

# Machine Learning For Design

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Lecture 3 - Machine Learning for Images

Alessandro Bozzon

16/02/2022

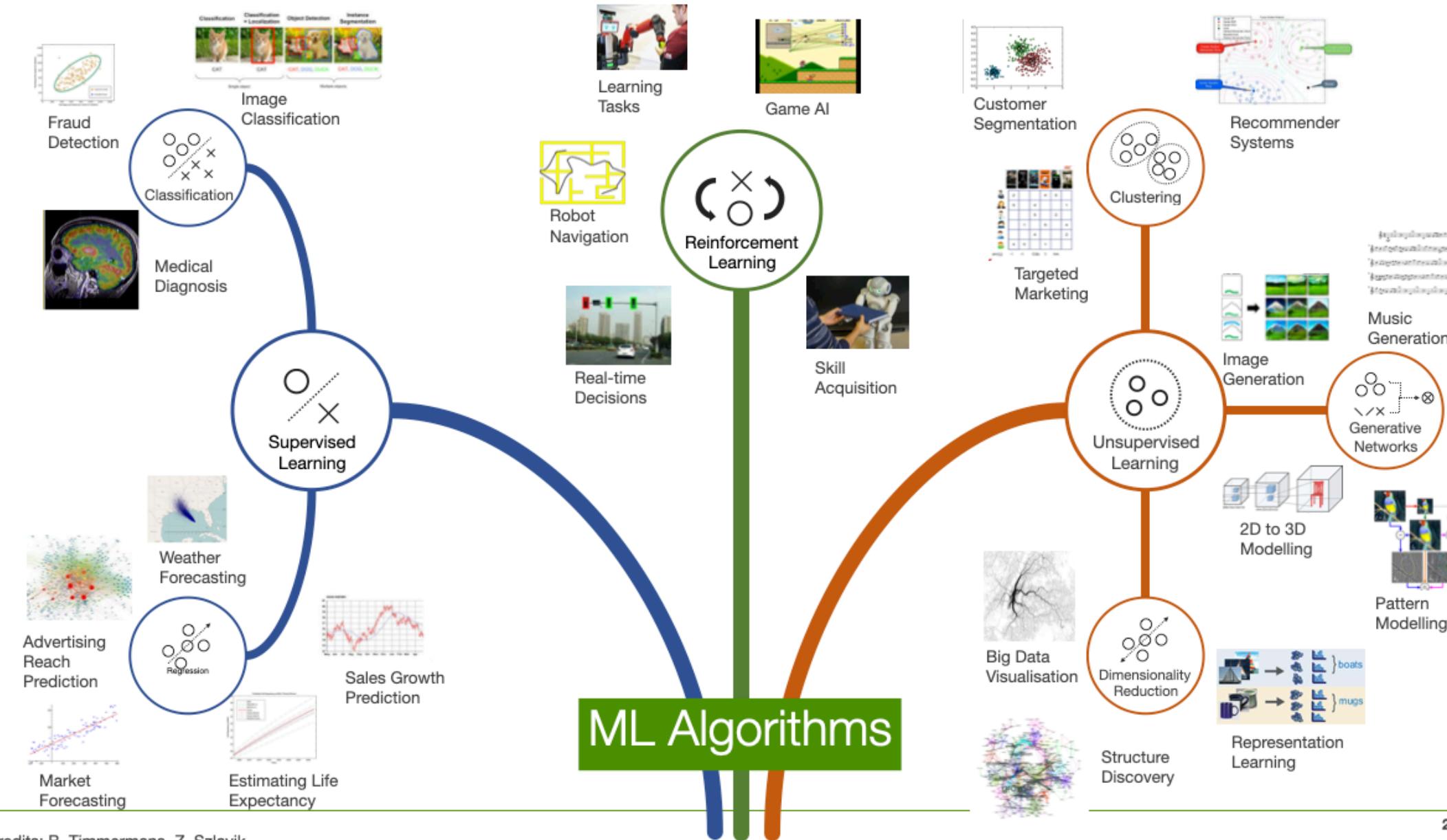
[mlfd-io@tudelft.nl](mailto:mlfd-io@tudelft.nl)  
[www.ml4design.com](http://www.ml4design.com)

# Admin

# Week 2 Tasks



**Previously,  
on ML4D. . .**

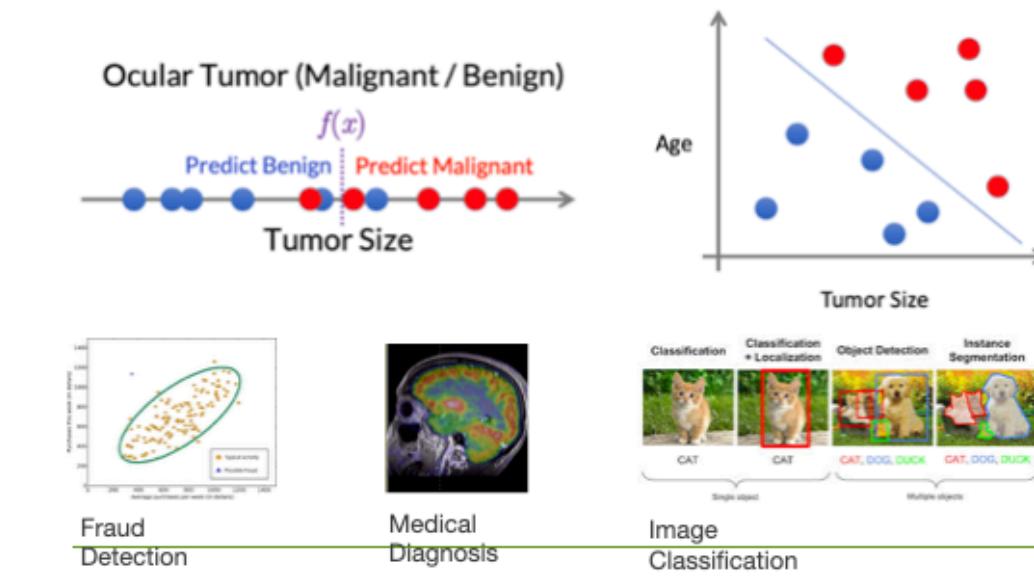


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# Classification / Regression

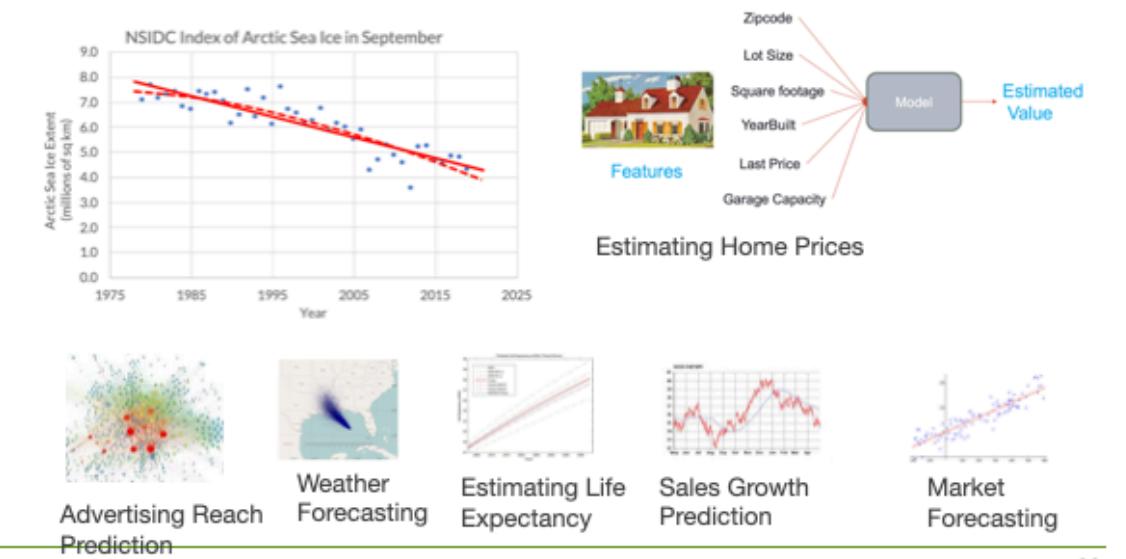
## Classification

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



## Regression

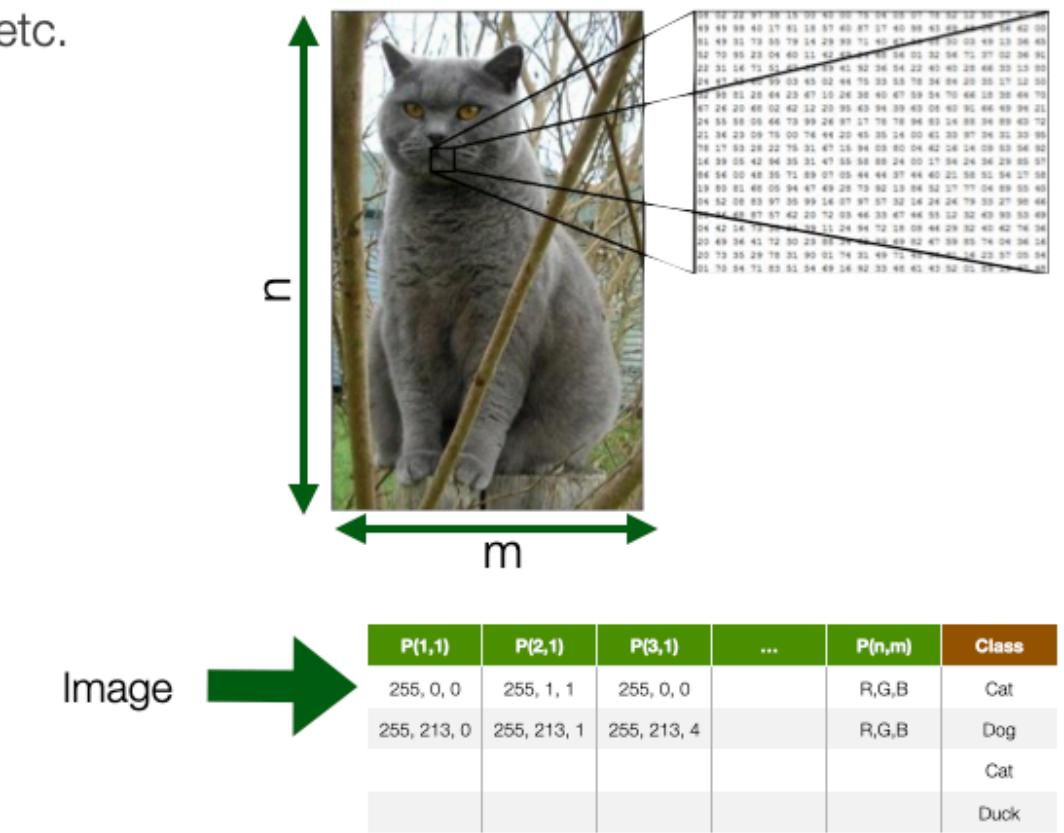
- Learn to guess one or more numbers
  - e.g. value of a share, number of stars in a review



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## 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 1

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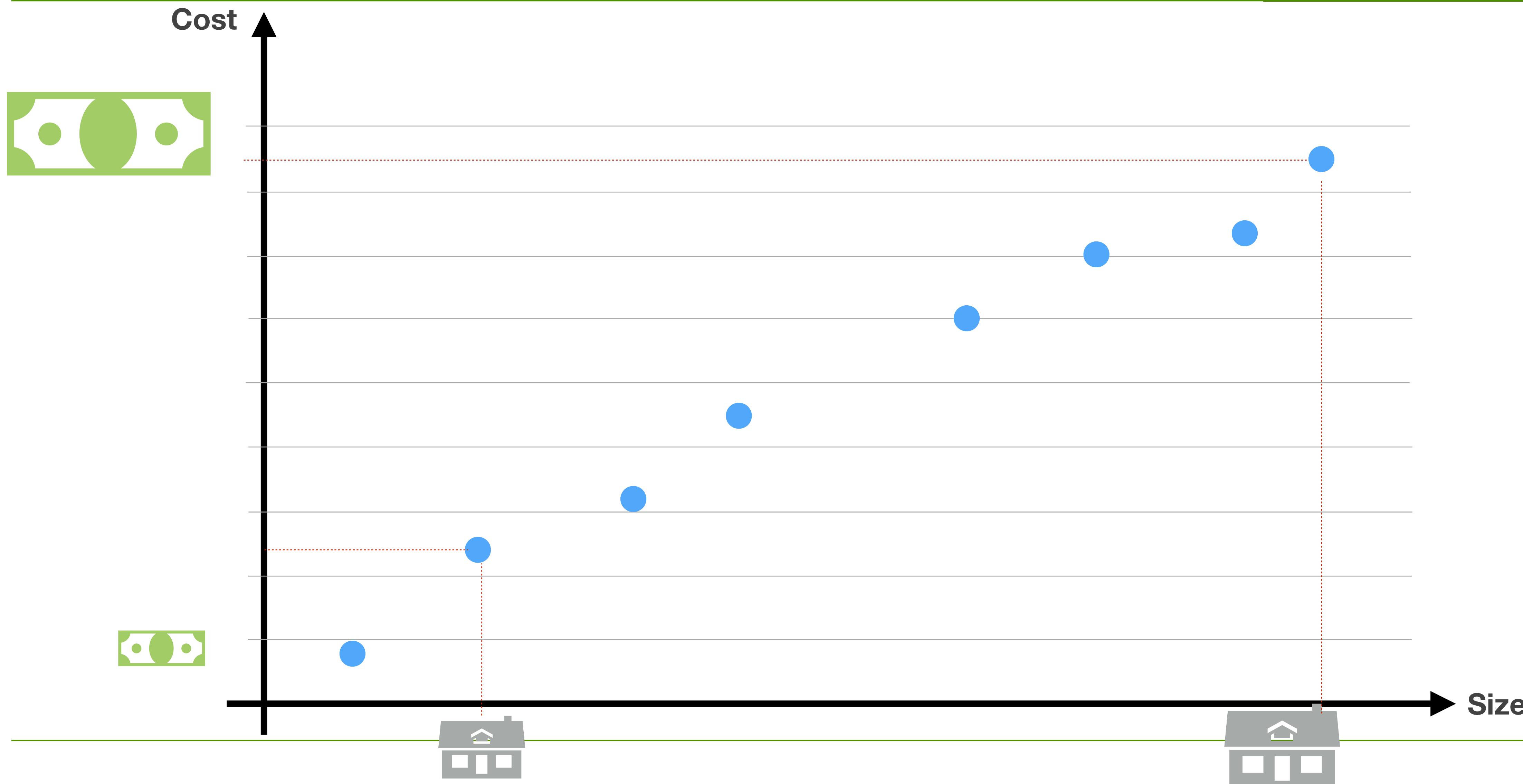
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# A bit more on regression and classification

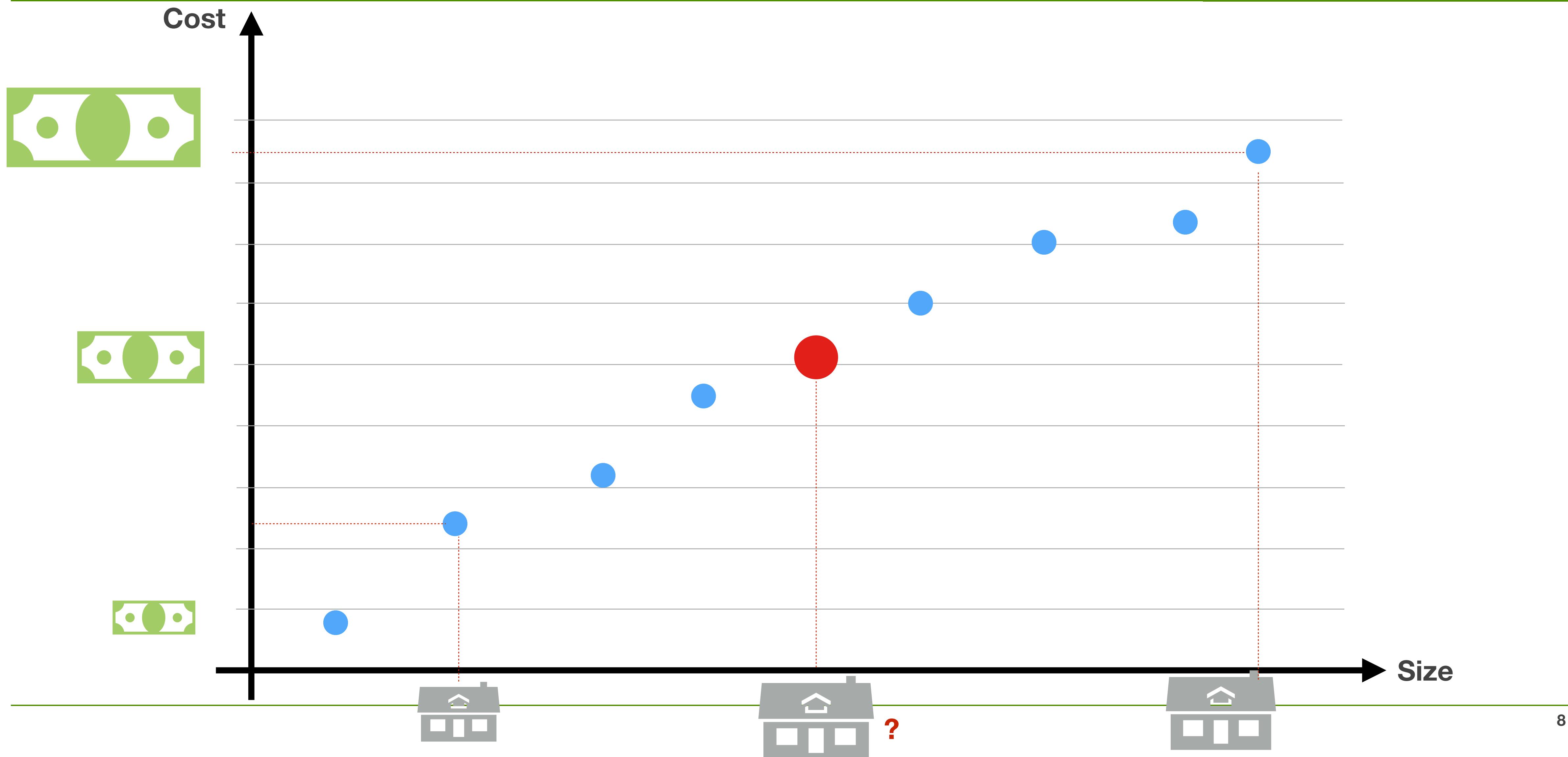
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**And your very first contact with (deep) neural networks**

# Linear Regression /1



# Linear Regression /2



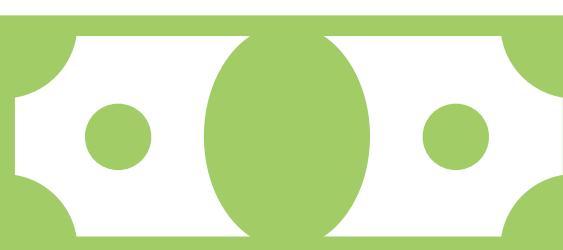
# Linear Regression

Intercept (parameter, or **weight**)

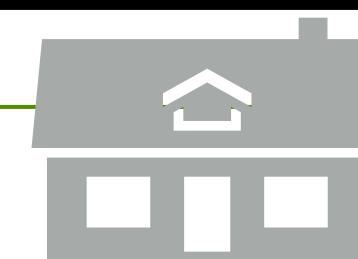
Slope (parameter, or **weight**)

