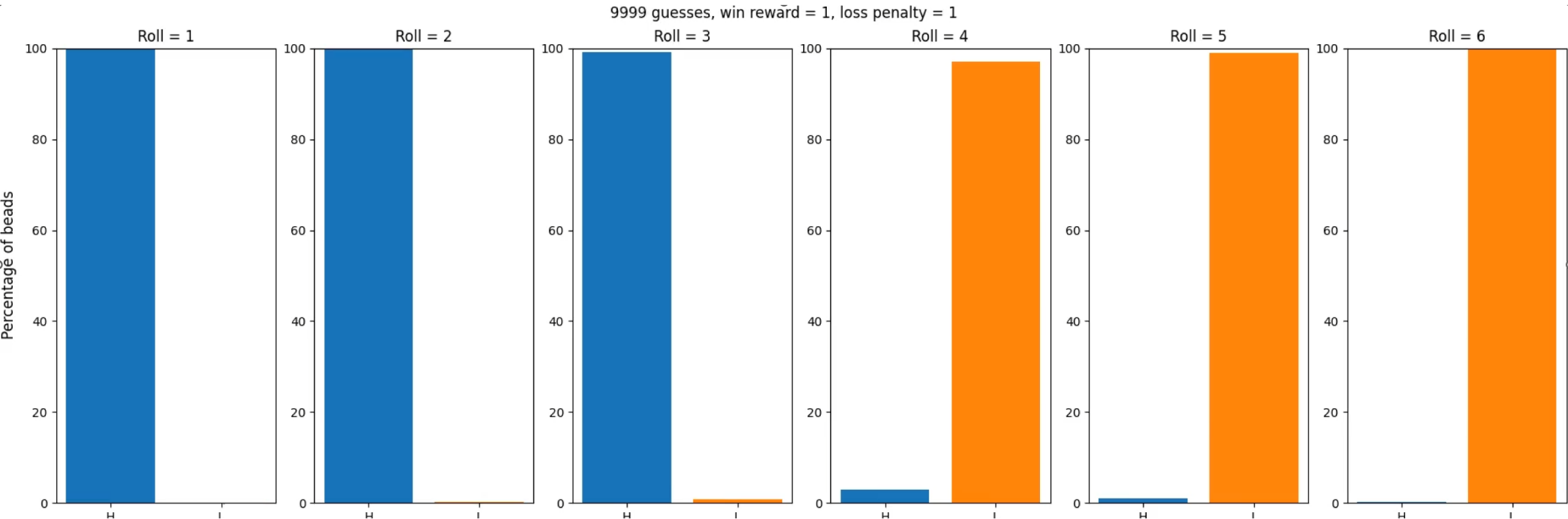
‘Reward’ by replacing the bead
and adding another bead of the same
colour

e.g. A ‘4’ is rolled

- Randomly pick a bead from the 4 box
- Guess accordingly (Blue = Higher)
 - a) If the next roll is higher, reward it
 - b) Otherwise, punish it

