

Advanced Machine Learning For Design

Lecture 4 - Machine Learning
for Image Processing (part 1)

Module 2

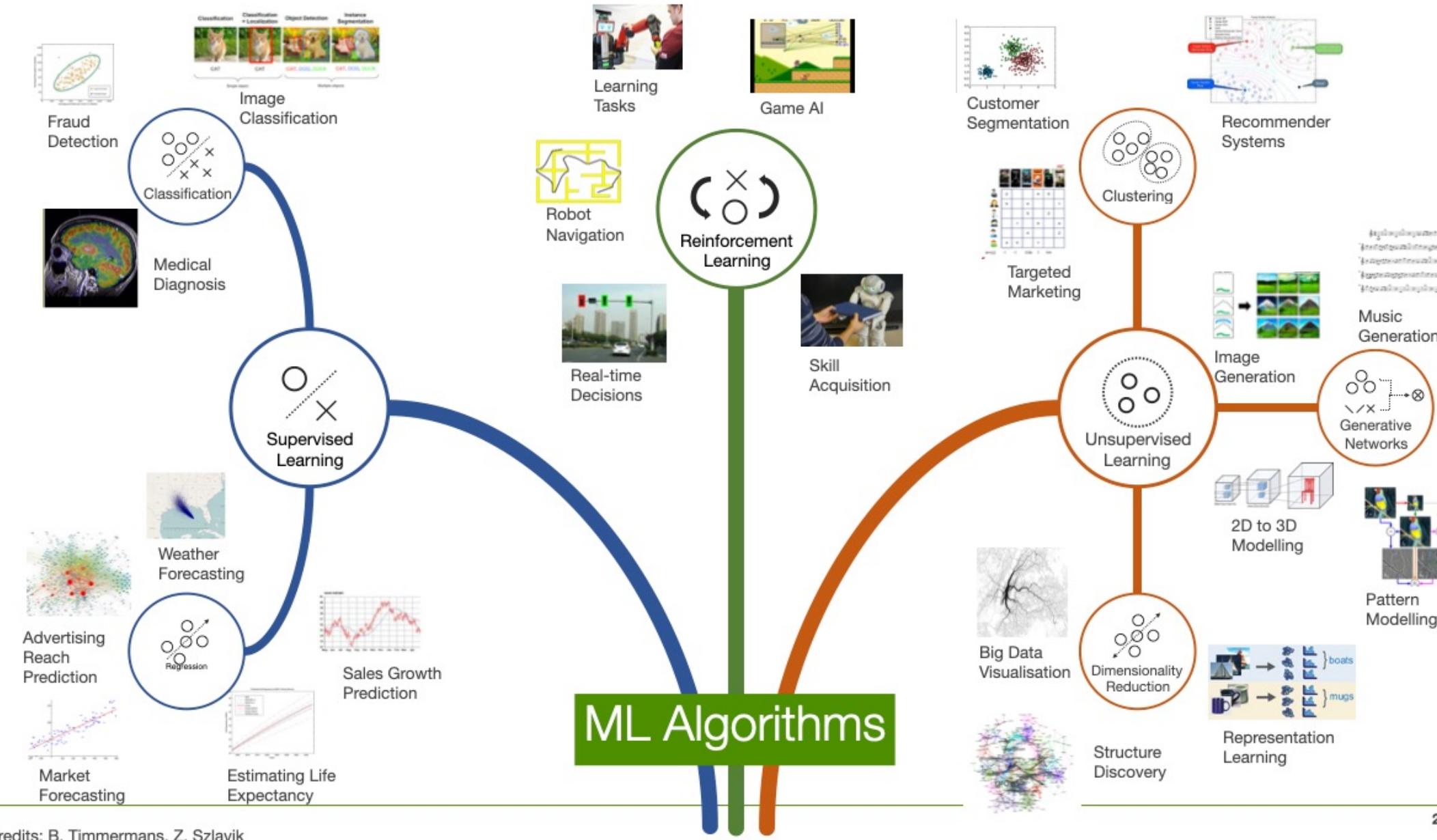
Evangelos Niforatos
12/10/2022

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<https://aml4design.github.io/>

Admin

Recap on ML/AI.

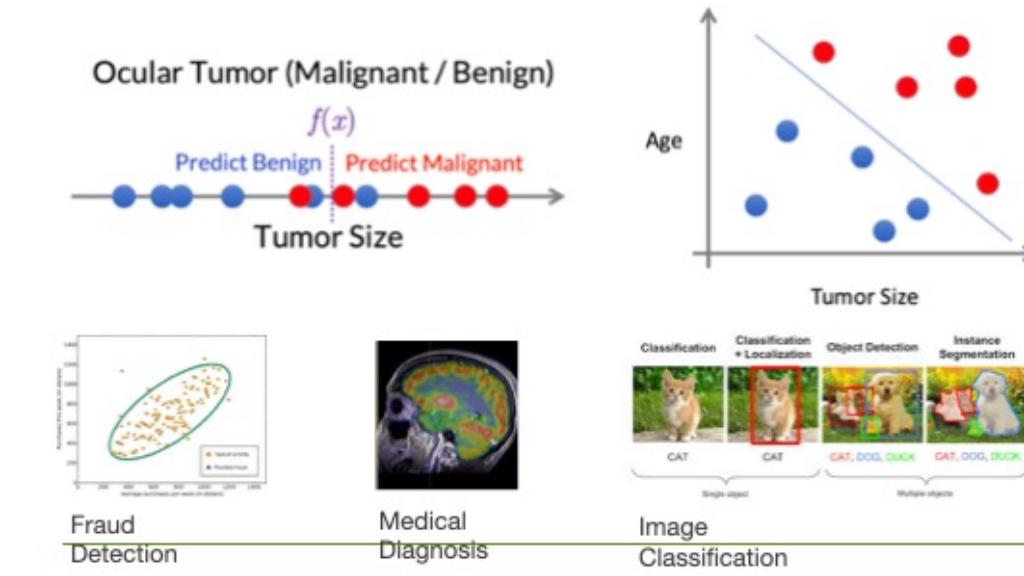




Classification / Regression

Classification

- Learn to output a **category** label
 - Binary (e.g. Spam / not Spam, Cat / not cat)
 - Multi-class (e.g. cat, dog, bird)



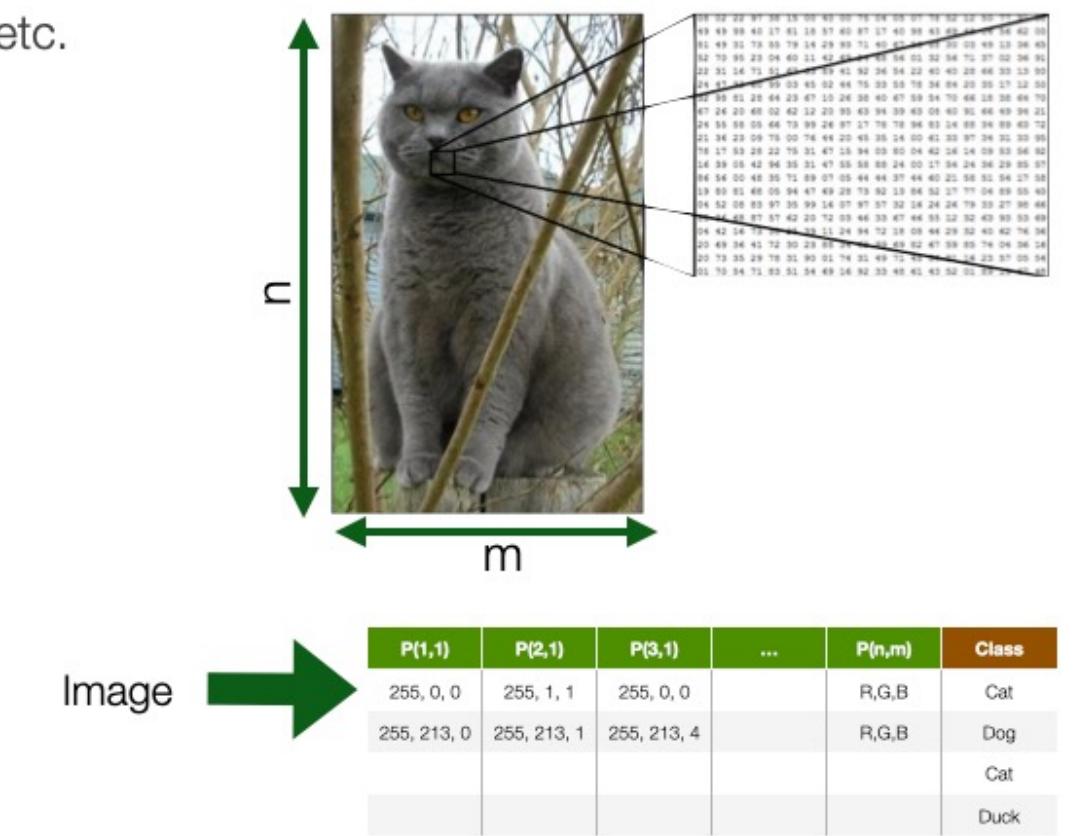
Regression (prediction)

- Learn to guess one or more numbers
 - e.g. value of a share, number of stars in a review



Images

- Visual content acquired through cameras, scanners, etc.
- Each pixel in an image is a feature
 - But spatially and geometrically organised
 - e.g. edges, corners
- Feature values are numerical values across channels
 - e.g. R,G,B
- Dimensionality $\rightarrow n \times m$

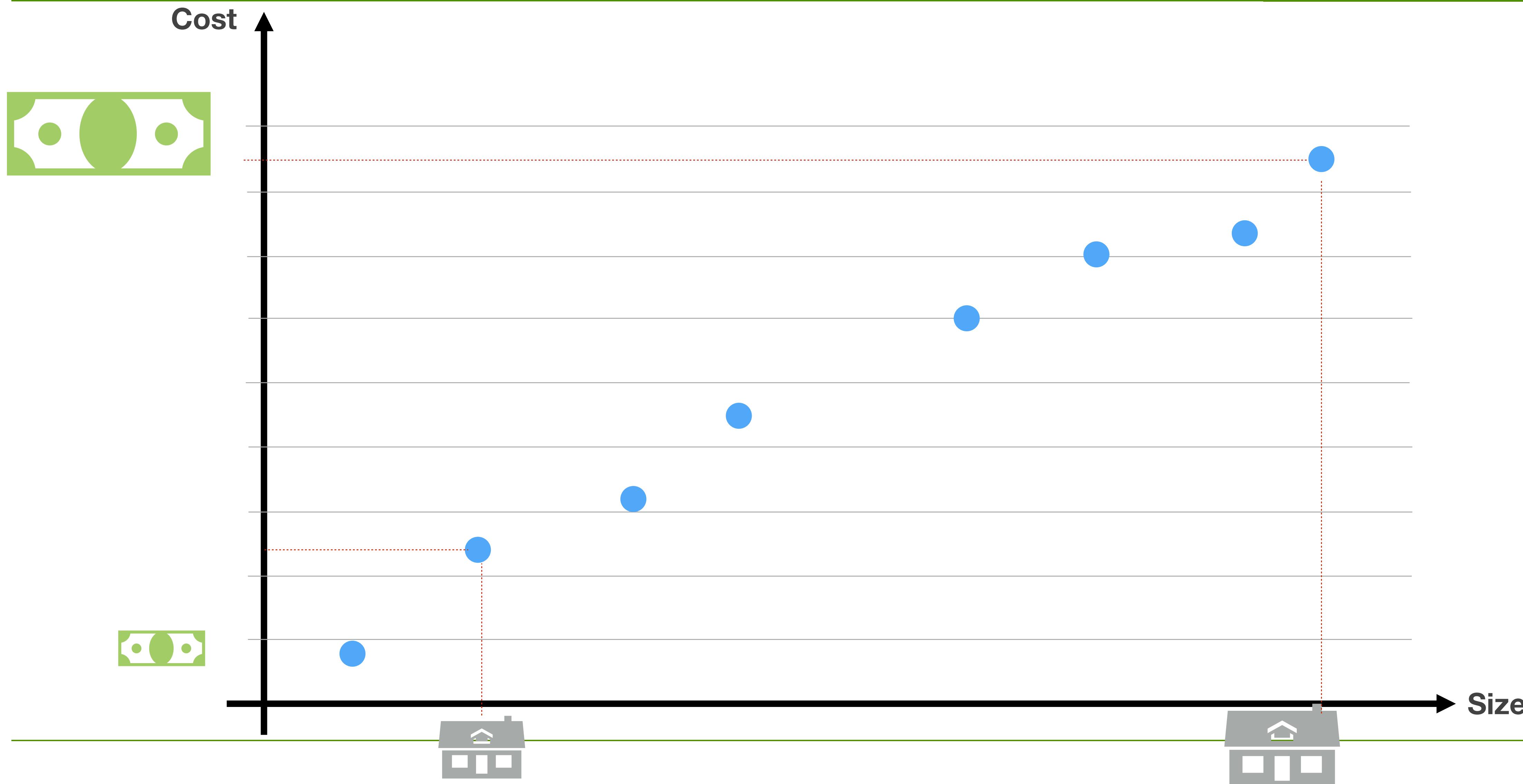


More in Module 2

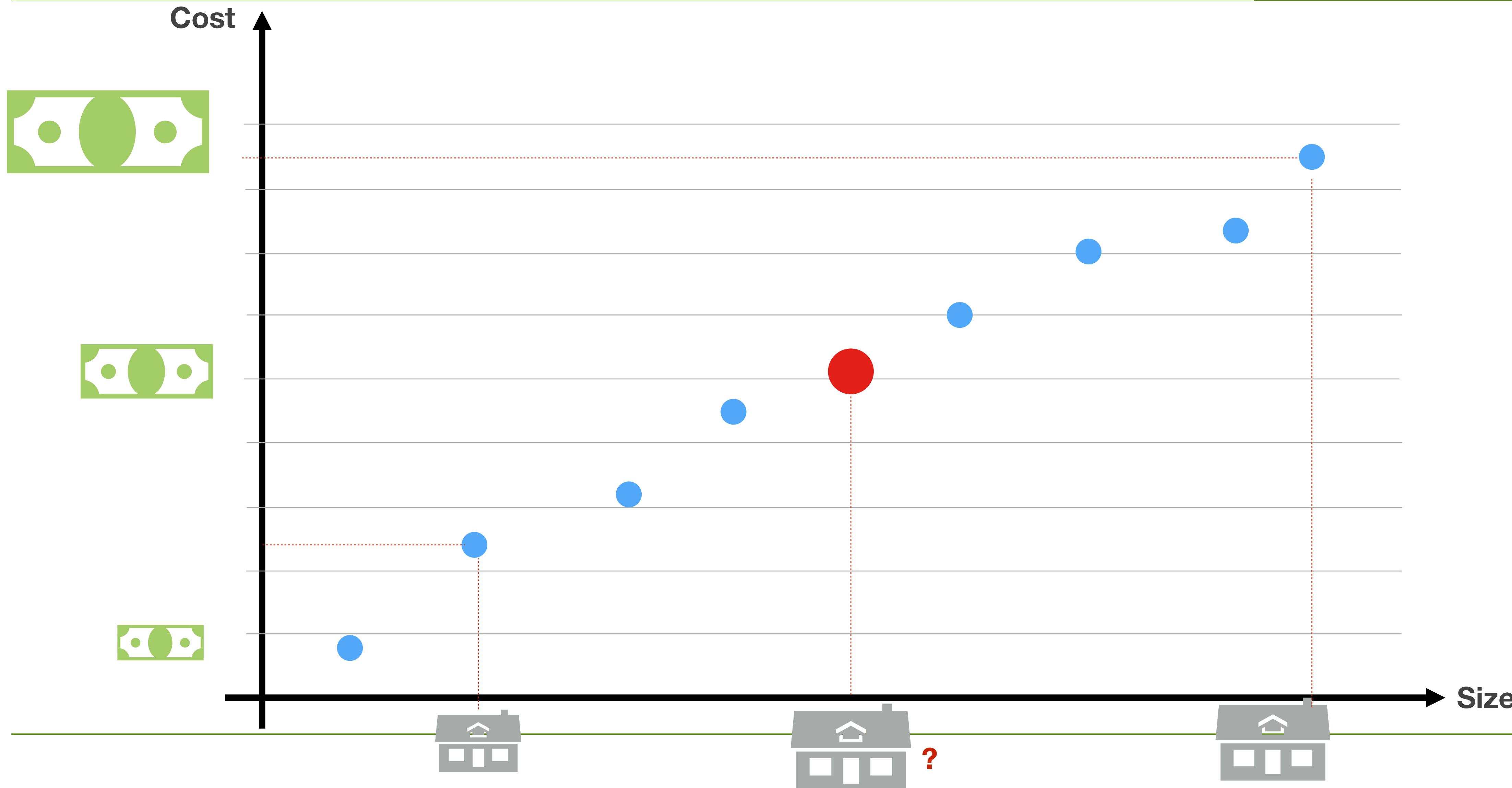
A bit more on regression and classification

And your 2nd contact with (deep) neural networks

Linear Regression /1



Linear Regression /2



Linear Regression

Intercept (parameter, or **bias**)

Slope (parameter, or **weight**)

Dependent Variable

Cost = $b + w^*Size$

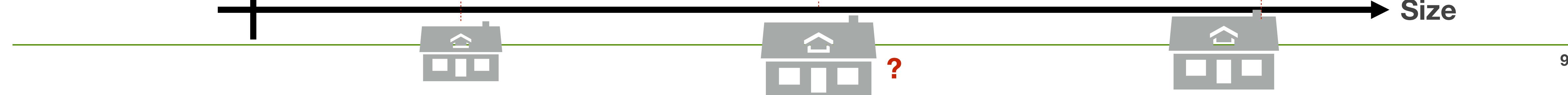


Cost



Independent Variable

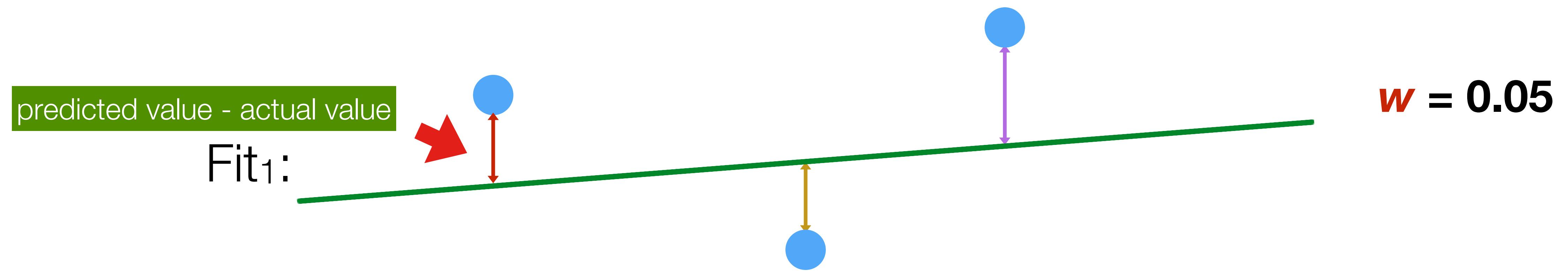
Size



Cost = w^*Size

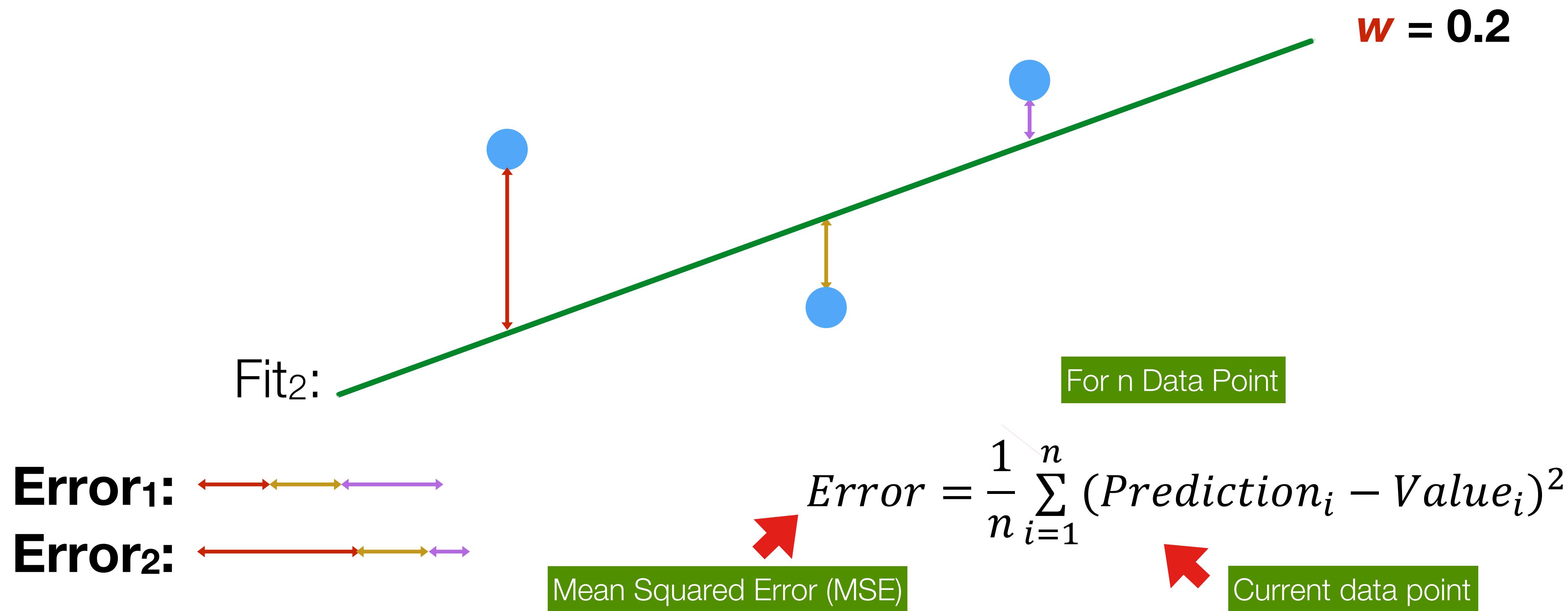


$$\text{Cost} = \mathbf{w}^* \mathbf{Size}$$

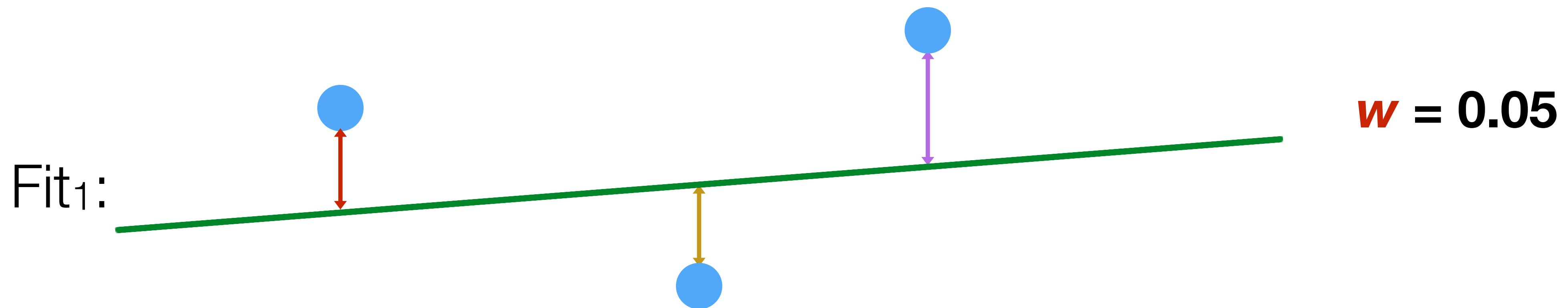


Error₁: ←→

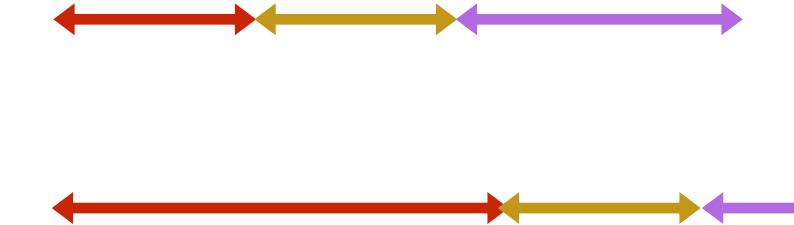
$$\text{Cost} = \mathbf{w}^* \mathbf{Size}$$