Dependent Variable

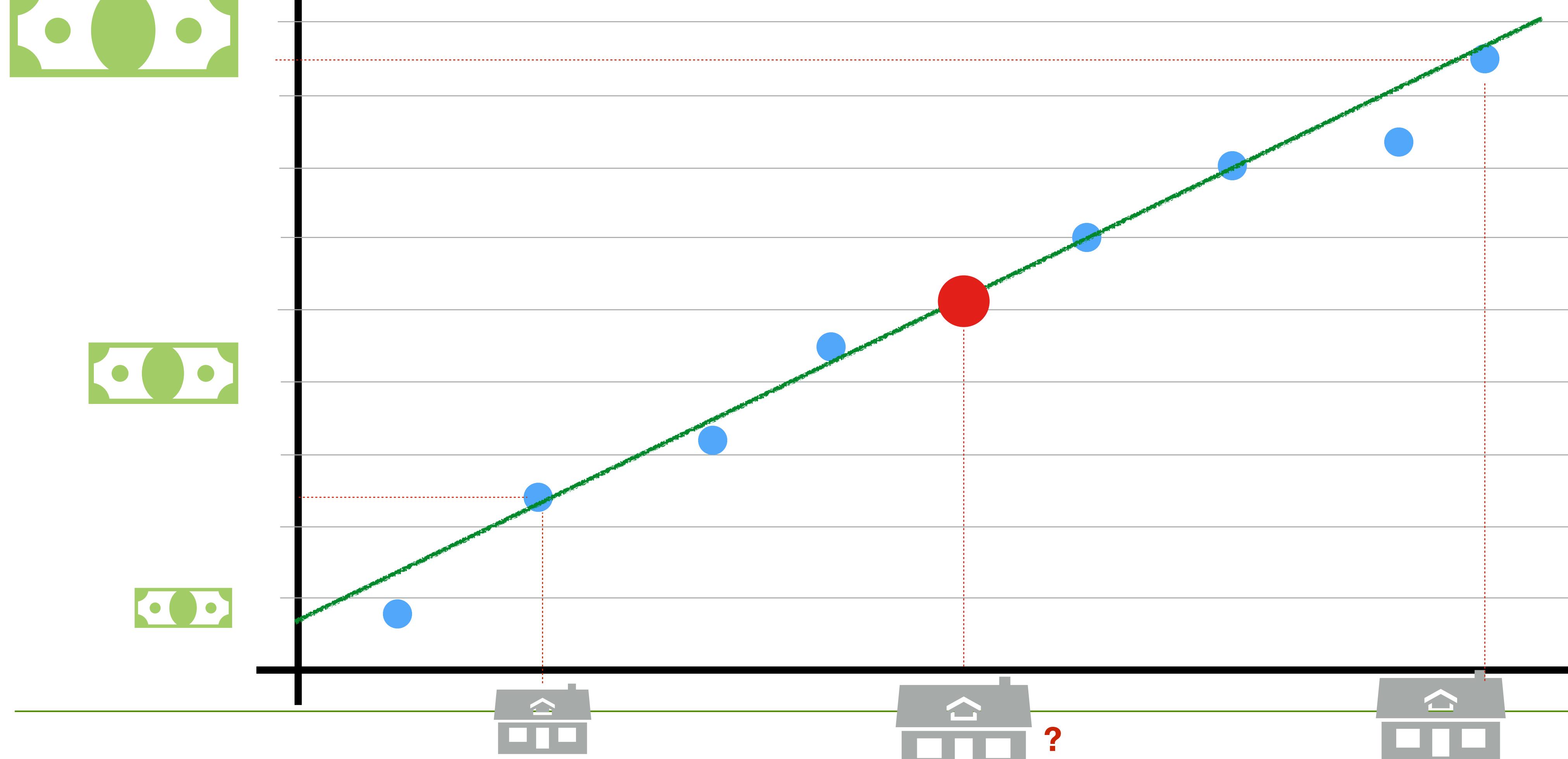
Cost =  $a + b \text{ Size}$



Cost



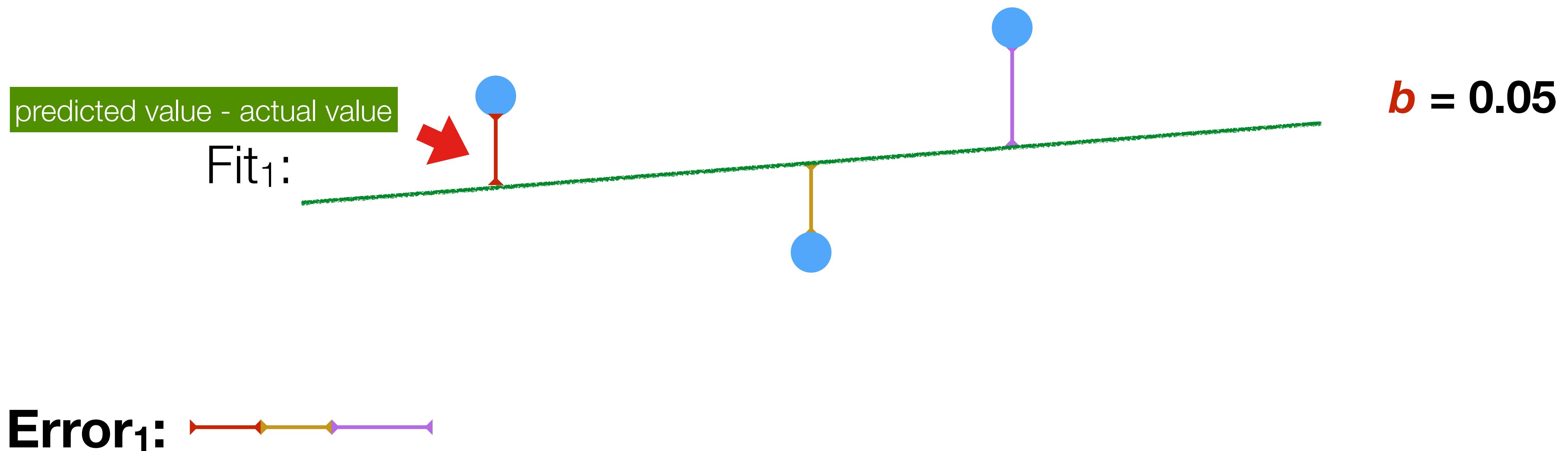
Size



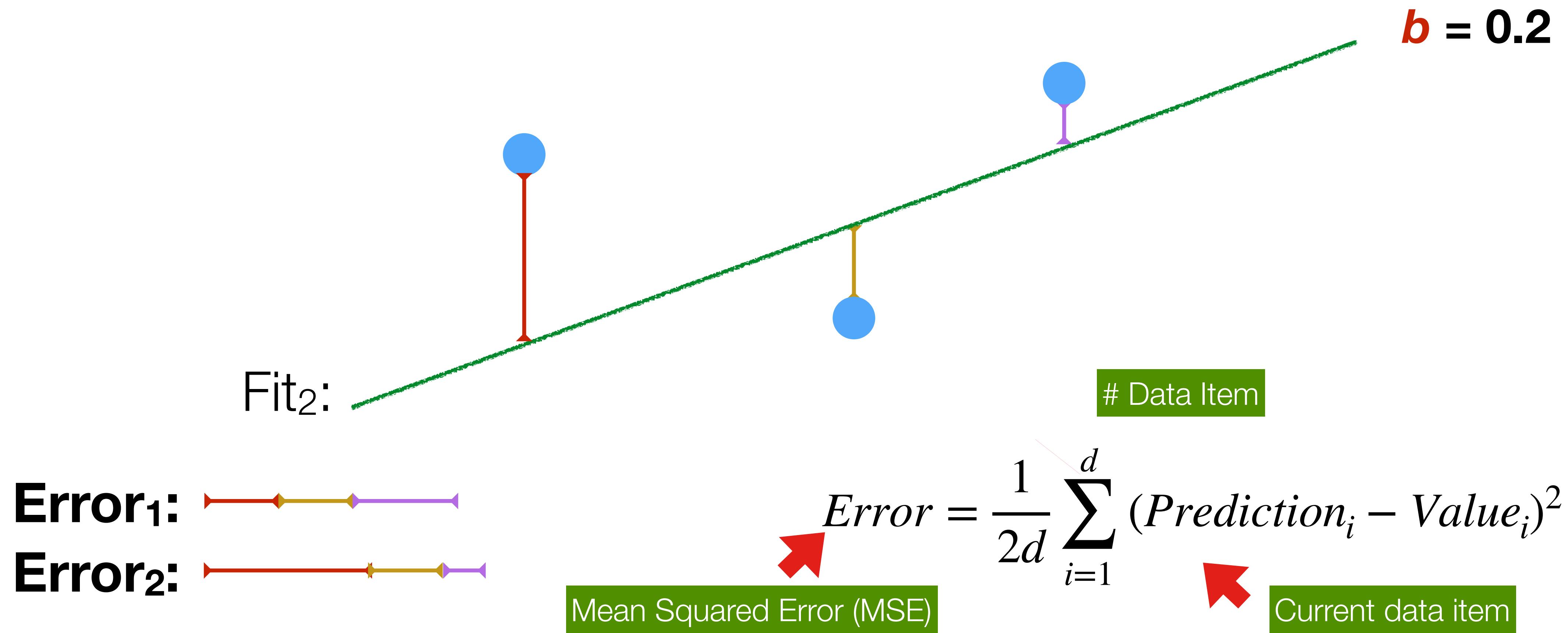
**Cost = *b* Size**



**Cost =  $b$  Size**

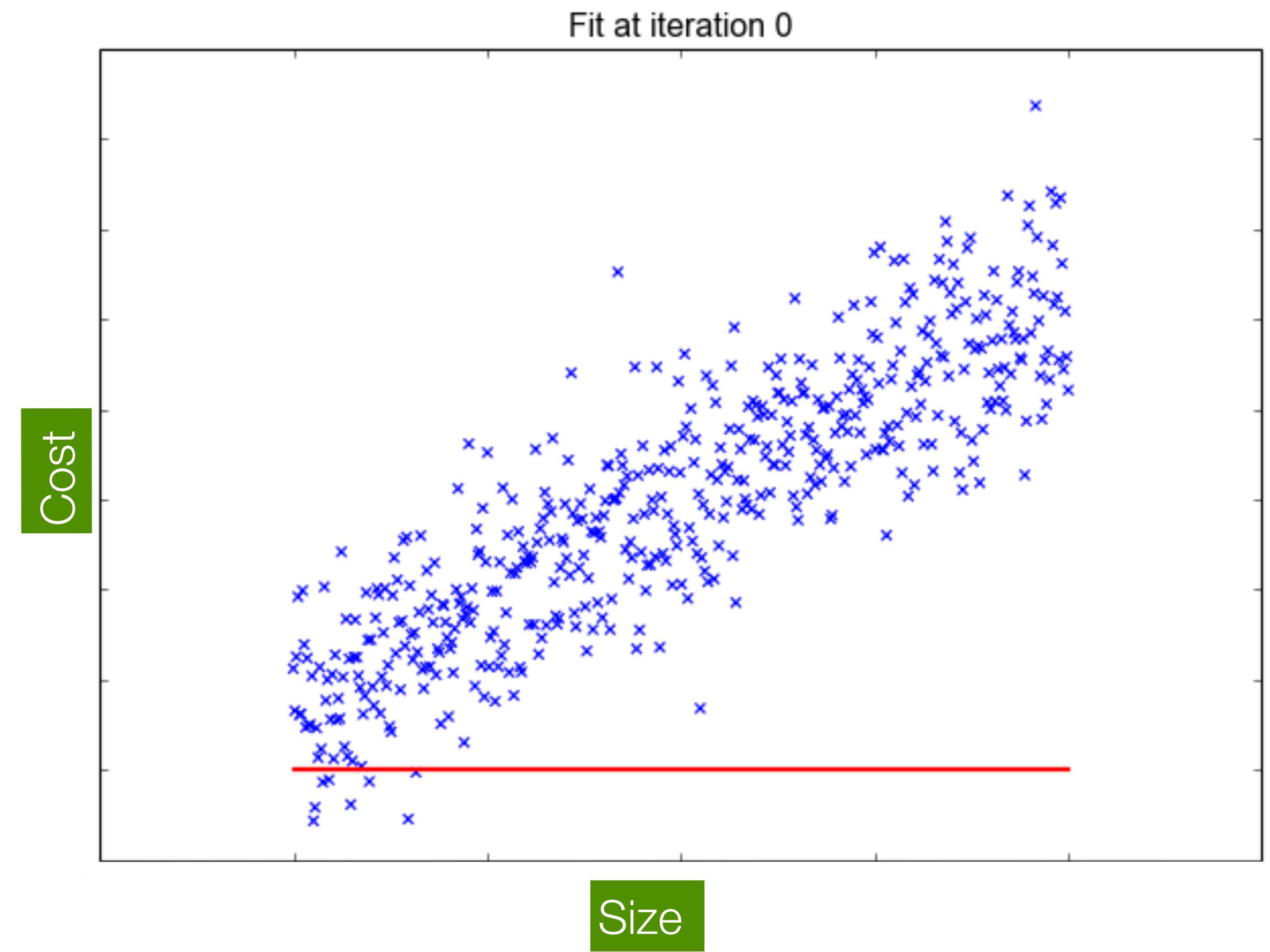
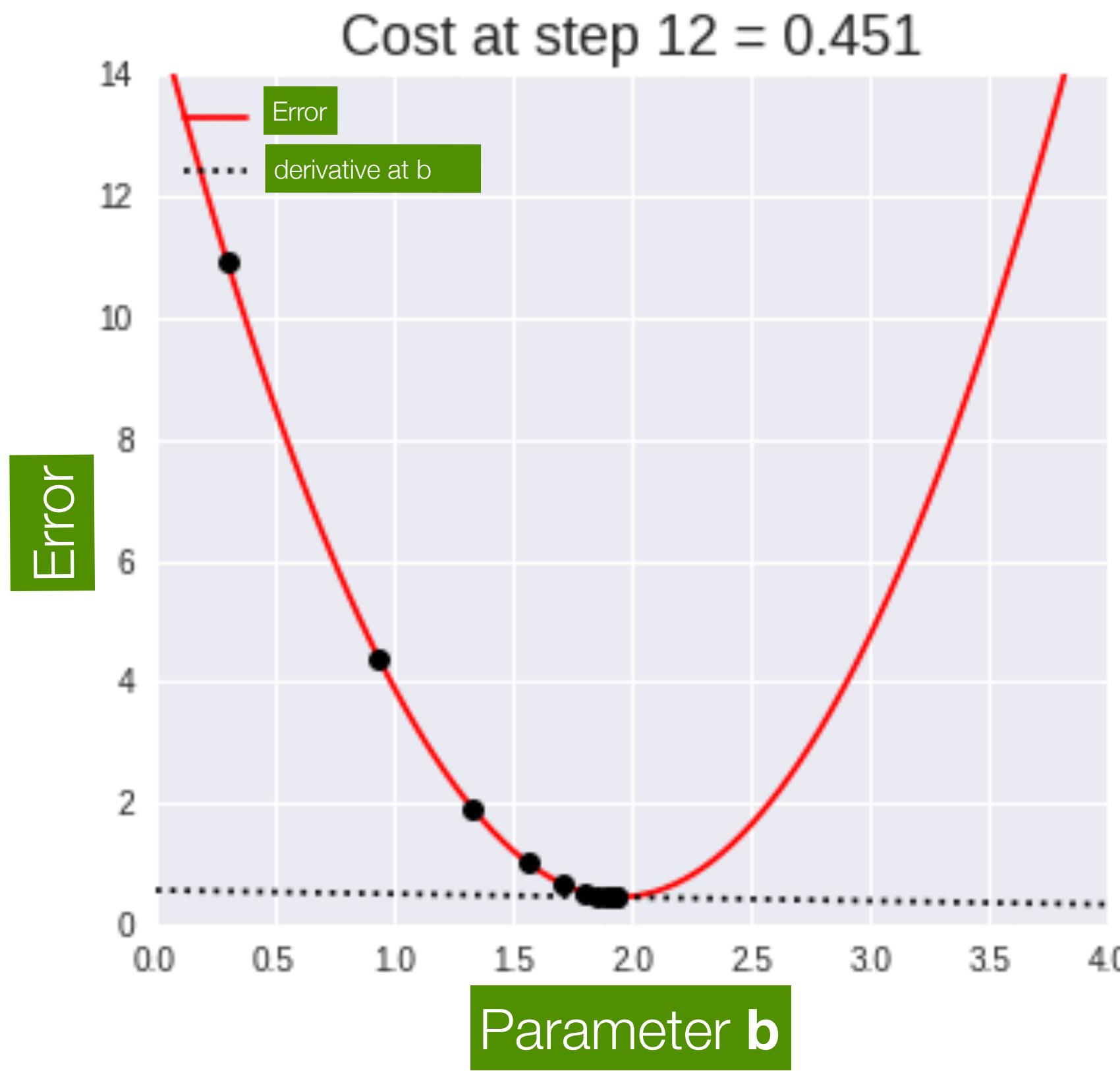


**Cost =  $b$  Size**



# Finding the best parameter values (training the model)

## ■ Gradient descent



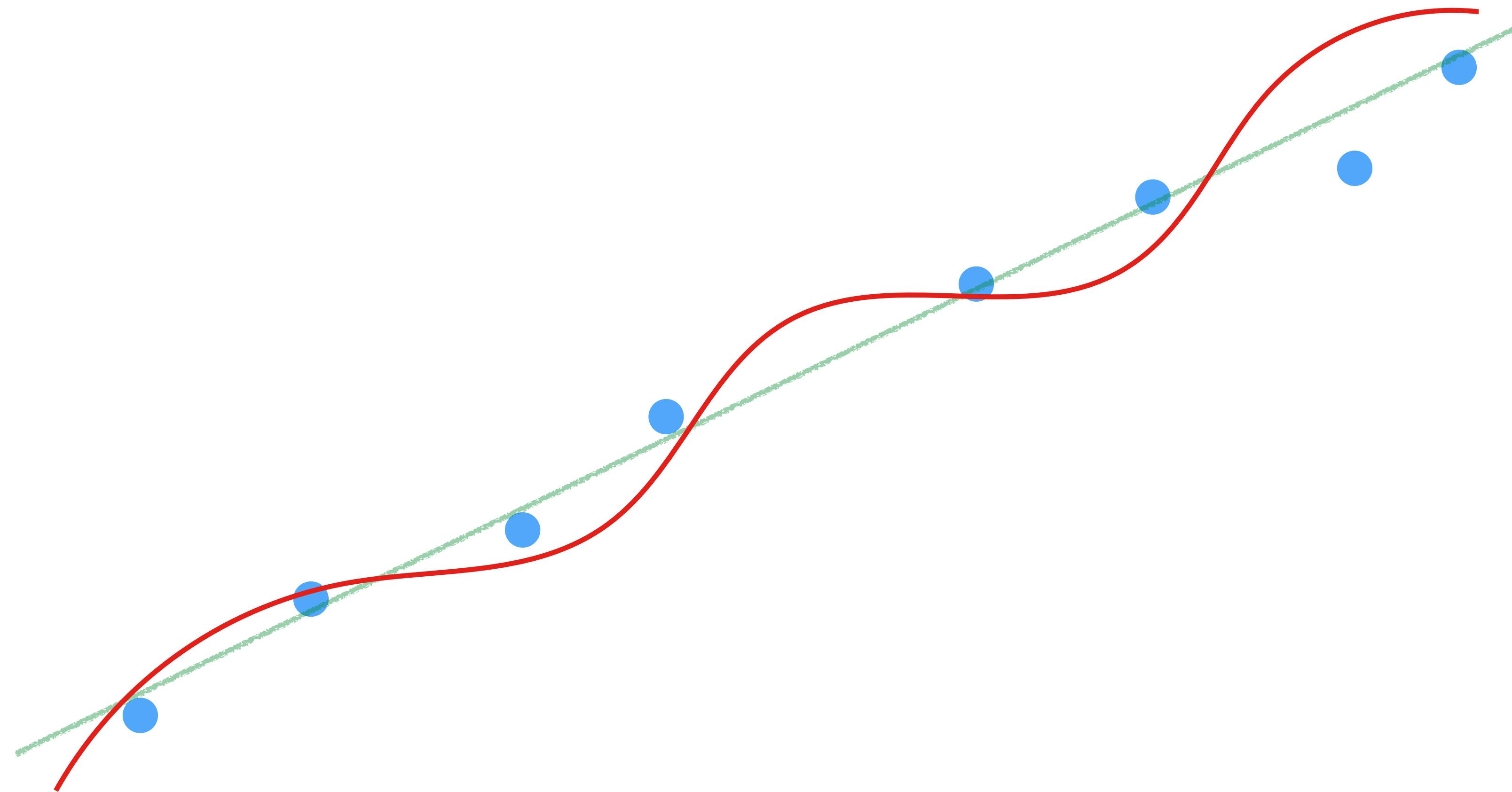
## ■ Hyperparameters

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

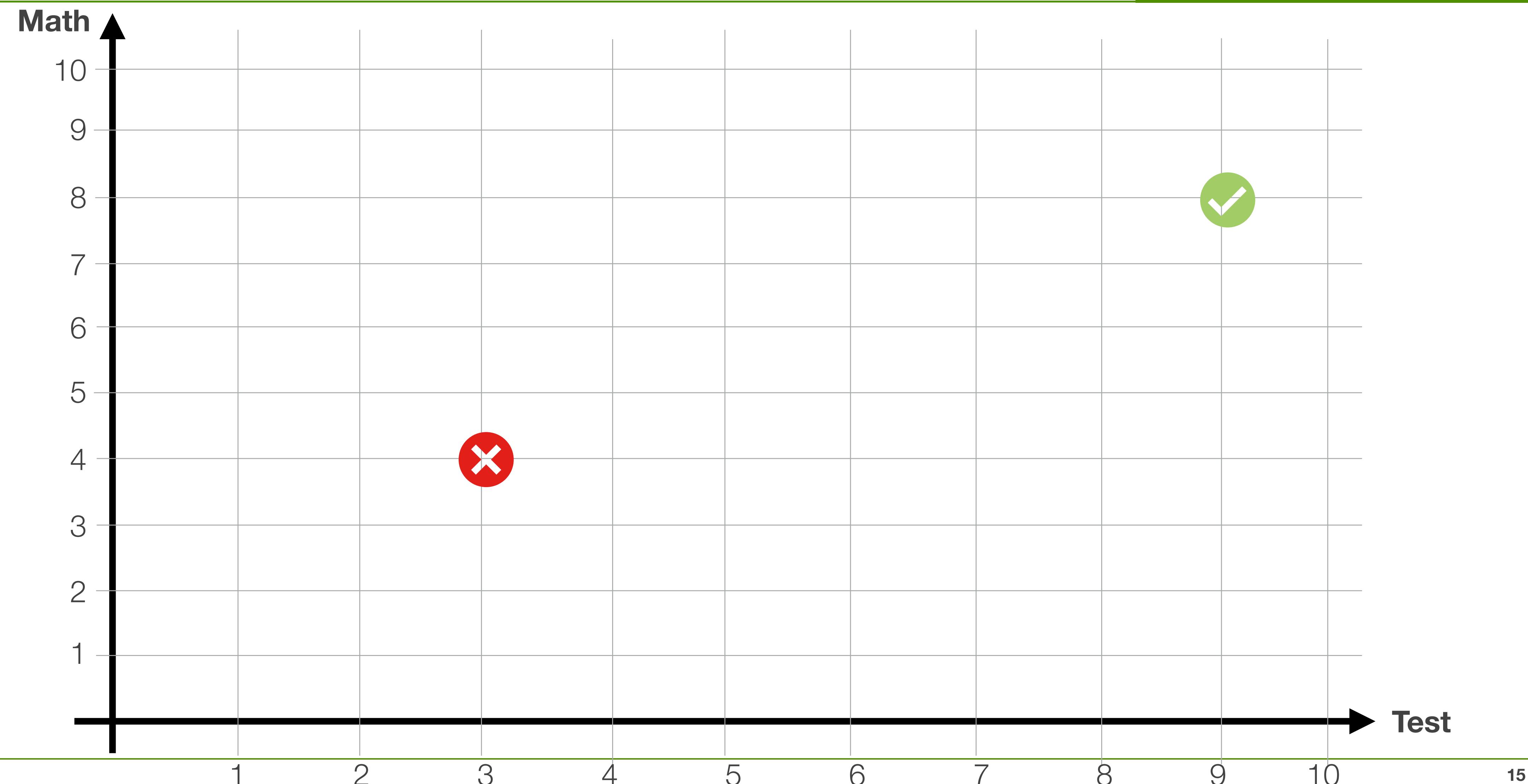
# Polynomial Regression

N<sup>th</sup> degree polynomial

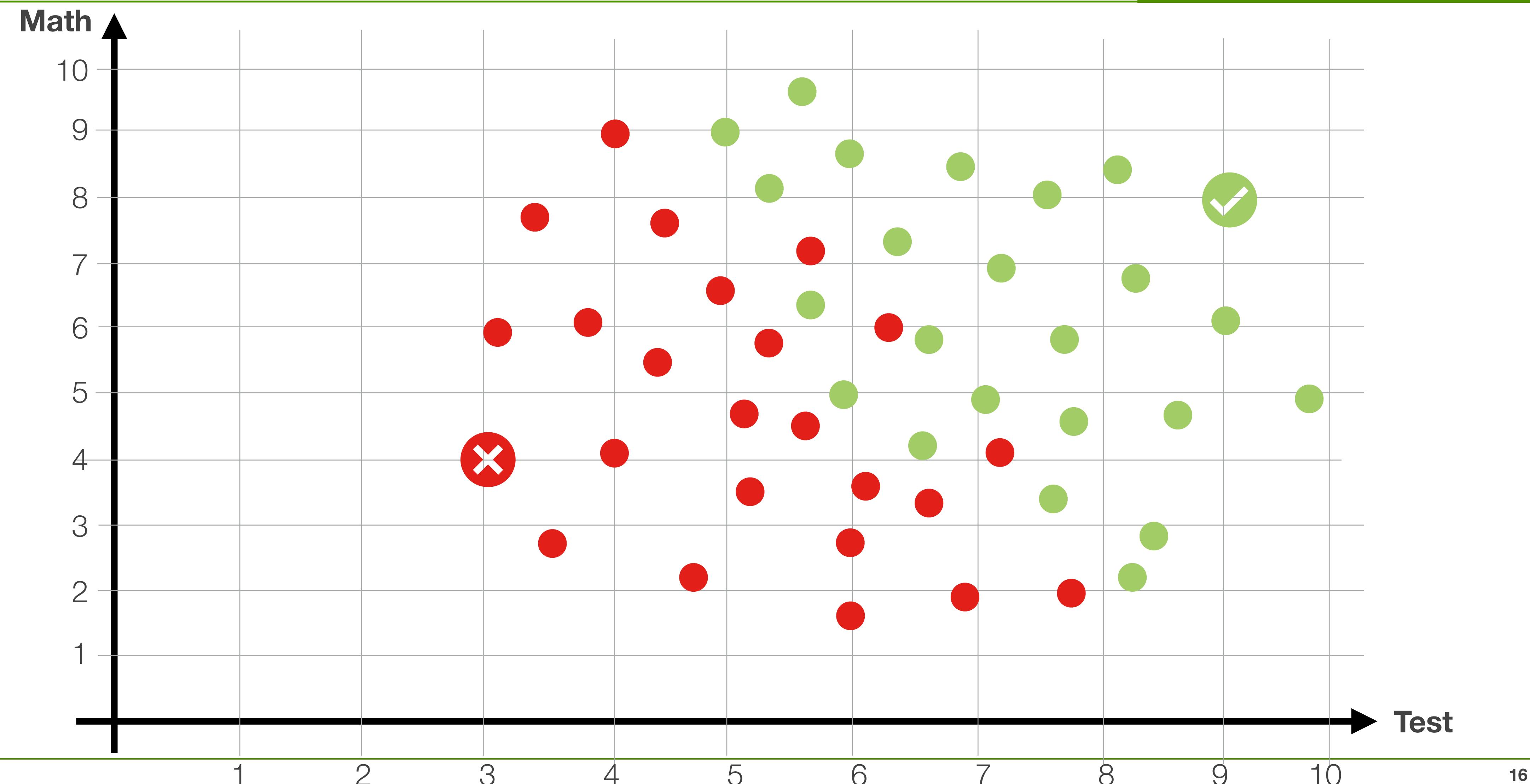
→ Cost =  $a + b \text{ Size} + c \text{ Size}^2 + \dots + w \text{ Size}^n$