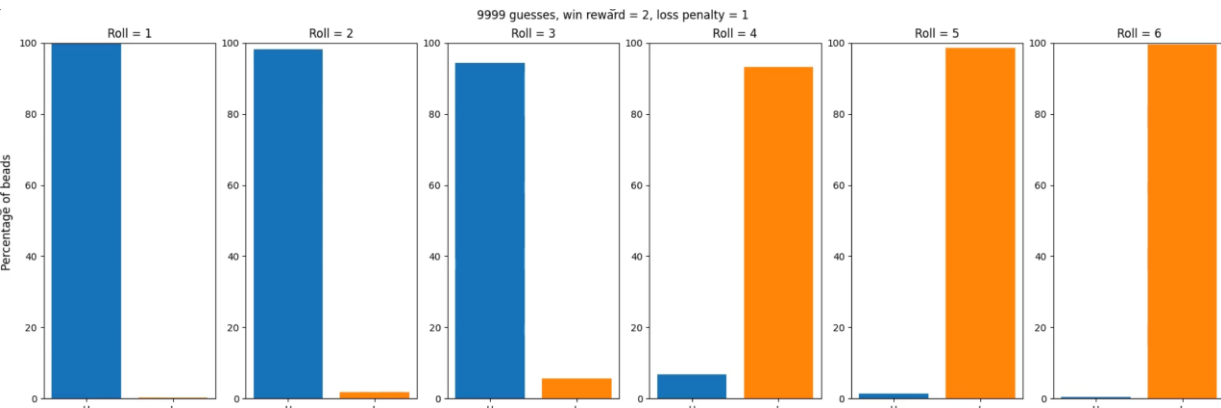


The matchboxes after 10,000 guesses

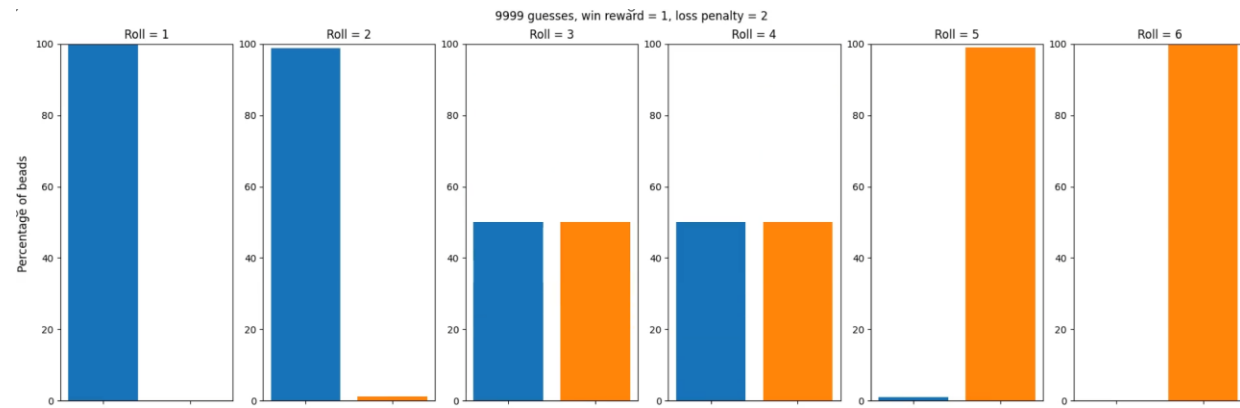


The bars represent the proportions of each colour bead in the box as a percentage