$$\text{Cost} = \mathbf{w}^* \mathbf{Size}$$



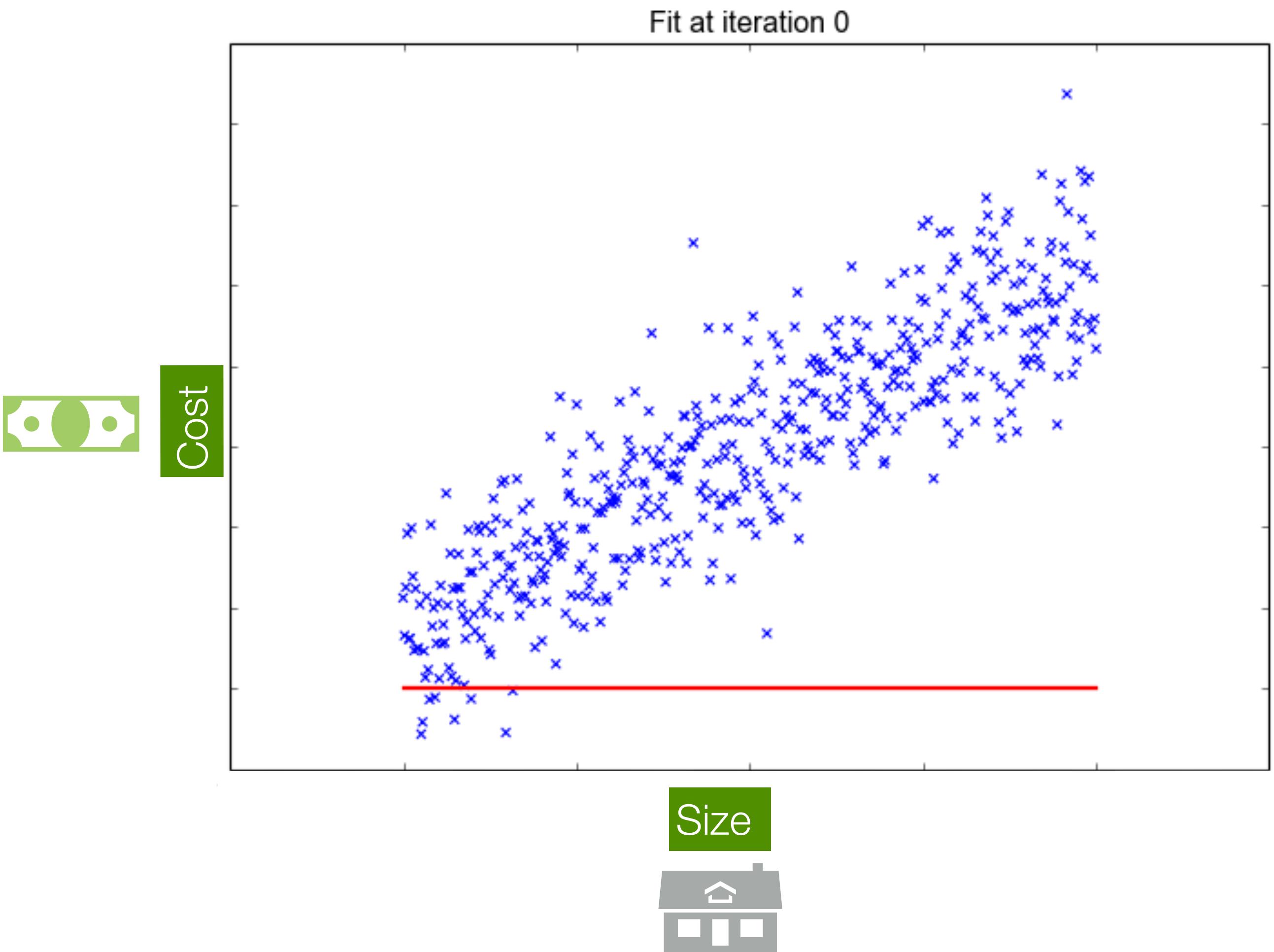
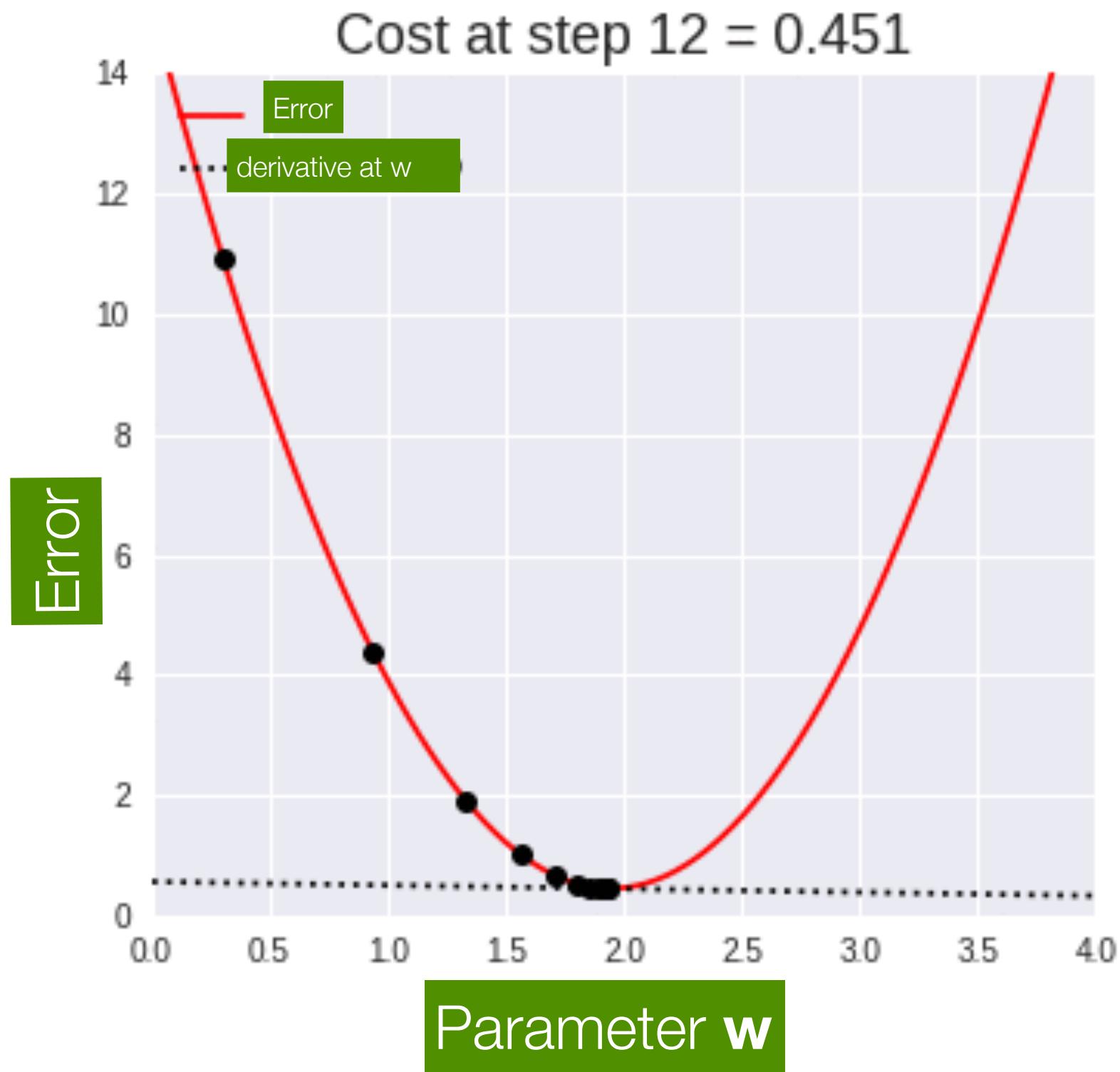
Error₁:
Error₂:



Fit₁ is a better fit on the training data than Fit₂
We select $\mathbf{w} = 0.05$

Finding the best parameter values (training the model)

■ Gradient descent



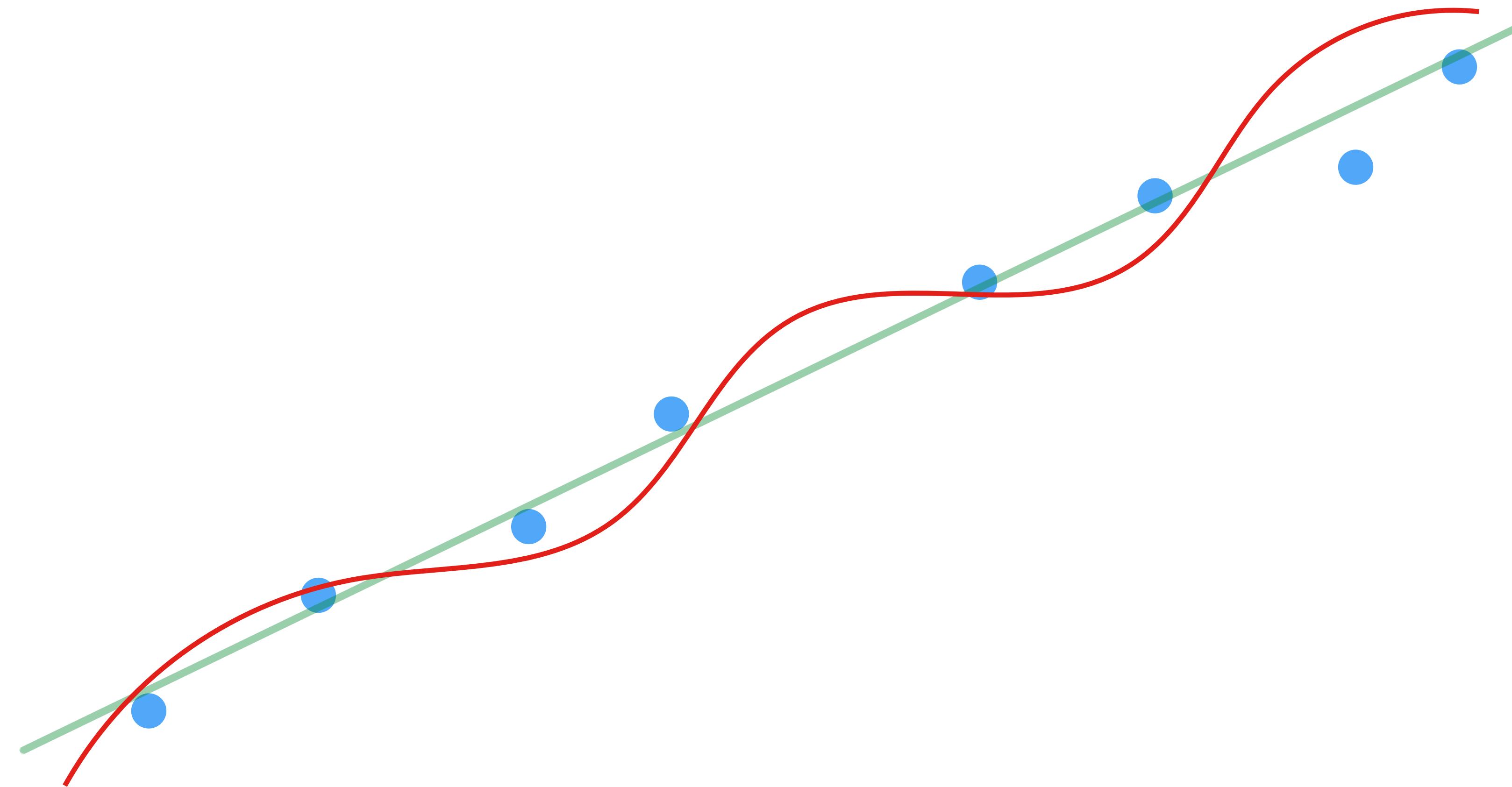
■ Hyperparameters

- **Learning Rate**: “speed” of descent
- **Epochs**: max number of steps

Polynomial Regression

Nth degree polynomial

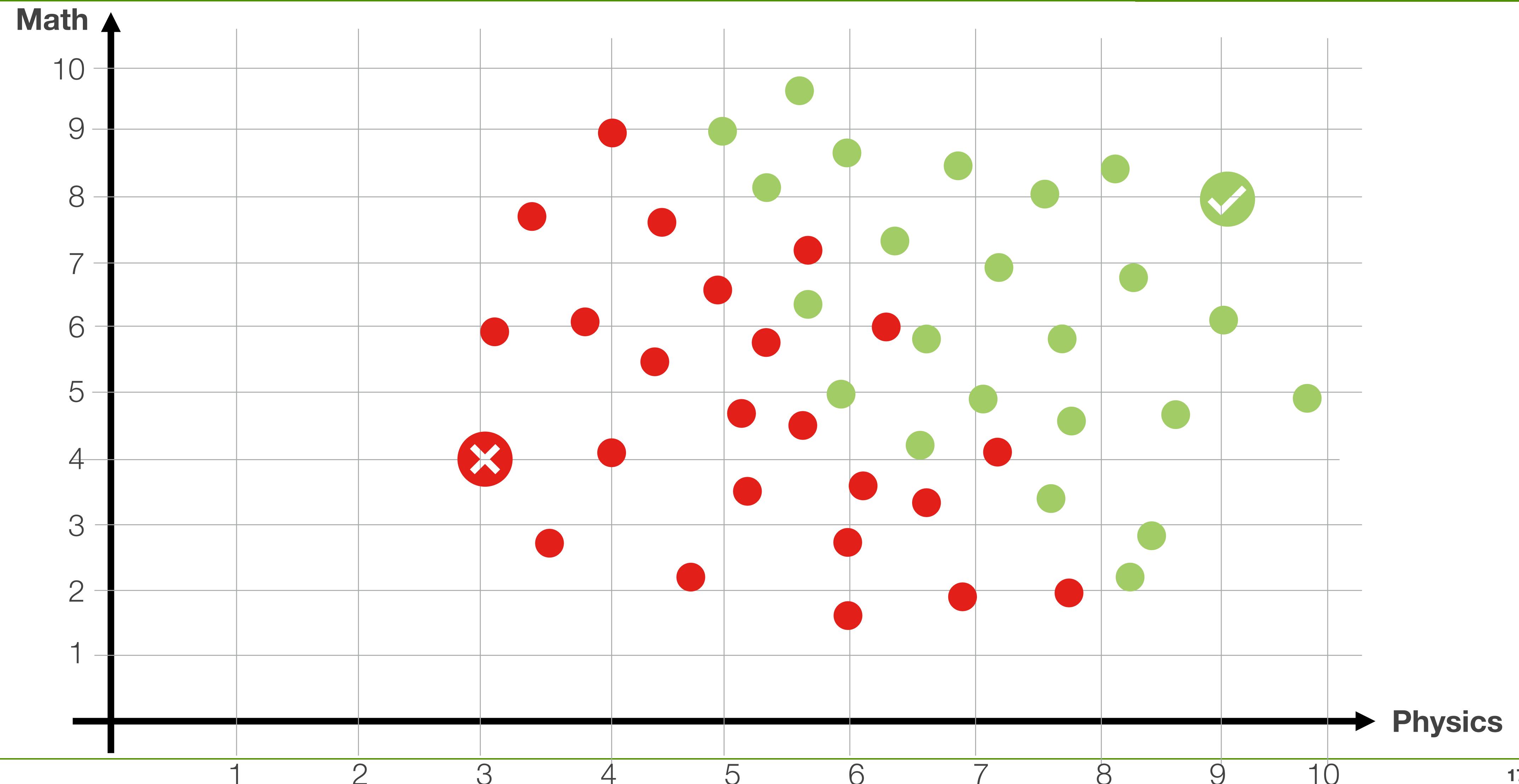
$$\rightarrow \text{Cost} = b + w_1 \text{ Size} + w_2 \text{ Size}^2 + \dots + w_n \text{ Size}^n$$



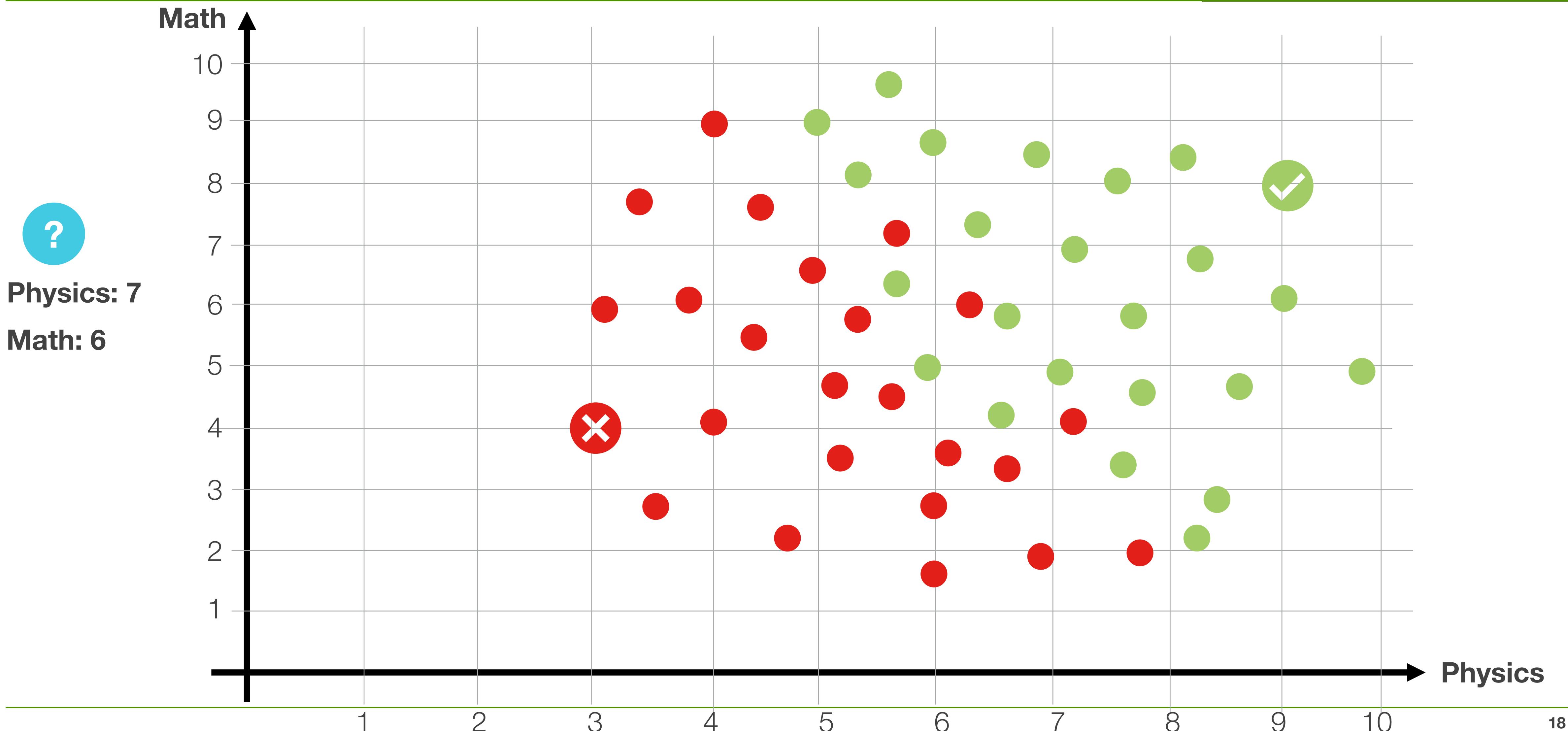
Classification



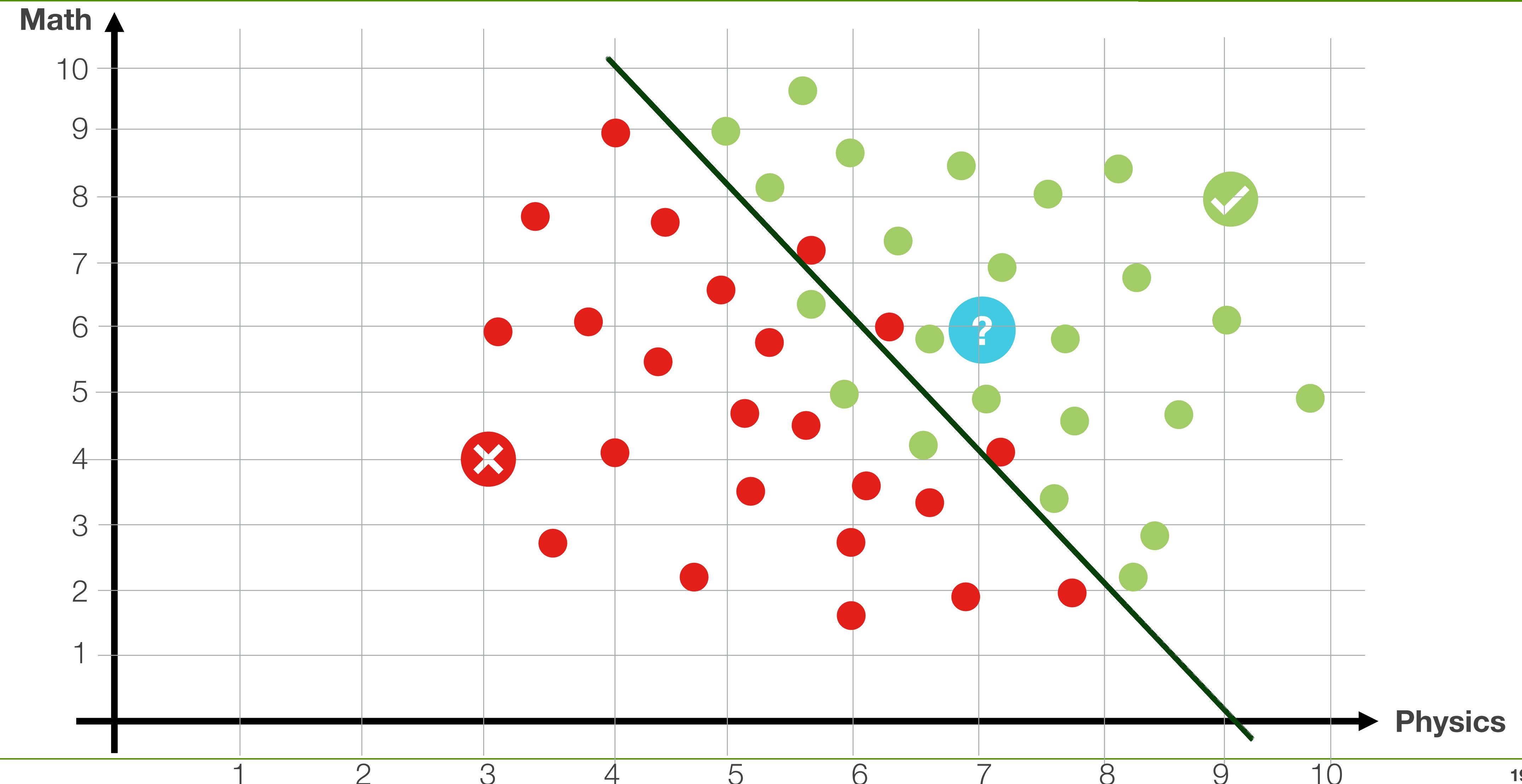
Classification

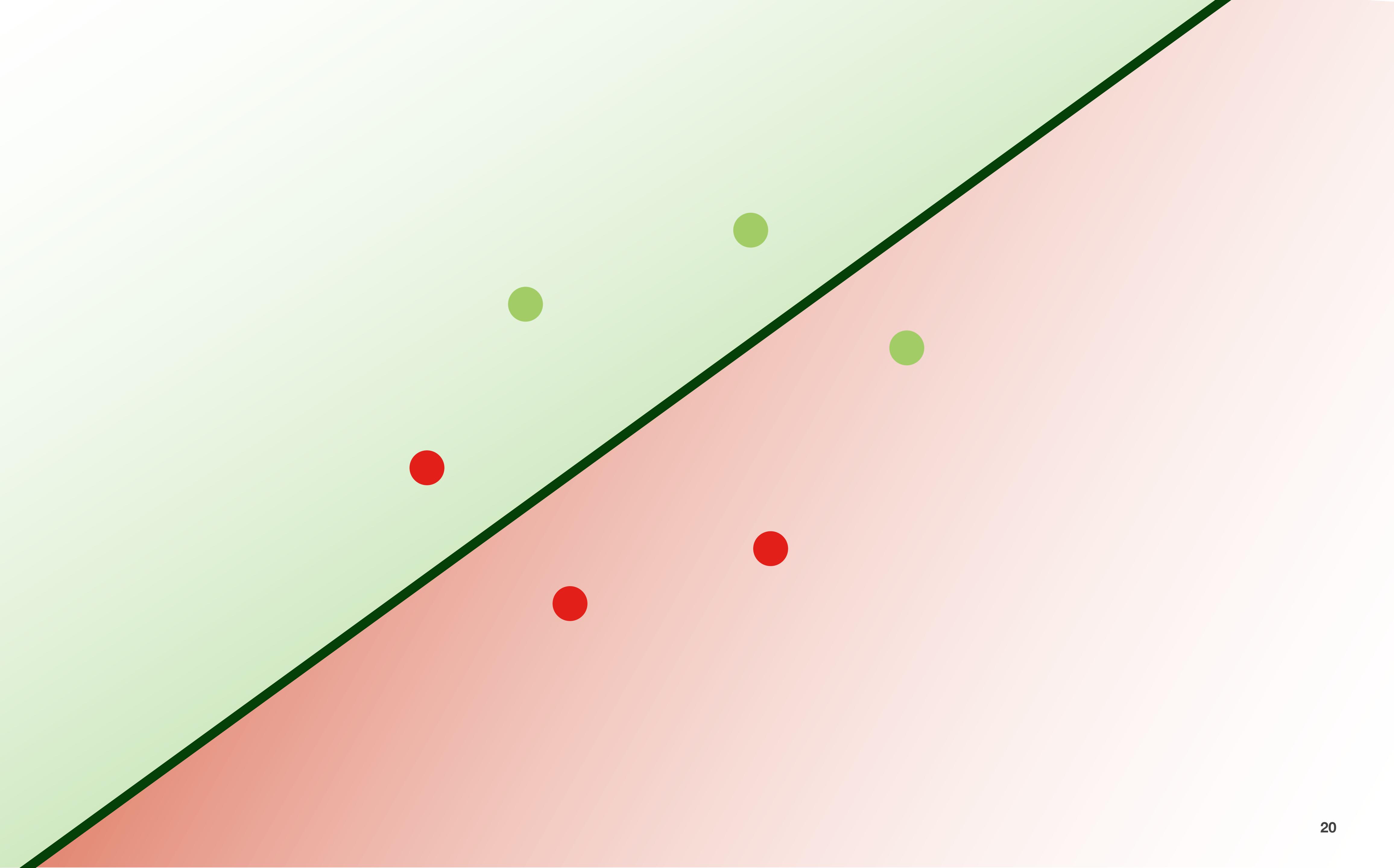


Classification



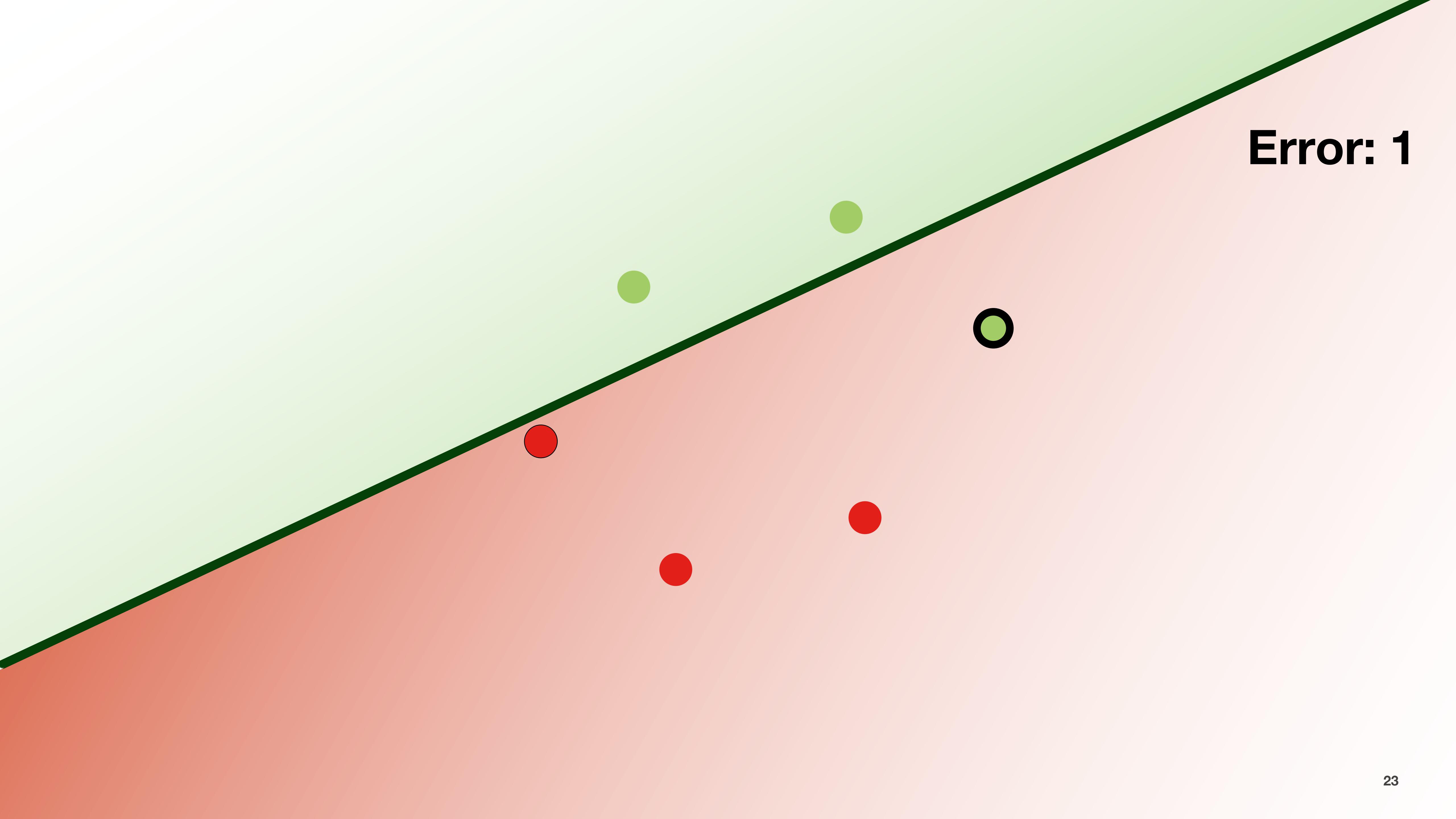
Logistic Regression





Error: 2

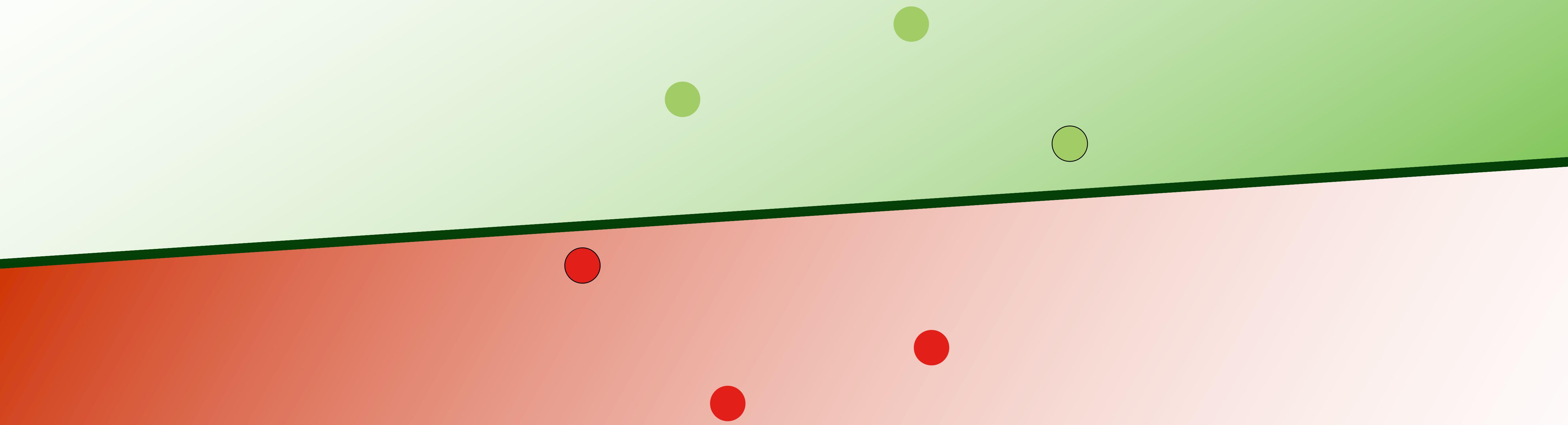
Error: 2



A scatter plot illustrating a linear classification model. The background is divided into two regions by a diagonal green line with a black border. The upper-right region is light green, while the lower-left region is light orange. There are six data points: one green point in the green region, three red points in the orange region, and two black-outlined green points. The text "Error: 1" is located in the top right corner.