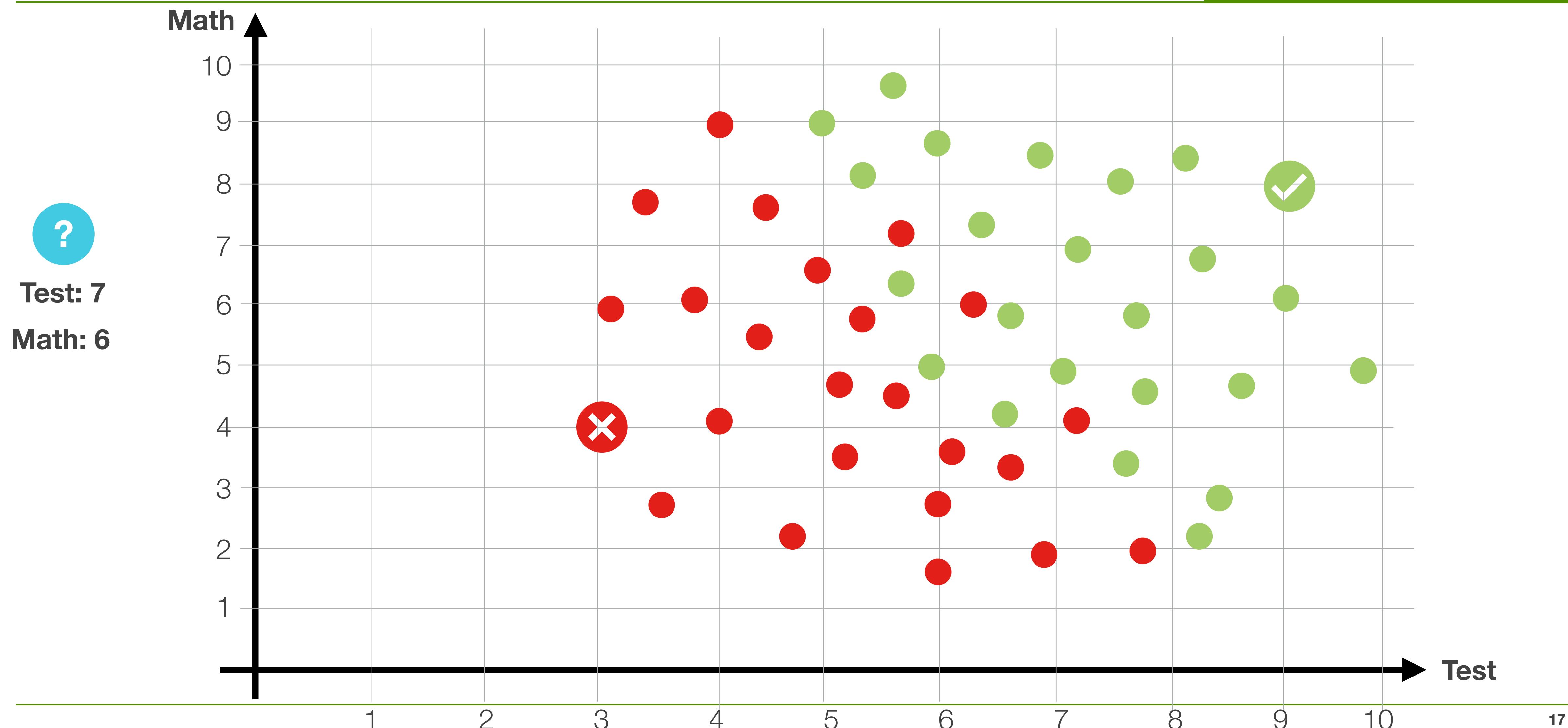
# Classification



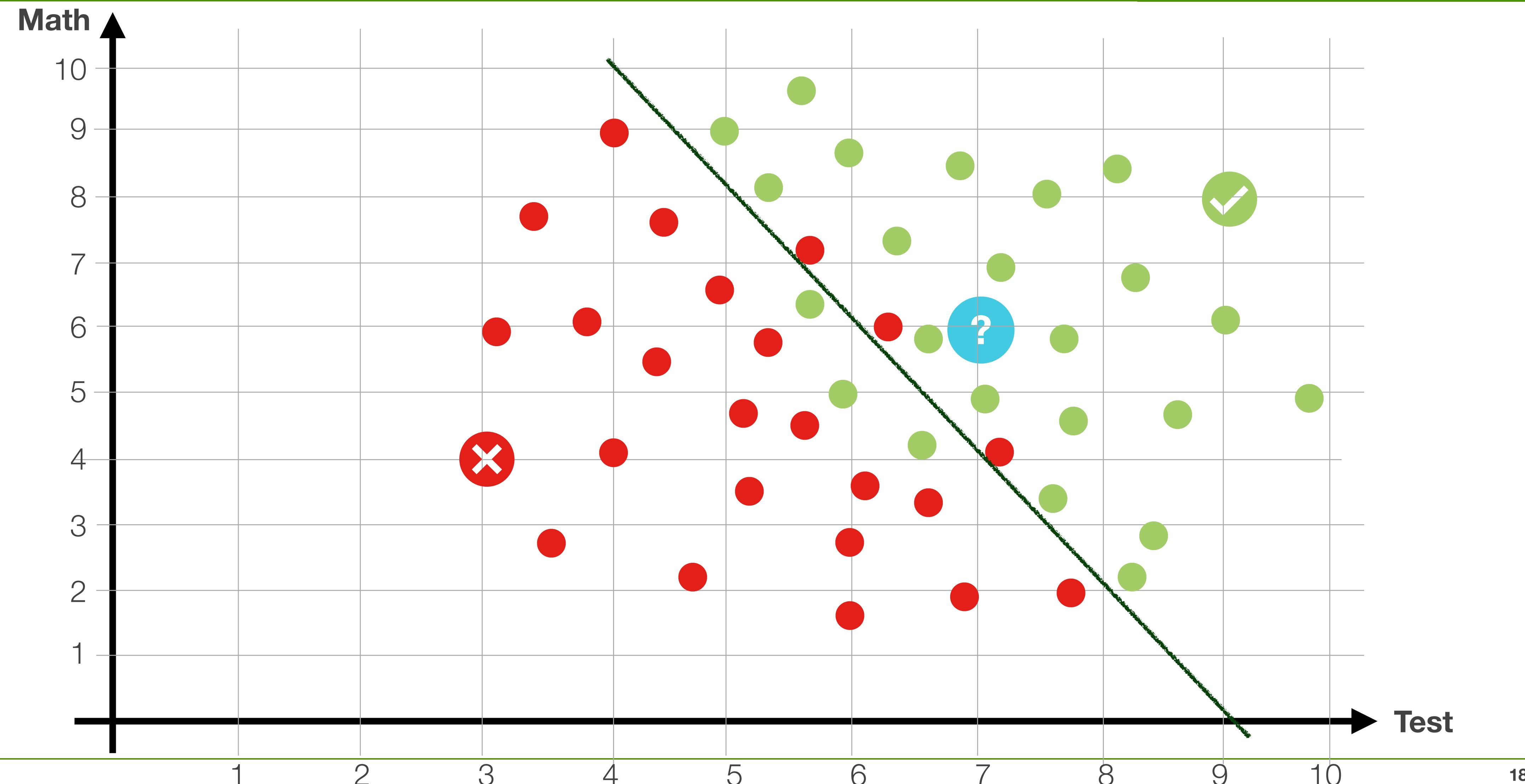
# Classification

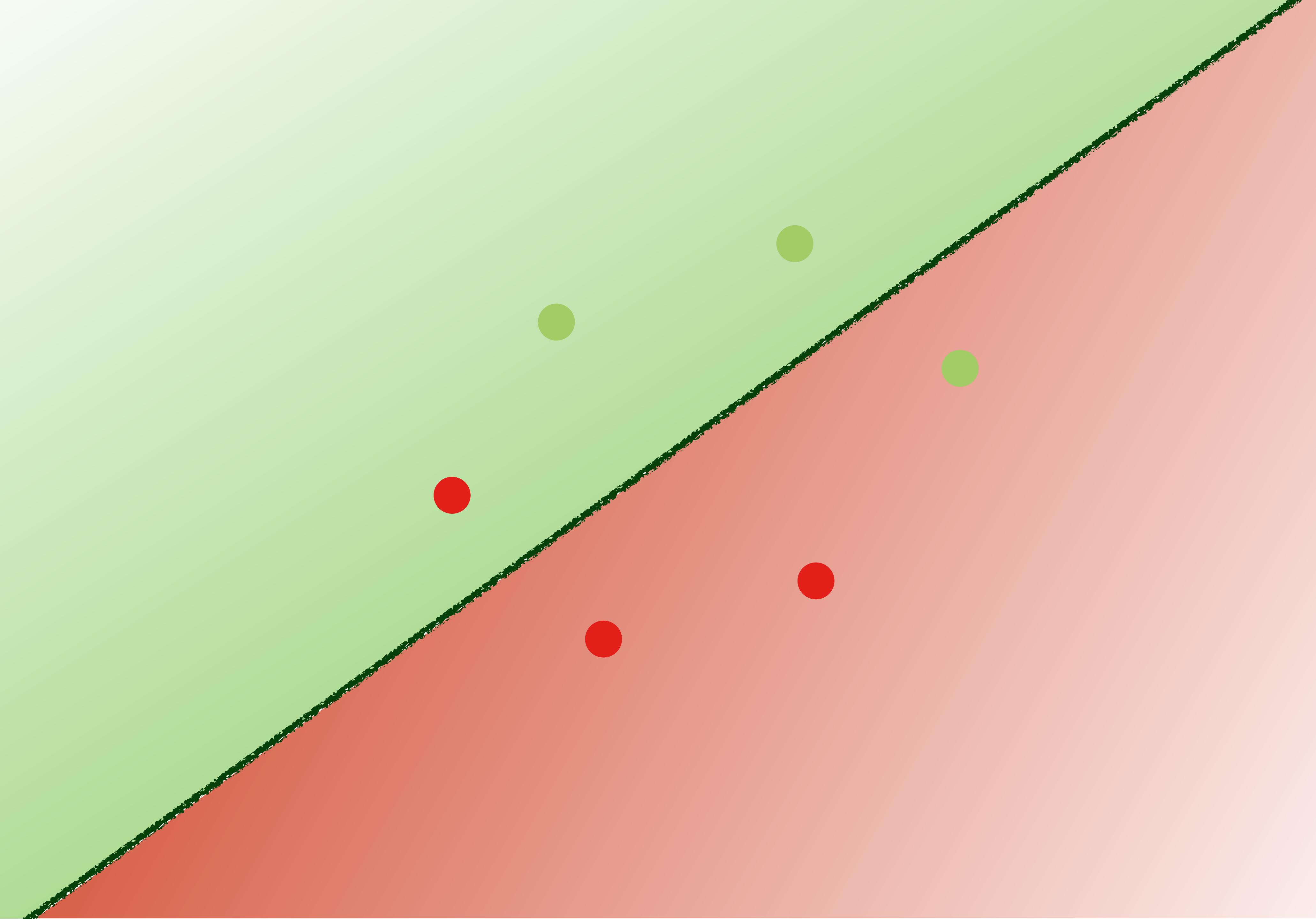


# Classification

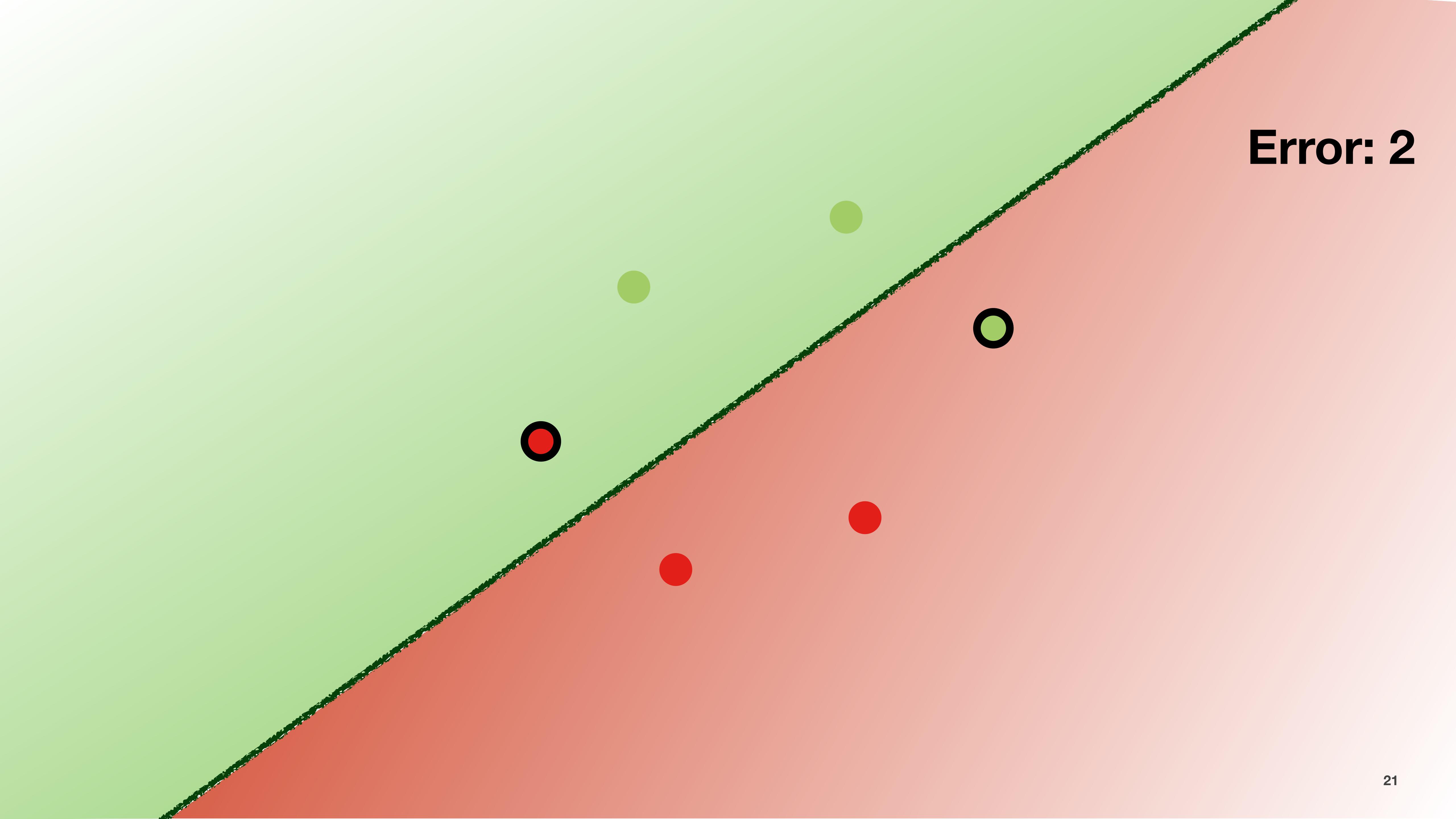


# Logistic Regression

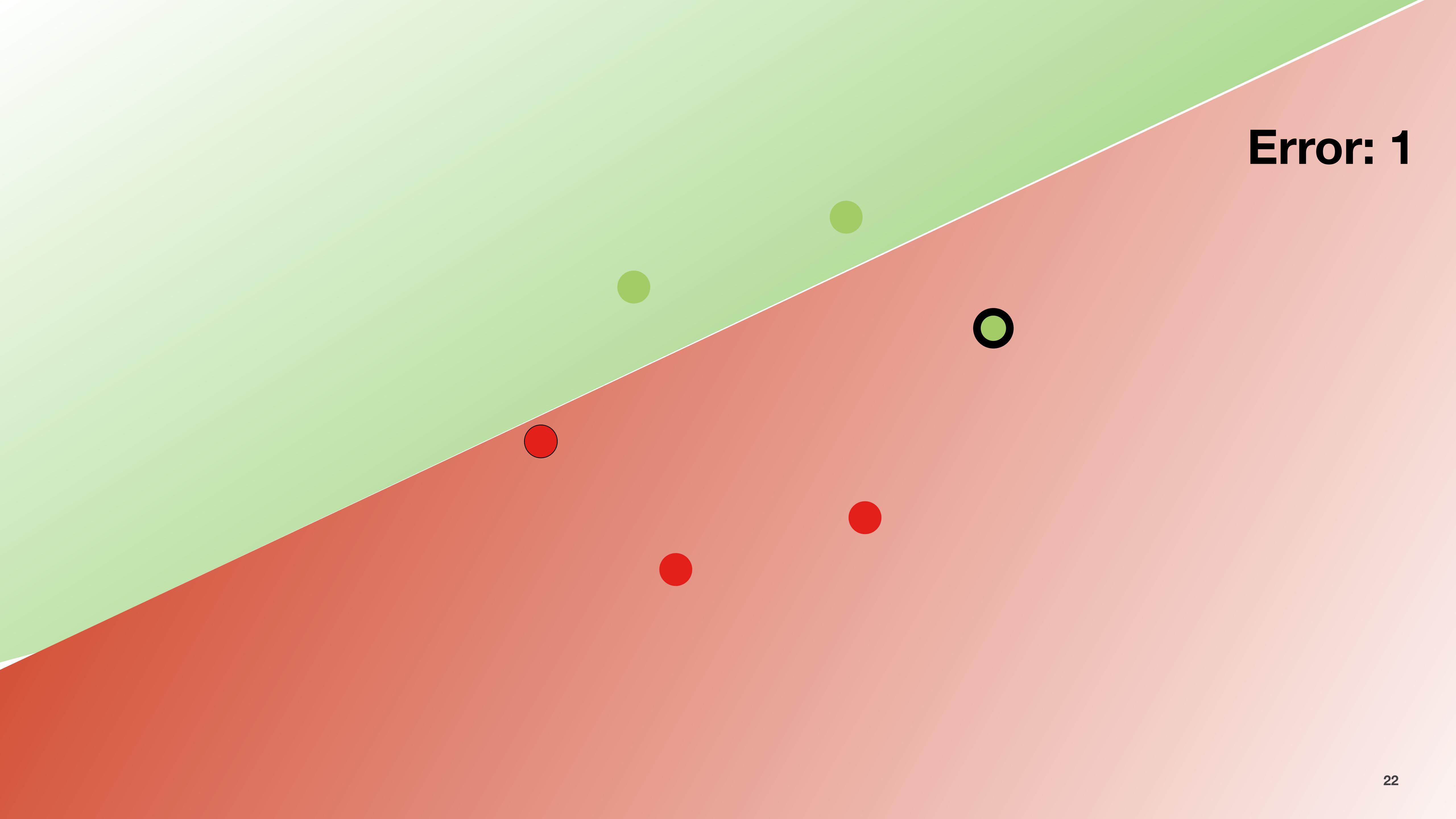




Error: 2

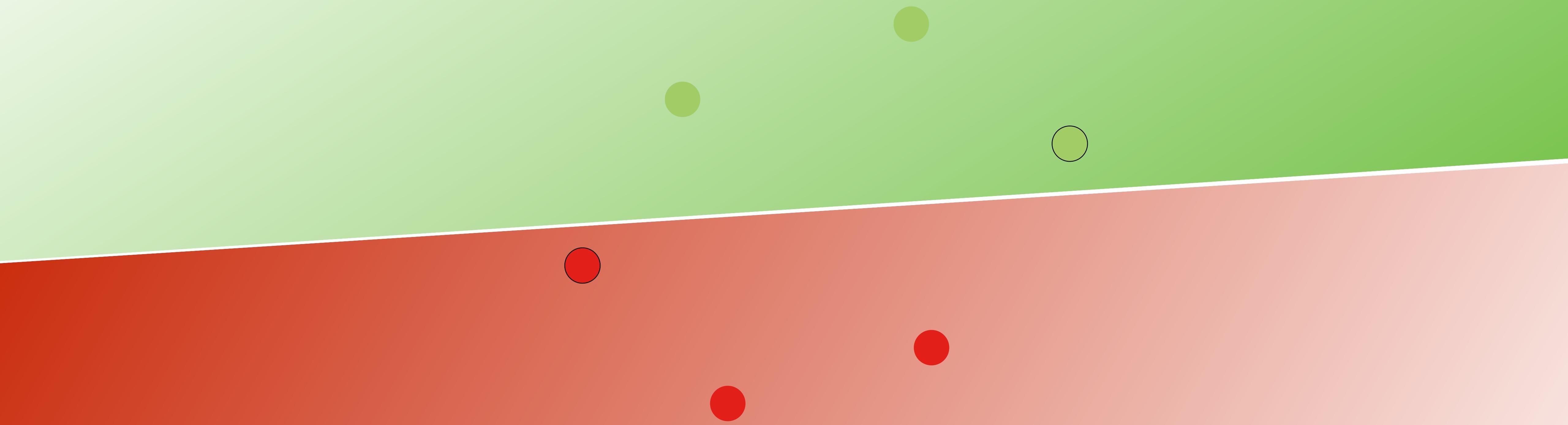


Error: 2



**Error: 1**

**Error: 0**



**Math**

10

9

8

7

6

5

4

3

2

1

**Test**

1

2

3

4

5

6

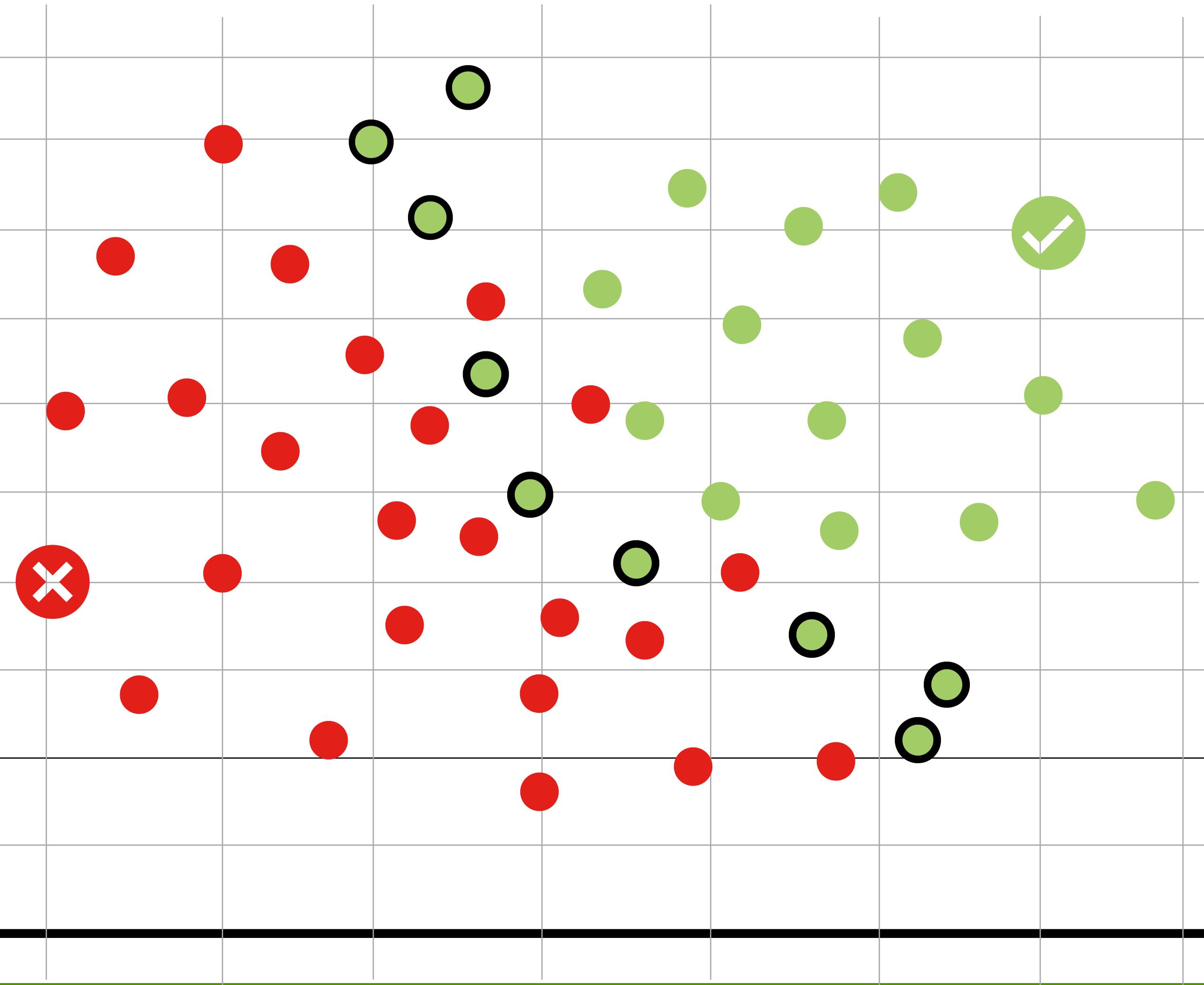
7

8

9

10

24



**Math**

10

9

8

7

6

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2

1

**Test**

1

2

3

4

5

6