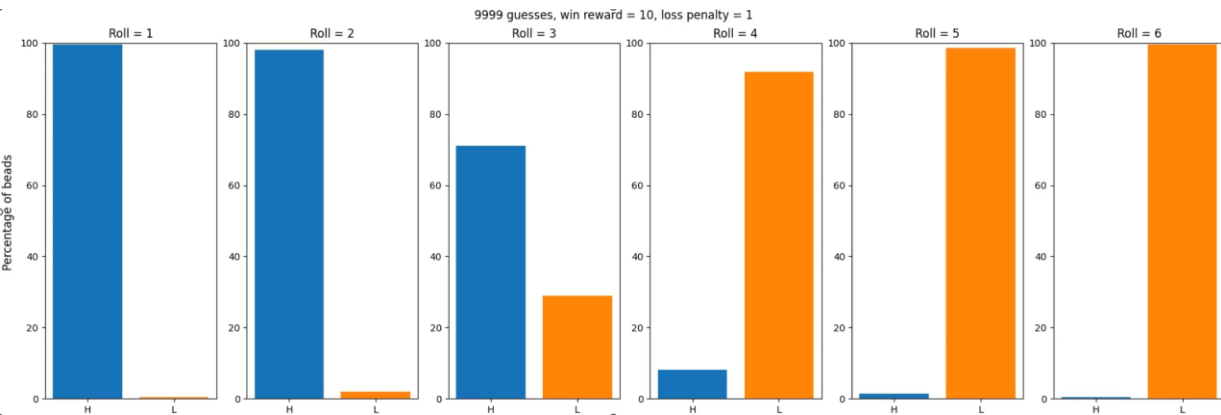
Reward/Punish parameters



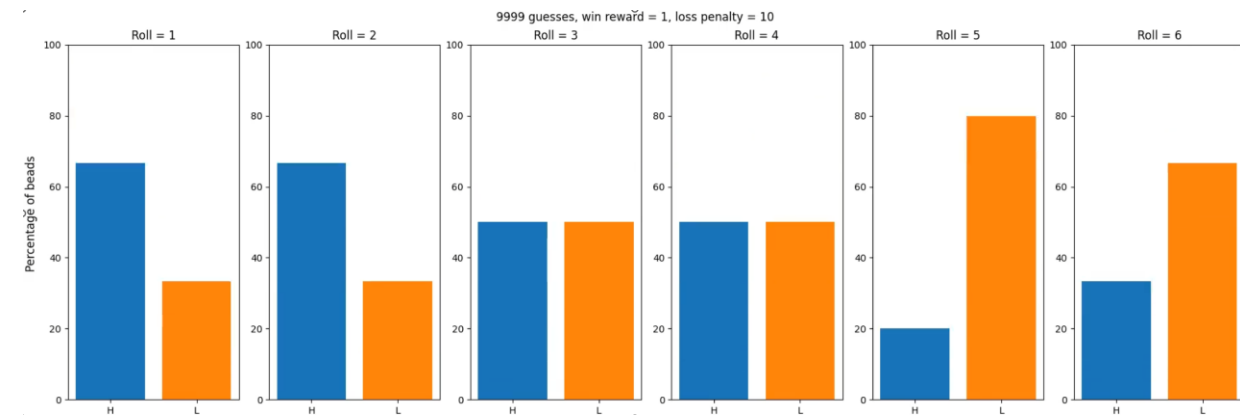
Win reward = 2, Loss penalty = 1



Win reward = 1, Loss penalty = 2