Error: 1

Error: 0



Math

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Physics

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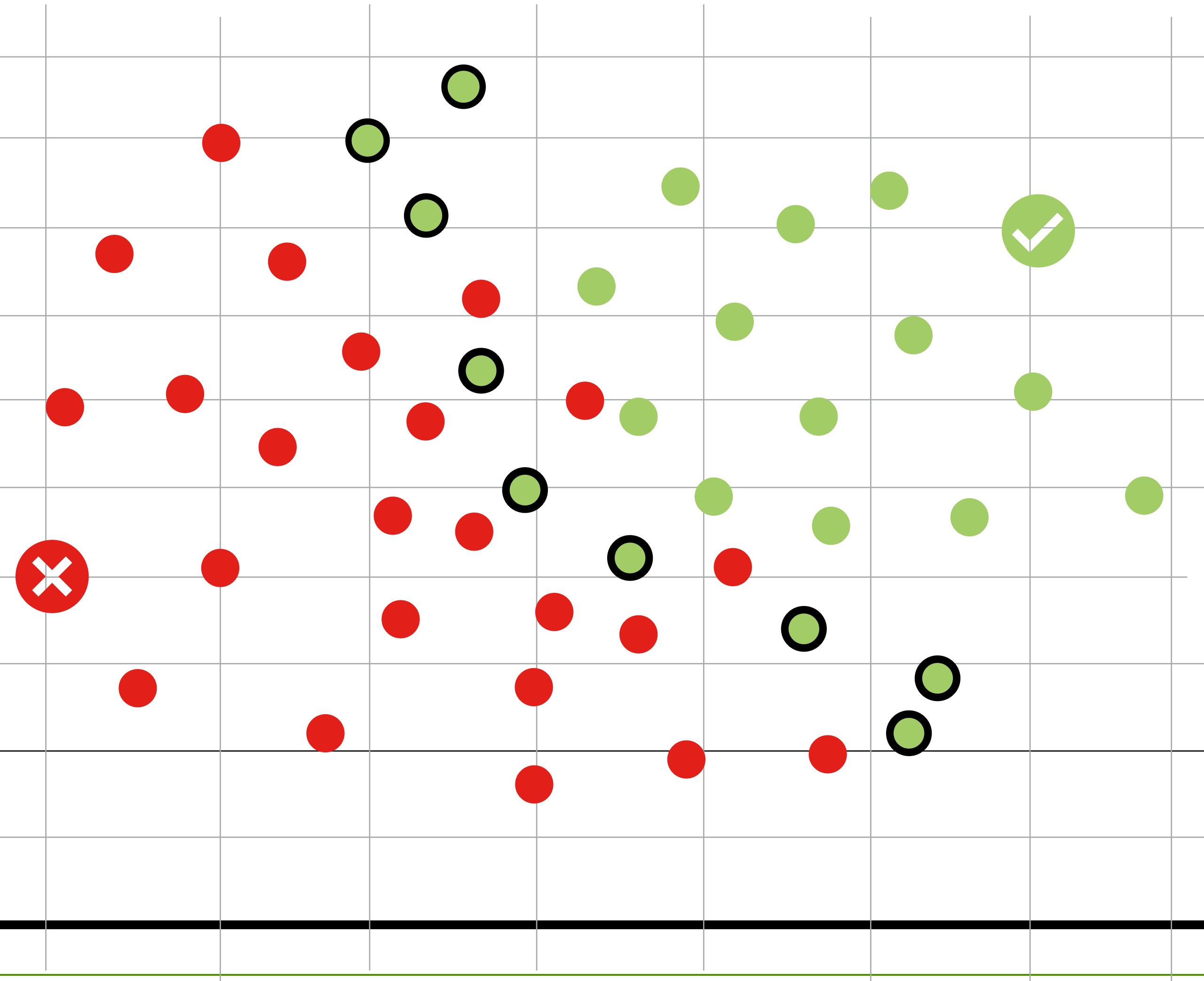
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Math

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Test

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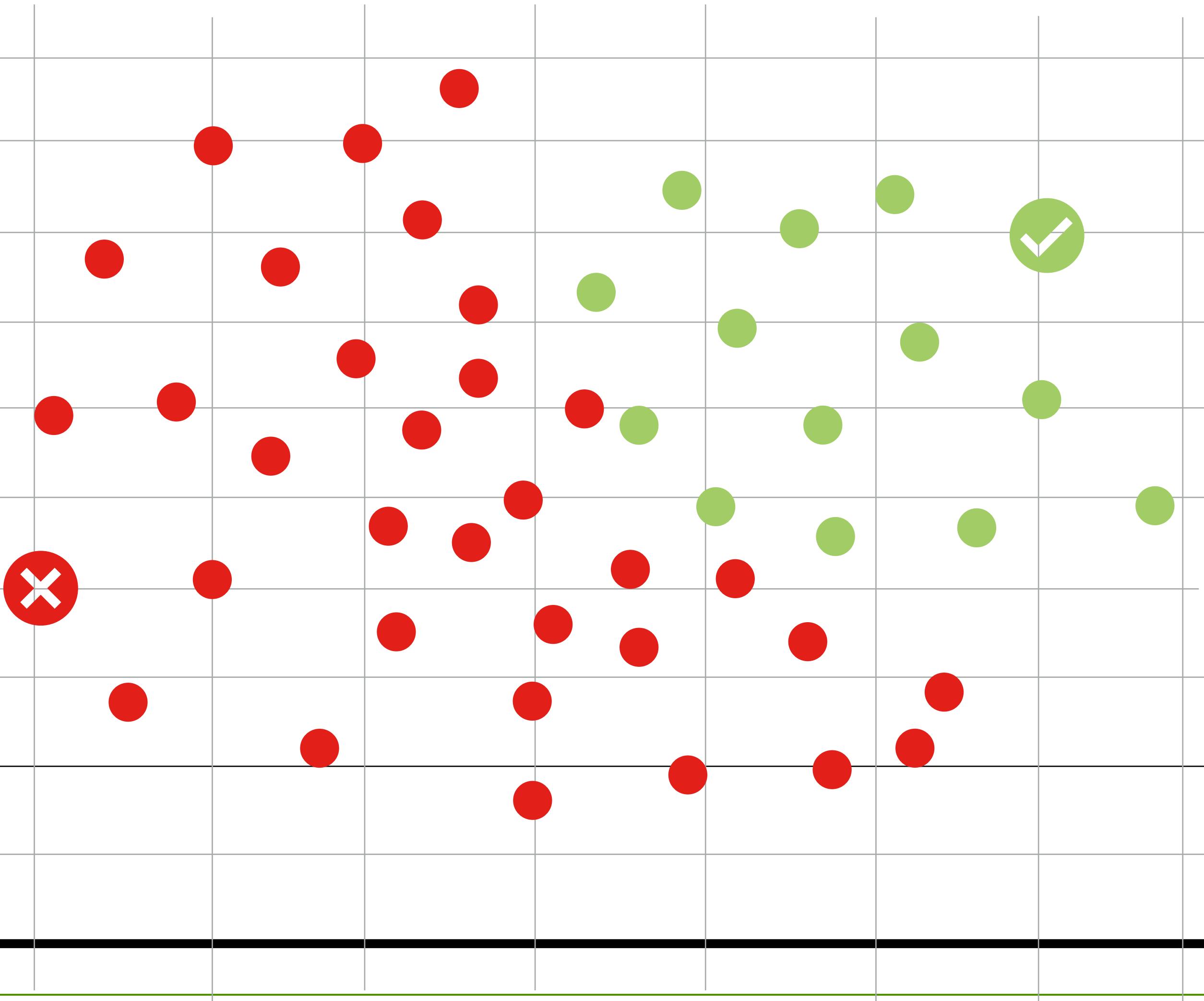
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Math

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Test

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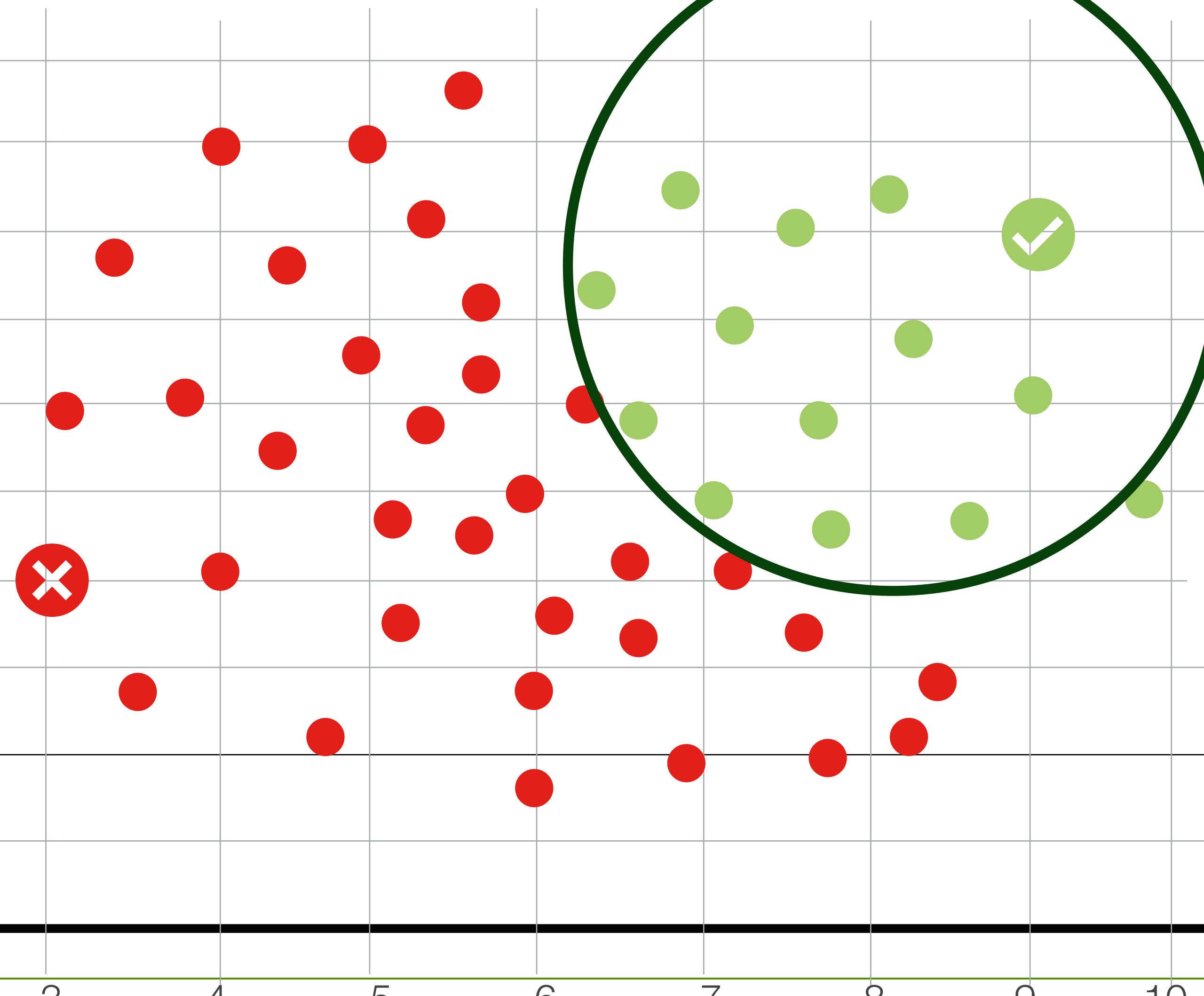
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Math

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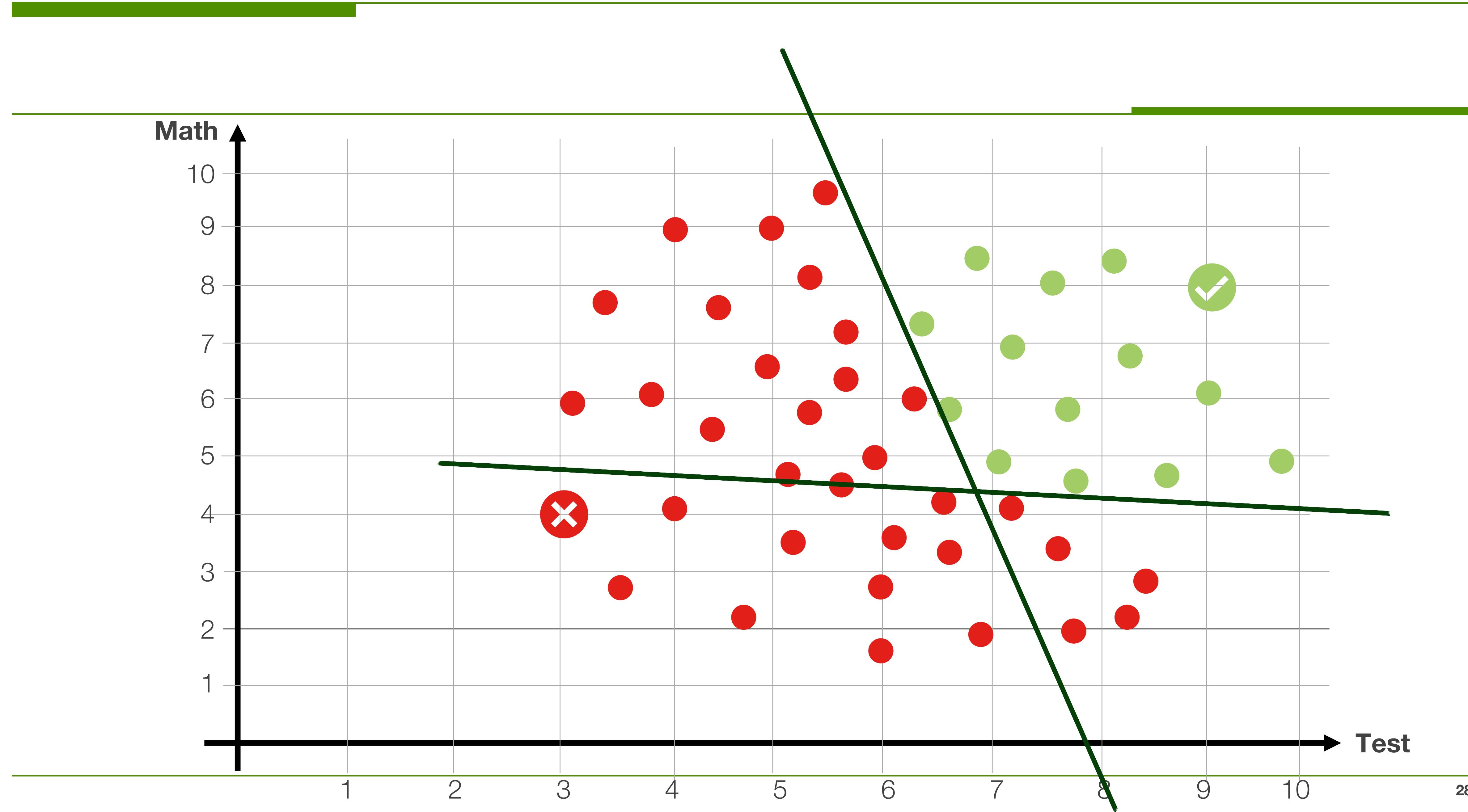
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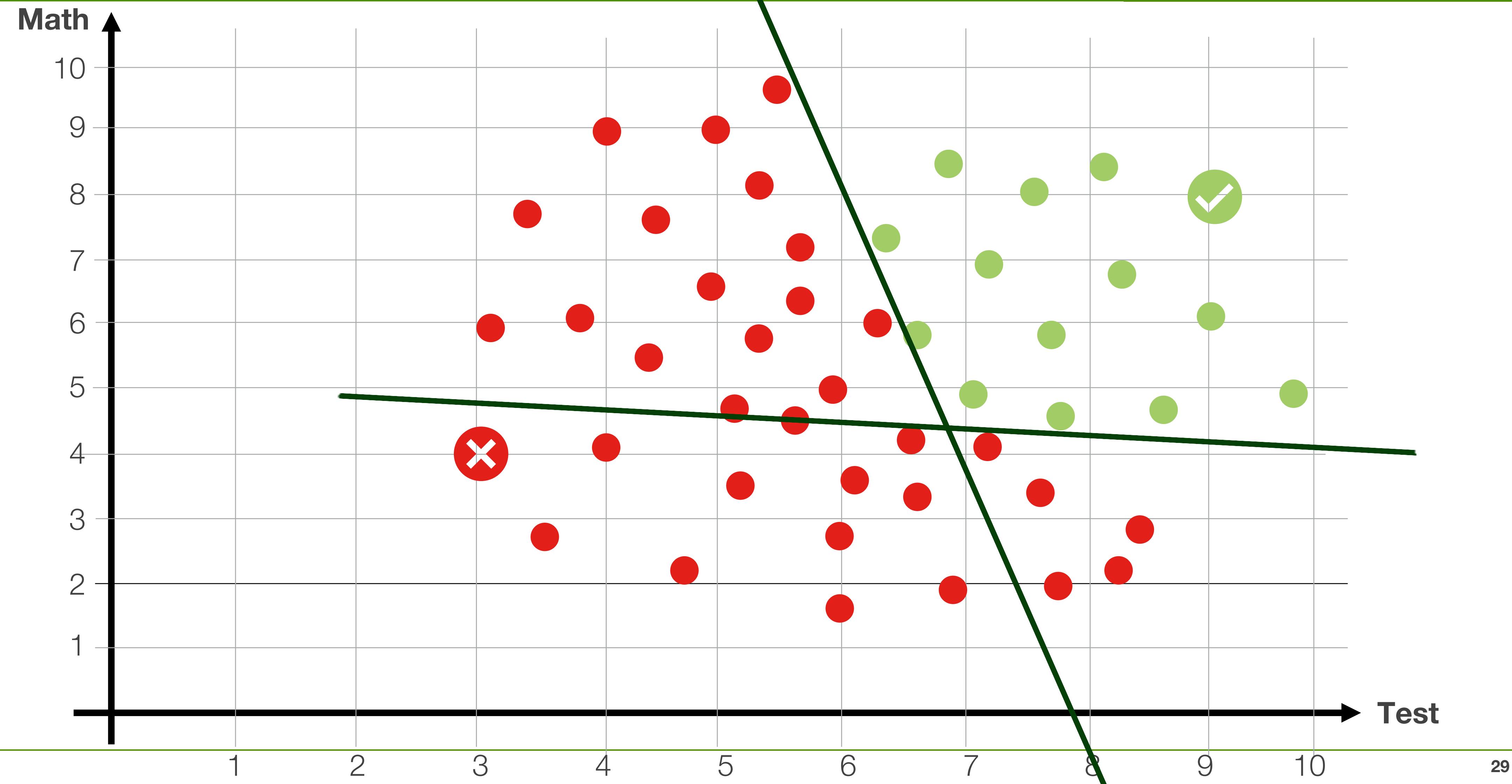
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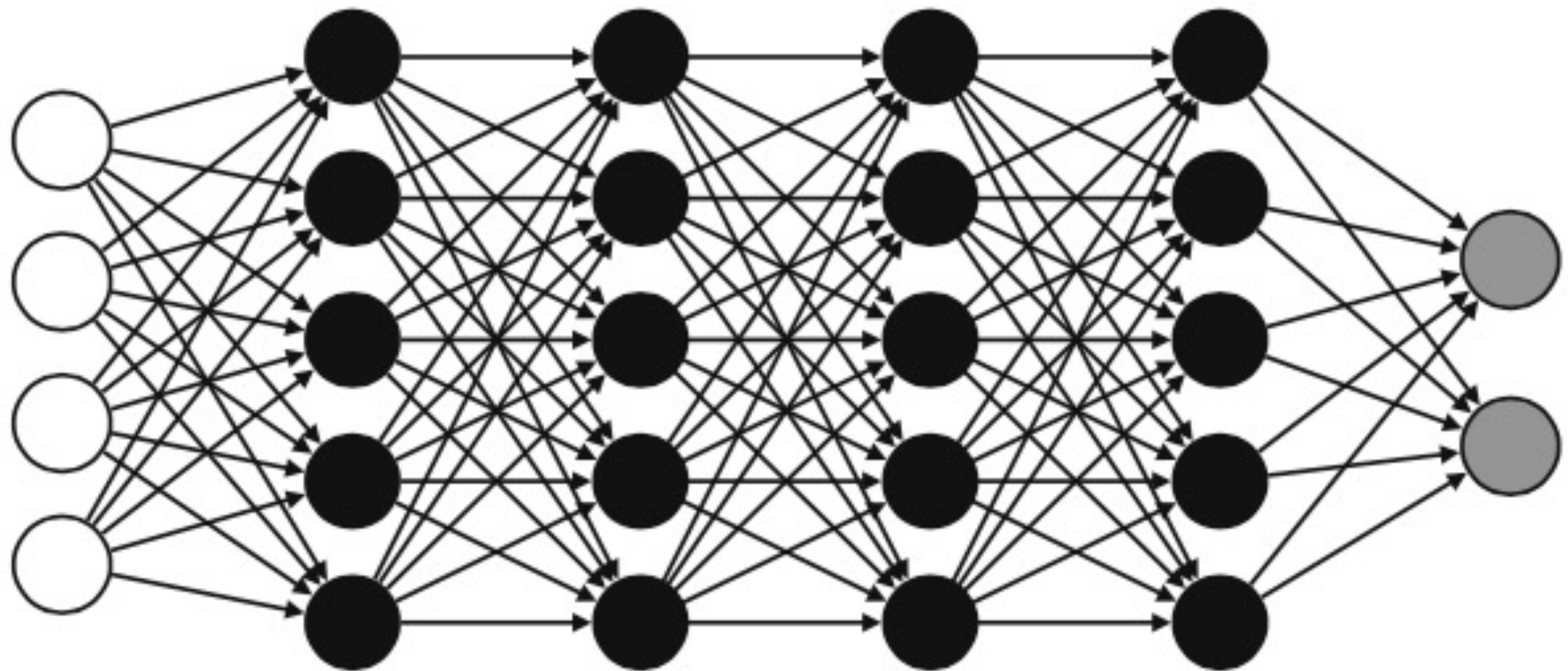
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Test