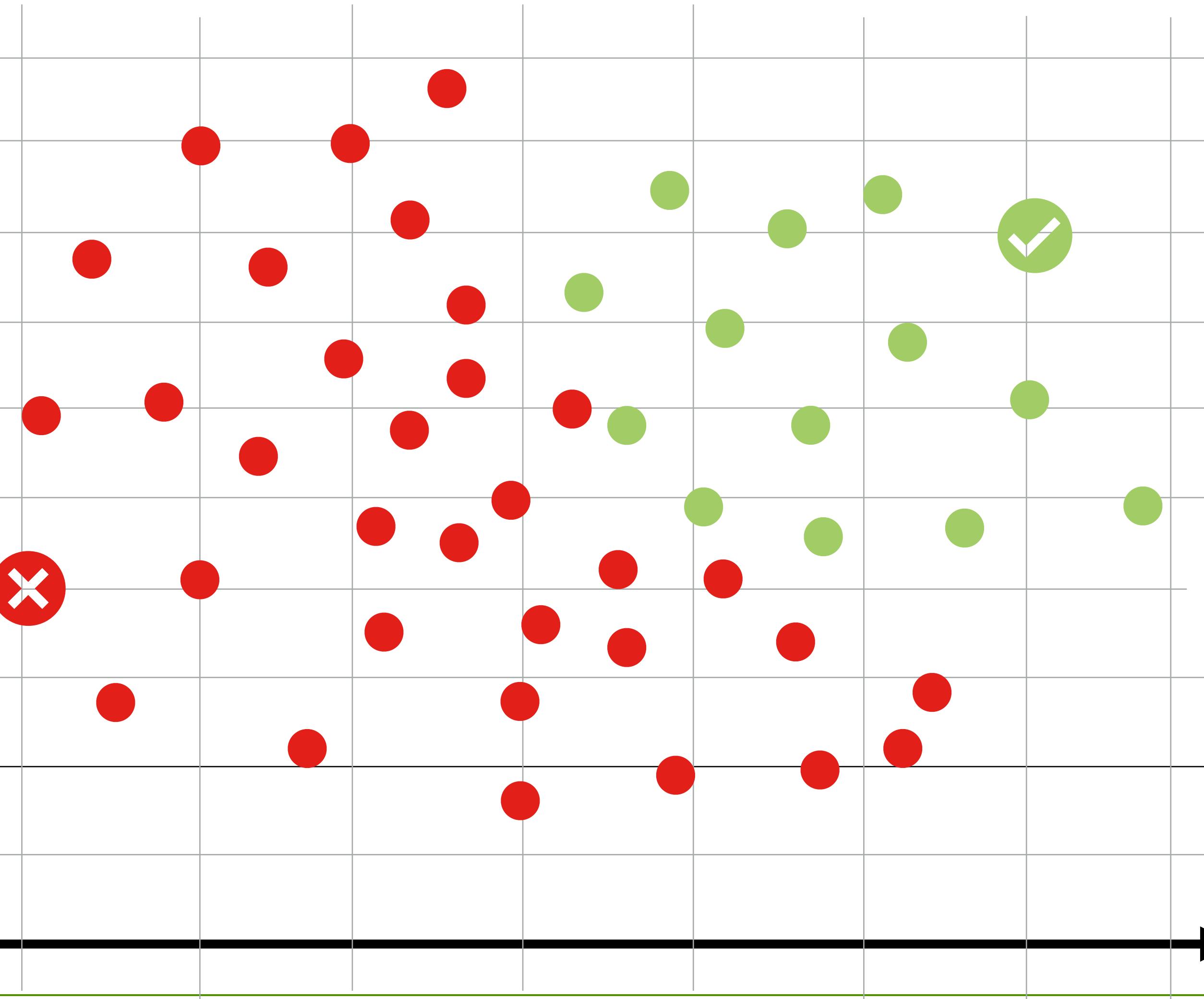
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8

9

10

25



**Math**

10

9

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3

2

1

**Test**

1

2

3

4

5

6

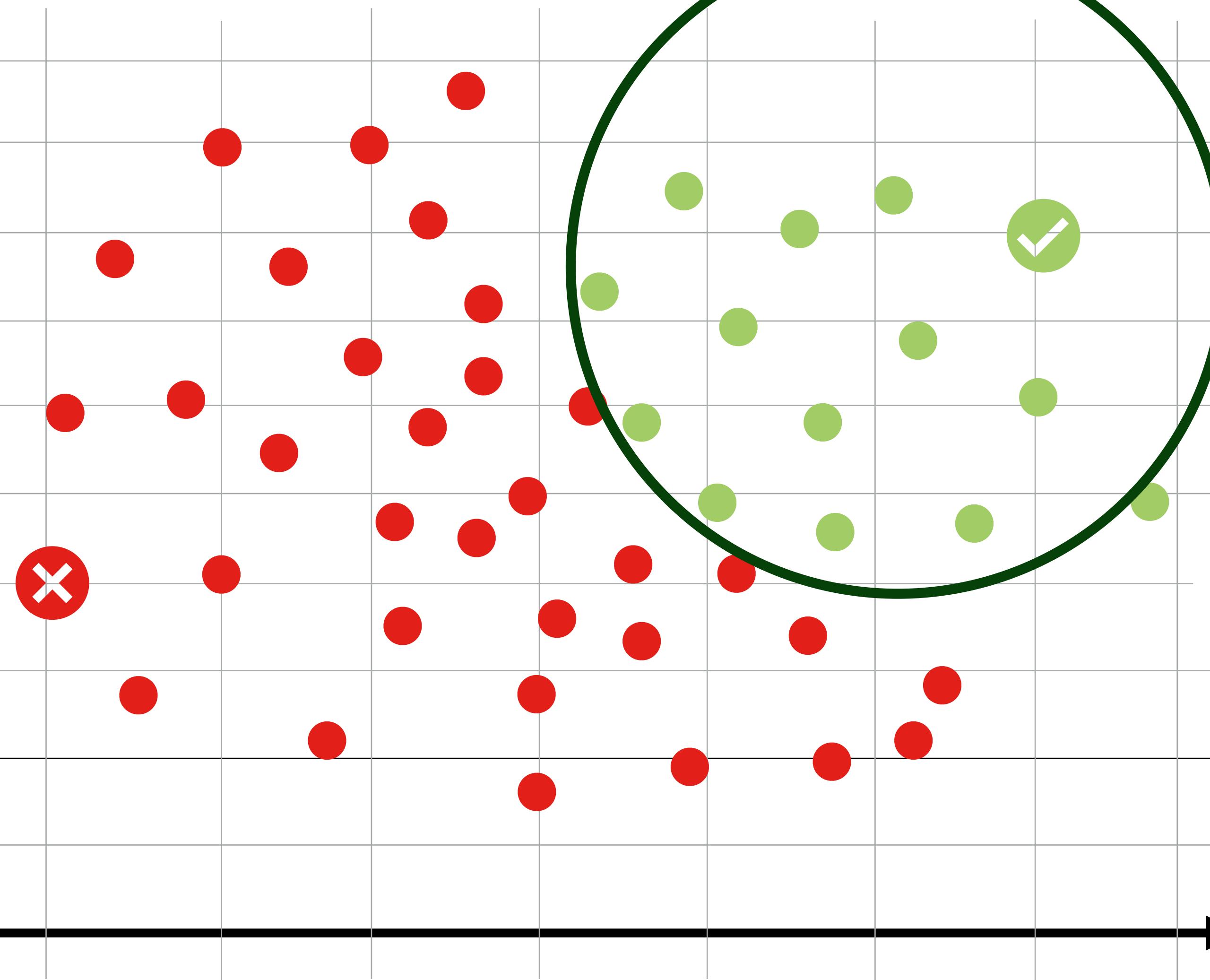
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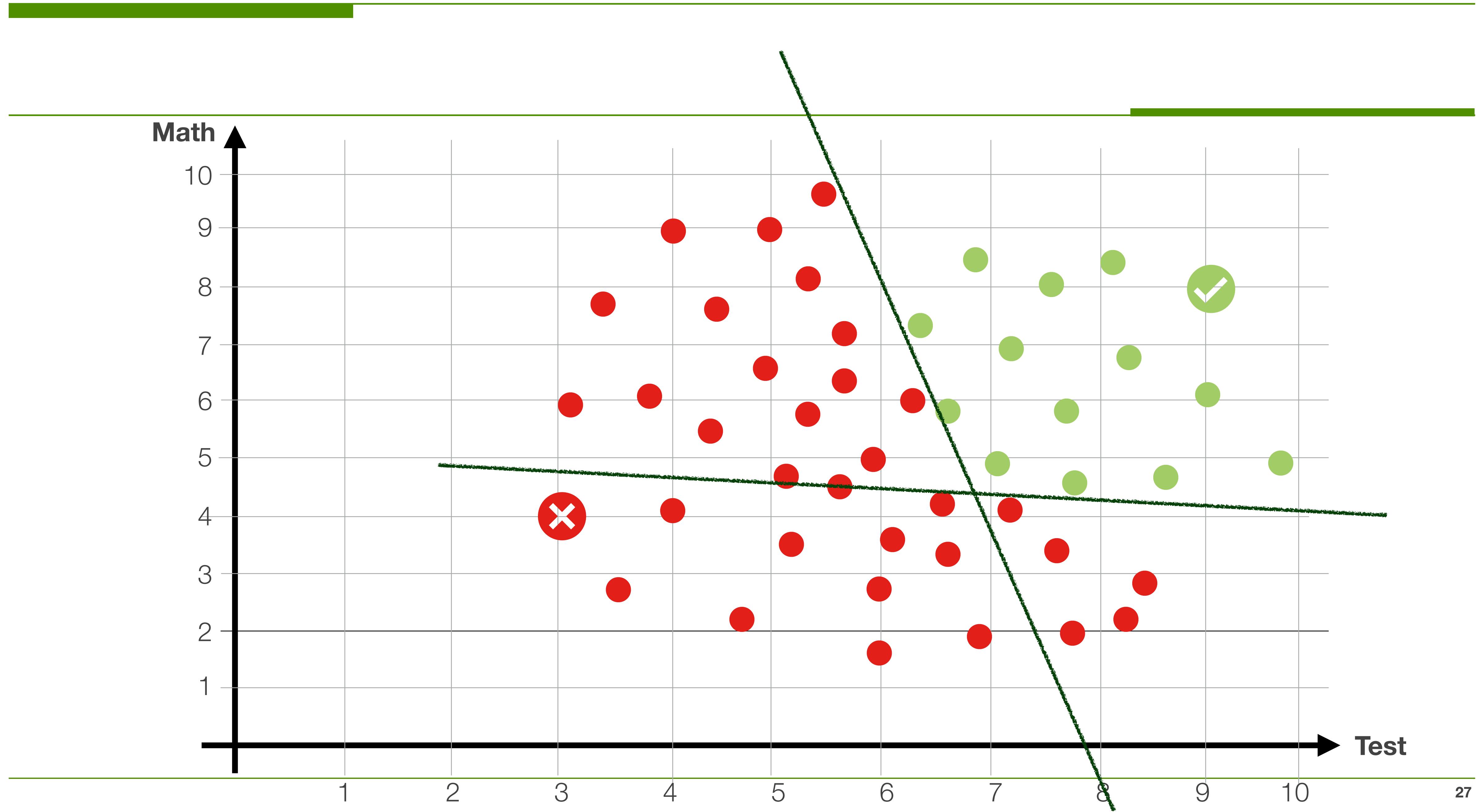
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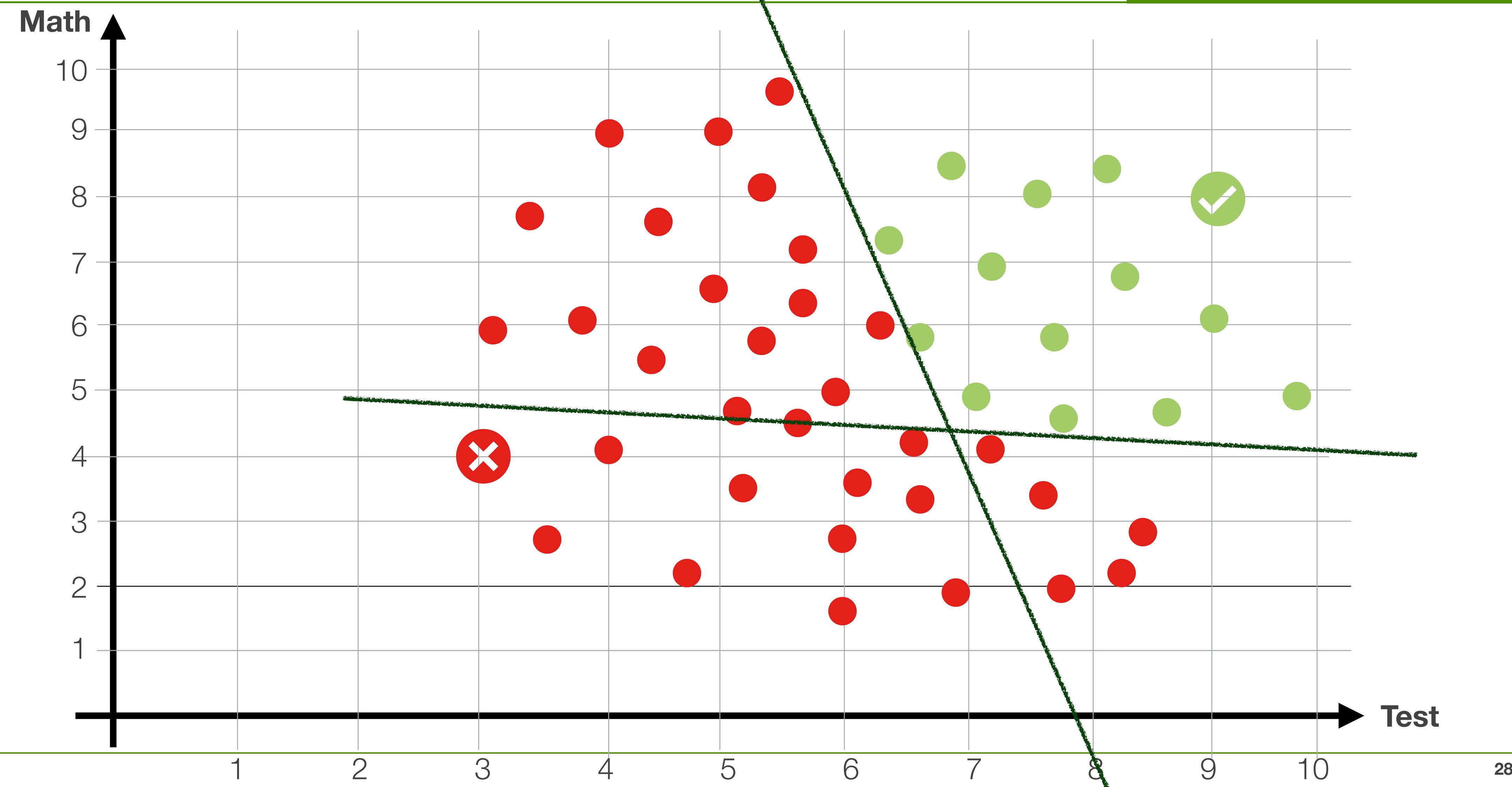
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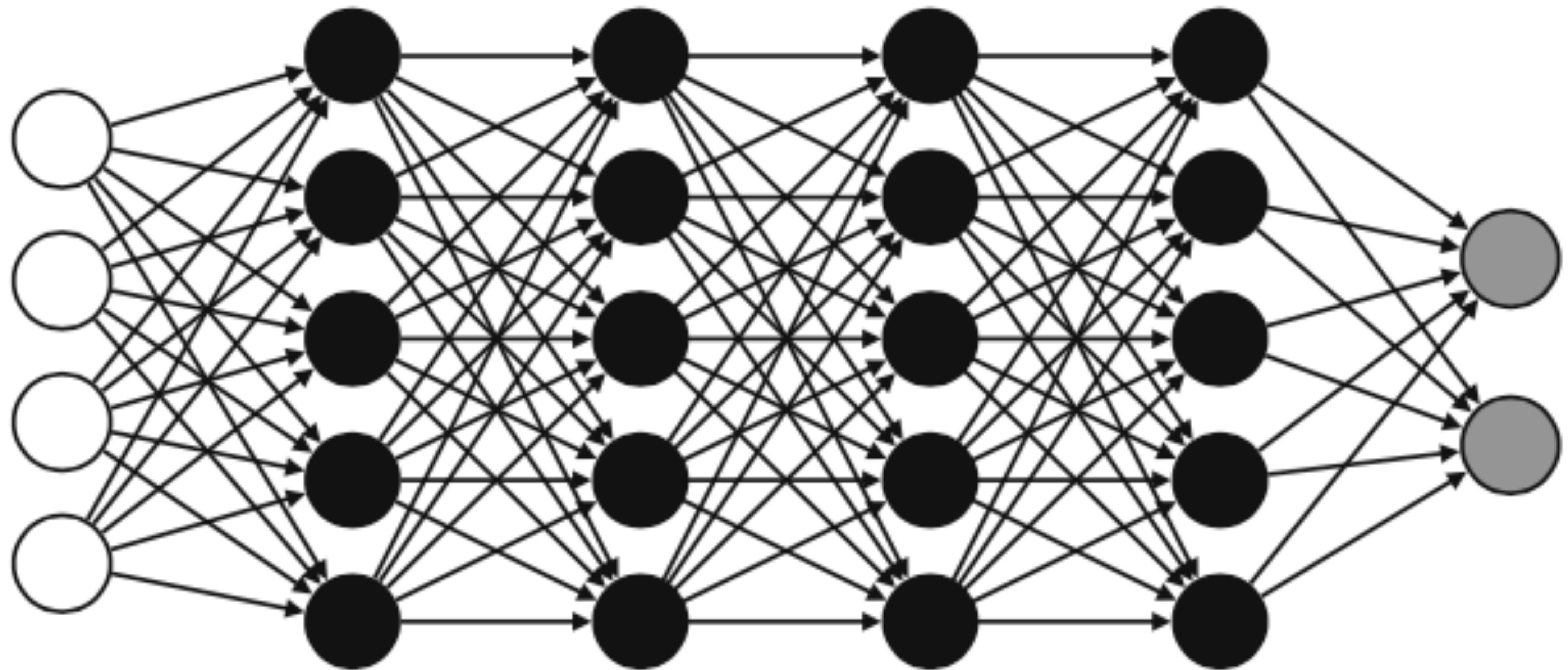
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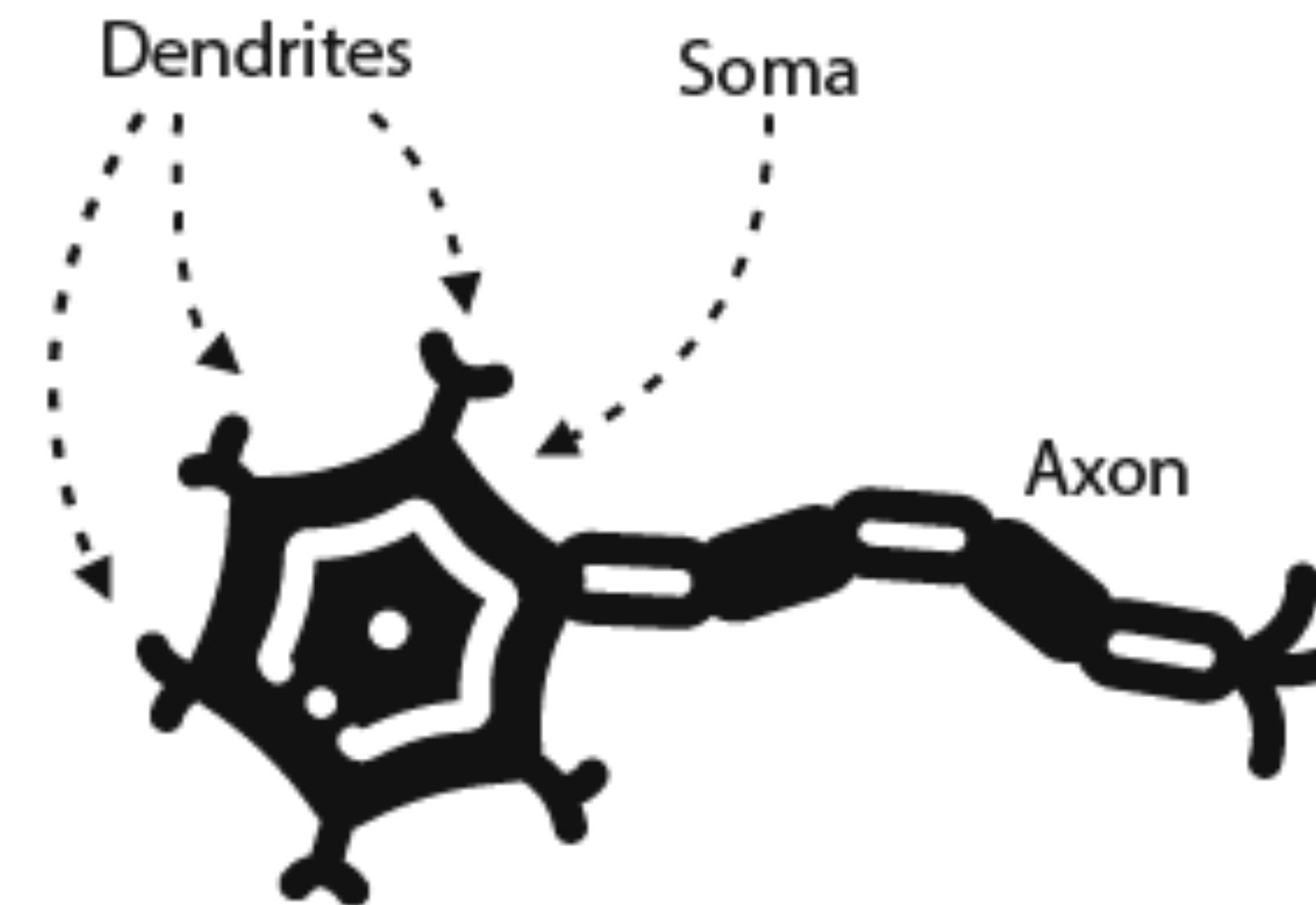