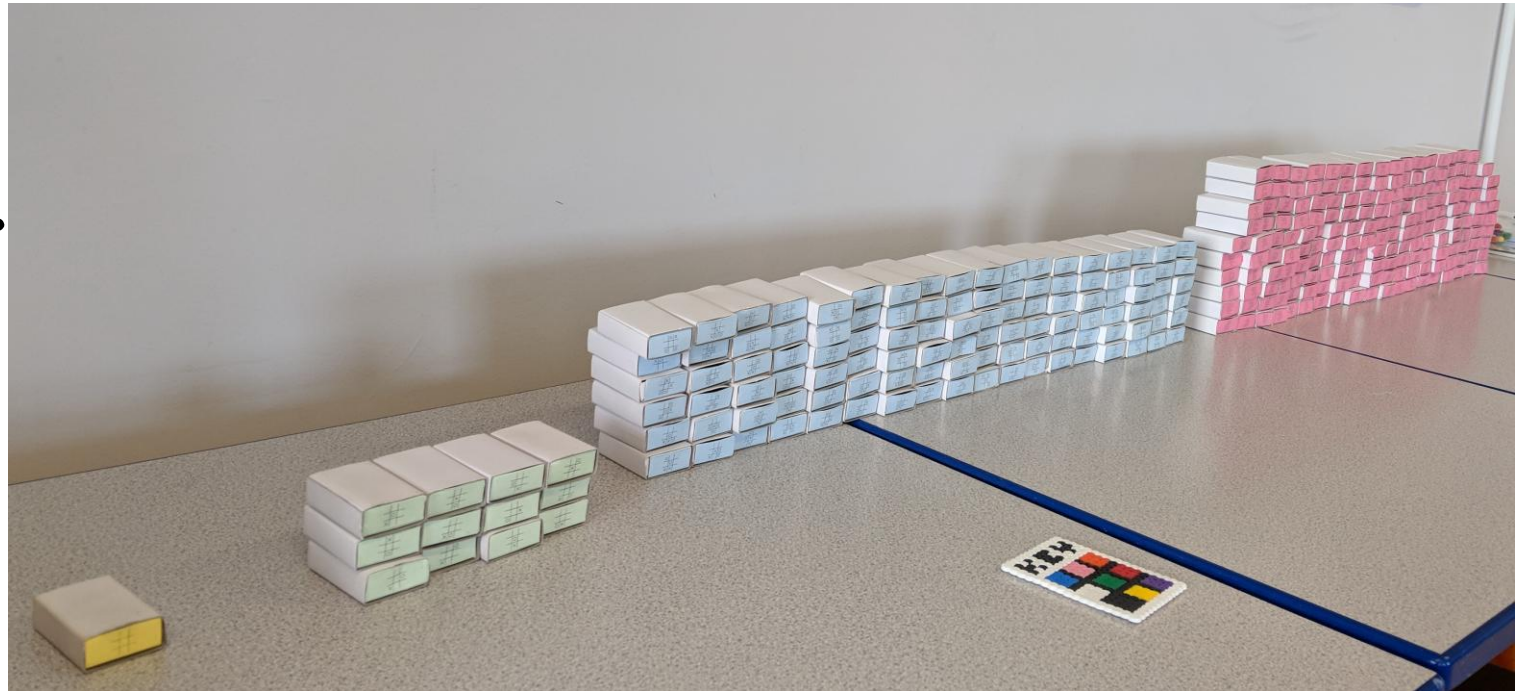
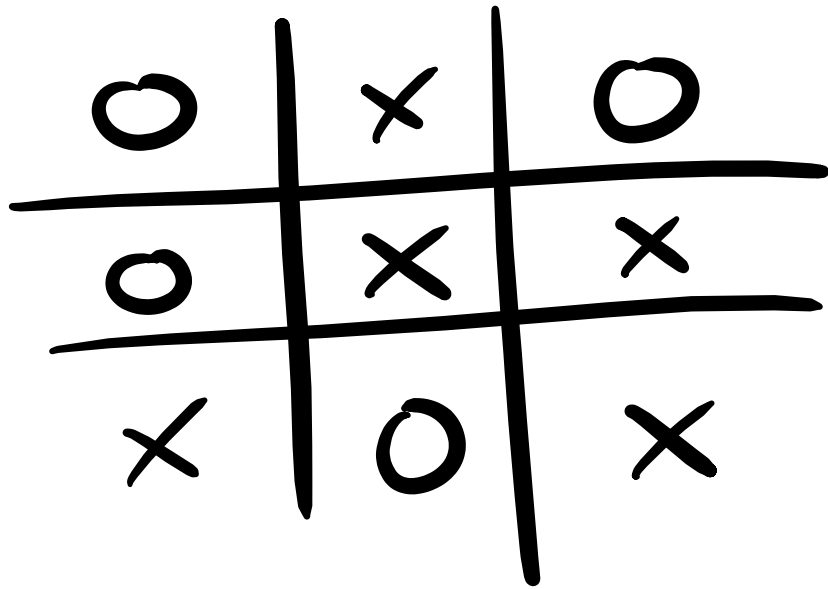
Win reward = 10, Loss penalty = 1



Win reward = 1, Loss penalty = 10

M.E.N.A.C.E.