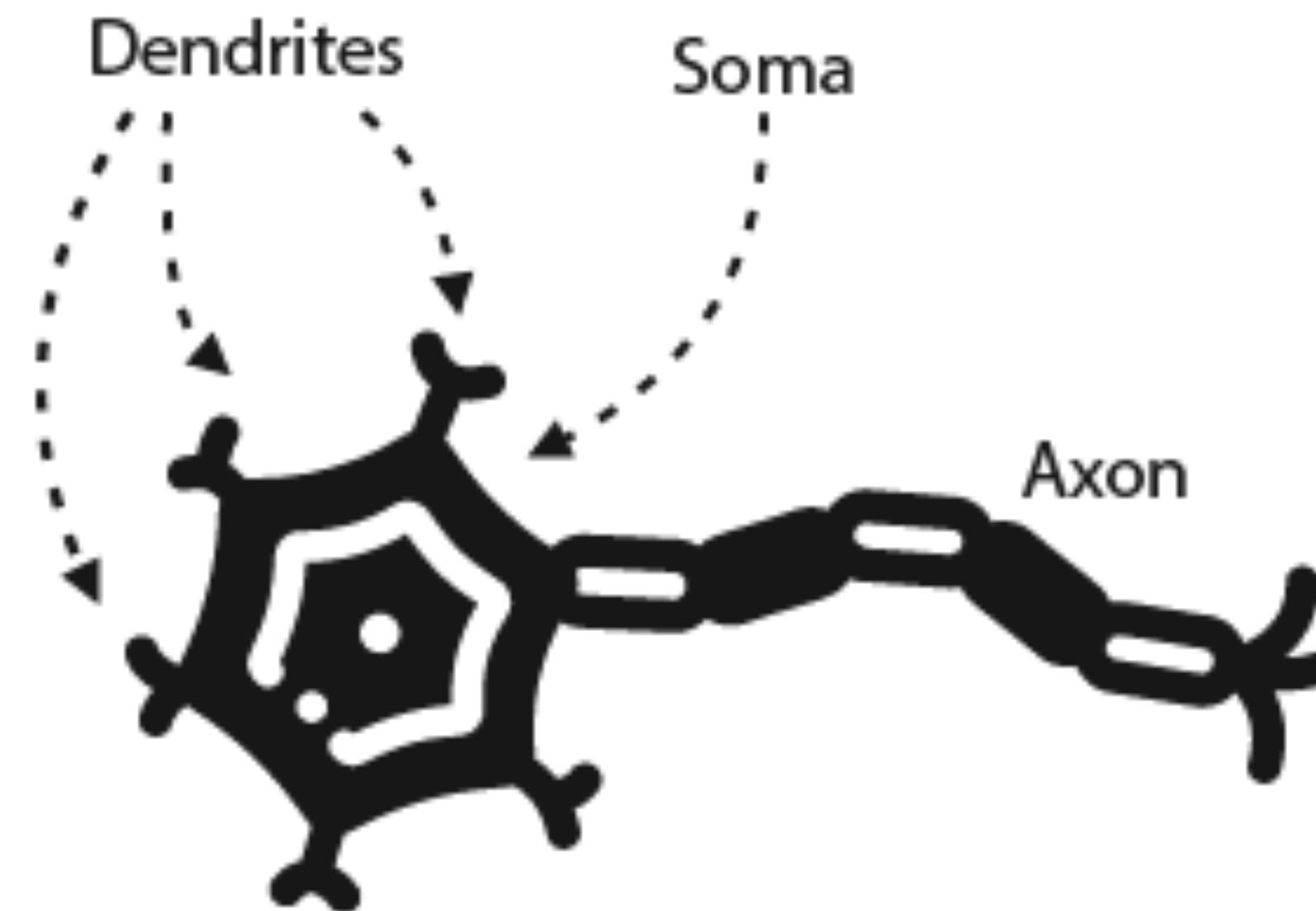


Neural Network

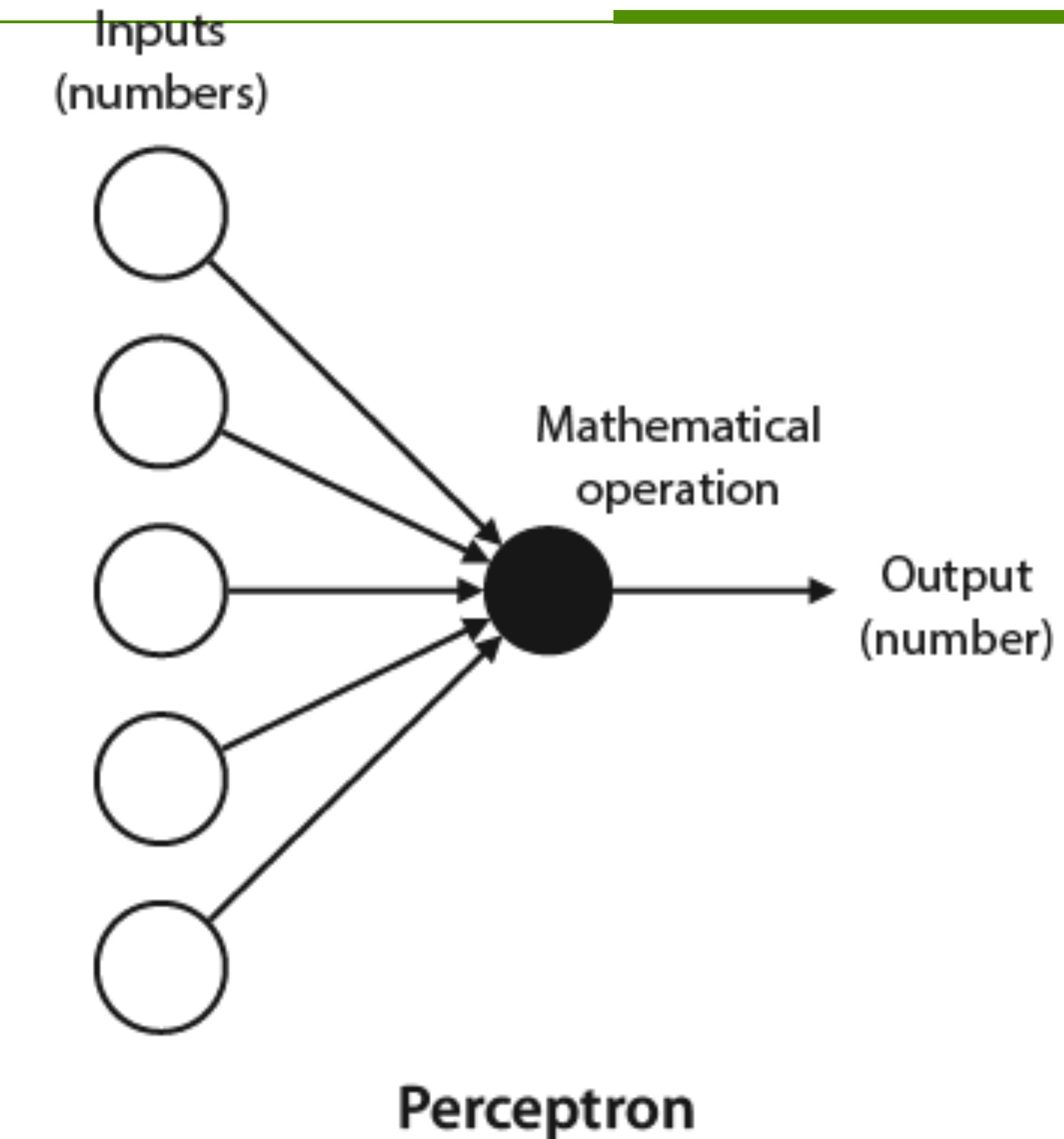




Neural Network

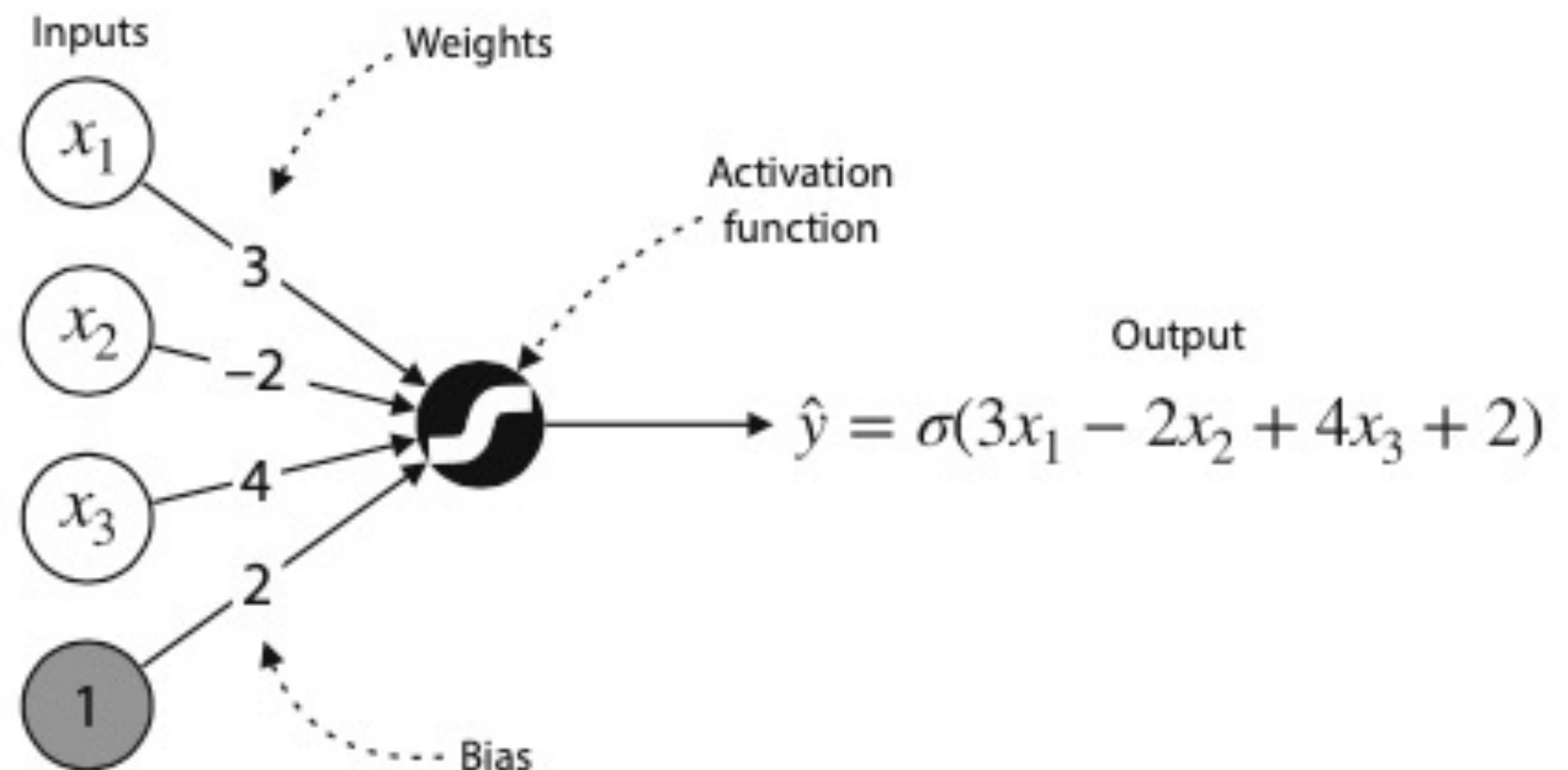


Neuron



Perceptron

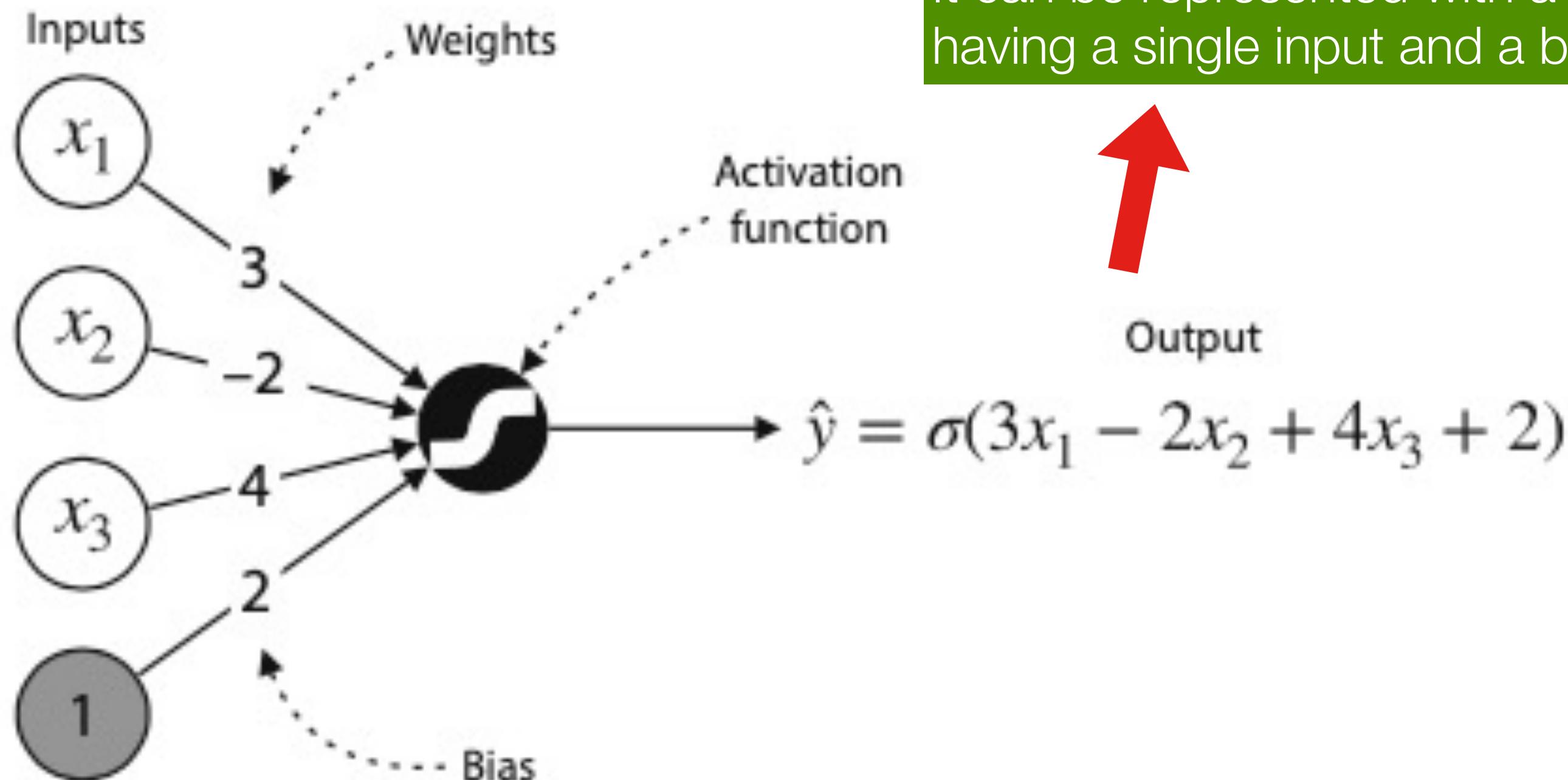
Perceptron



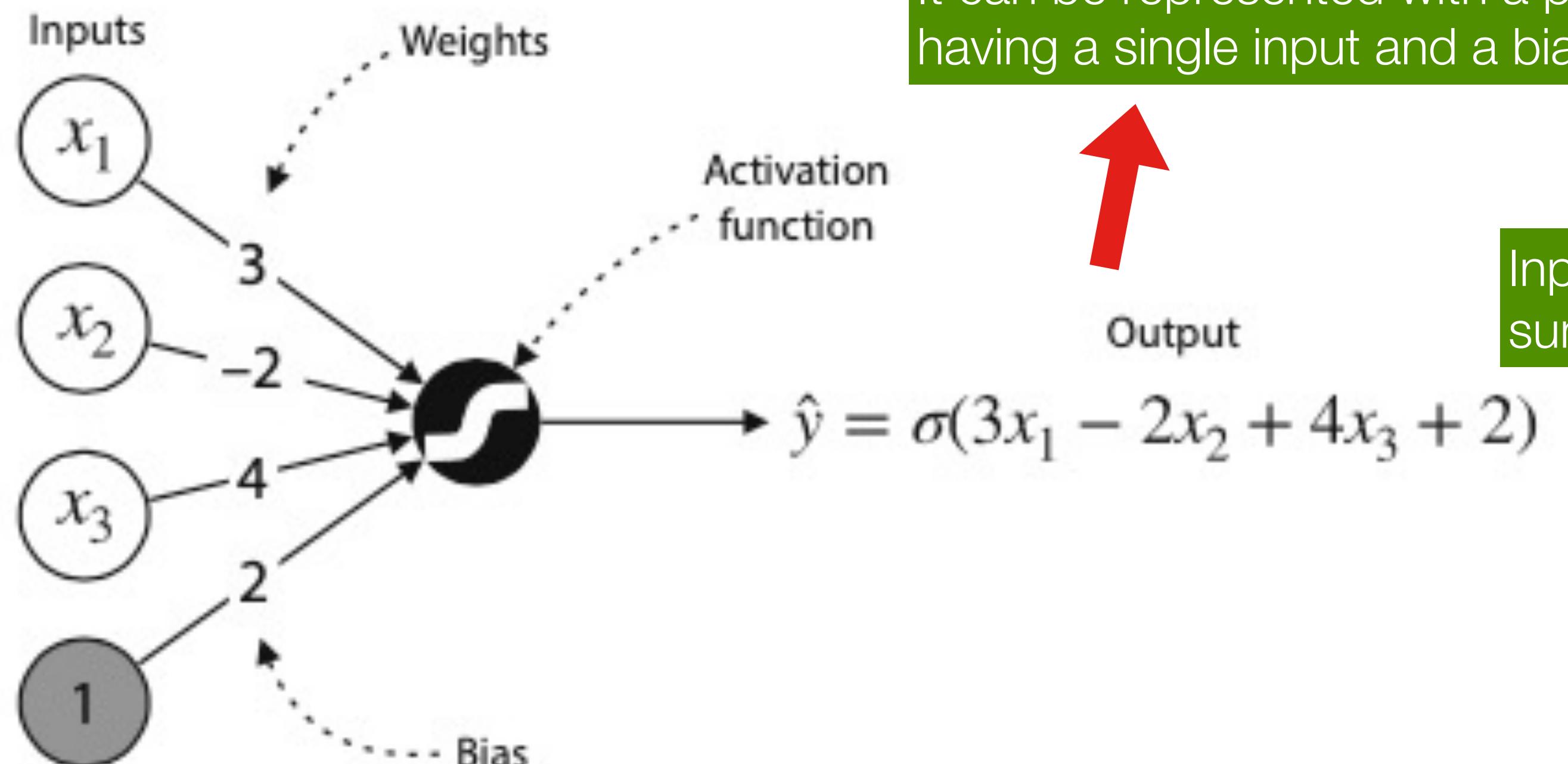
Perceptron

$$\text{Cost} = \mathbf{b} + \mathbf{w}^* \mathbf{Size}$$

Remember our linear regression function?
It can be represented with a perceptron
having a single input and a bias



Perceptron



$$\text{Cost} = a + w^* \text{Size}$$

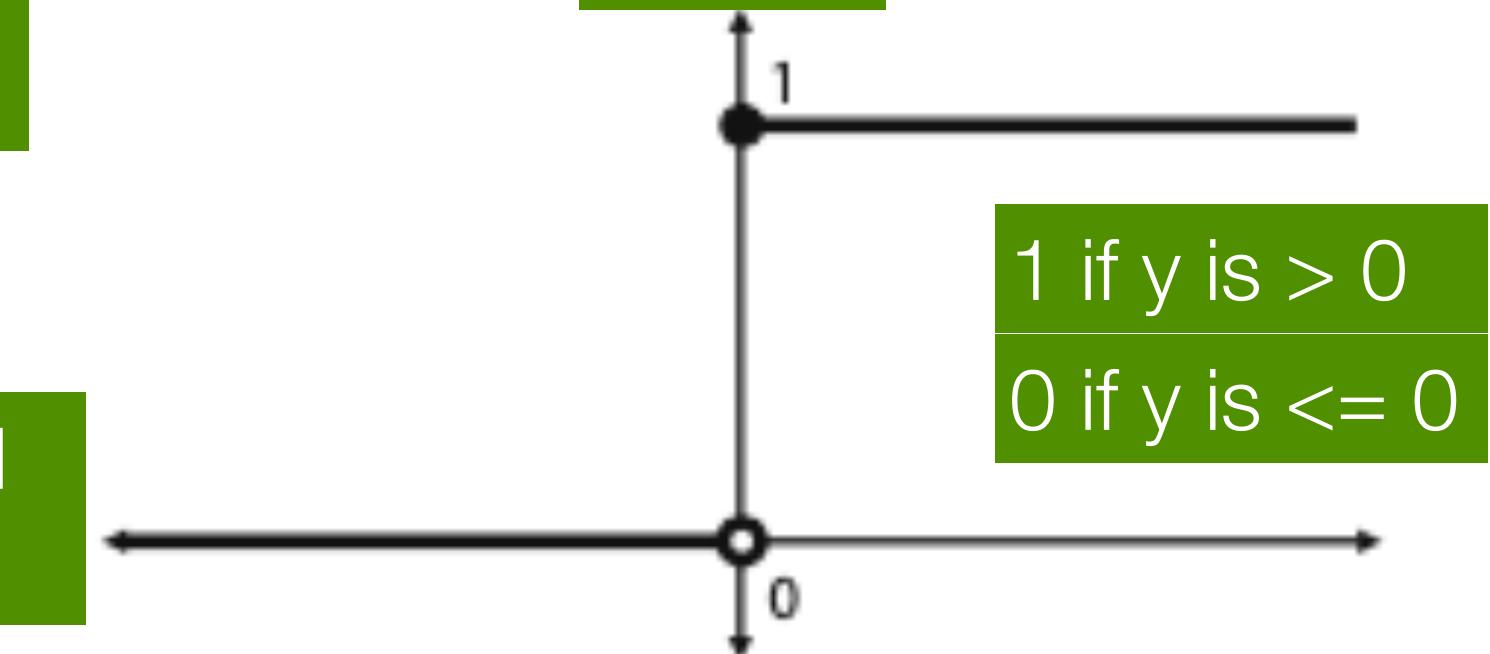
Remember our linear regression function?
It can be represented with a perceptron
having a single input and a bias

Input (weighted
sum of values)

Input (weighted
sum of values)

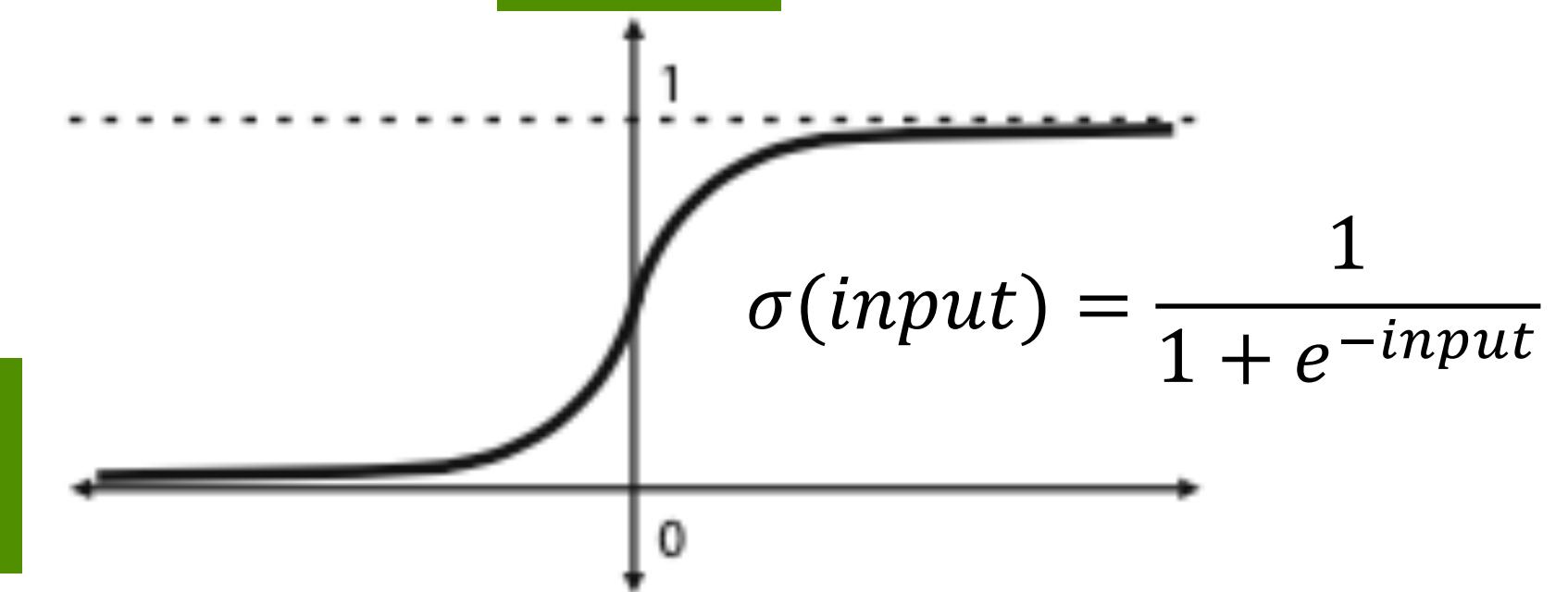
Step function
(discrete)

Output

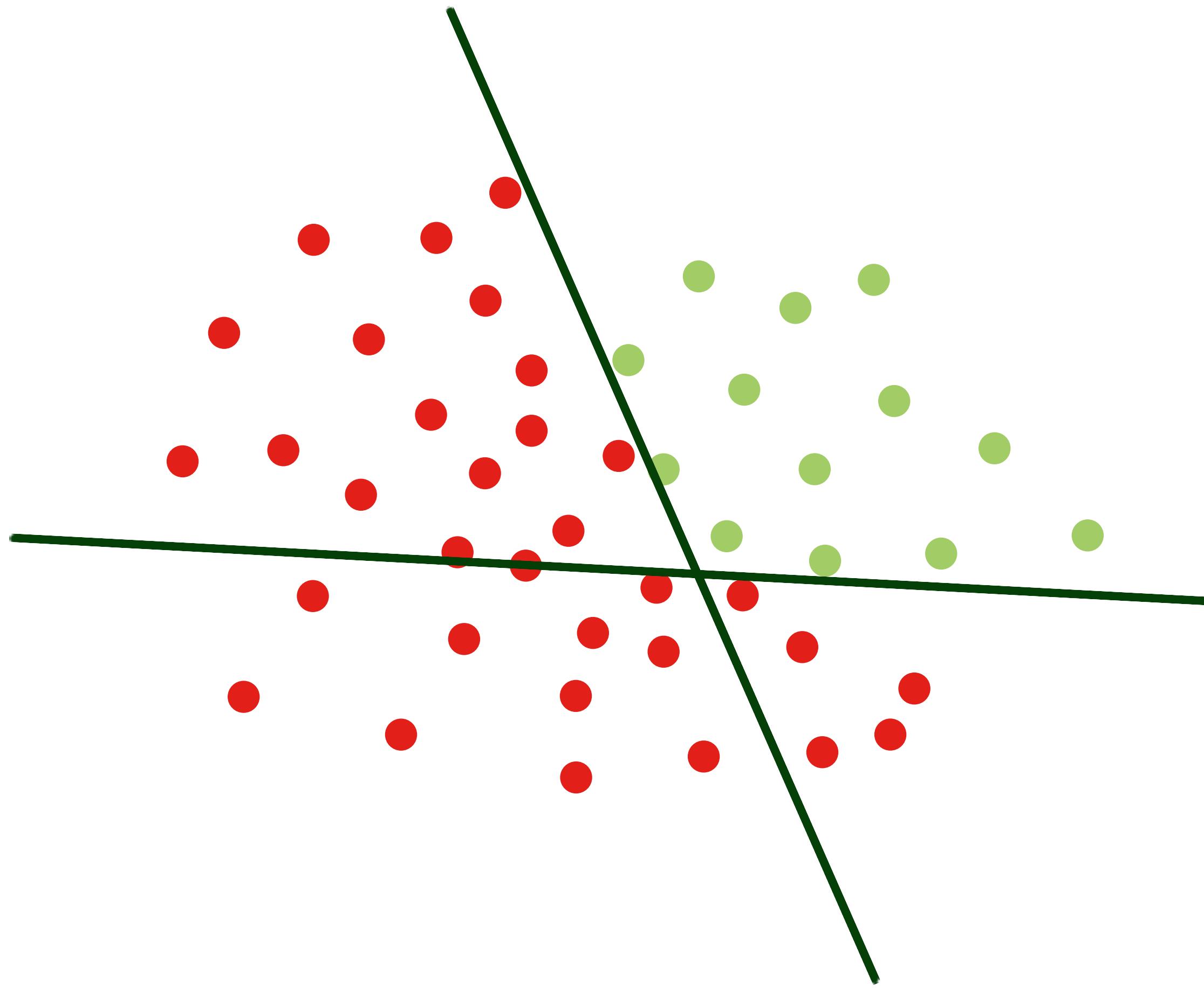


Sigmoid function
(continuous)