# Neural Network

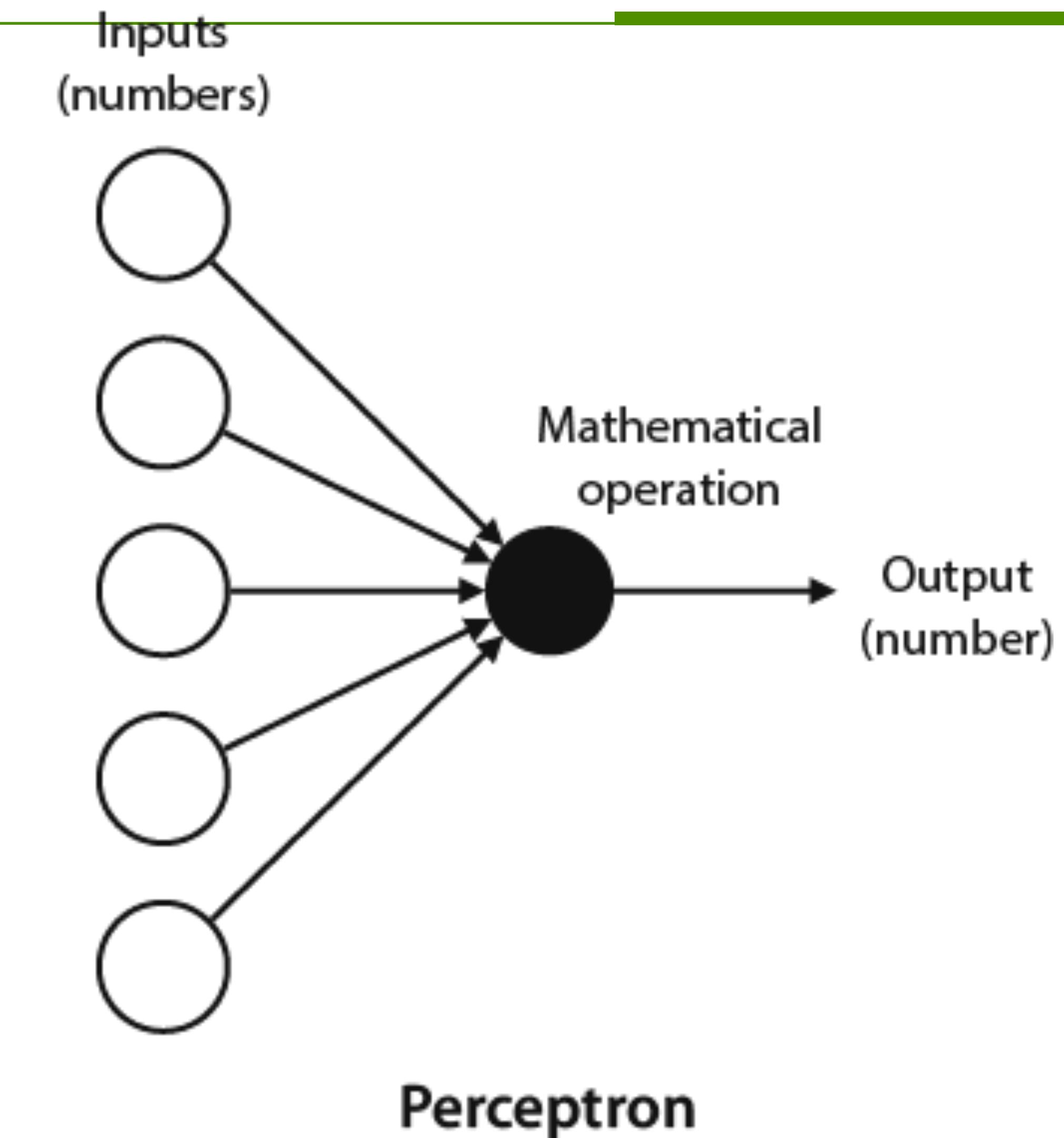




# Neural Network

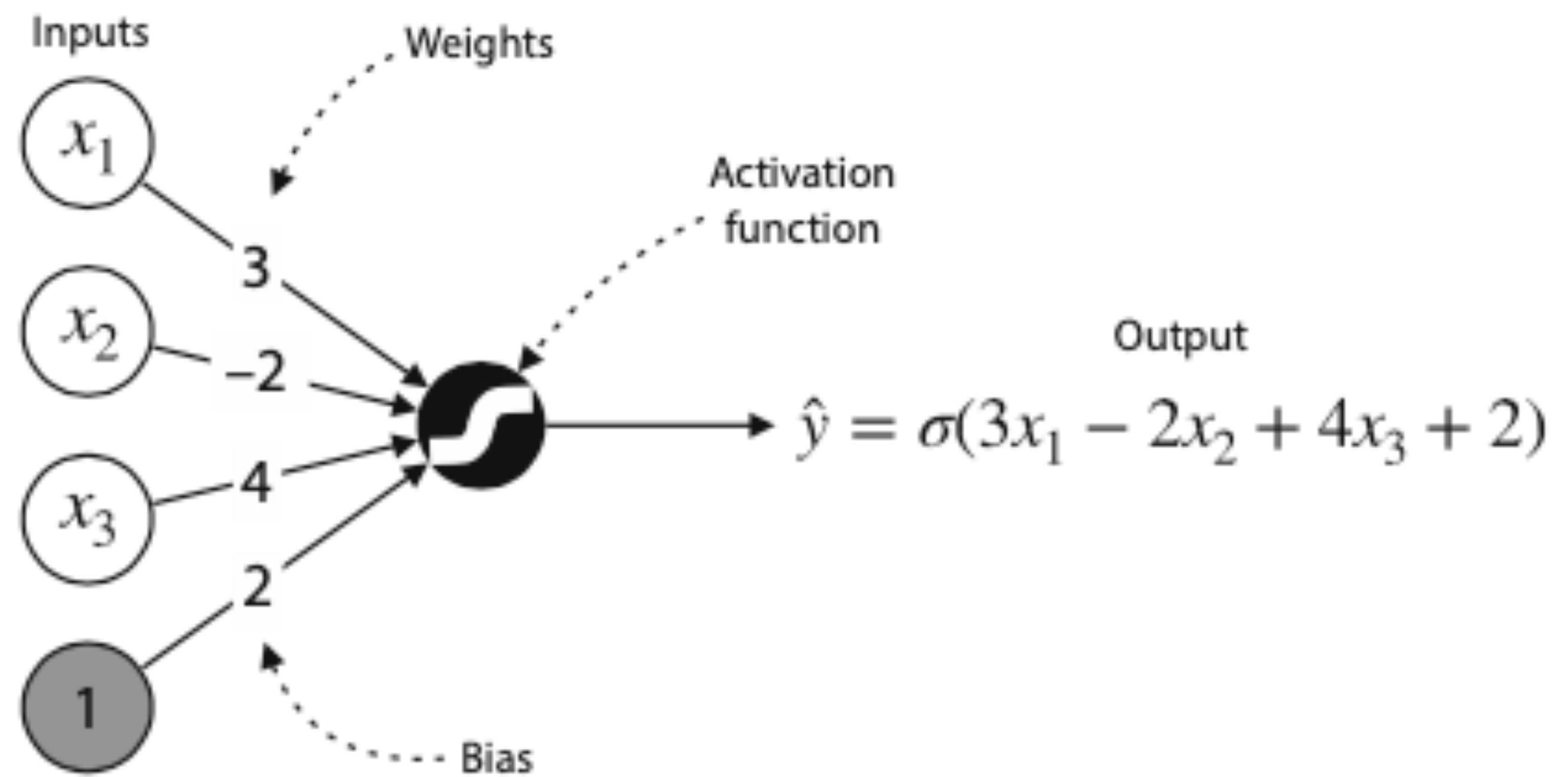


Neuron



Perceptron

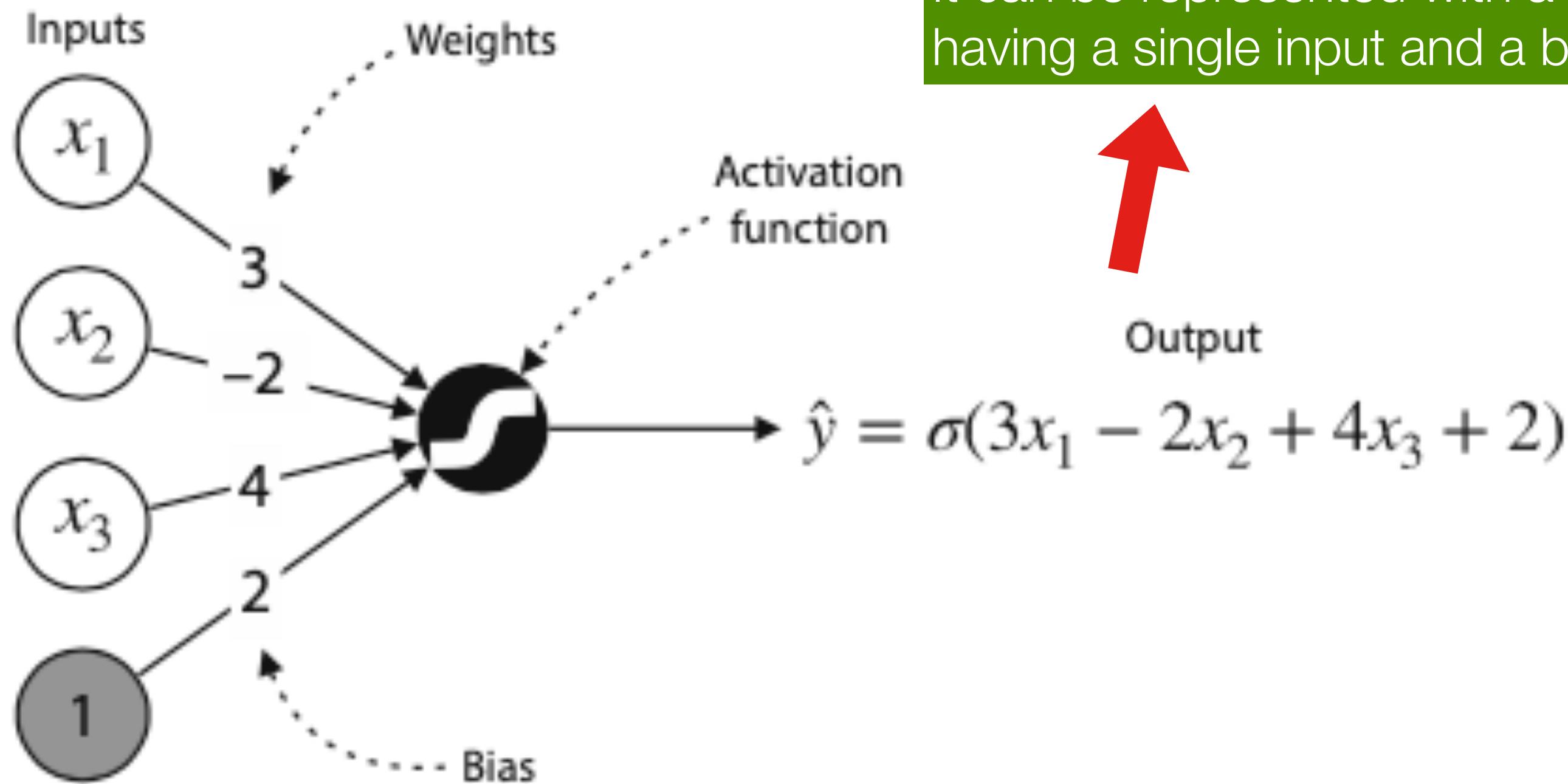
# Perceptron



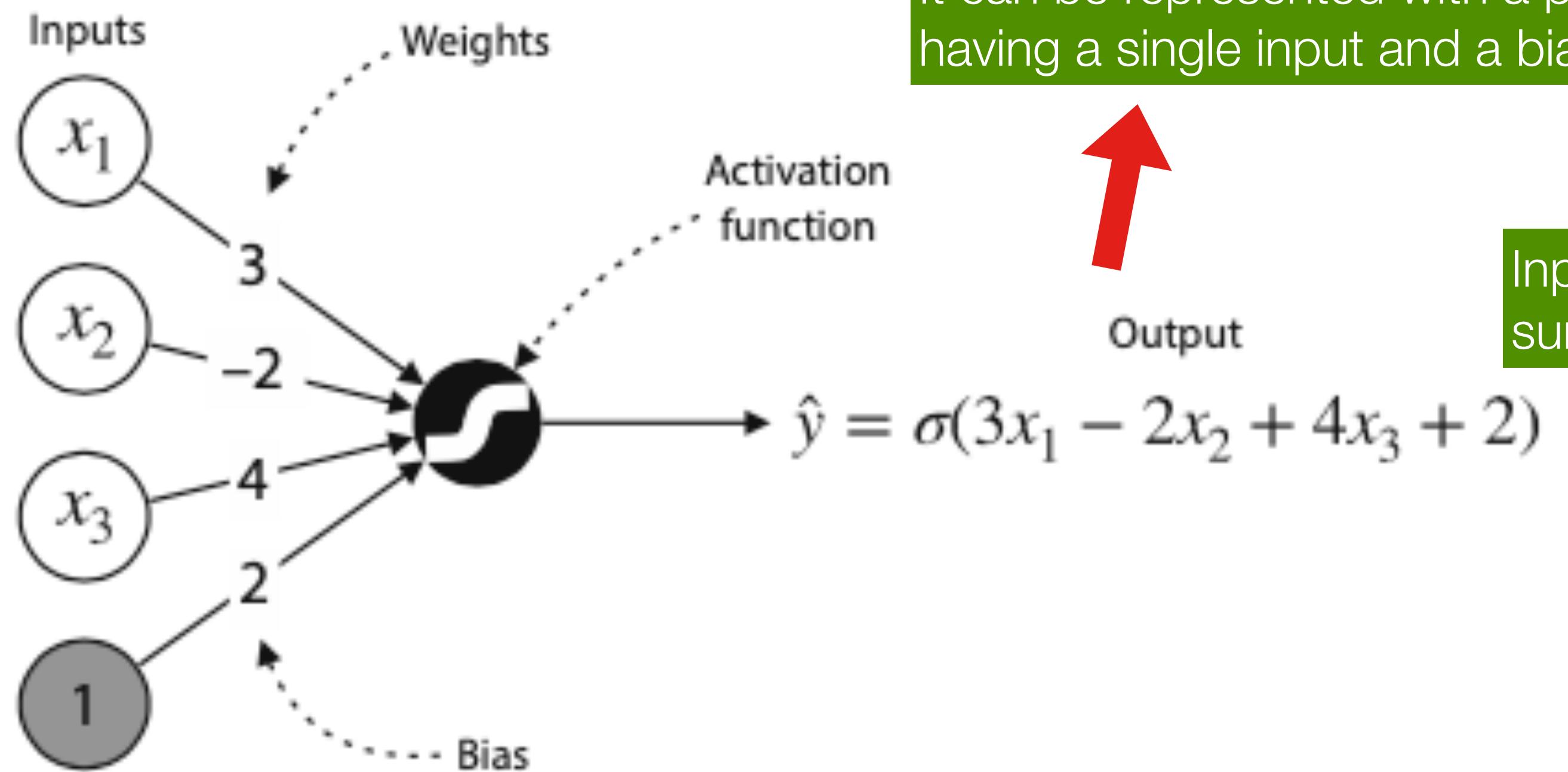
# Perceptron

**Cost =  $a + b$  Size**

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



# Perceptron



**Cost =  $a + b$  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

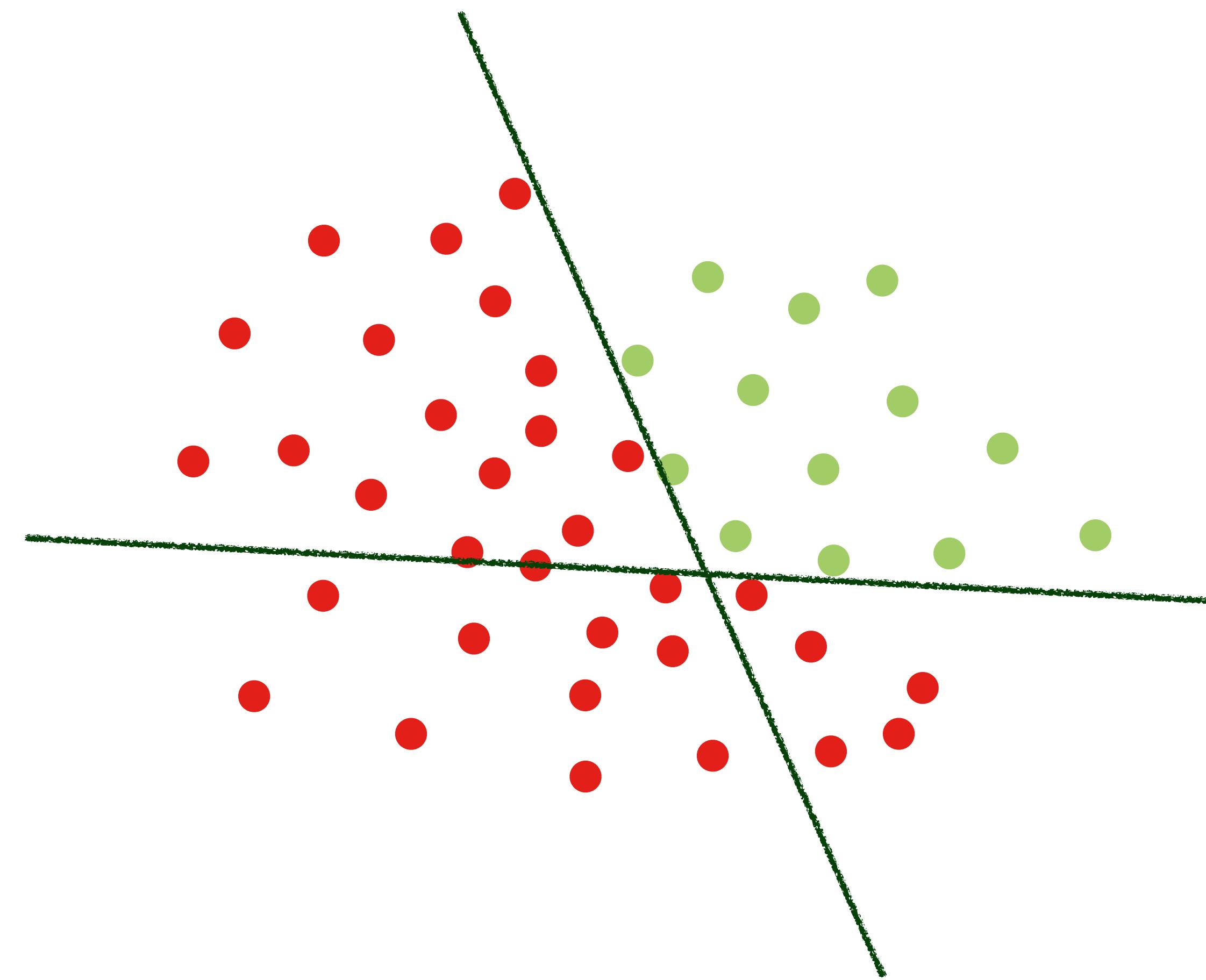
1 if  $y > 0$   
0 if  $y \leq 0$

Sigmoid function  
(continuous)