Machine (or Matchbox) Educable Noughts And Crosses Engine



<https://www.msccroggs.co.uk/menace/>

Intro

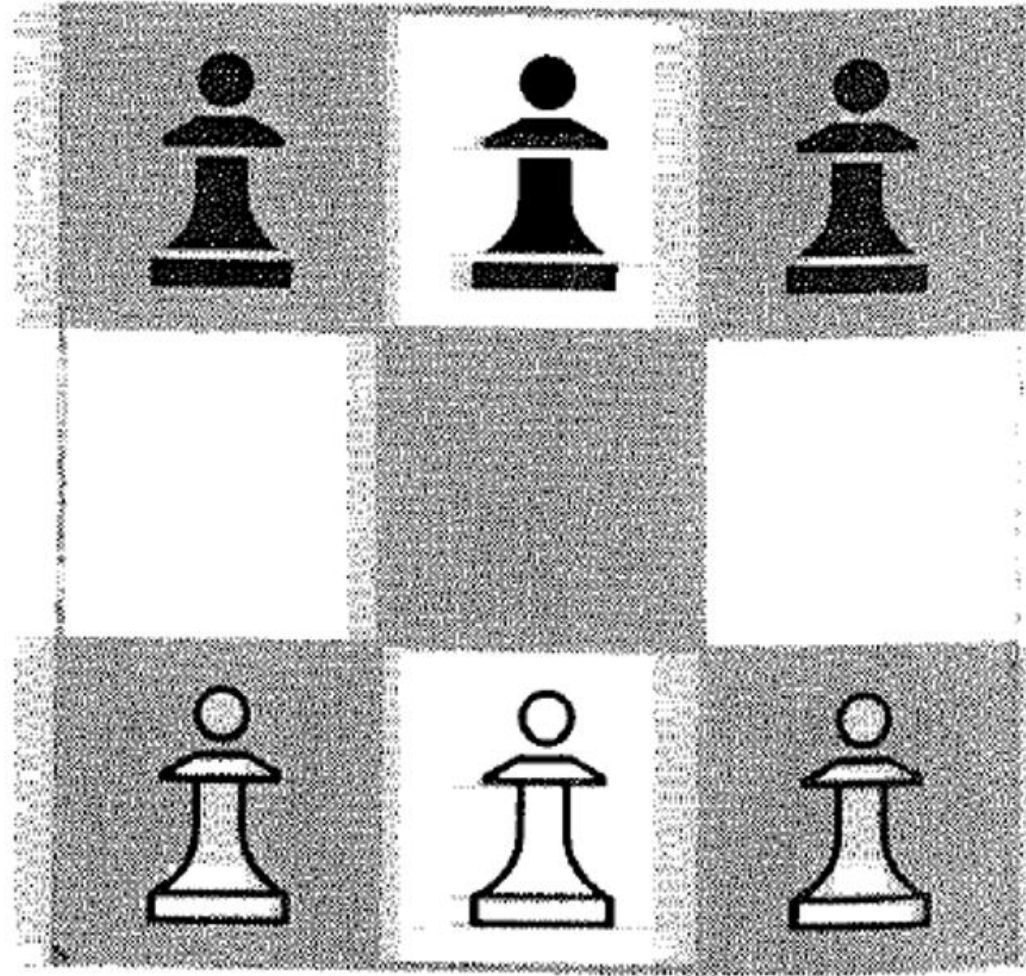
Supervised

Unsupervised

Reinforcement

Summary

Martin Gardner – Hexapawn – 1962 Scientific American



The game of hexapawn

Intro

Supervised

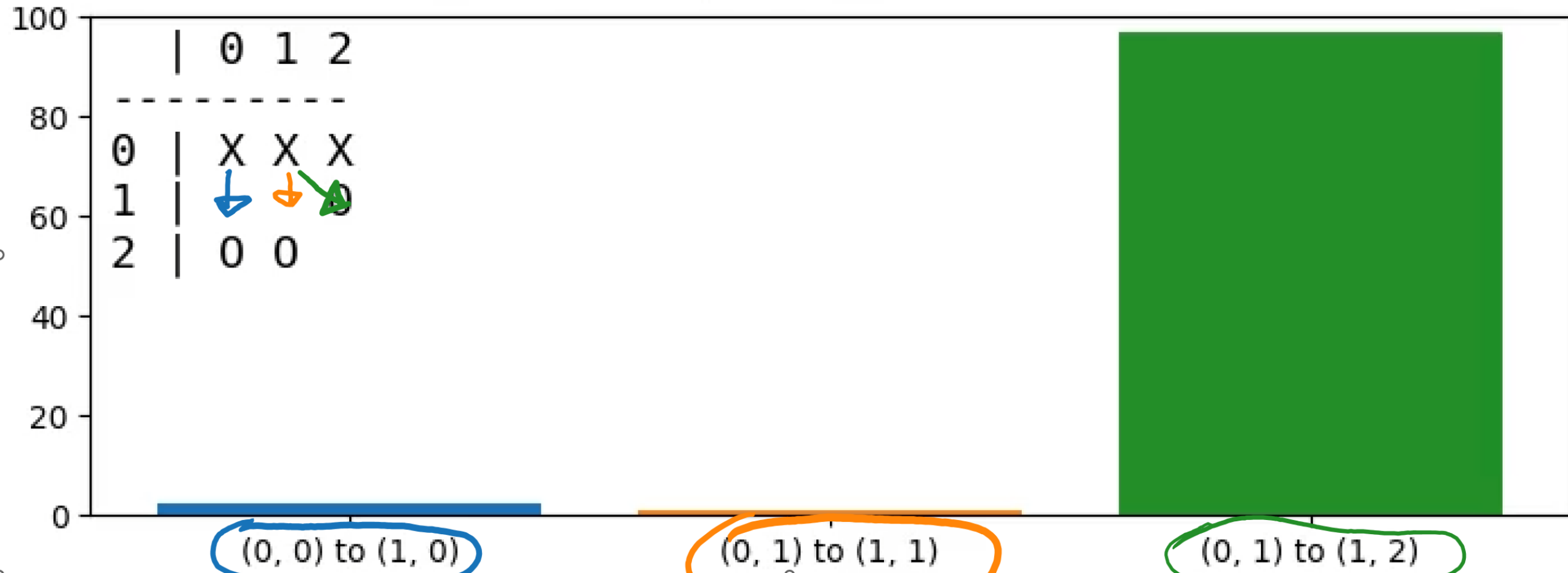