Output

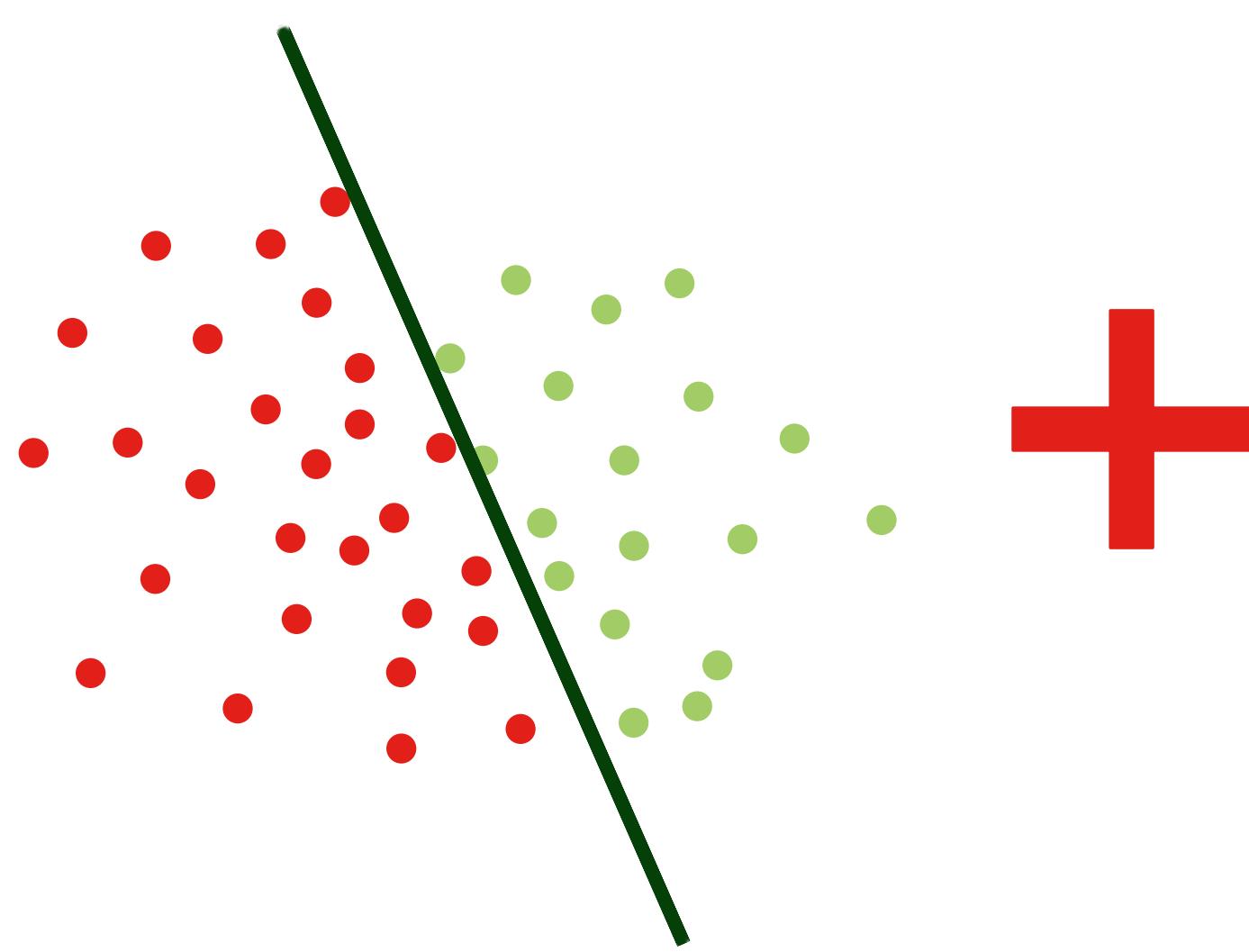


Neural Network

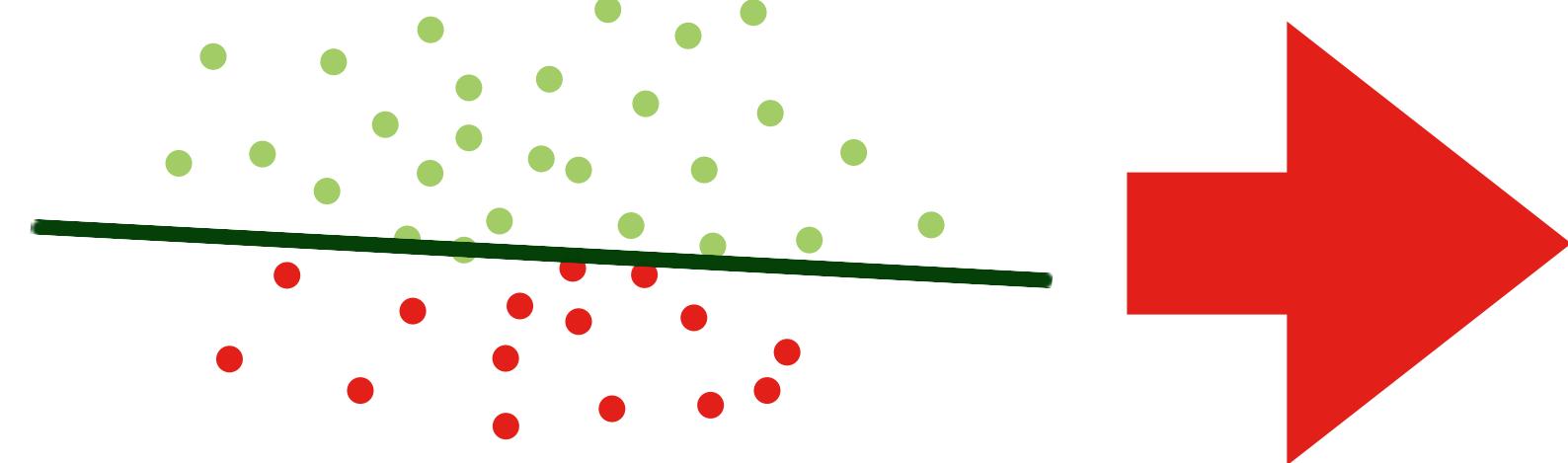


Neural Network

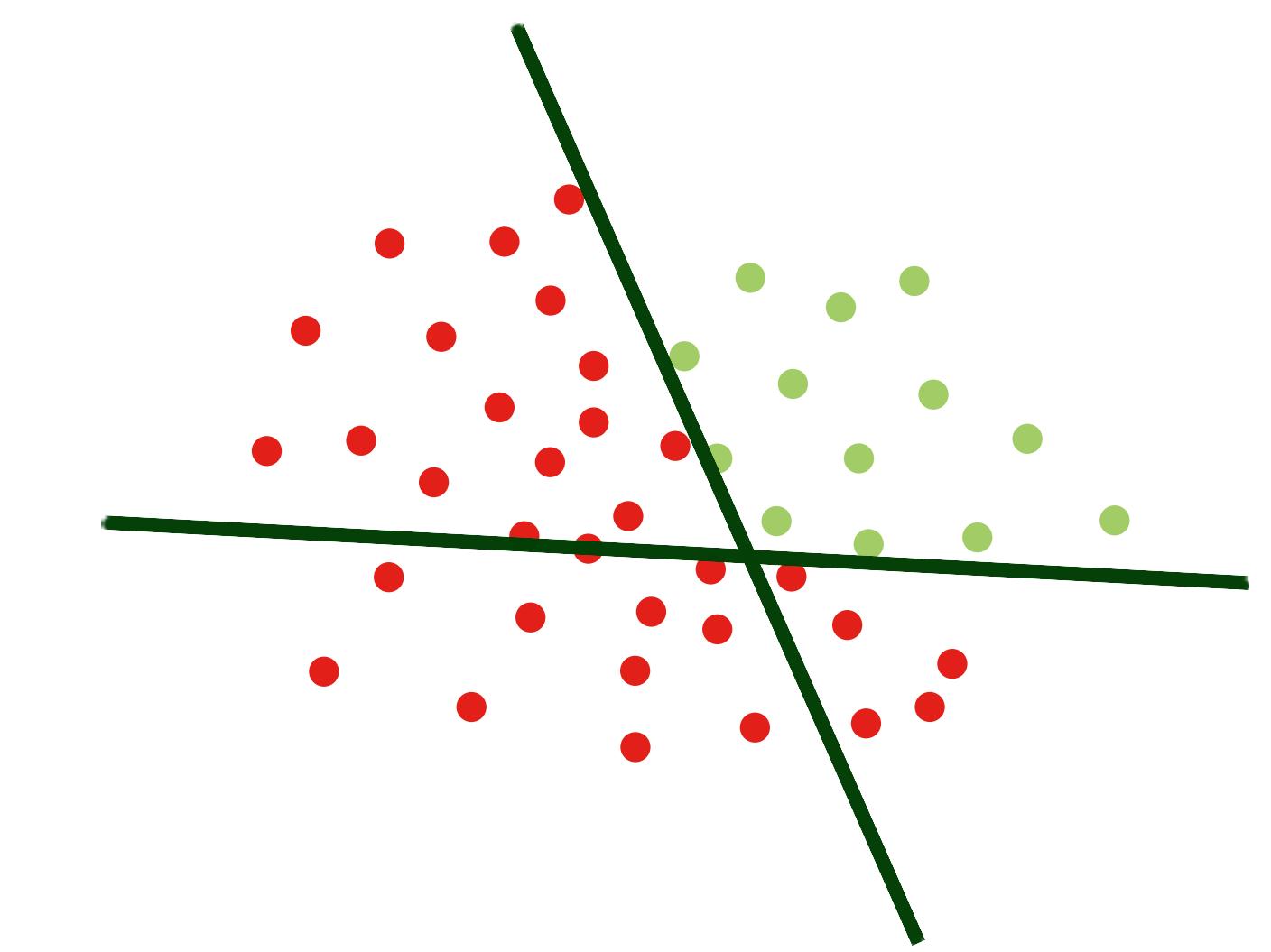
Physics Grade
Classifier



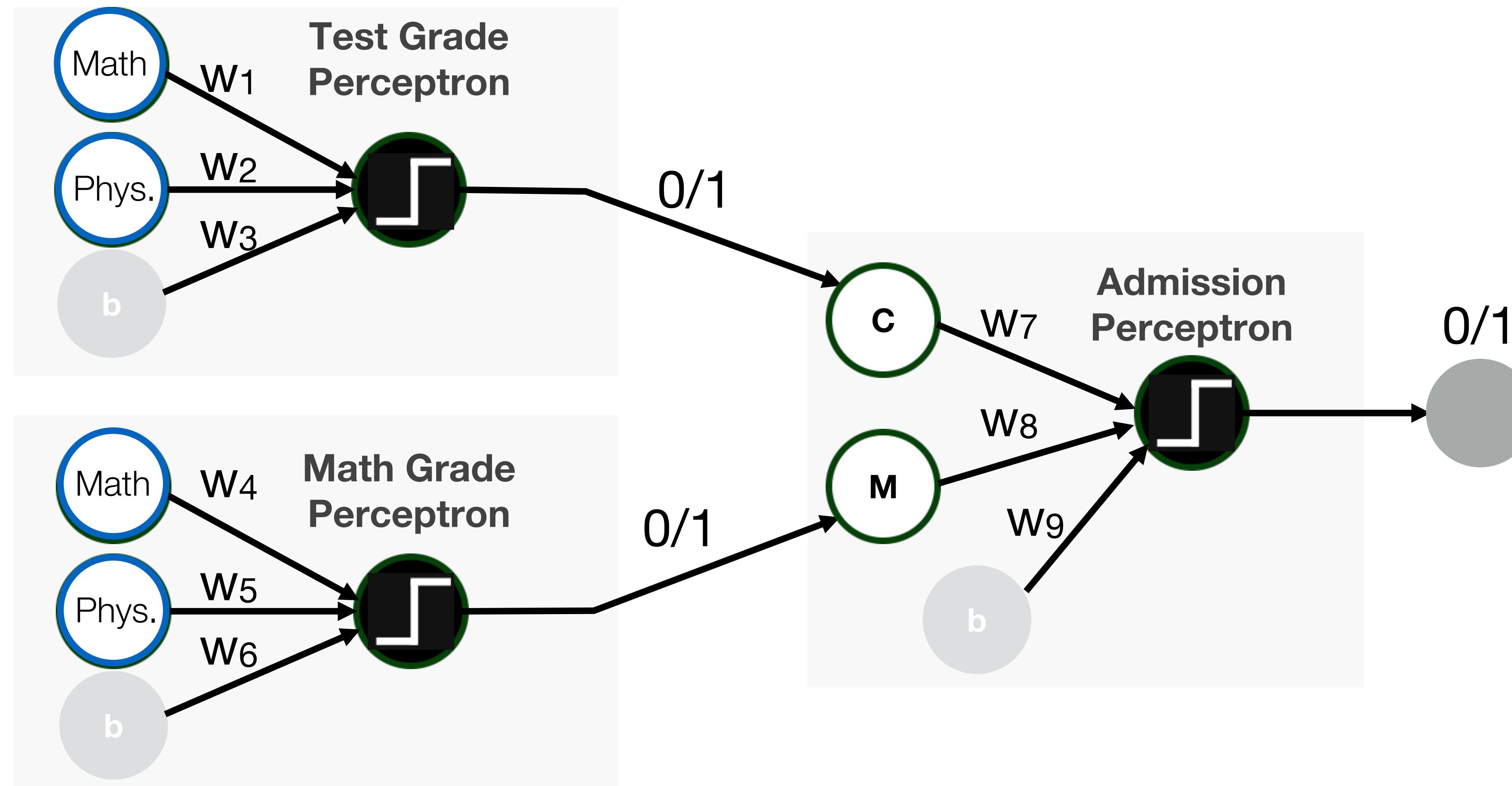
Math Grade
Classifier



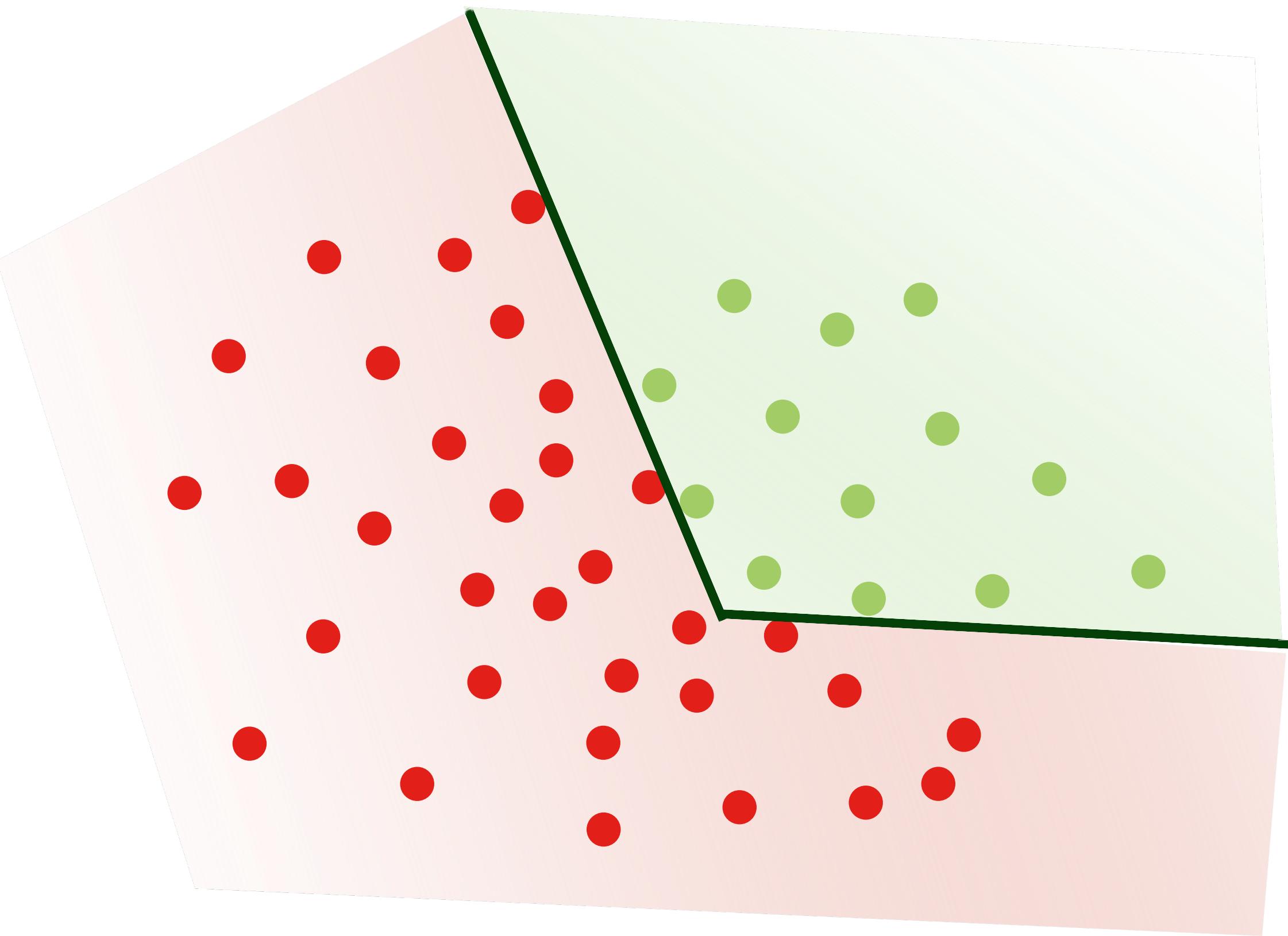
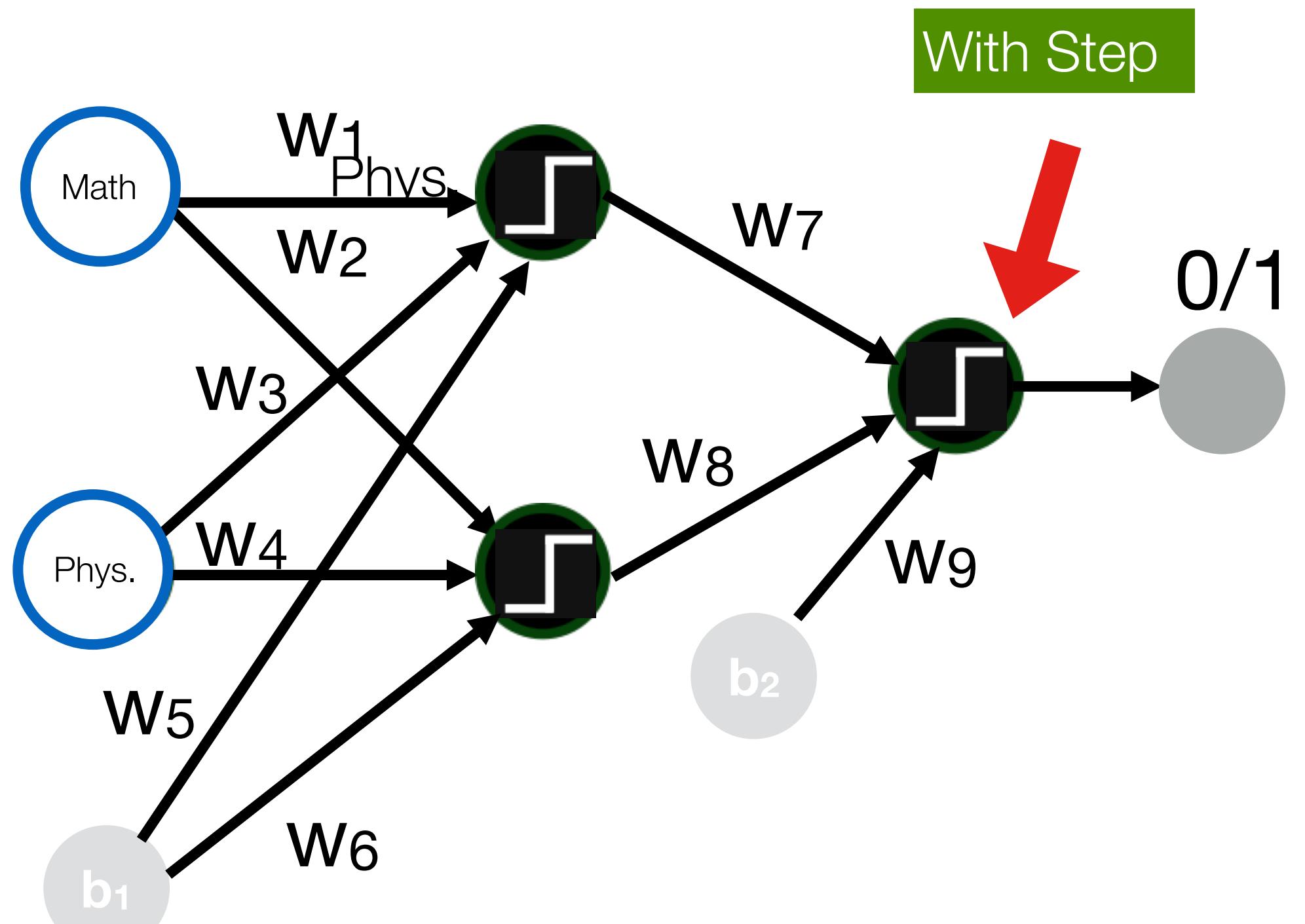
Admission
Classifier



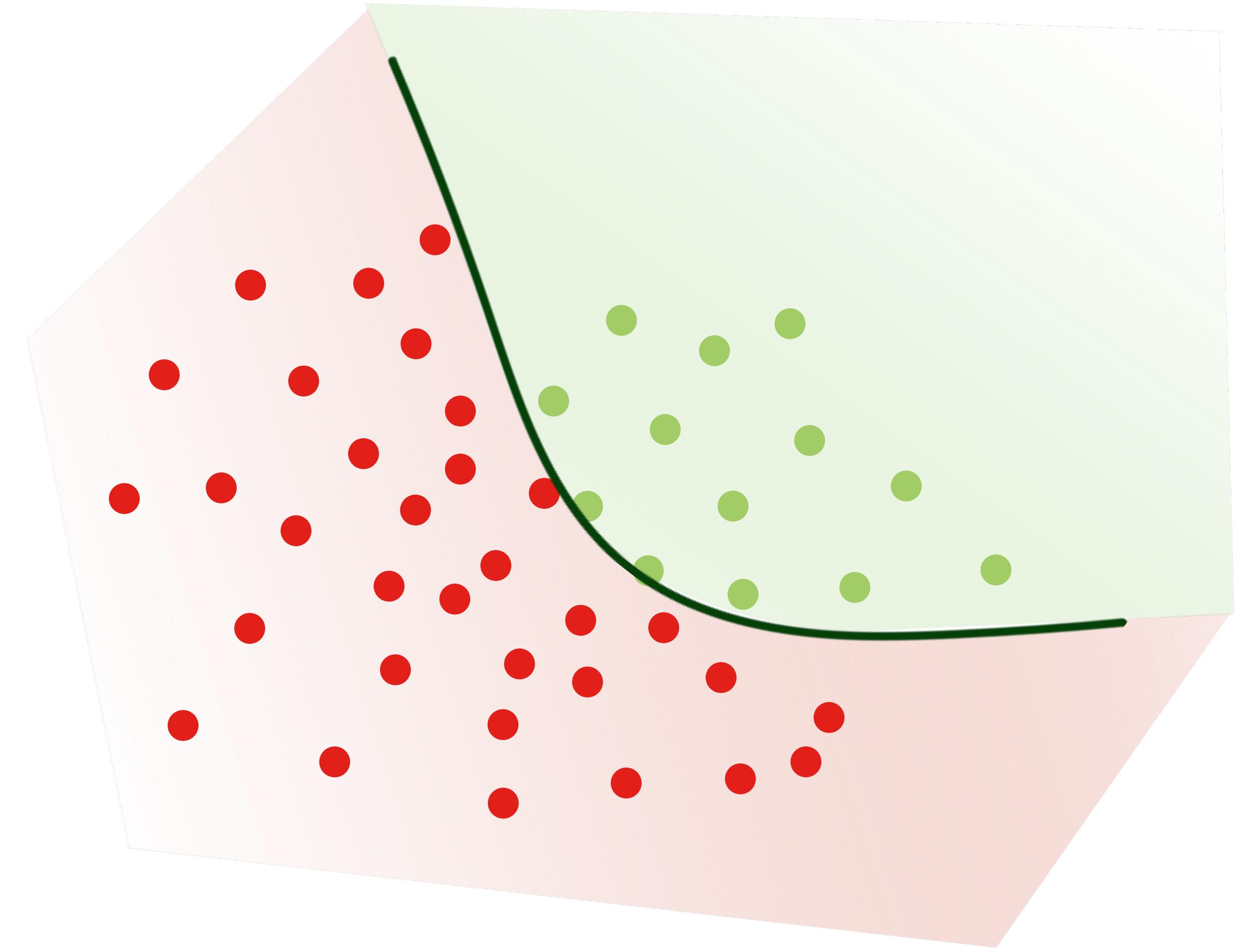
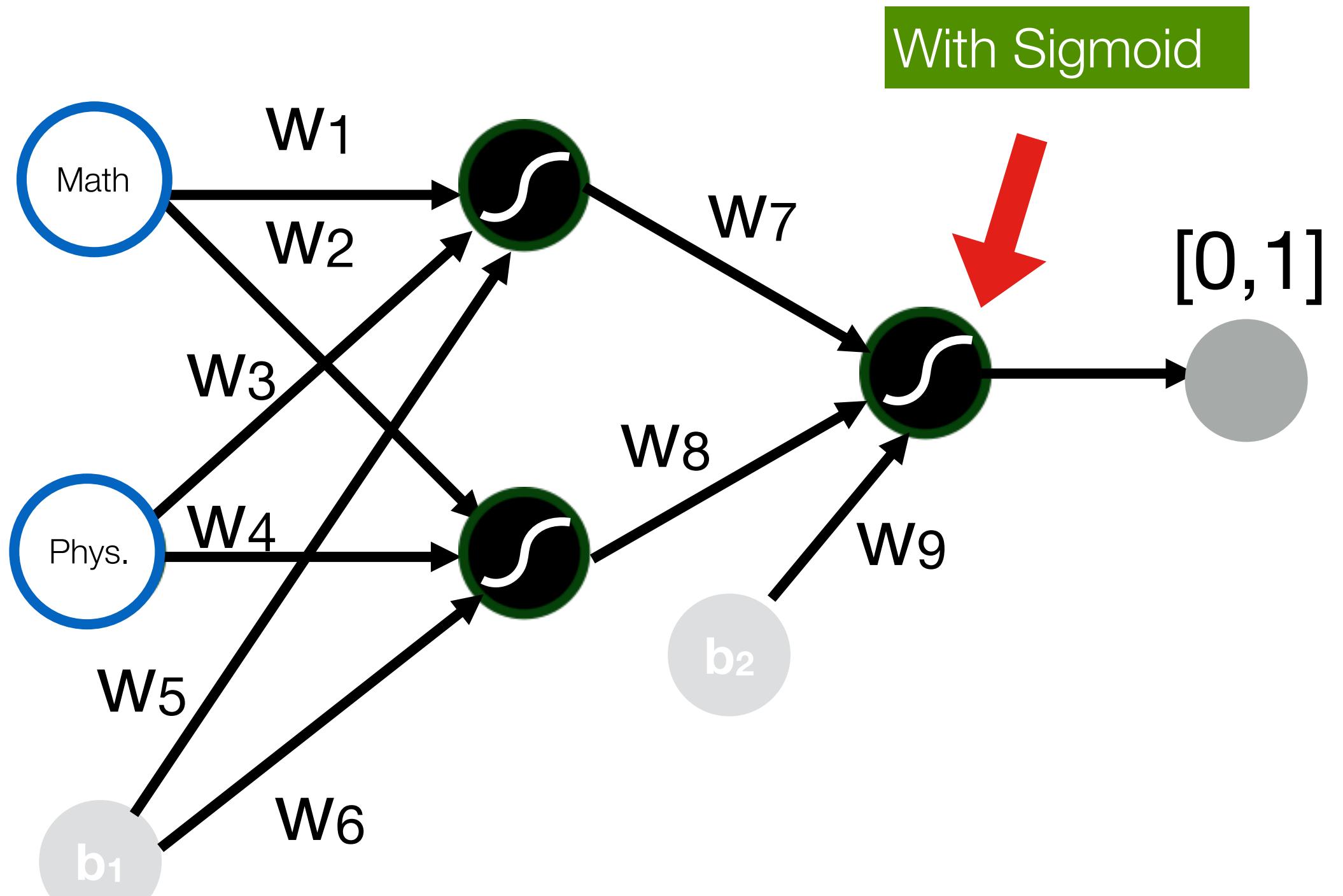
Neural Network