Output

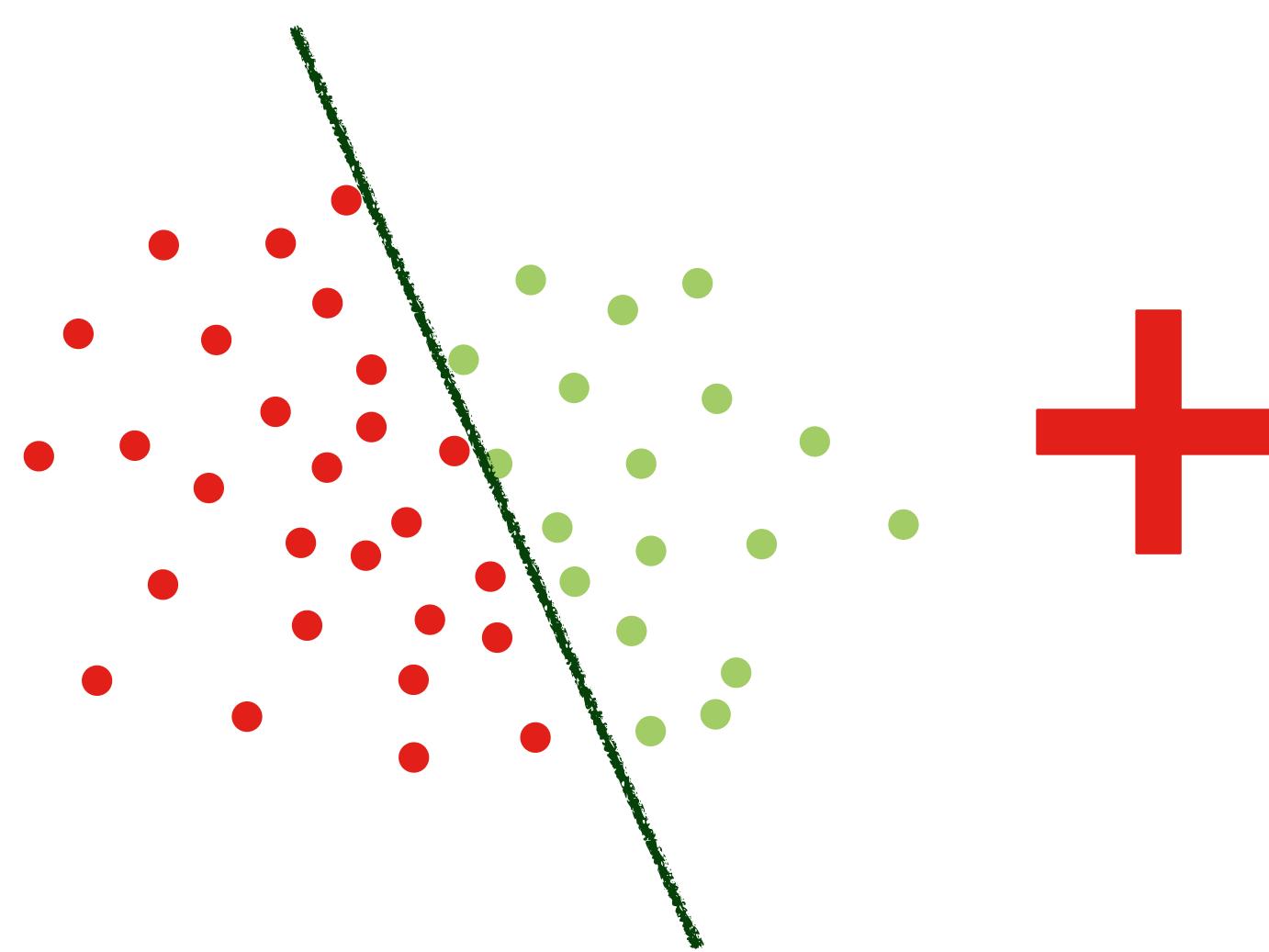
$$\sigma(\text{input}) = \frac{1}{1 + e^{-\text{input}}}$$

# Neural Network

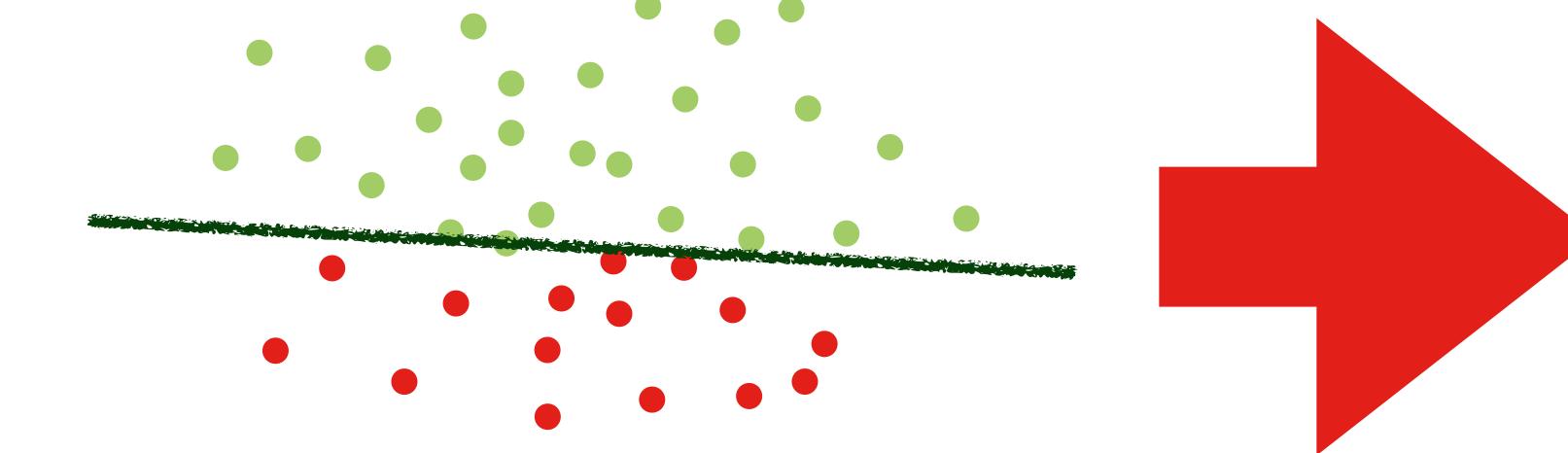


# Neural Network

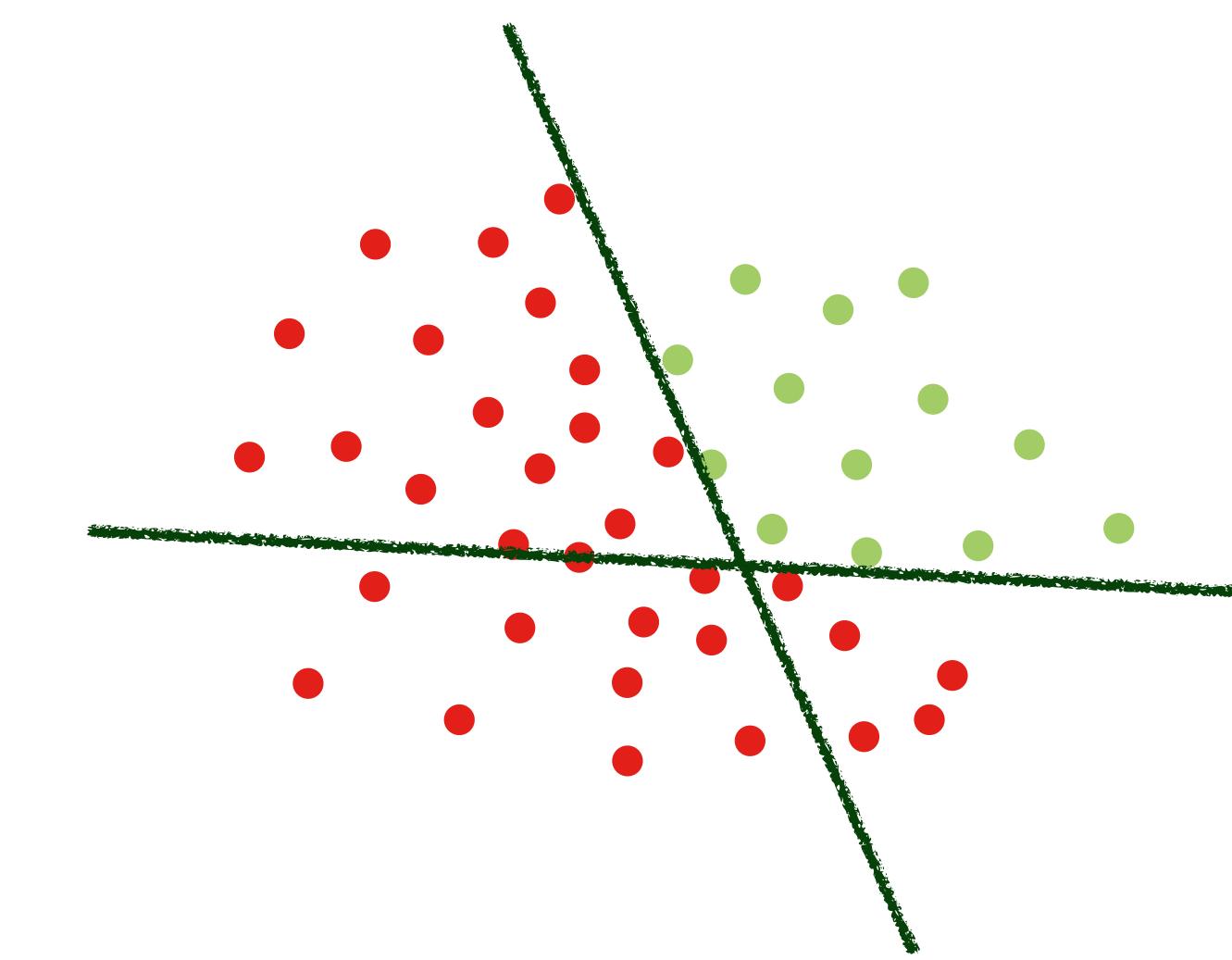
**Test Grade  
Classifier**



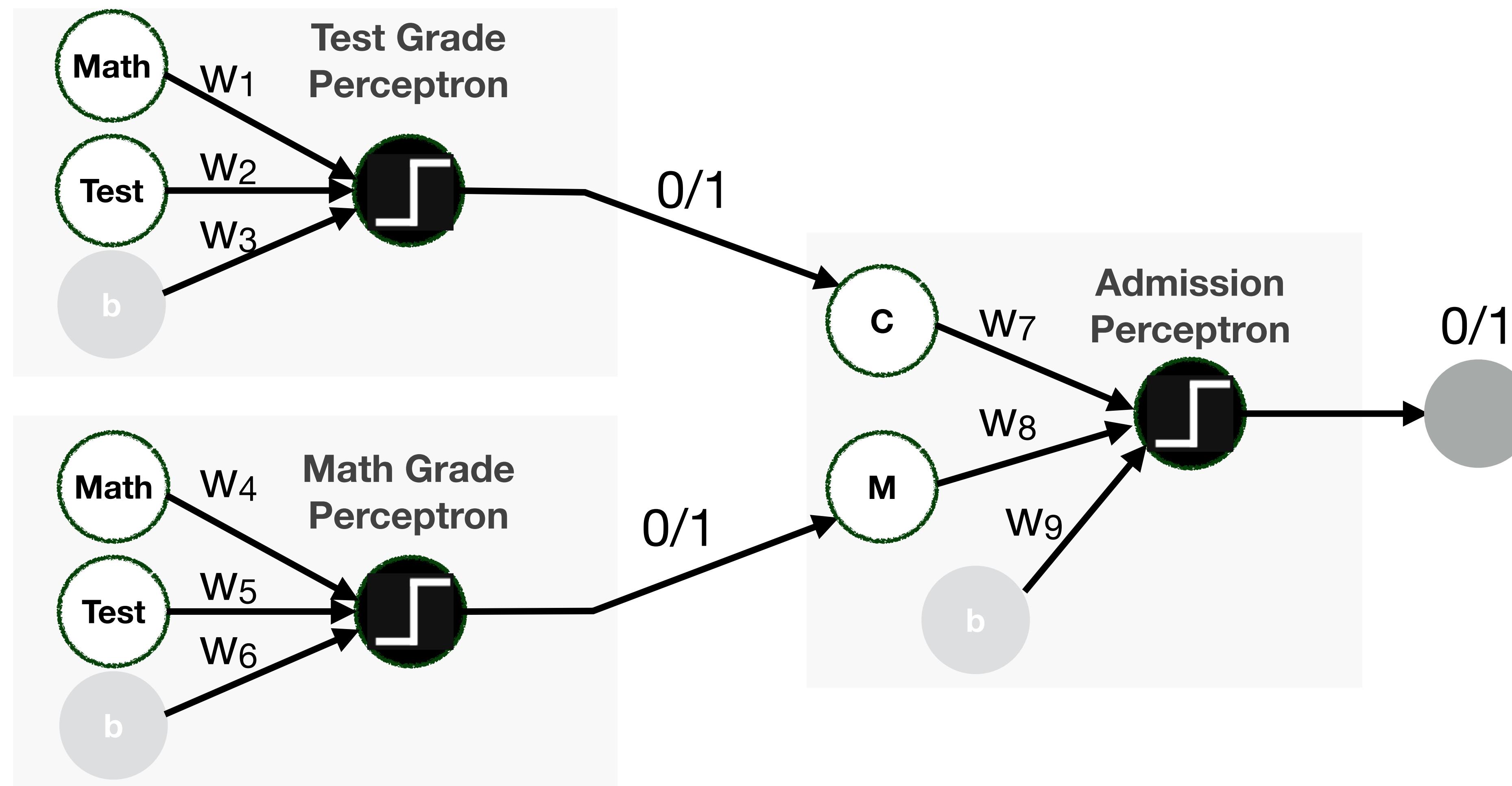
**Math Grade  
Classifier**



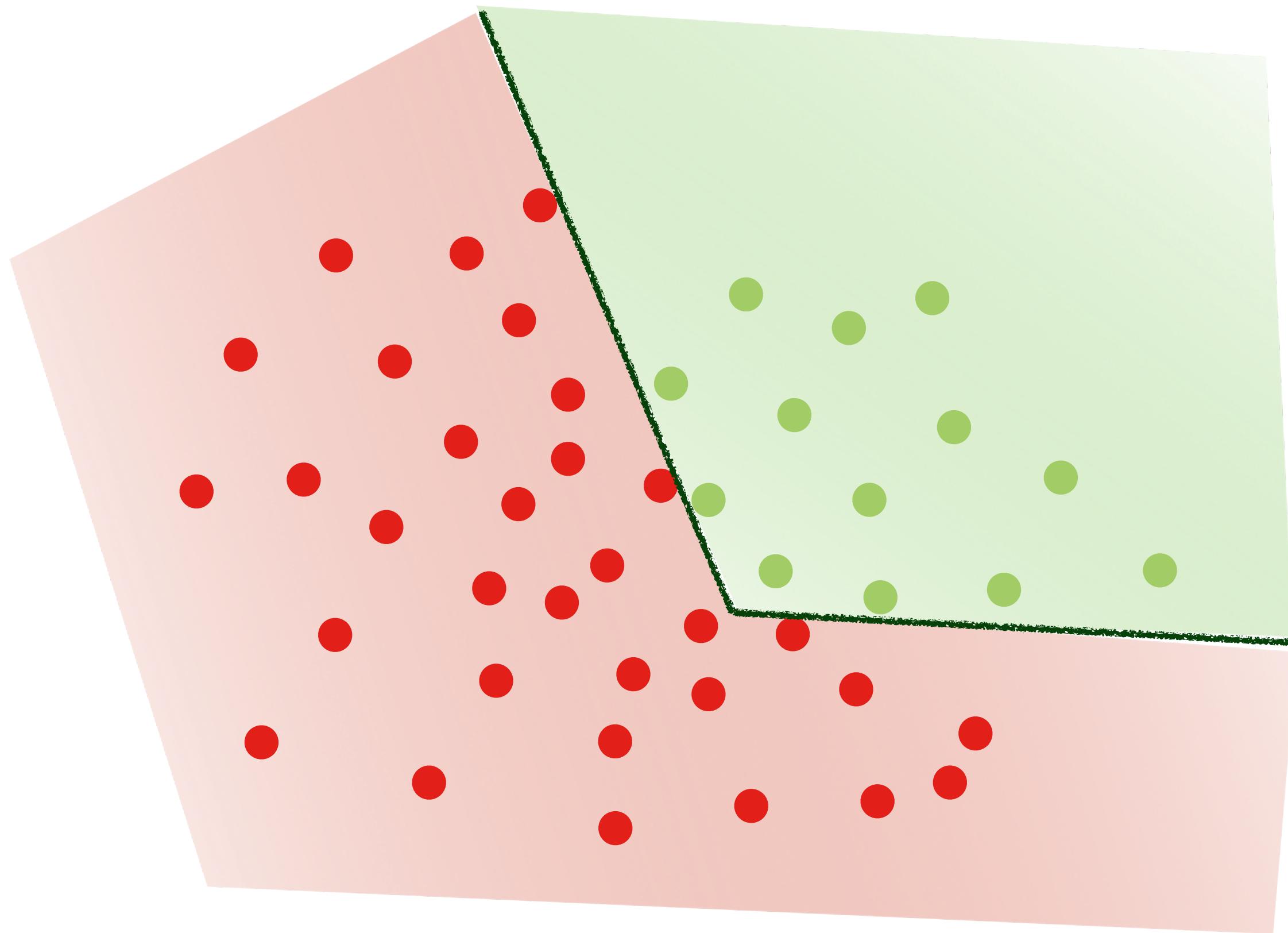
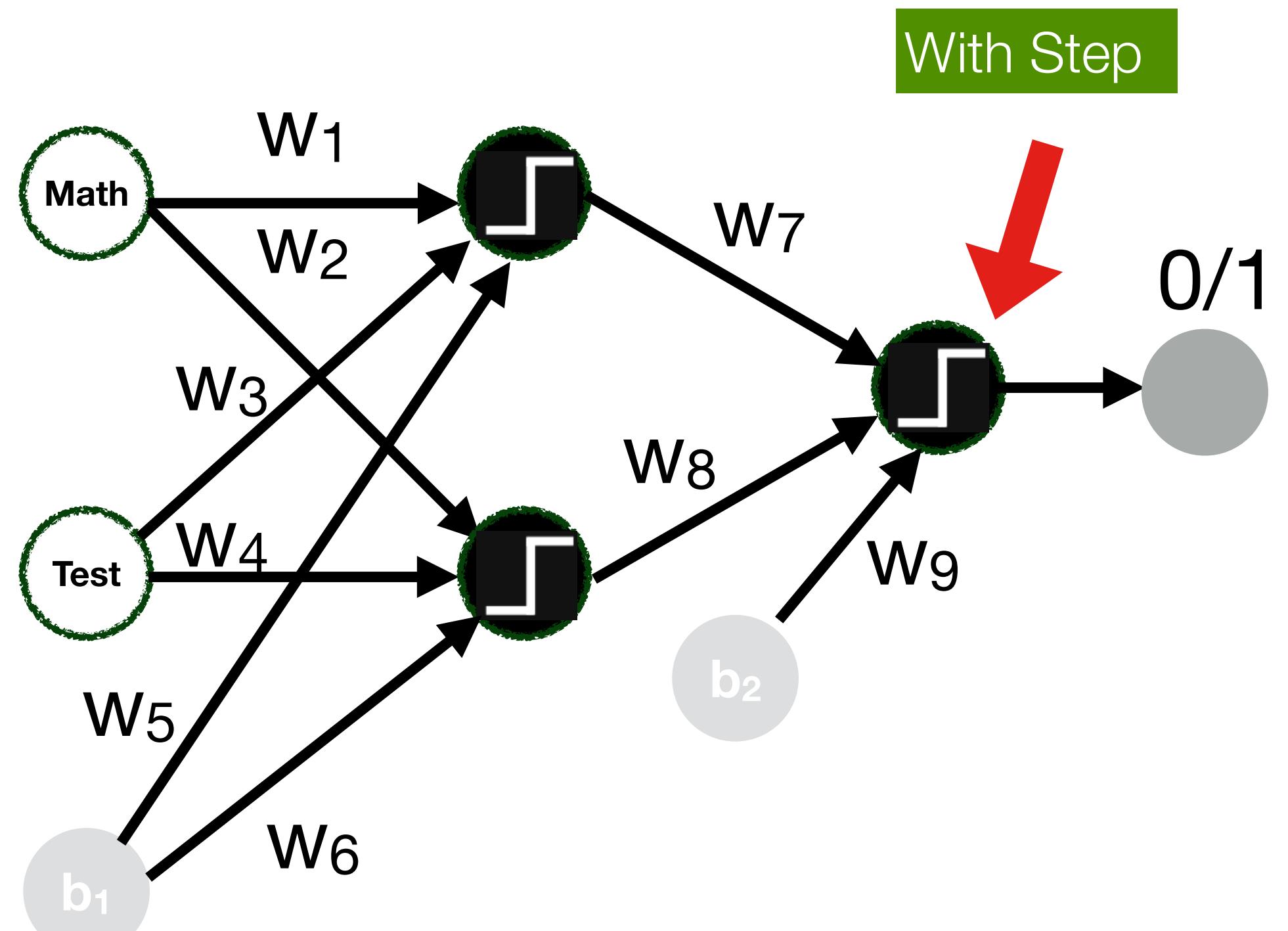
**Admission  
Classifier**