Unsupervised

Reinforcement

Summary

Hexapawn 'Matchbox' Animation

After approx 10000 games, move weights for these boards playing as "X"



Intro

Supervised

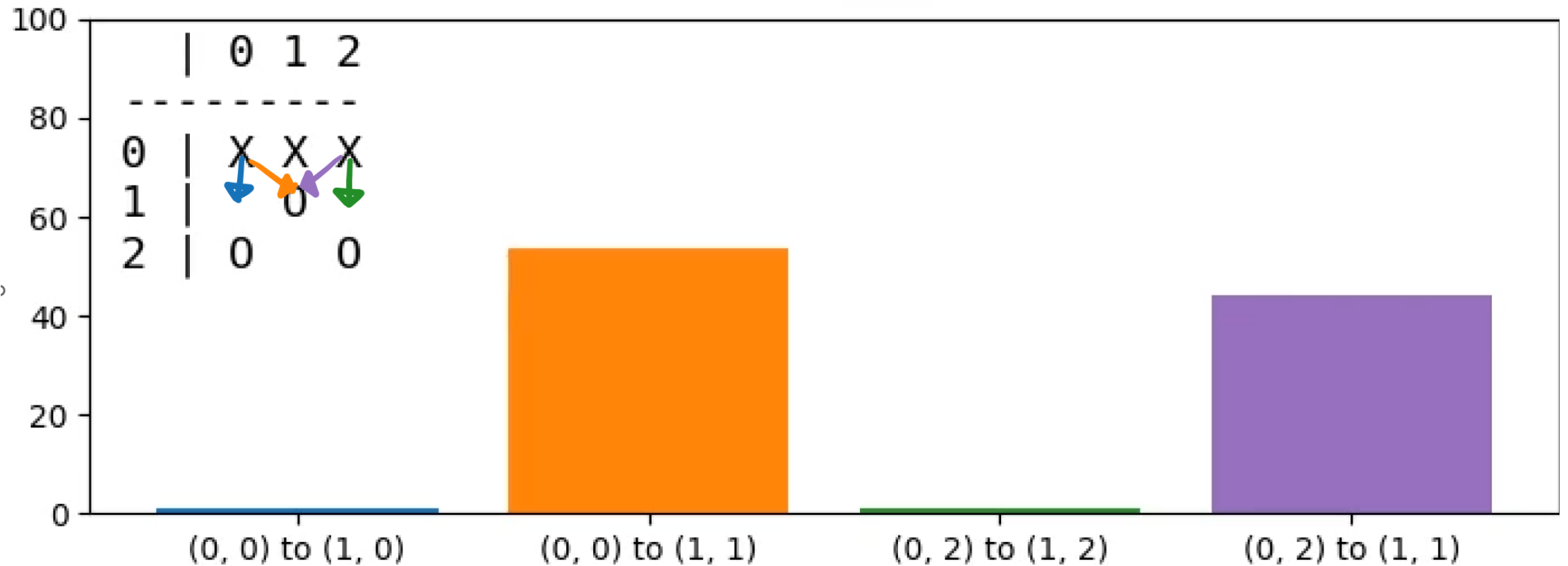
Unsupervised

Reinforcement

Summary

More Hexapawn Matchboxes

After approx 0 games, move weights for these boards playing as "X"