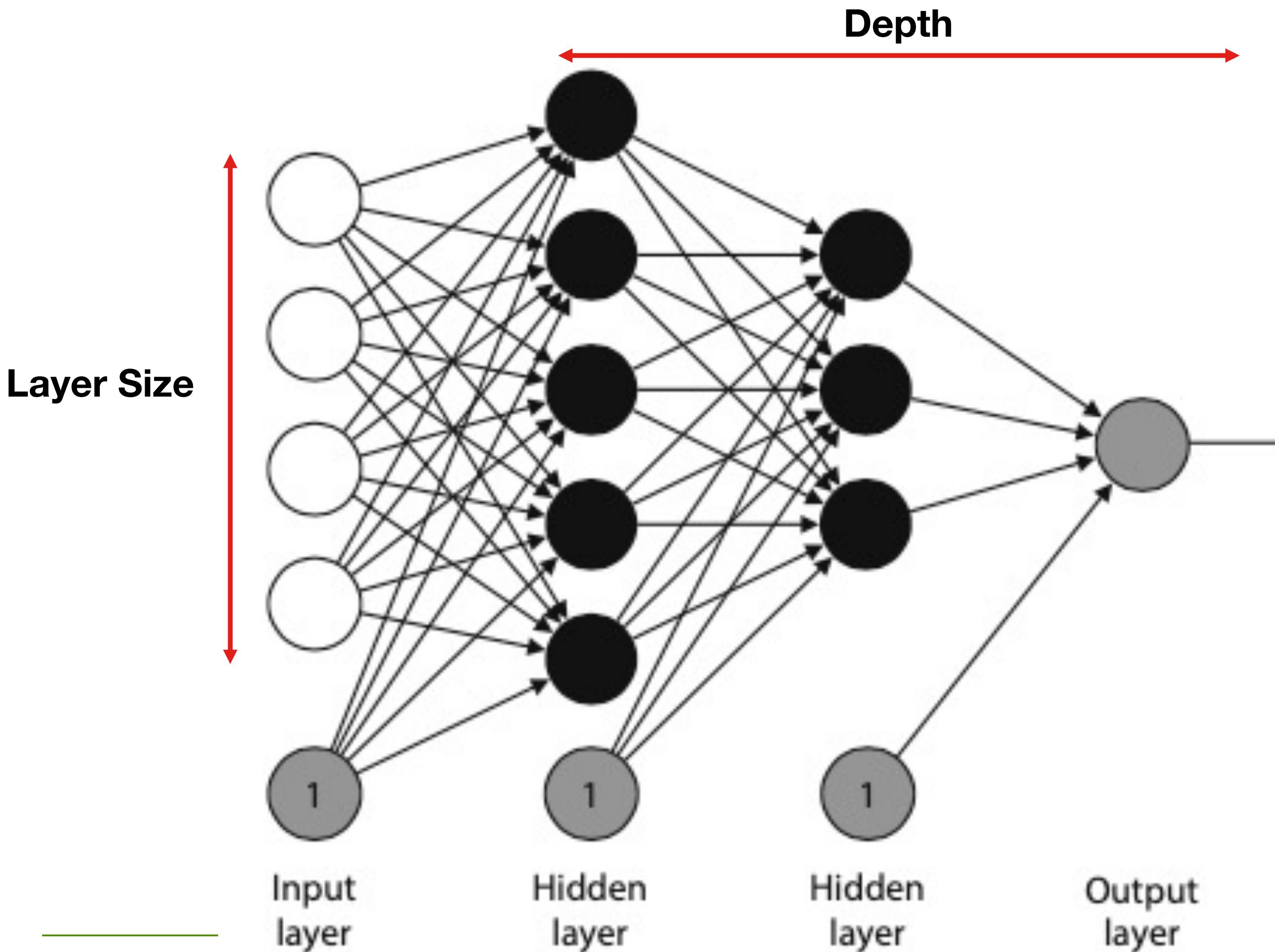
Neural Network



Neural Network



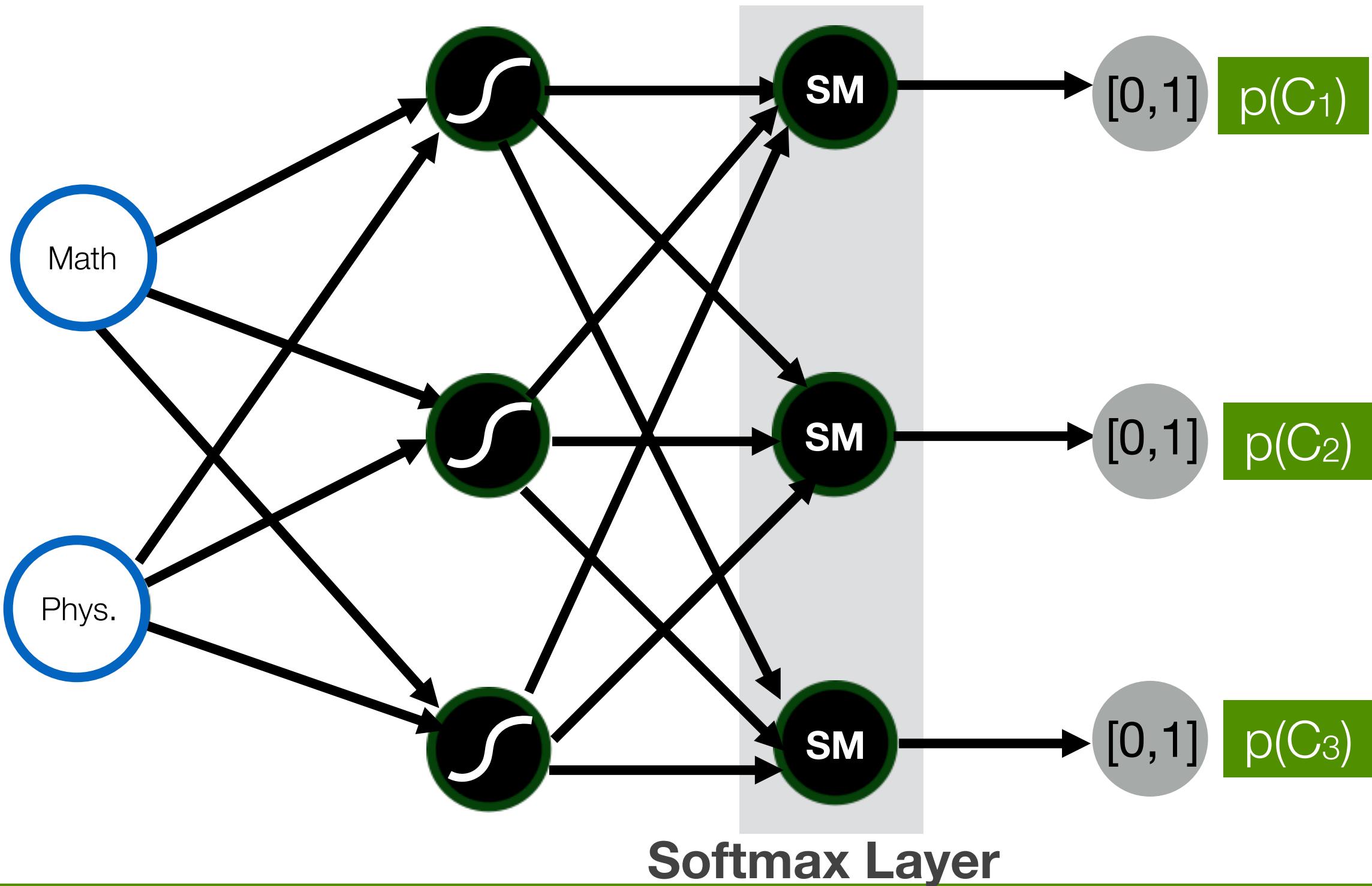
Fully Connected Neural Network



- **Hyperparameters**
 - Learning rate
 - Number of epochs
 - Architecture
 - # layers, #nodes, activation functions
 - Batch vs. mini-batch vs. stochastic gradient descent
 - Regularization parameters:
 - Dropout probability p
- Advanced → next lecture**

Classifying into multiple classes - softmax function

- Return a probability for **each class**
 - Imagine example $C_1 = \text{ADMITTED}$, $C_2 = \text{NOT ADMITTED}$, $C_3 = \text{TAKE TEST}$
 - $p(C_1) = 0.37$, $p(C_2) = 0.21$, $p(C_3) = 0.42$
- We use the Softmax activation function for the output layer

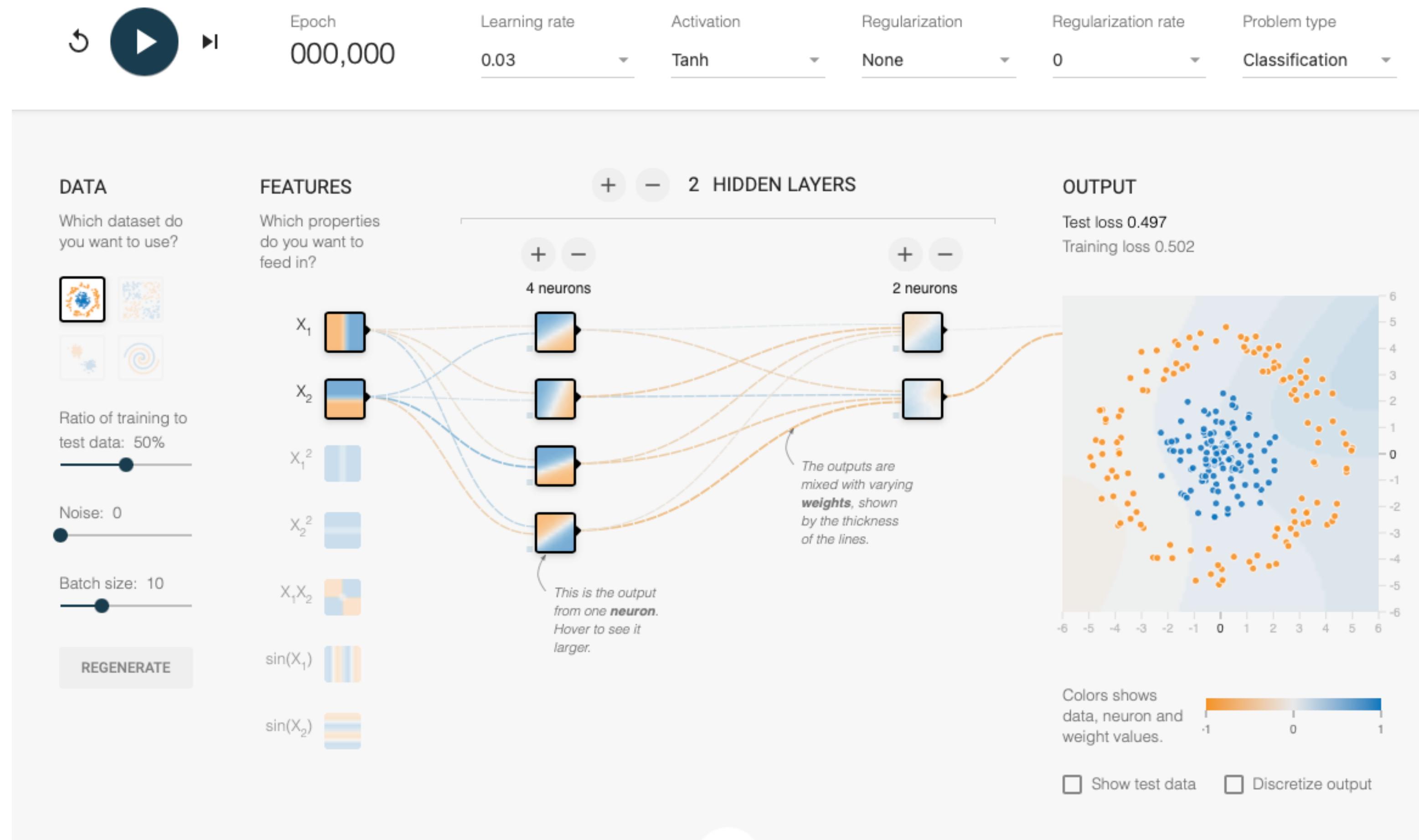


$$\text{Softmax}(x_i) = \frac{e^{(x_i)}}{\sum_j^K e^{(x_j)}}$$

Value of class i

Normalisation term on K classes

Tinker With a **Neural Network** Right Here in Your Browser.
Don't Worry, You Can't Break It. We Promise.



<https://playground.tensorflow.org/>

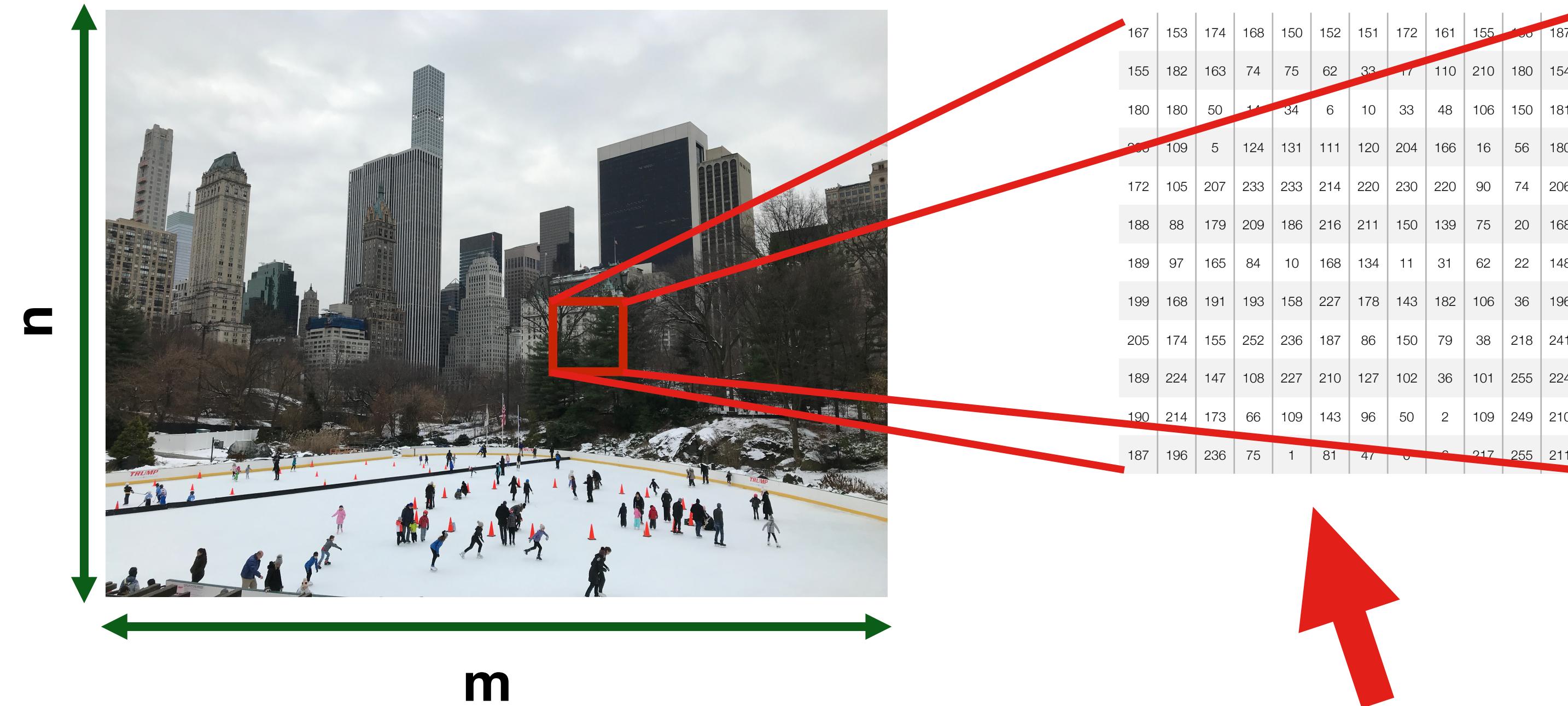
Machine Learning and Images

What do you see?



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189	224	147	108	227	210	127	102	36	101	255	22																																										

Images



- Each pixel in an image is a feature
- Dimensionality
 - $n \times m$
- Feature (pixel) values are numerical
 - 0 or 1 for Black and White
 - Between 0 and 255 for greyscale
 - 16M values for RGB

Computer Vision

- Building algorithms that can “understand” the content of images and use it for other applications
- It is a “Strong AI” problem
 - signal-to-symbol conversion
 - The **semantic gap**
- A general-purpose vision system requires
 - Flexible, robust visual representation
 - Updated and maintained
 - Reasoning
 - Interfacing with attention goals, and plans
- What specific tasks can we train a CV system to perform?

Strong vs. Weak Artificial Intelligence

- **Strong AI**
 - Artificial General Intelligence (AGI), human-level, general
 - The AI we see in movies
 - AI that can do everything we humans can do, and possibly much more
- **Weak AI**
 - Narrow AI
 - AI specialised in well-defined tasks
 - e.g. speech recognition, chess-playing, autonomous driving
- No AI program has been created yet that could be called intelligent in any general (Strong AI) sense
 - "A pile of narrow intelligence will never add up to a general intelligence. General intelligence isn't about the number of abilities, but about the integration between those abilities?"
- Superintelligence doesn't really mean anything - a basic calculator far exceeds any human benchmark for performing basic arithmetic

“Easy problems are hard”

Marvin Minsky

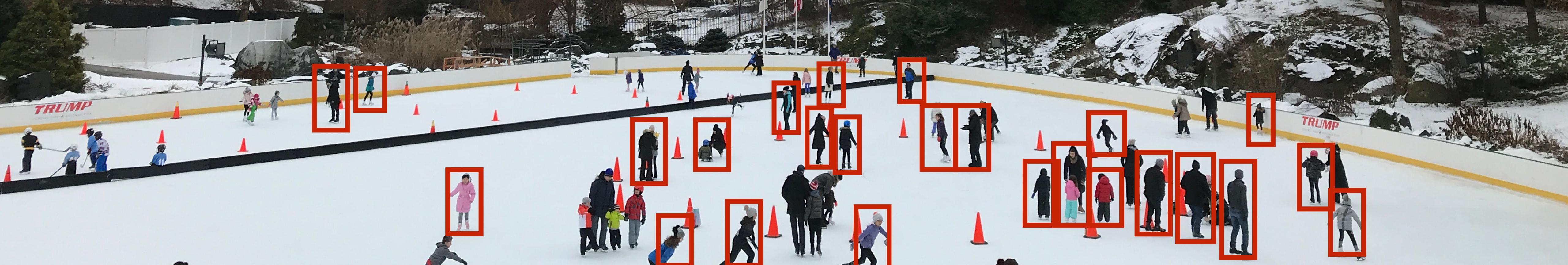


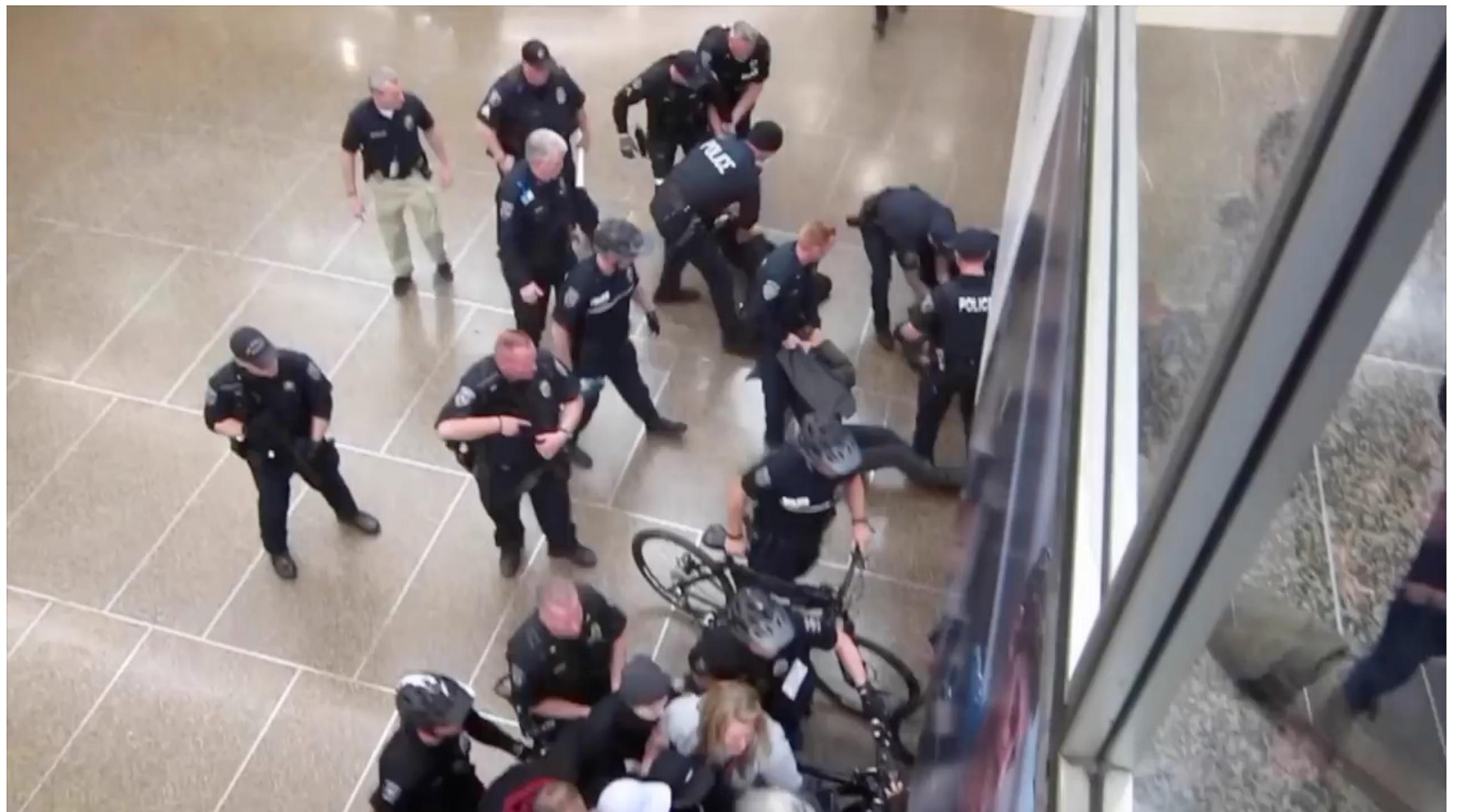
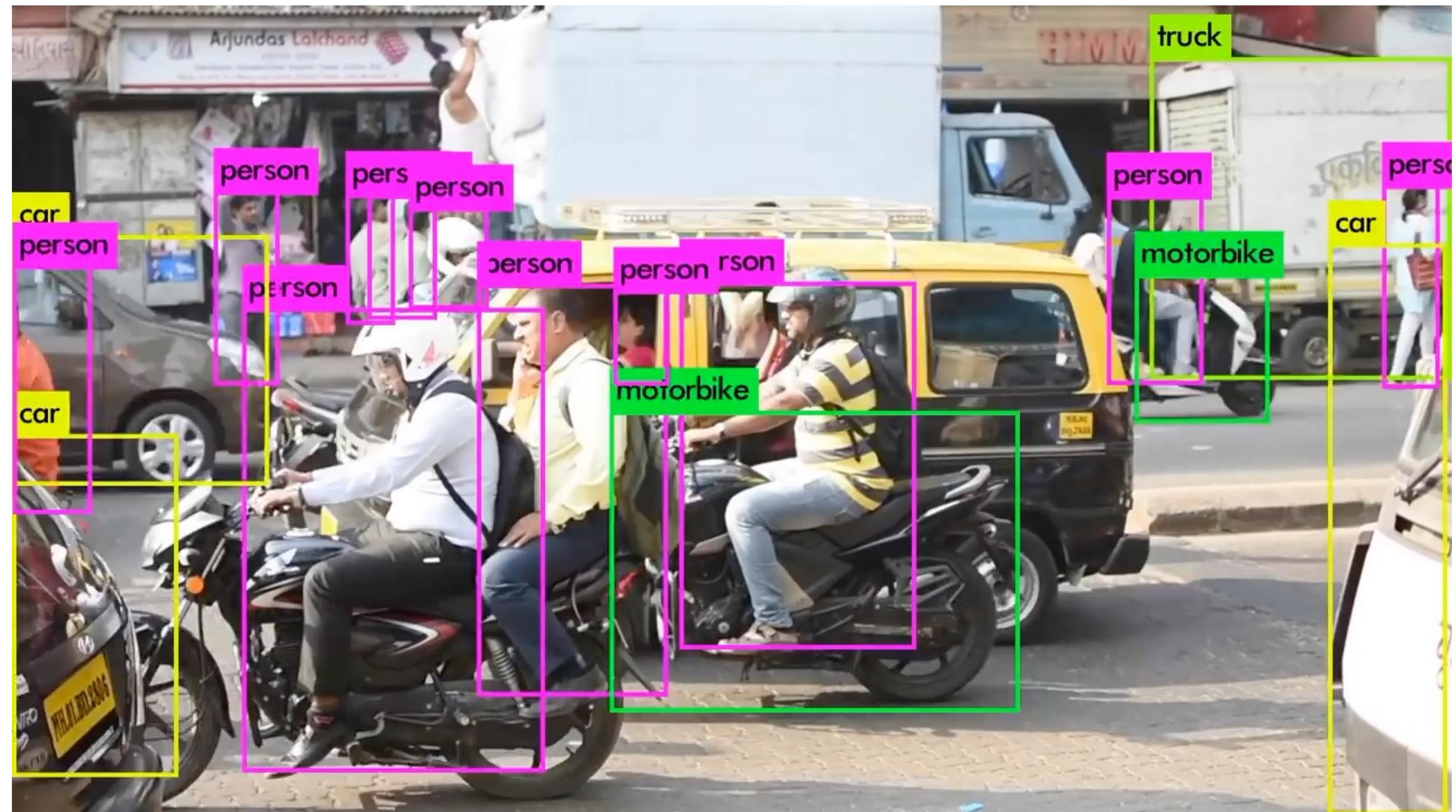
Is this a flag?
(Recognition / Classification)