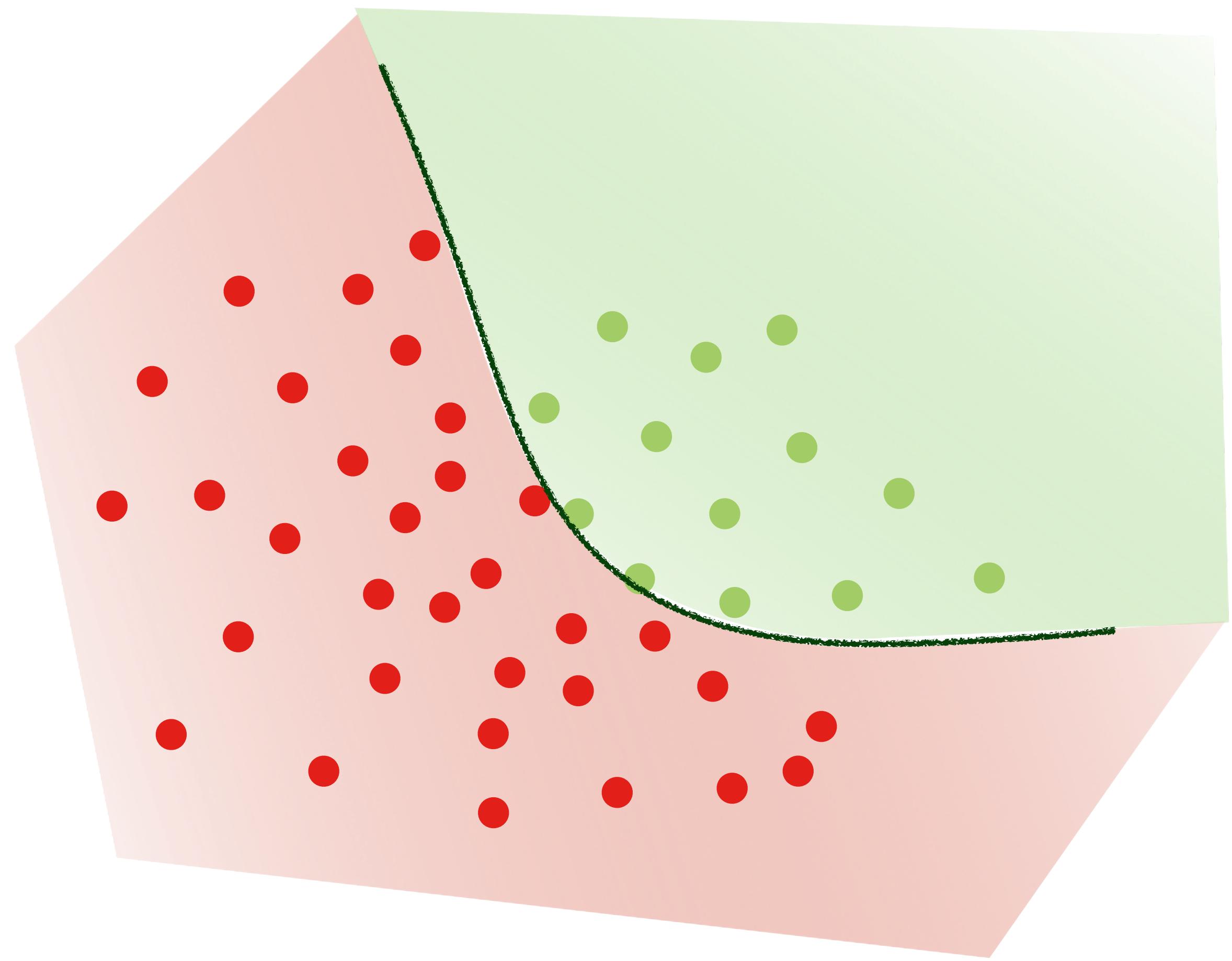
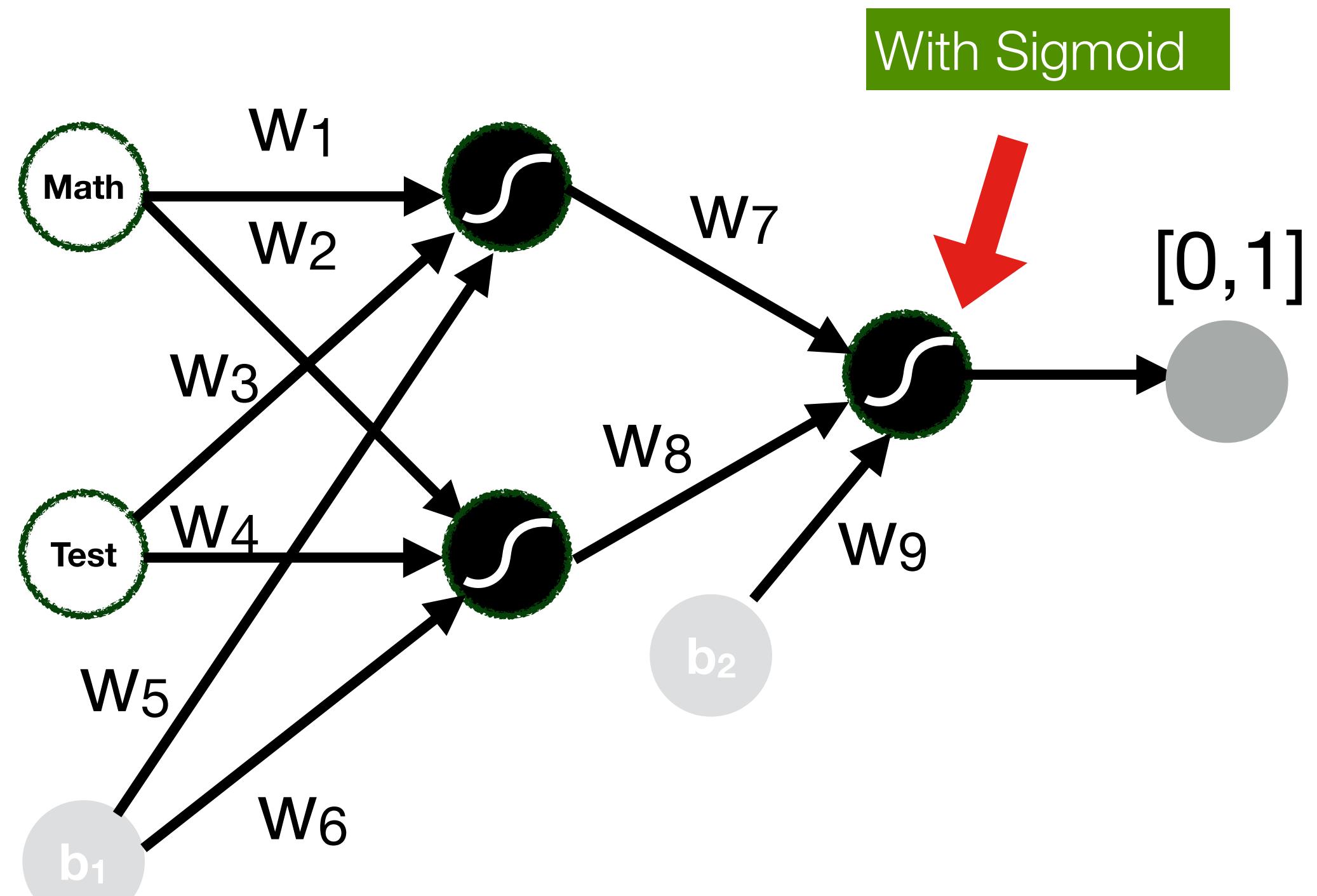
# Neural Network



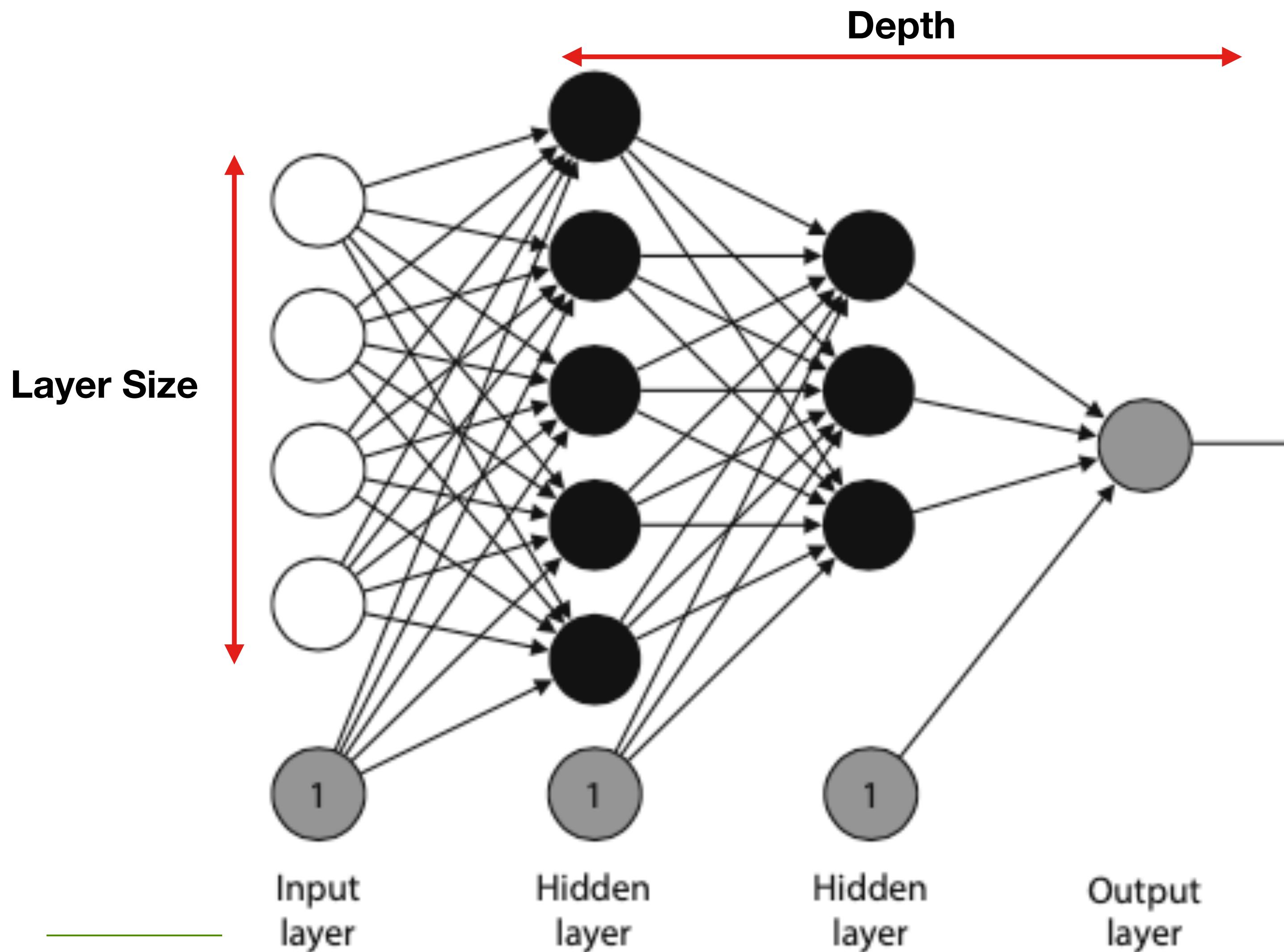
# Neural Network



# Neural Network



# Fully Connected Neural Network



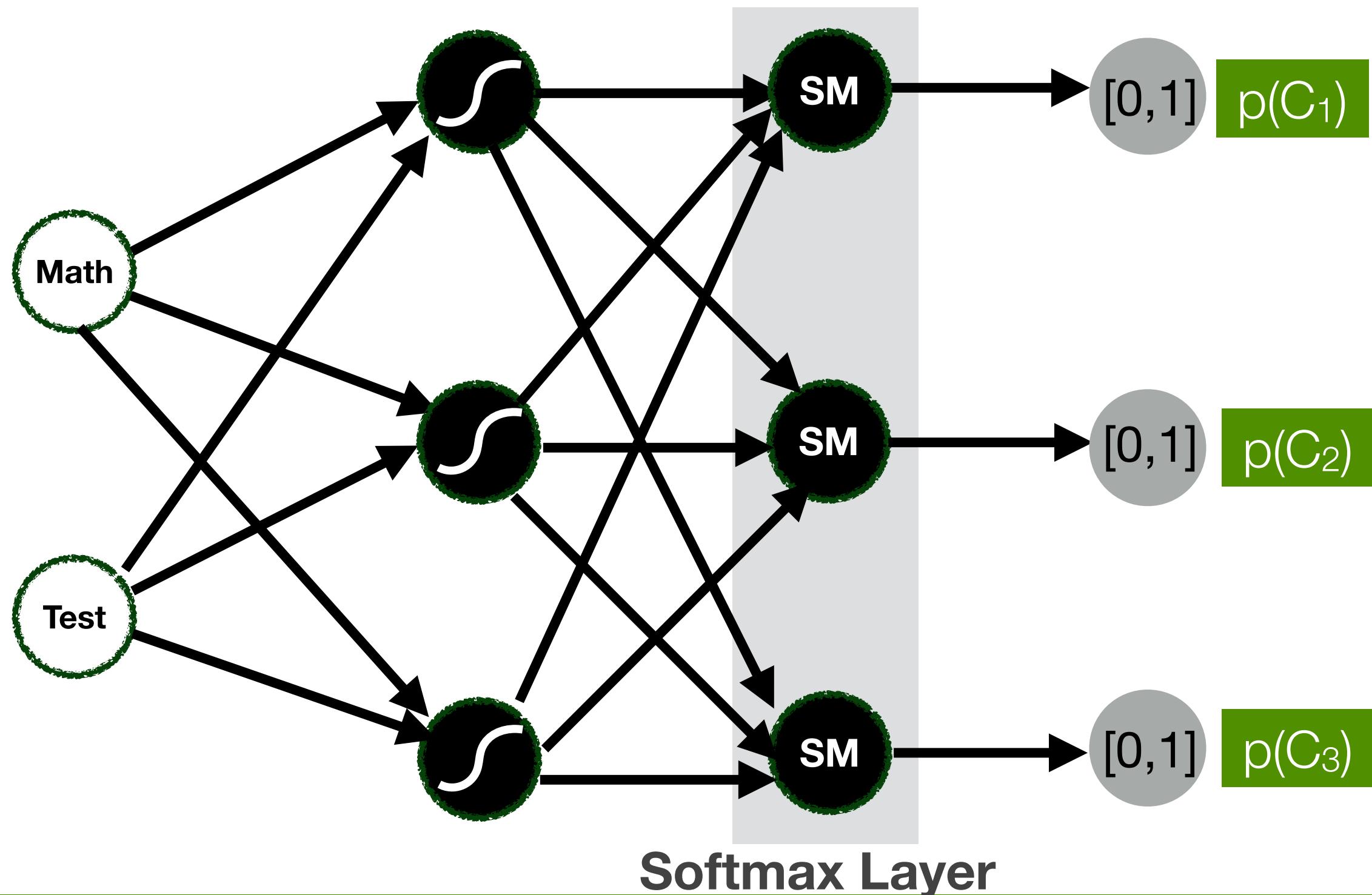
## ■ Hyperparameters

- Learning rate
- Number of epochs
- Architecture
  - # layers, #nodes, activation functions
- Batch vs. mini-batch vs. stochastic gradient descent
- Regularization parameters:
- Dropout probability  $p$

**Advanced, not covered**

# Classifying into multiple classes - softmax function

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



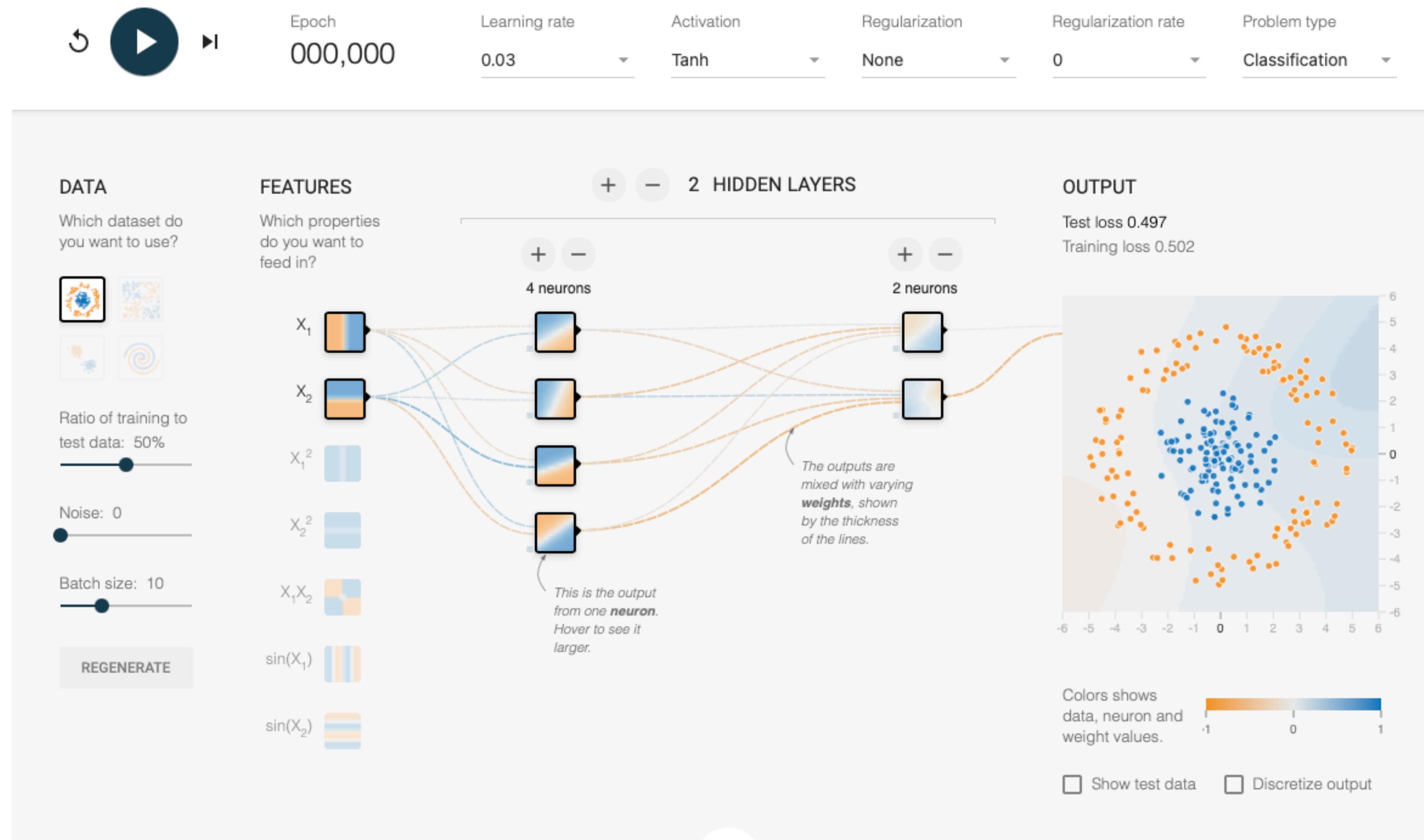
$$\text{Softmax}(x_i) = \frac{e^{(x_i)}}{\sum_j^K e^{(x_j)}}$$

Value of class  $i$

Normalisation term on K classes

The equation defines the Softmax function. It takes a value  $x_i$  as input and produces a probability  $p(C_i)$  as output. The output is calculated by dividing the exponential of  $x_i$  by the sum of the exponentials of all values  $x_j$  for  $j$  from 1 to  $K$ . Red arrows point from the labels "Value of class  $i$ " and "Normalisation term on K classes" to the respective terms in the equation.

# 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?



167	153	174	168	150	152	151	172	161	155	156	187	167	153	174	168	150	152	151	172	161	155	156	187	167	153	174	168	150	152	151	172	161	155	156	187								
155	182	163	74	75	62	33	17																																				
180	180	50	14	34	6	10	33																																				

# This is what a computer “sees”

75	62	33	17	110	210	180	154
34	6	10	33	48	106	150	181

206	109	5	124	131	111	120	204	166	16	56	180	206	109	5	124	131	111	120	204	166	16	56	180	206	109	5	124	131	111	120	204	166	16	56	180
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# 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

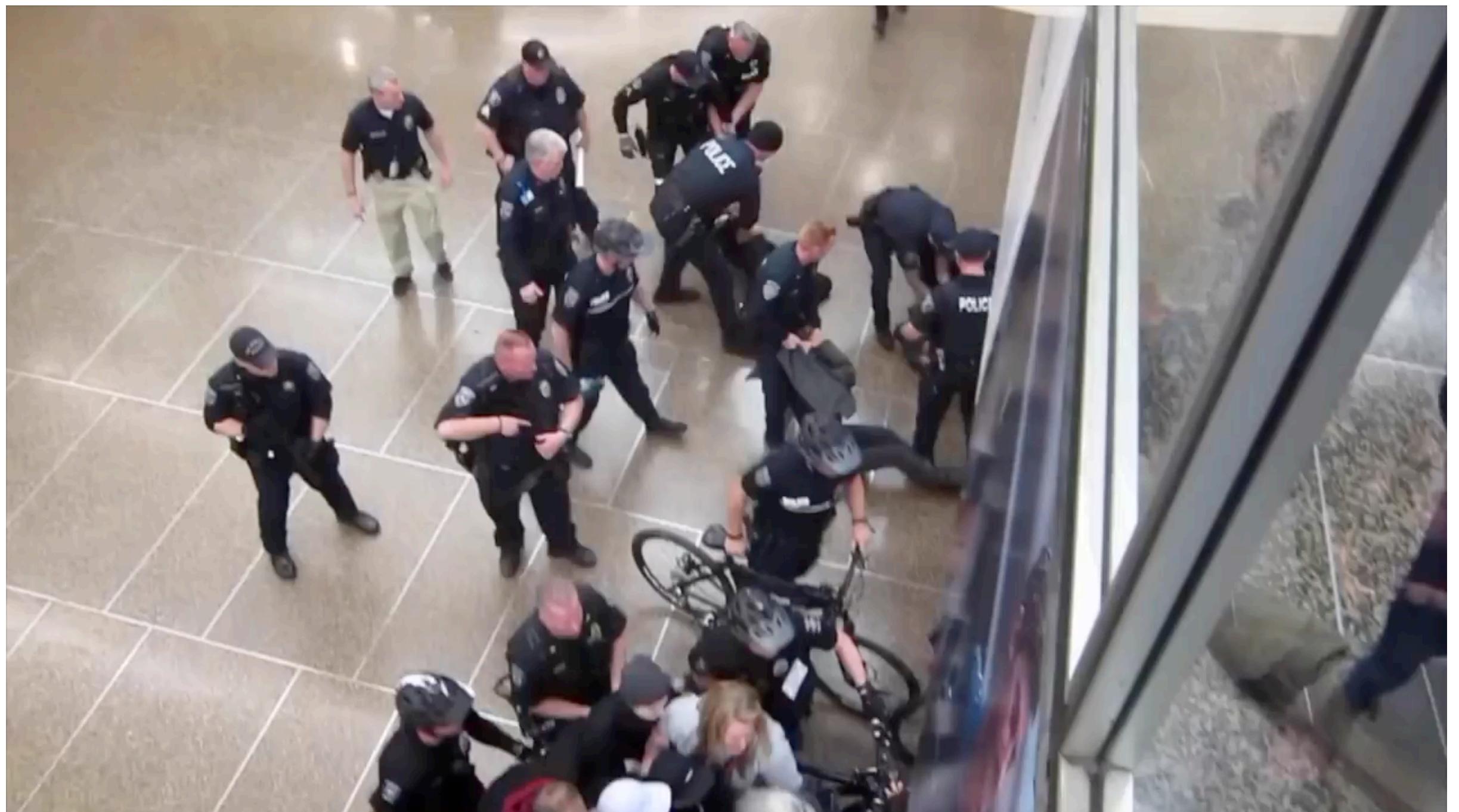
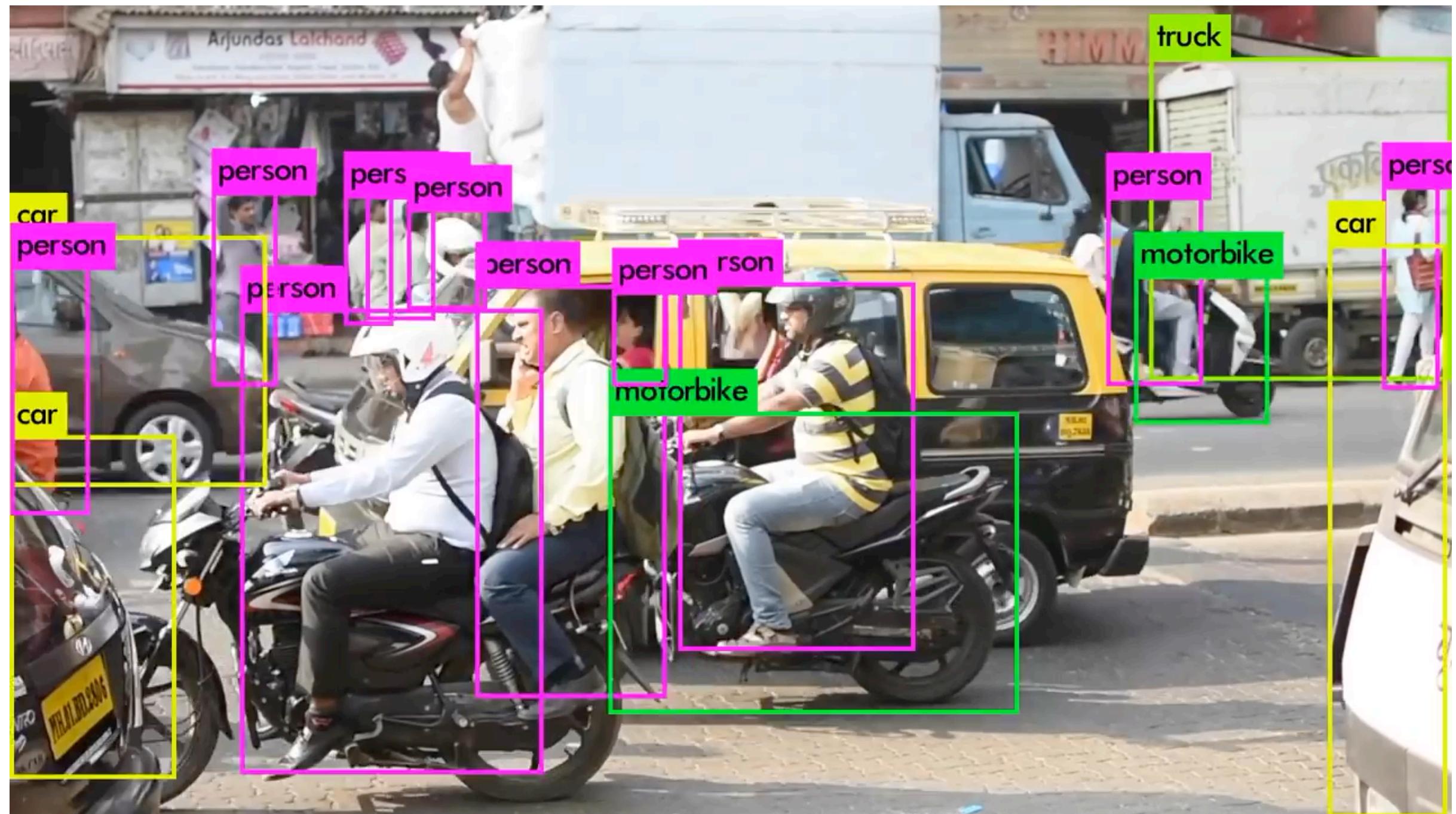