Intro

Supervised

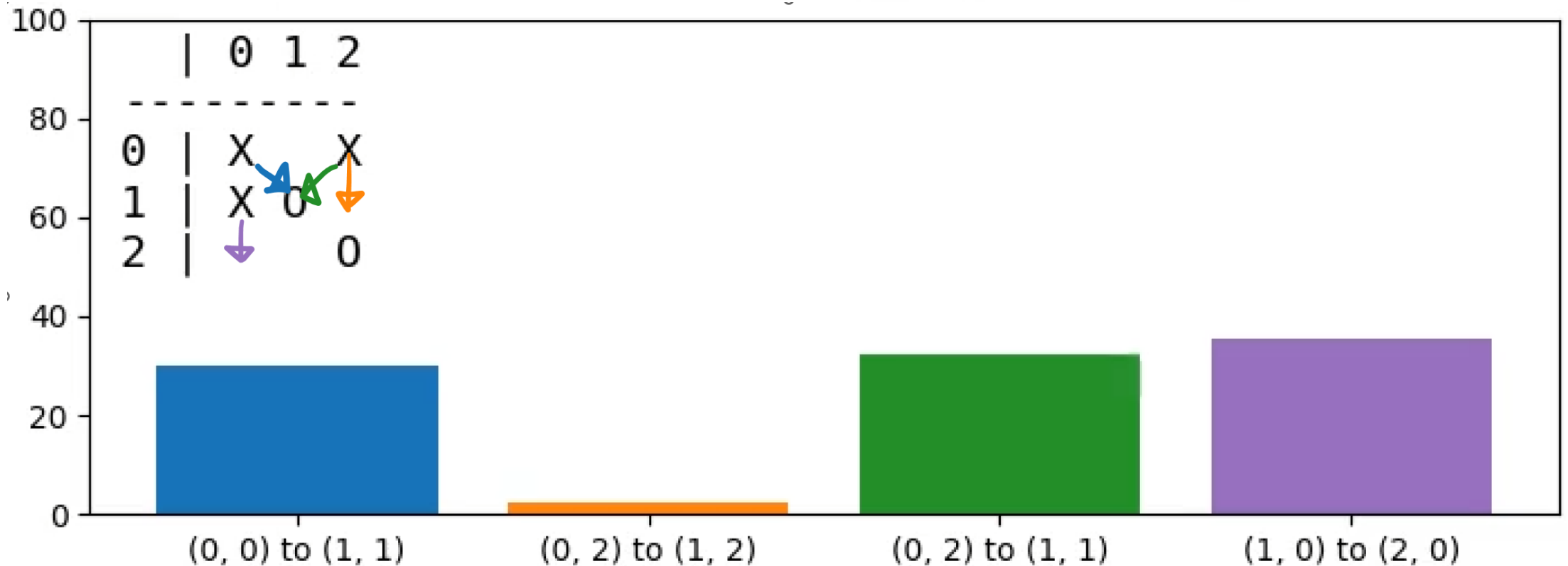
Unsupervised

Reinforcement

Summary

More Hexapawn Matchboxes

After approx 0 games, move weights for these boards playing as "X"