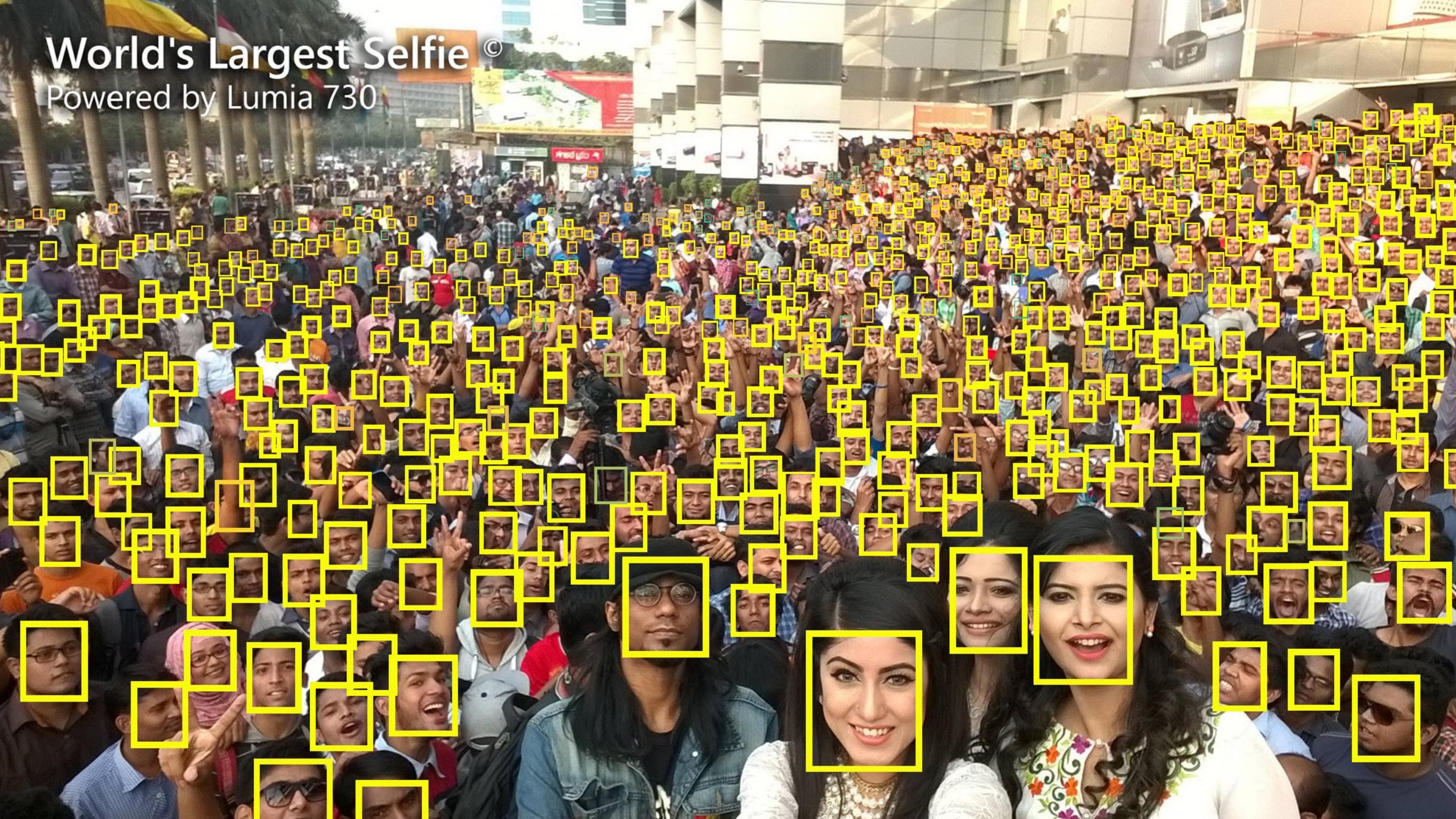
Where are the people?
(Recognition/Classification)





World's Largest Selfie ©

Powered by Lumia 730



MORPHCAST



AFFECT

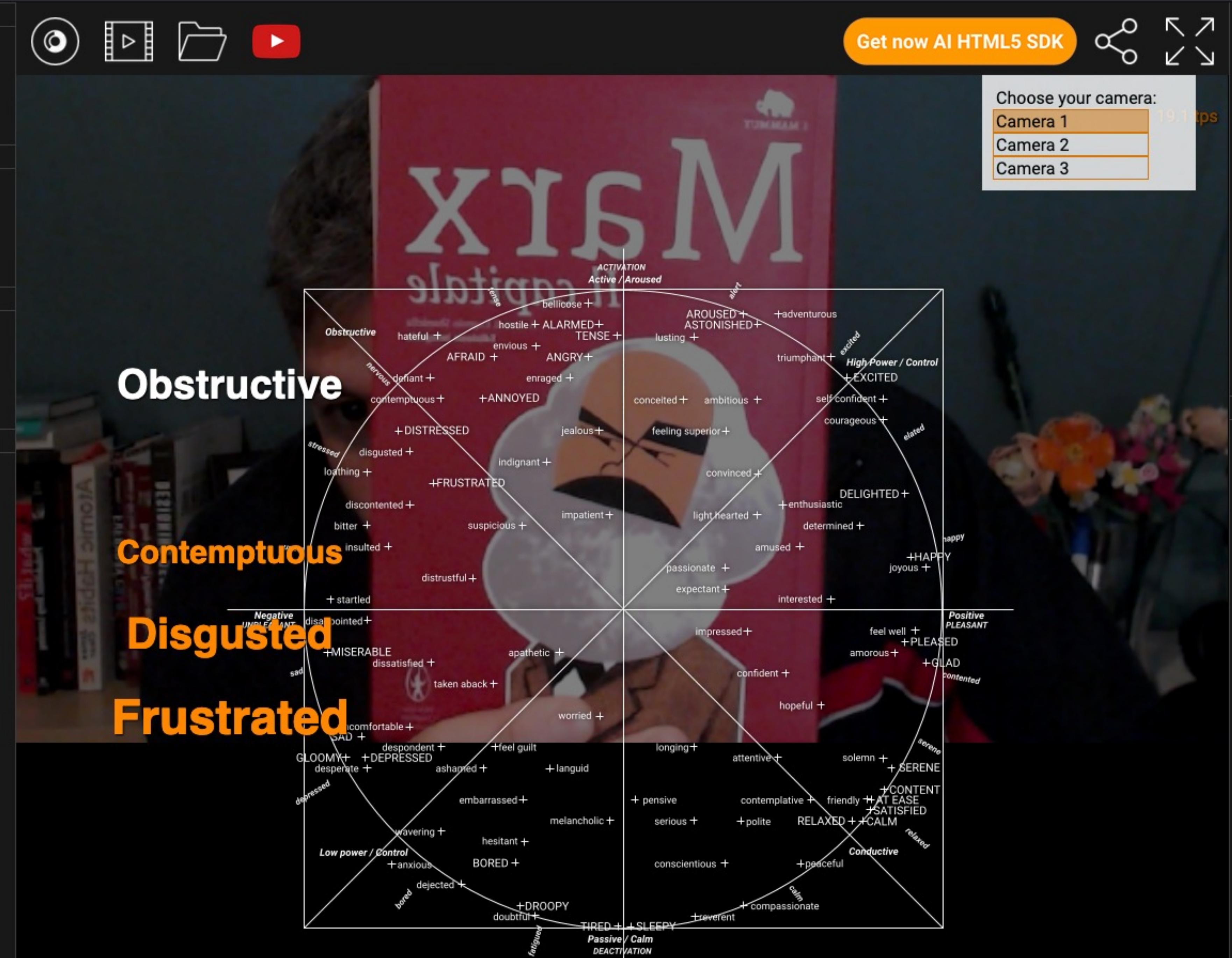


EMOTION



LIKELY AGE

-100
- 90
- 80
- 70
- 60
- 50
- 40
- 30
- 20
- 10
- 0

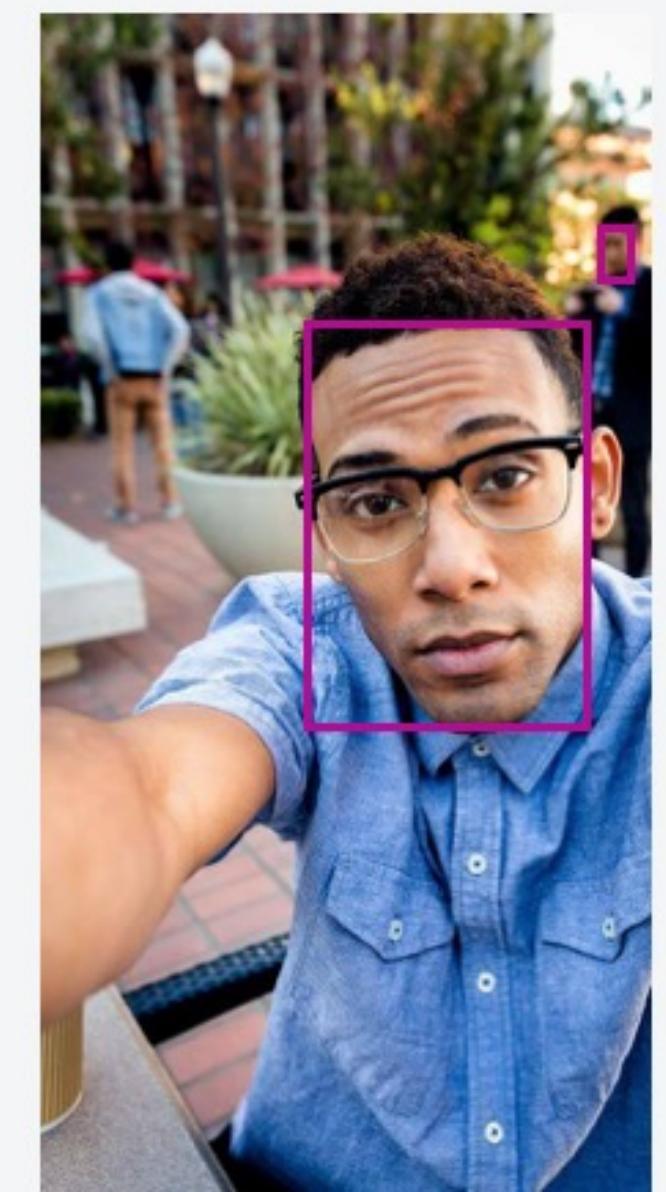


Fully protects
your privacy
No recording
No storing of
biometric data
All images
overwritten
in 100 MS



**Is this Jeff?
(Identification)**

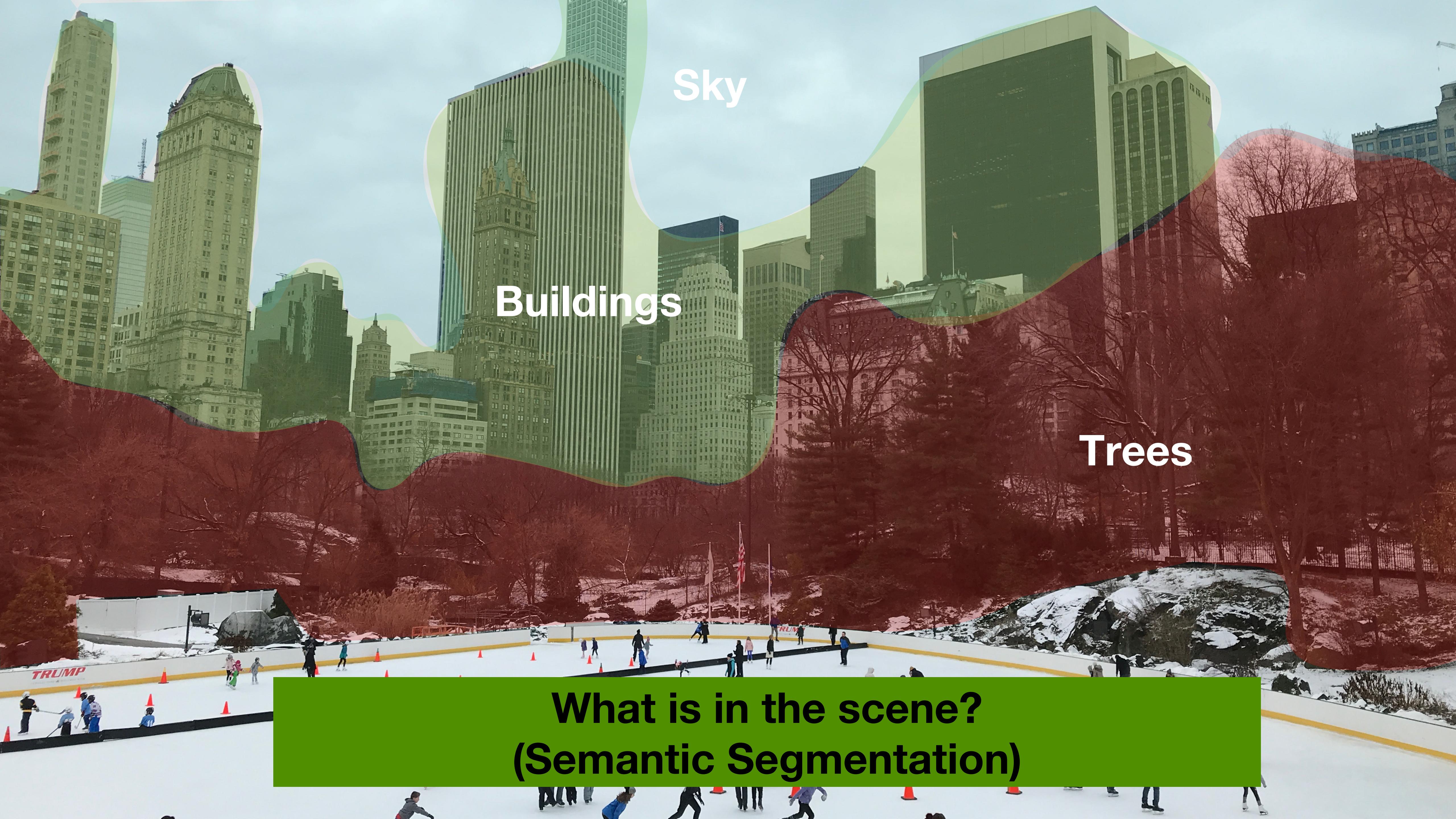




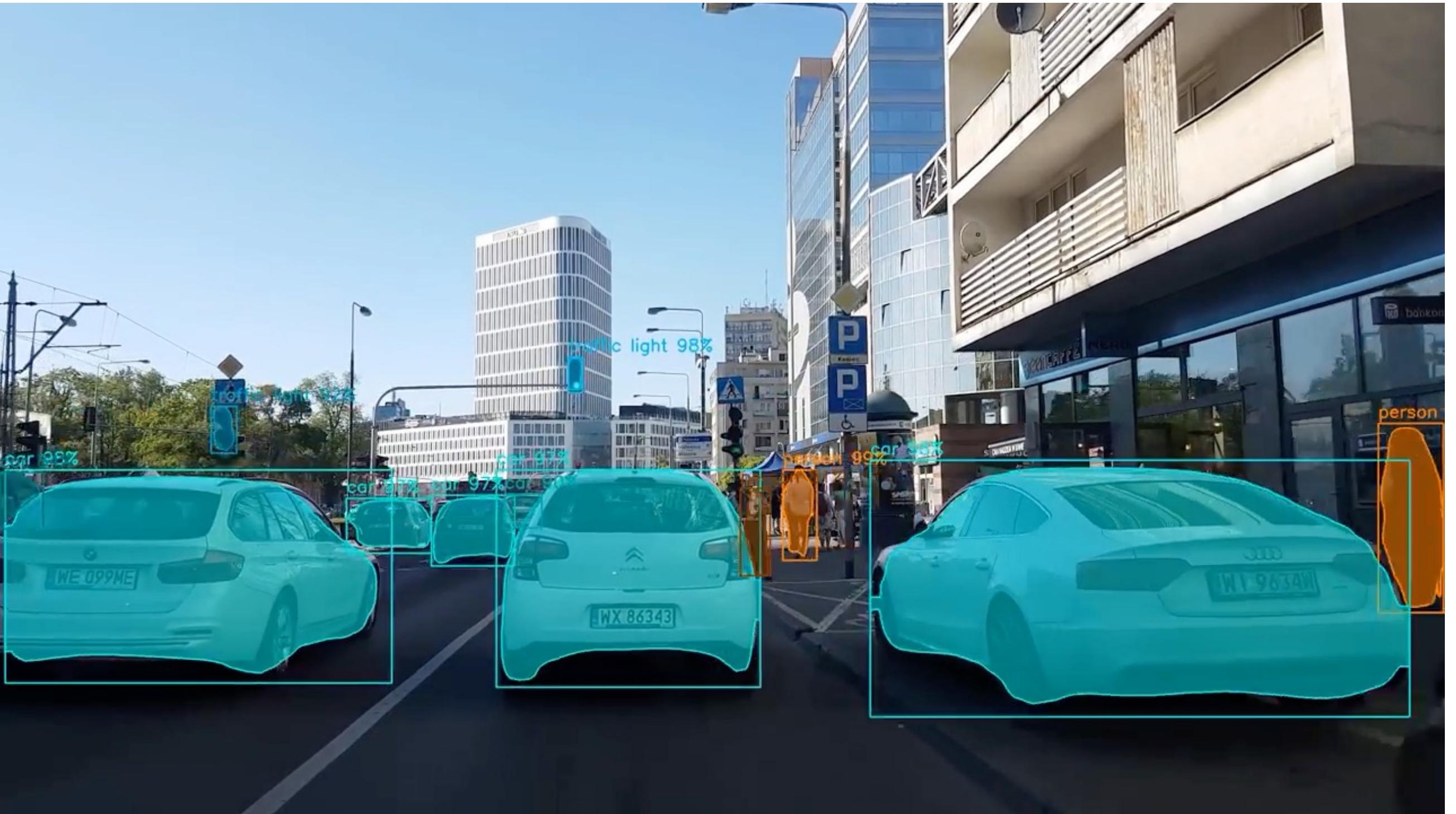


**Is this the Wollman Rink?
(Identification)**





**What is in the scene?
(Semantic Segmentation)**



https://github.com/matterport/Mask_RCNN

Project Sunroof

≡ Google Project Sunroof

Savings estimator

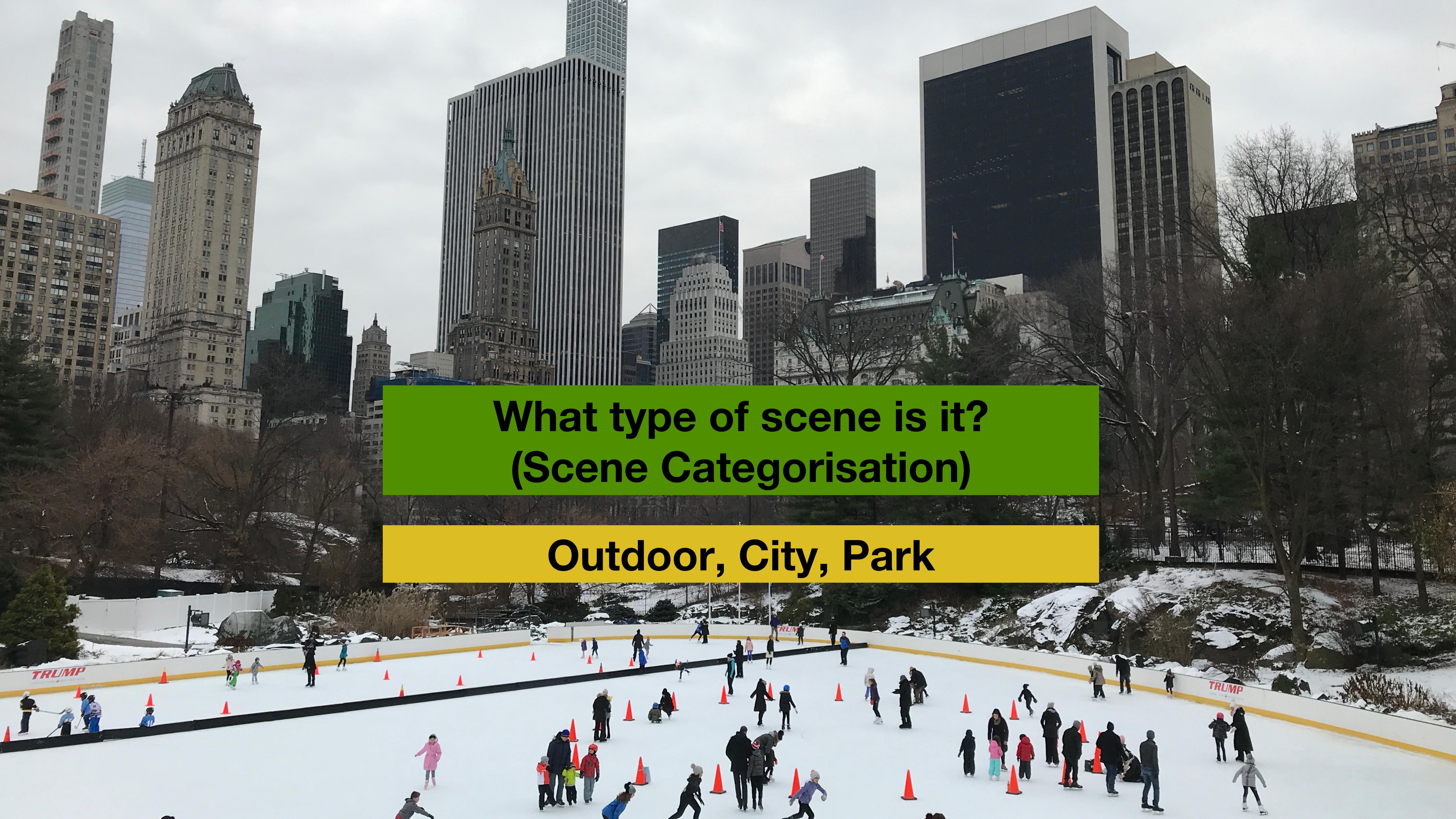
Data explorer

Solar 101

FAQ

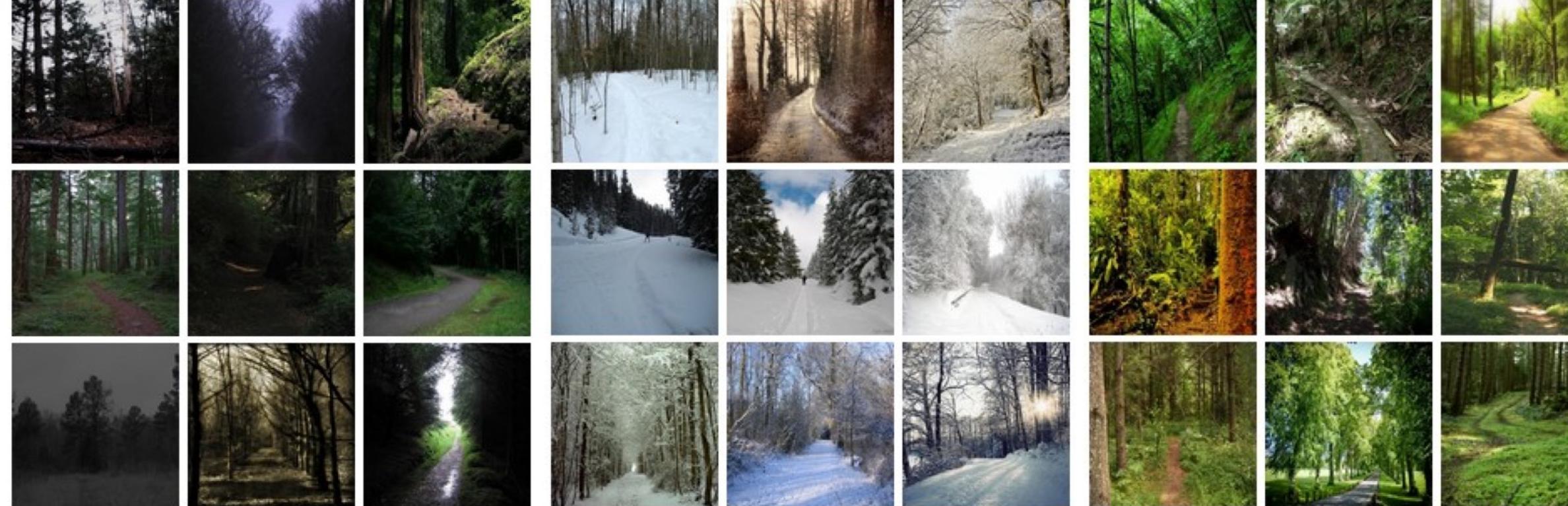


<https://www.google.com/get/sunroof>

The background image shows a wide-angle view of an outdoor ice skating rink in a park. The rink is filled with people skating. In the foreground, there's a large green rectangular area containing text. Behind the rink, a dense city skyline with many skyscrapers is visible against a cloudy sky.