Intro

Supervised

Unsupervised

Reinforcement

Summary

Other games

Chess



Go



Intro

Supervised

Unsupervised

Reinforcement

Summary

Reinforcement learning: Applications



AlphaGo

youtu.be/WXuK6gekU1Y



Self-driving cars

youtu.be/ixloDYVfKA0



Large language models

youtu.be/-dvIK9zt8qM

Intro

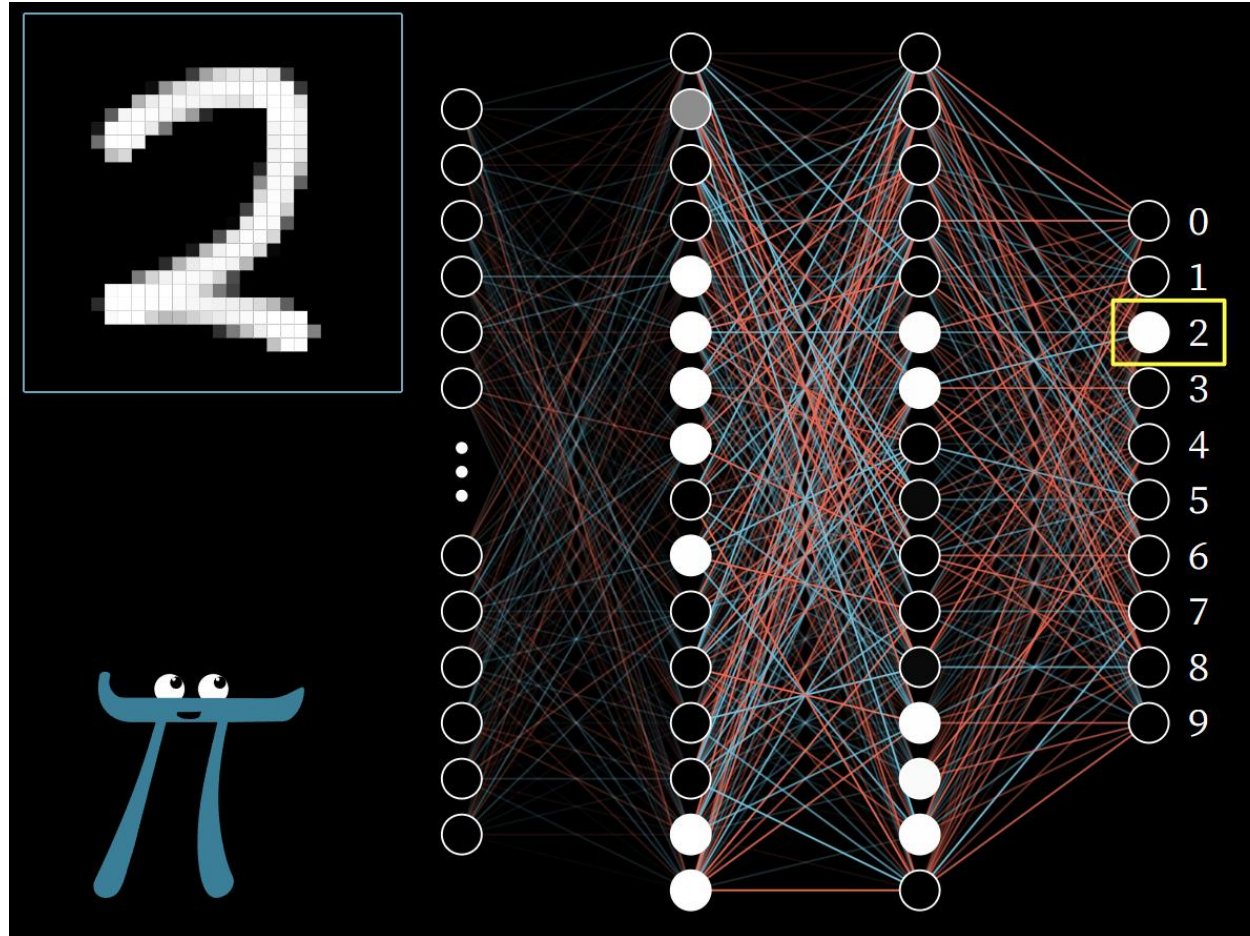
Supervised

Unsupervised

Reinforcement

Summary

Handwritten digits and Neural Networks



<https://www.3blue1brown.com/lessons/neural-networks>

AMSP Maths into AI courses



Maths into AI Student Programme

Maths into Data Science and AI aims to develop students' knowledge of how maths and statistics underpin machine learning and AI. It supports students' future study and careers by developing practical skills in modelling with real data.

EXPLORE THE STUDENT COURSES >>



Maths into AI Teacher Programme

Our professional development aims to develop teachers' understanding of how maths and statistics underpin machine learning and AI, as well as demonstrating how these link to the existing curriculum. We offer an introductory webinar as well as online and blended sustained courses.

EXPLORE THE TEACHER COURSES >>

amsp.org.uk/teachers/maths-into-ai-programme/