**What type of scene is it?
(Scene Categorisation)**

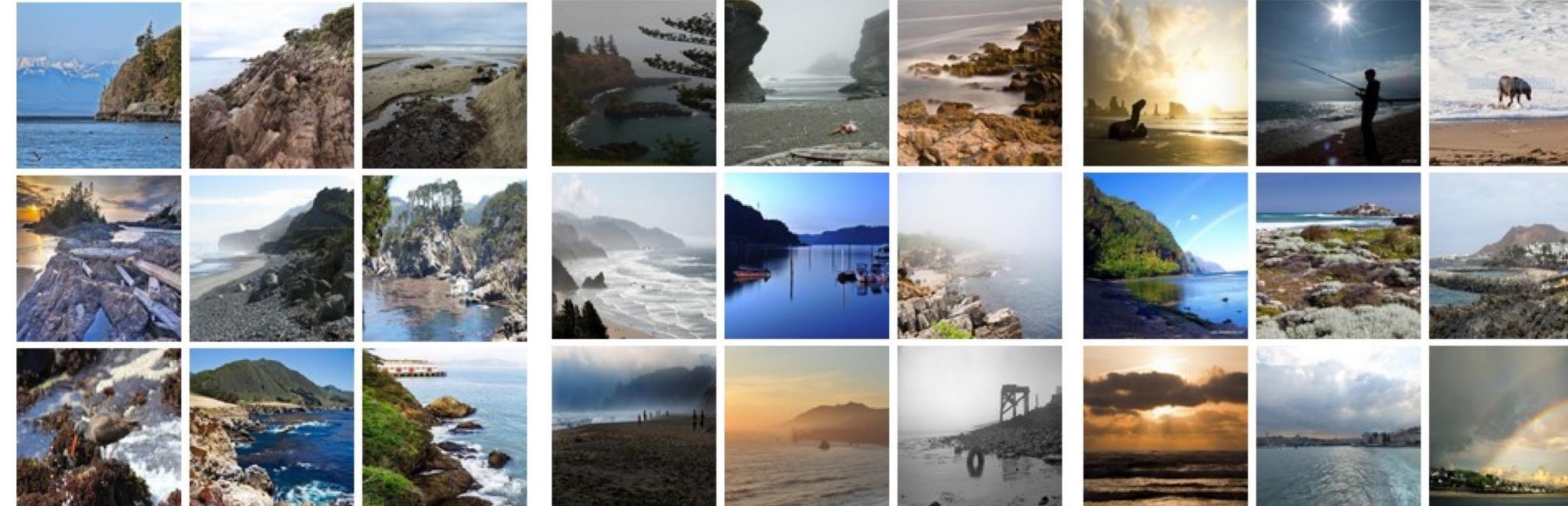
Outdoor, City, Park



darkest forest path

wintering forest path

greener forest path



rocky coast

misty coast

sunny coast



Predictions:

- **Type of environment:** outdoor
- **Scene categories:** skyscraper (0.704), downtown (0.211)
- **Scene attributes:** man-made, vertical components, open area, natural light, clouds, no horizon, metal, glass, sunny
- **Informative region for predicting the category *skyscraper* is:**





Spaces: OFA-Sys/OFA-Image_Caption

like 12

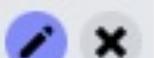
Running

App Files and versions

OFA-Image_Caption

Gradio Demo for OFA-Image_Caption. Upload your own image or click any one of the examples, and click "Submit" and then wait for the generated caption.

Image



Caption

8.28s

people skating on a rink in a city park with skyscrapers

Clear

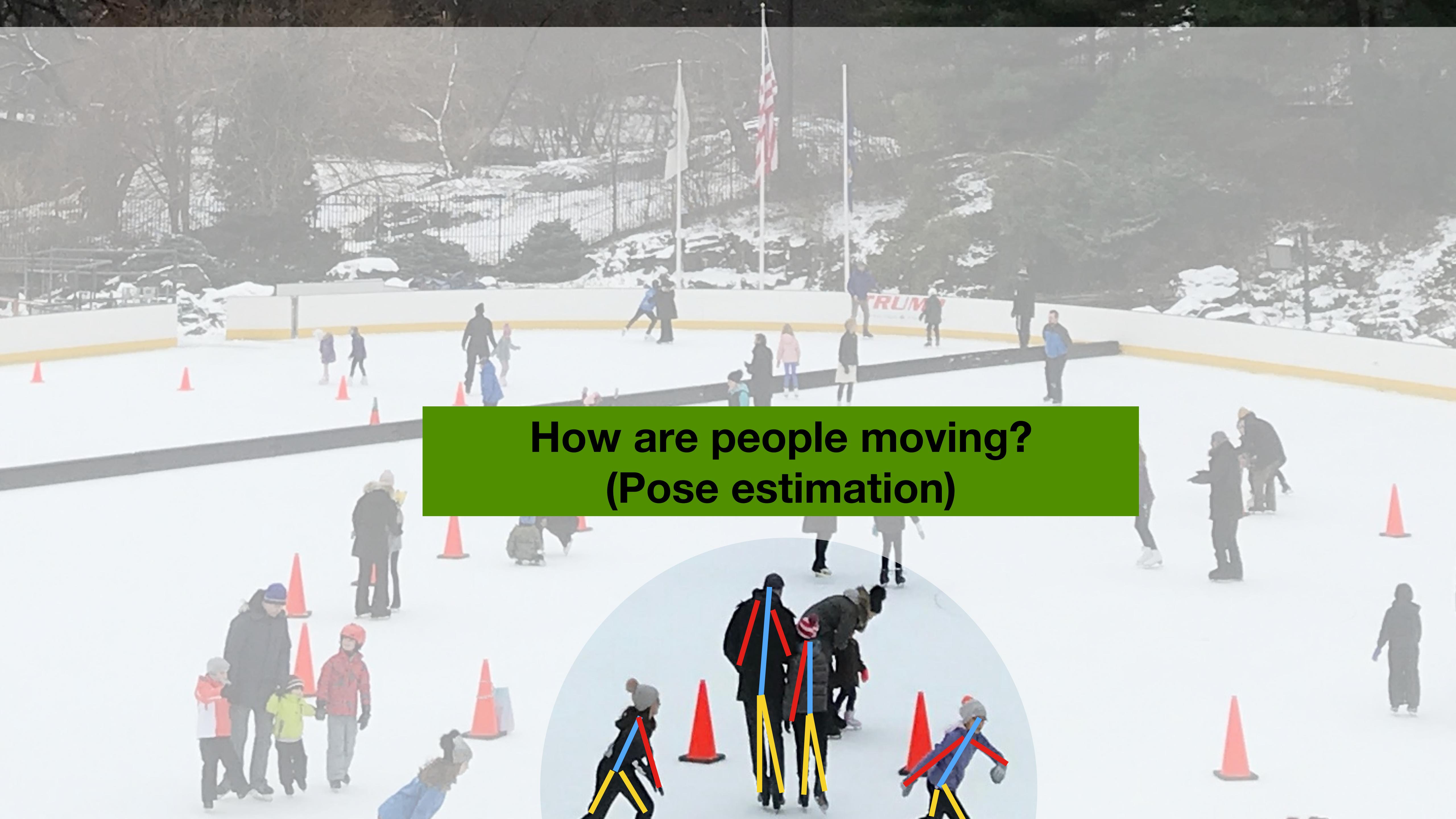
Submit

The background image shows a wide outdoor ice skating rink in a park. In the foreground, many people are skating on the ice. The rink is surrounded by a white fence. In the background, a dense city skyline with numerous skyscrapers is visible under a clear sky.

**What are these people doing?
(Activity / event recognition)**

Skating





**How are people moving?
(Pose estimation)**

X: -1.11702489838885962072754
Z: -5.3908080944445801
Work station

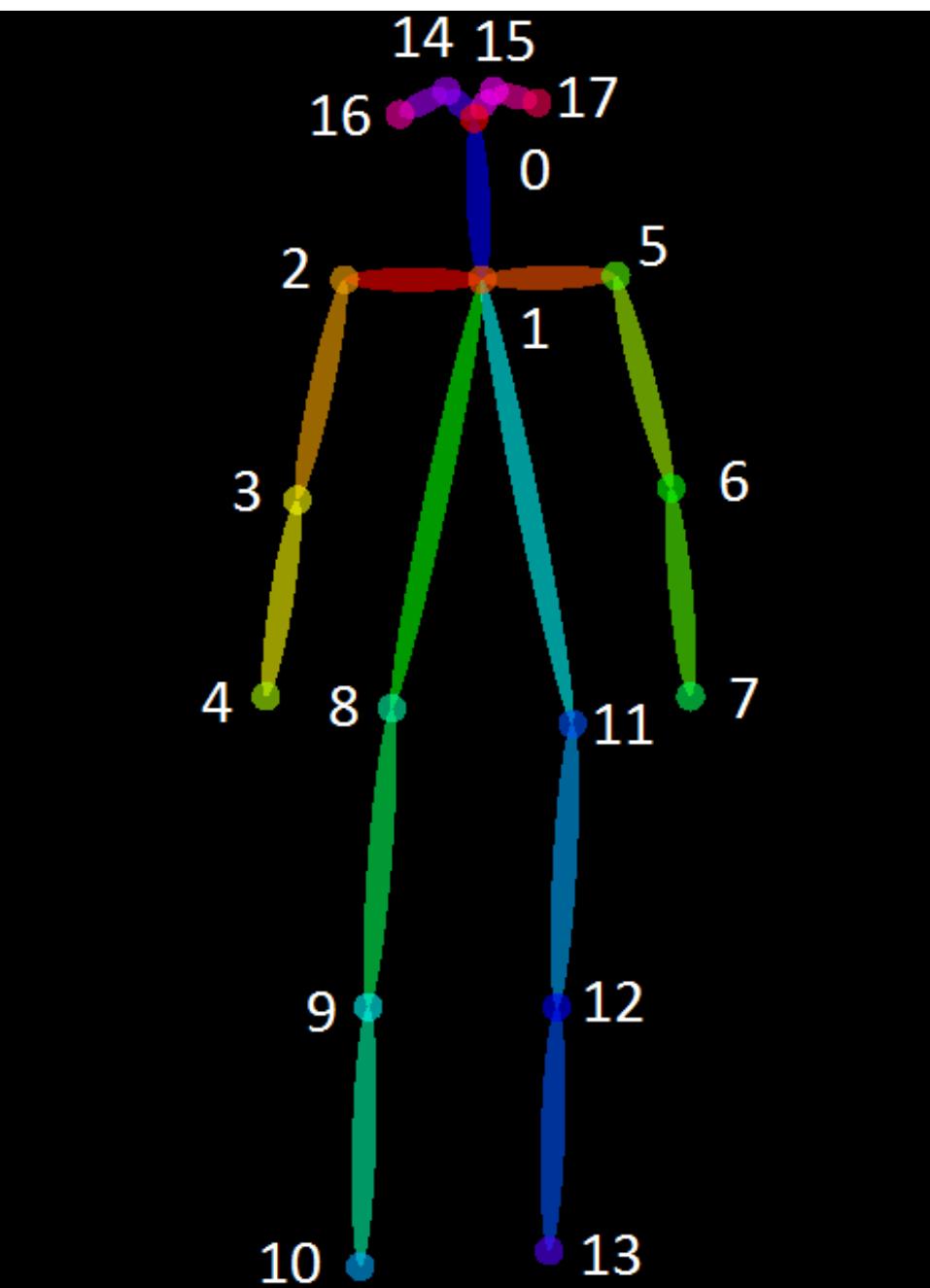


Ethical task tracking of operators in agile manufacturing

<http://resolver.tudelft.nl/uuid:3408e8c3-809b-436d-94eb-efb4f0532b17>

Stereolabs ZED Camera used in COALA

- 3D Object Detection
- Body tracking
- Positional tracking



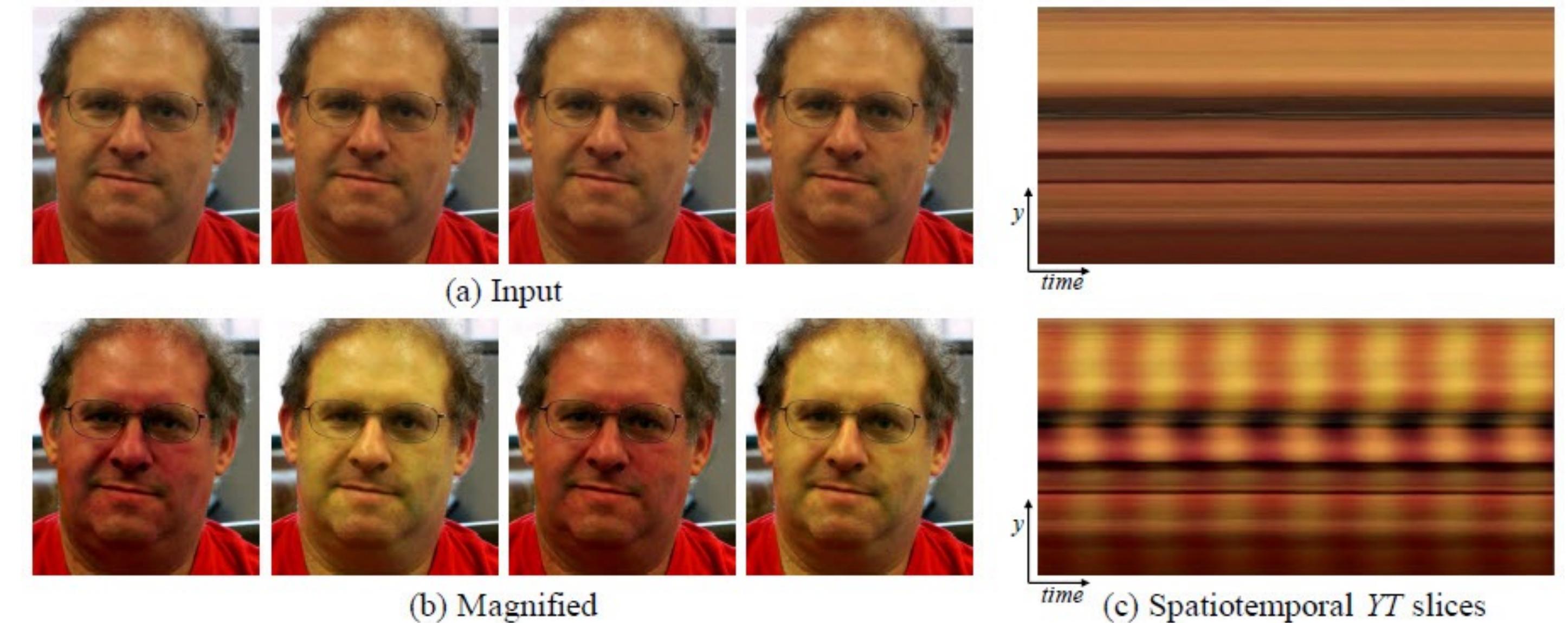
[https://www.stereolabs.com/
docs/object-detection/](https://www.stereolabs.com/docs/object-detection/)

[https://www.stereolabs.com/d
ocs/body-tracking/](https://www.stereolabs.com/docs/body-tracking/)



Eulerean Video Magnification

- Measure heart rate by magnifying and measuring the subtle changes in skin colour owed to blood vessels
- <https://people.csail.mit.edu/mrub/vidmag/>



Wu, H. Y., Rubinstein, M., Shih, E., Guttag, J., Durand, F., & Freeman, W. (2012). Eulerian video magnification for revealing subtle changes in the world. ACM transactions on graphics (TOG), 31(4), 1-8.

Google images

Search by image
Search Google with an image instead of text. Try dragging an image here.

Drop image here

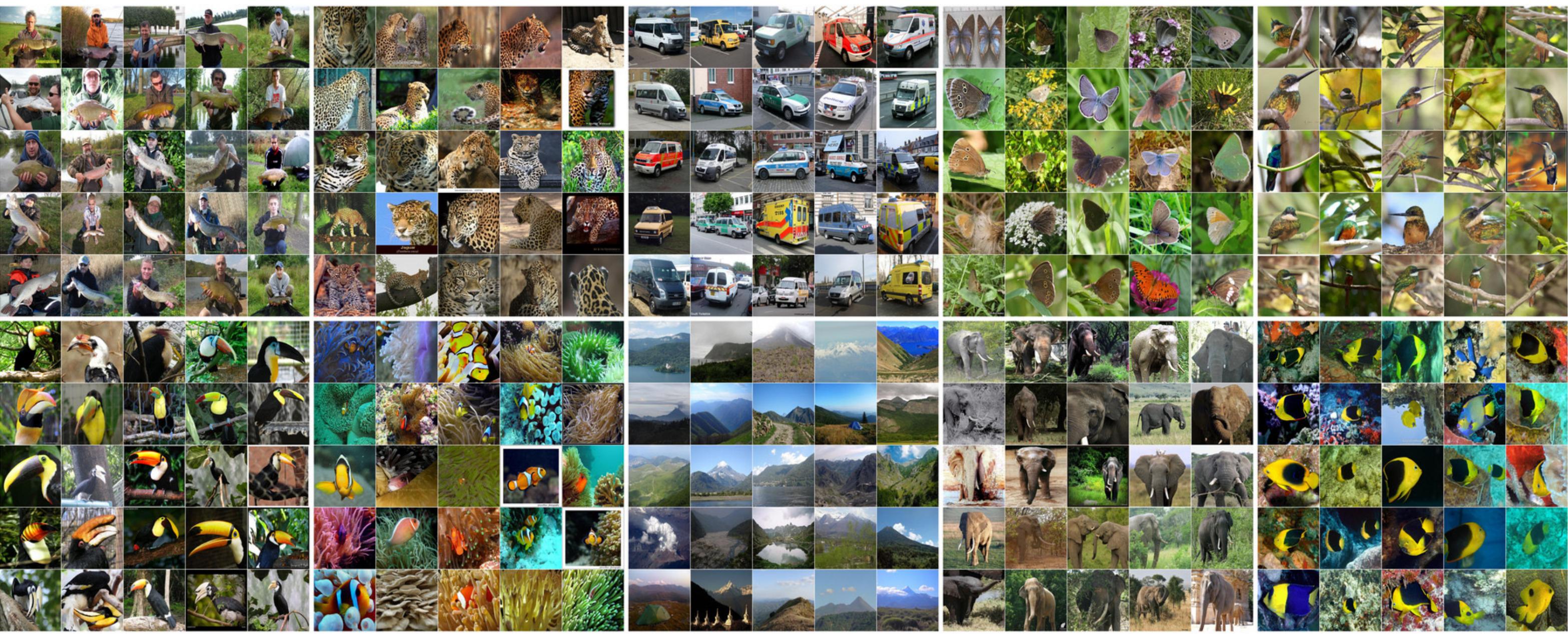


Image search results

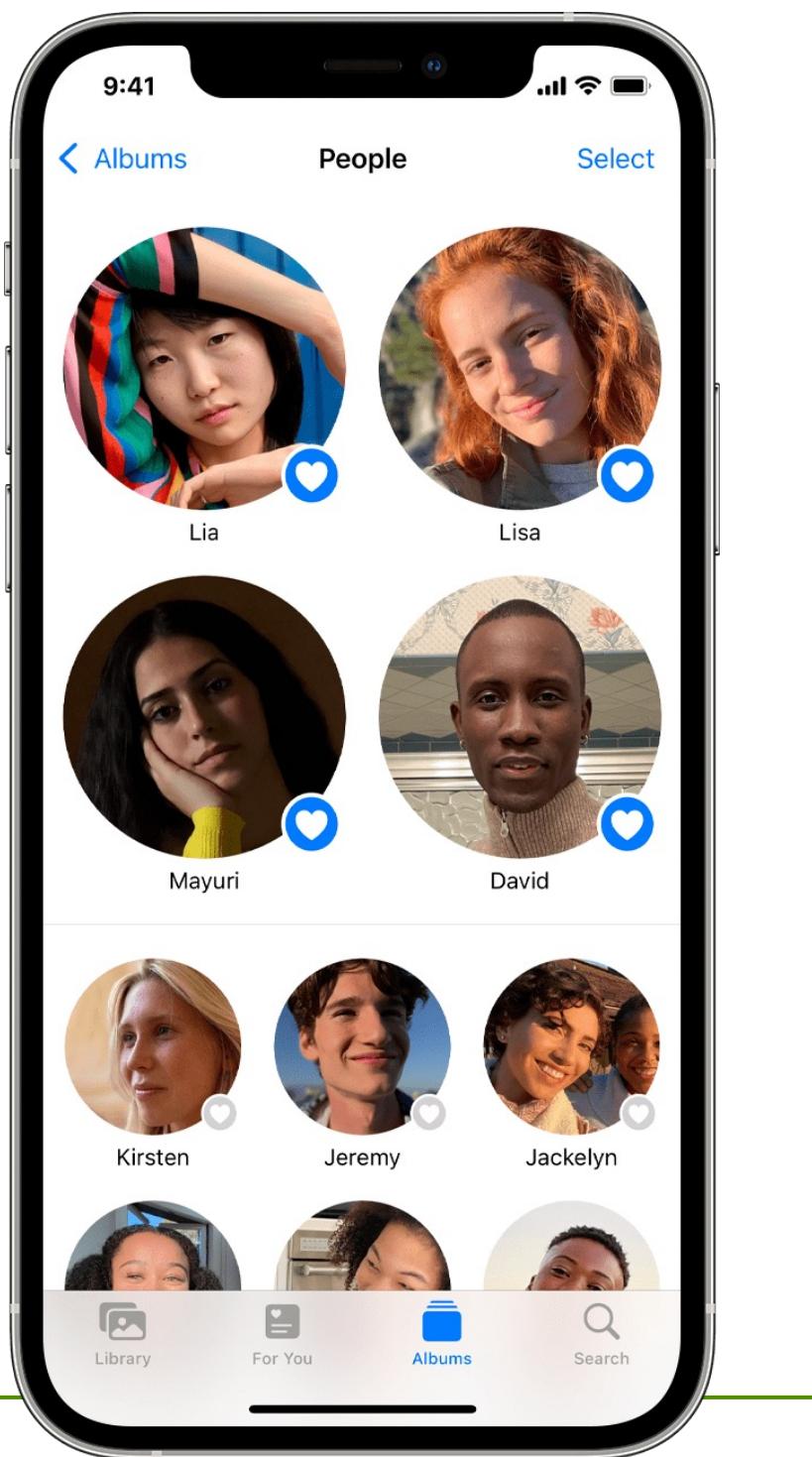
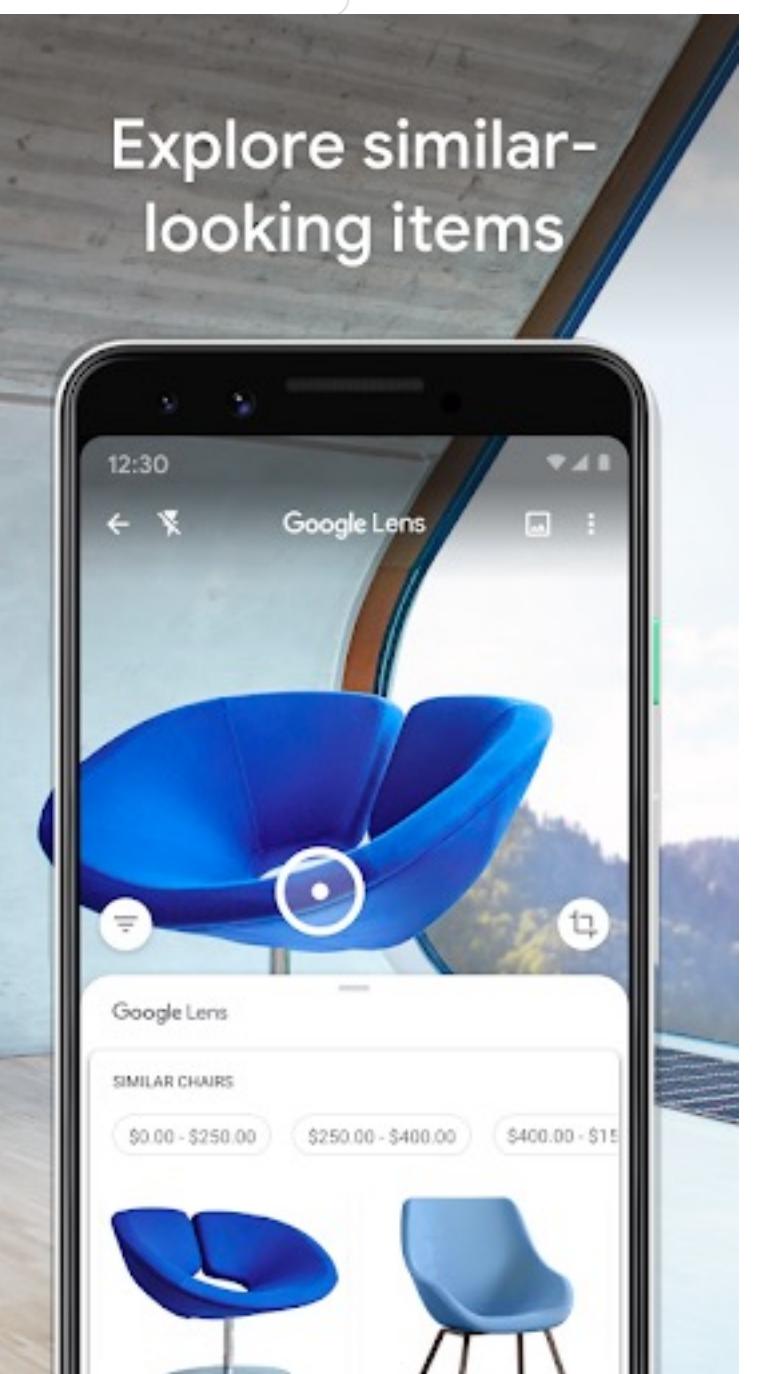


1 88432.jpg X Lamp

Visually similar images



Report images



Admin

Advanced Machine Learning For Design

Lecture 4 - Machine Learning
for Image Processing (part 1)

Module 2

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12/10/2022

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<https://aml4design.github.io/>

Sources

- CMU Computer Vision course - Matthew O'Toole. <http://16385.courses.cs.cmu.edu/spring2022/>
- Grokking Machine Learning. Luis G. Serrano. Manning, 2021
- [What is back propagation really doing?](#)
 - YouTube walk-through
- “[Gradient descent, how neural networks learn](#)”
 - YouTube walk-through
- “[Neural networks and back-propagation explained in a simple way](#)”
 - Article on medium.com
- “[The MNIST DATABASE](#)” of handwritten digits
- [“Neural Networks from Scratch”](#)
 - YouTube tutorial
- [“Neural Networks from Scratch in Python”](#)
 - New book/e-book