Is this a flag?  
(Recognition / Classification)



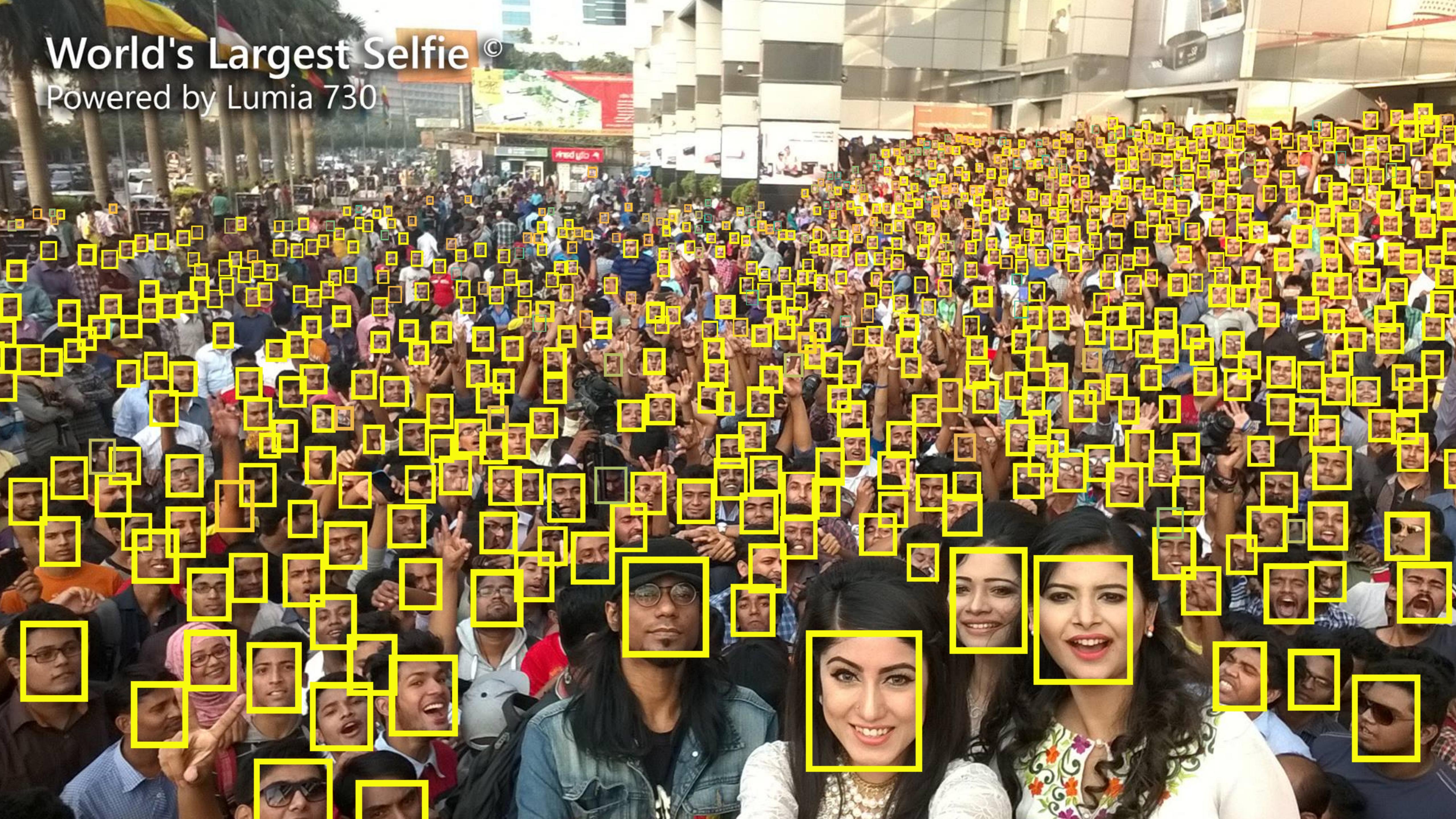


**Where are the people?  
(Recognition/Classification)**



# World's Largest Selfie ©

Powered by Lumia 730

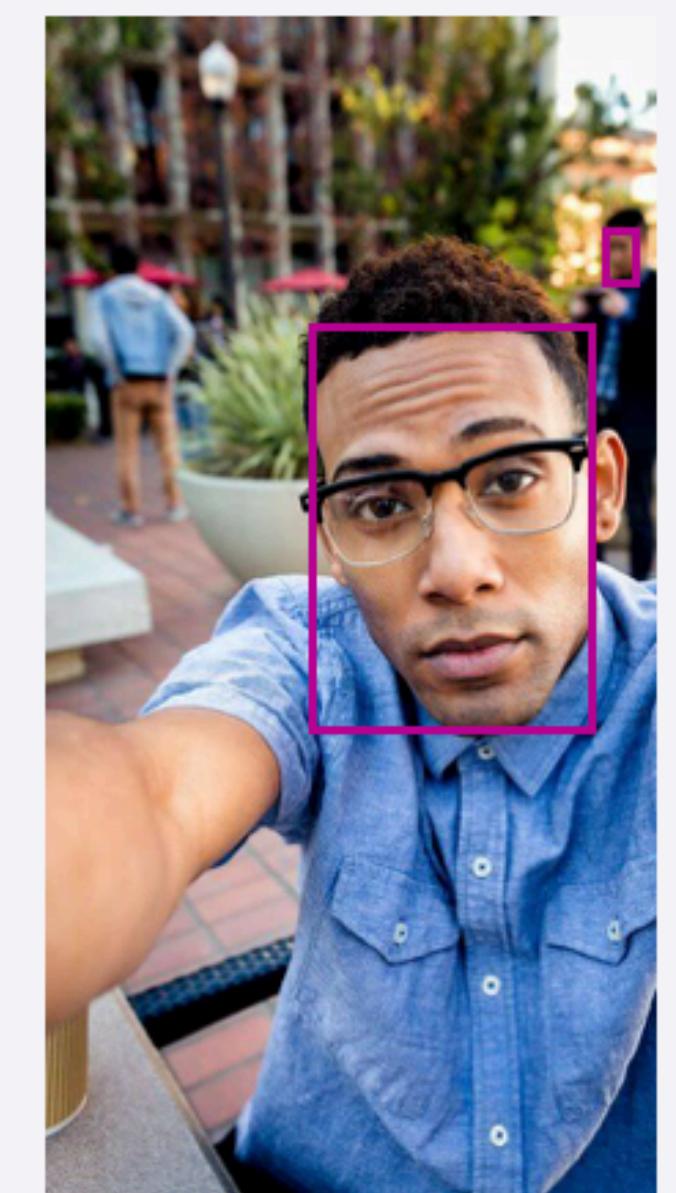






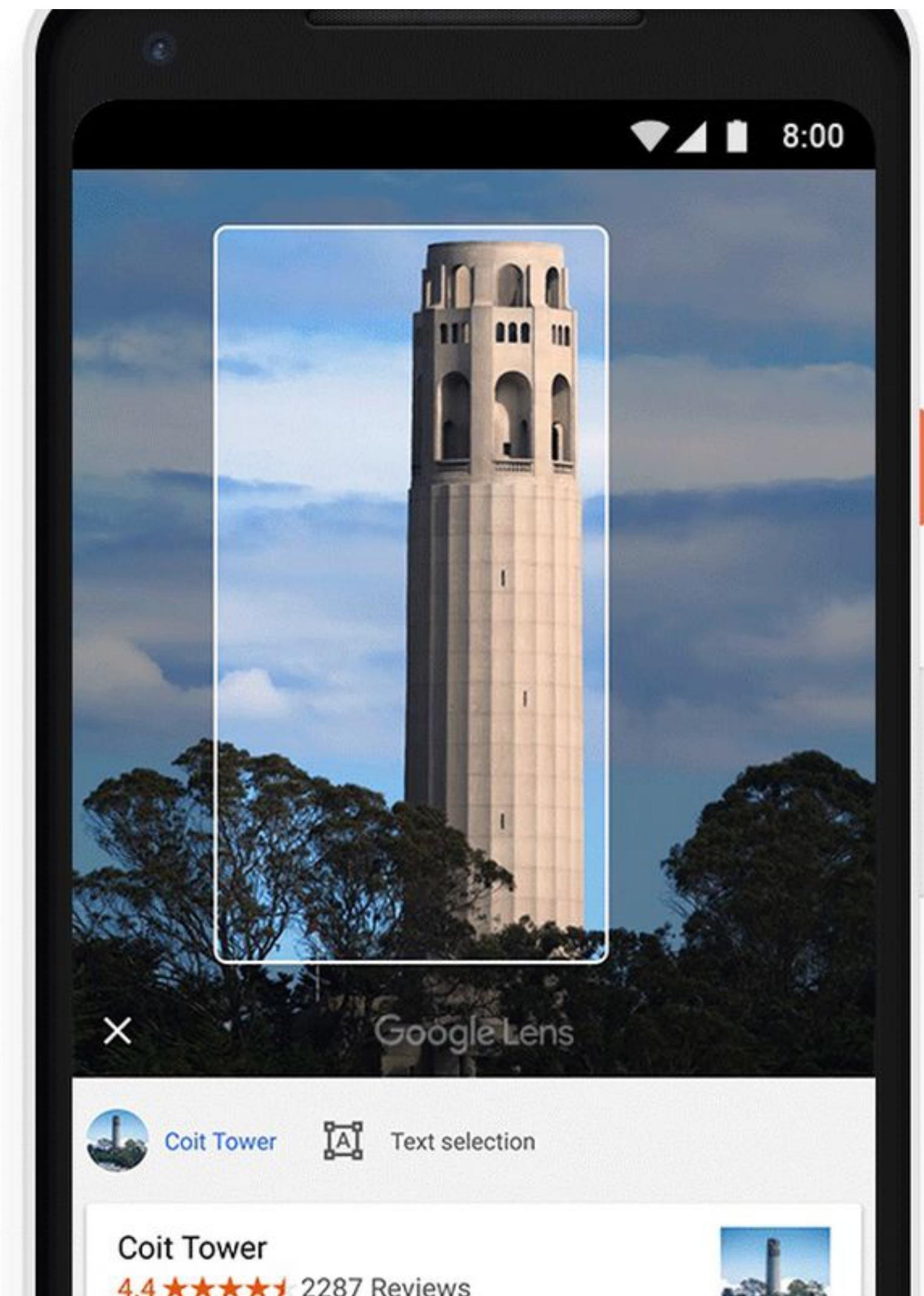
**Is this Jeff?  
(Identification)**

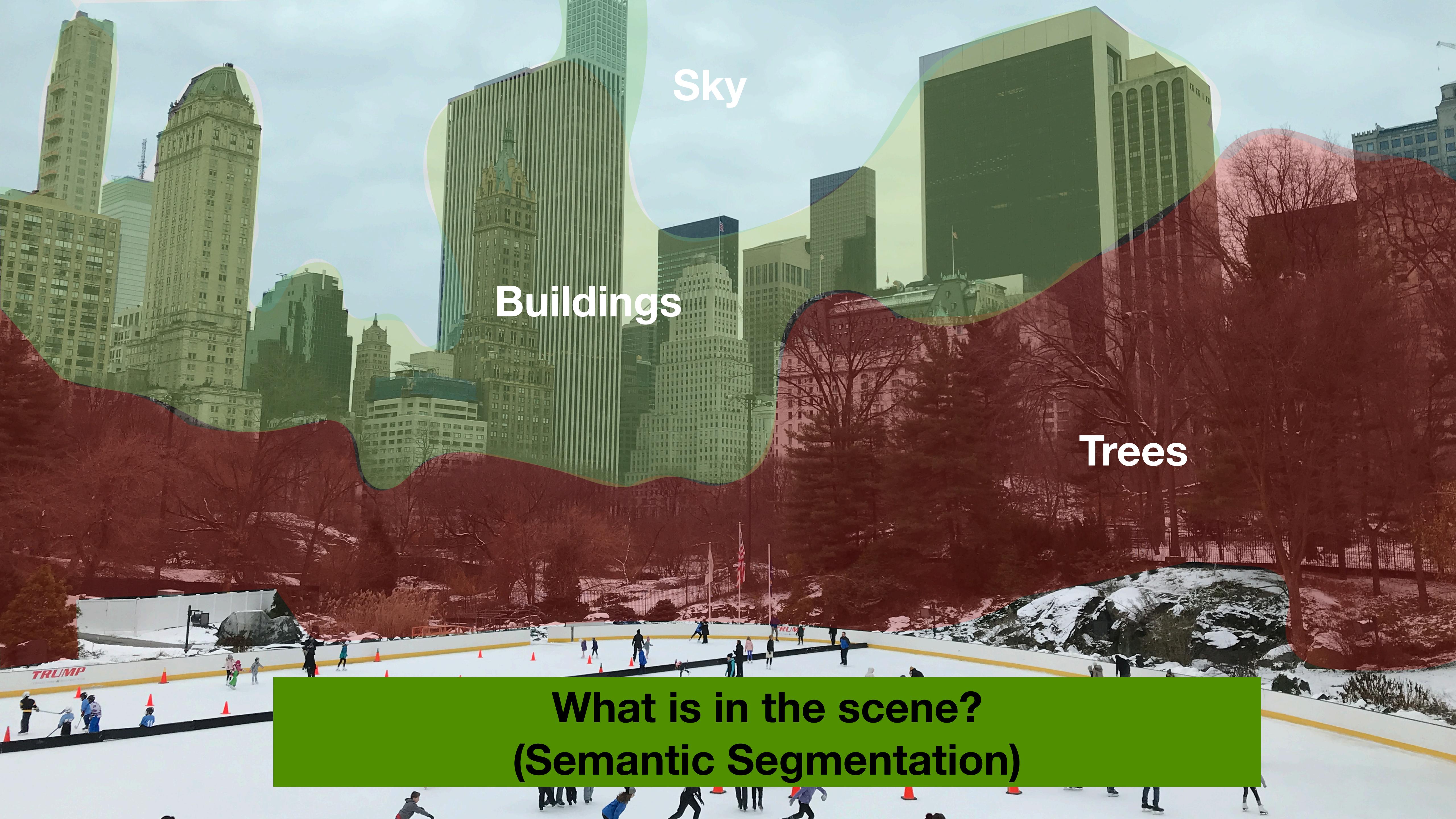




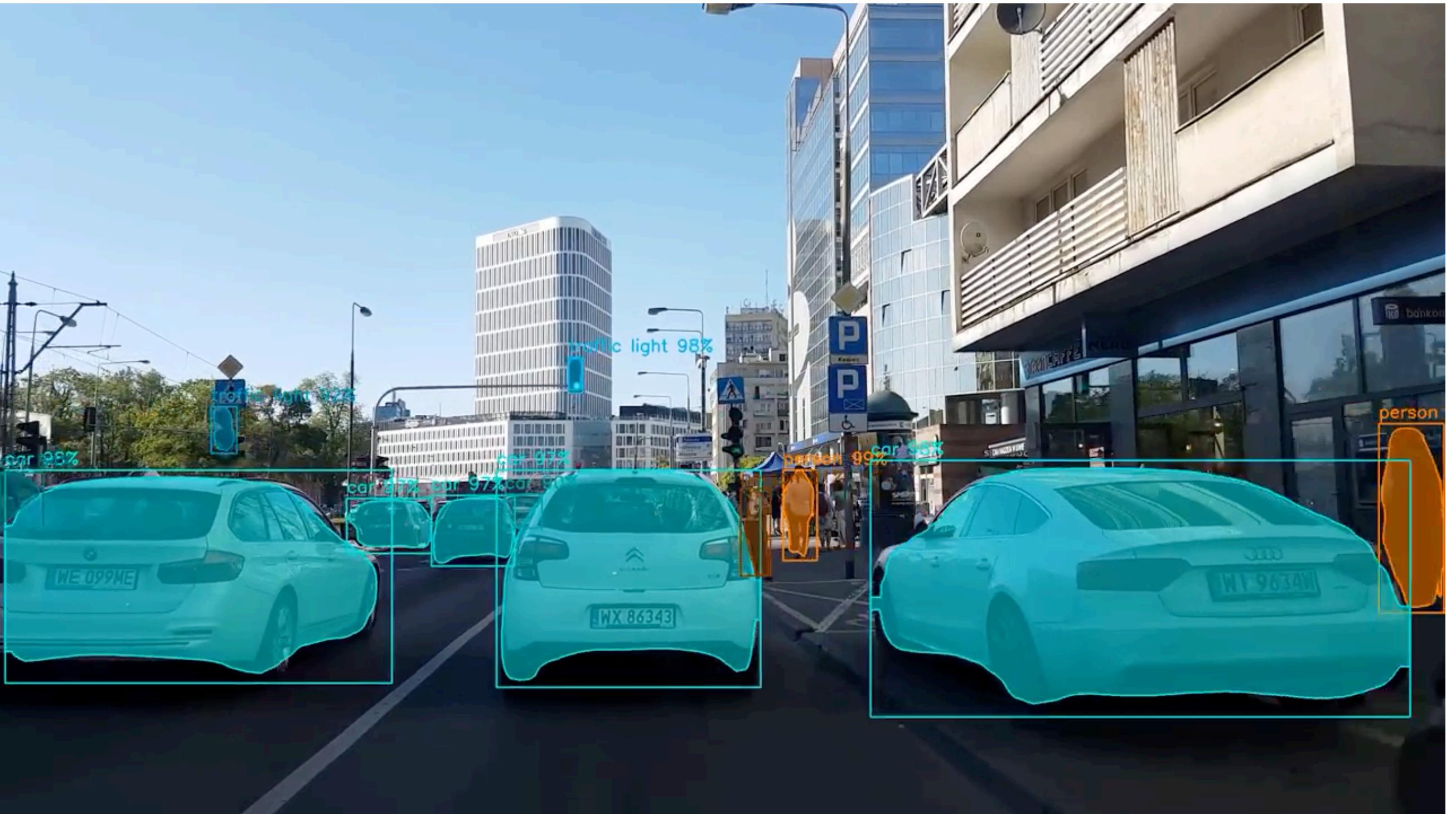
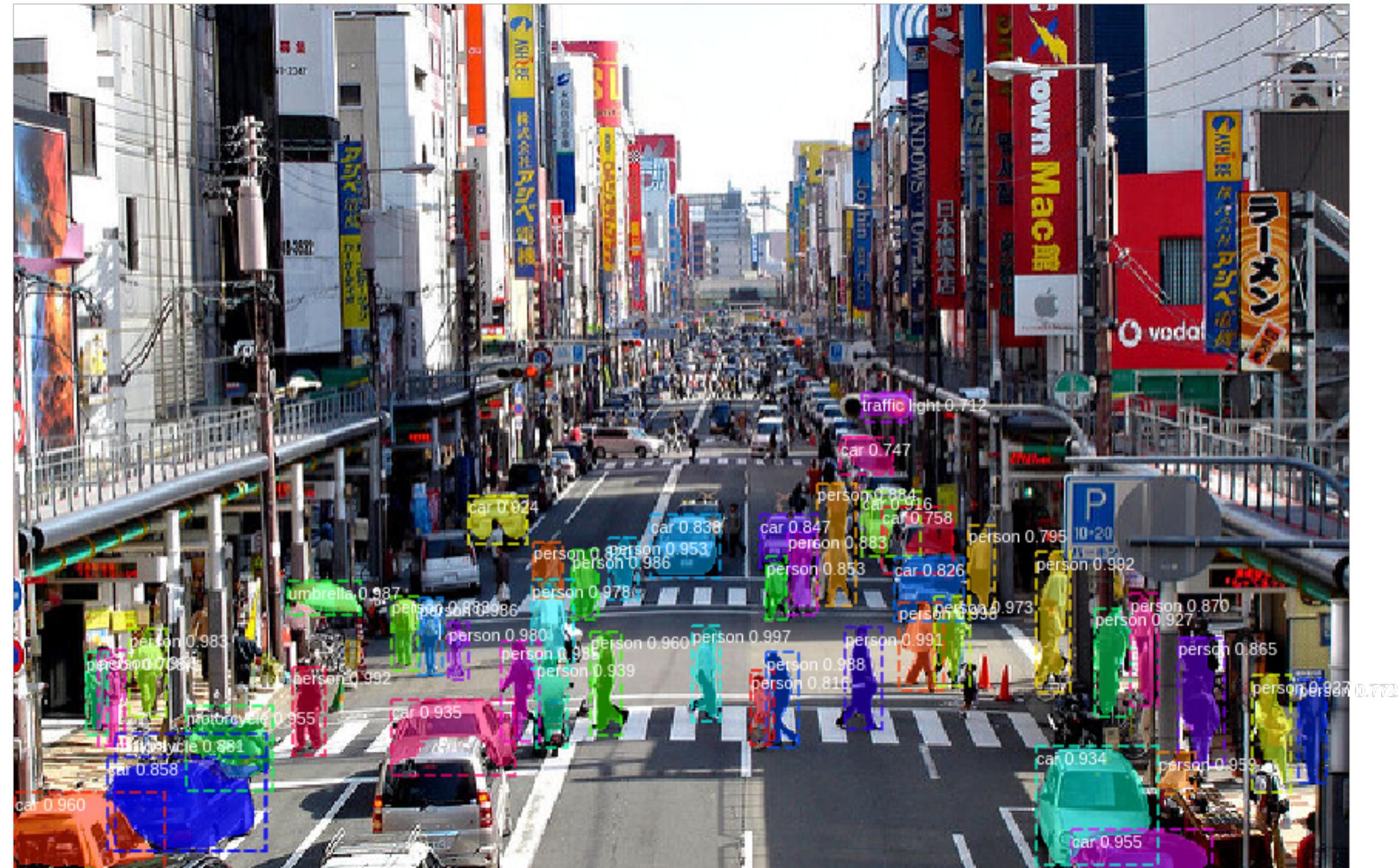


**Is this the Wollman Rink?  
(Identification)**





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



[https://github.com/matterport/Mask\\_RCNN](https://github.com/matterport/Mask_RCNN)

# Project Sunroof

≡ Google Project Sunroof

Savings estimator

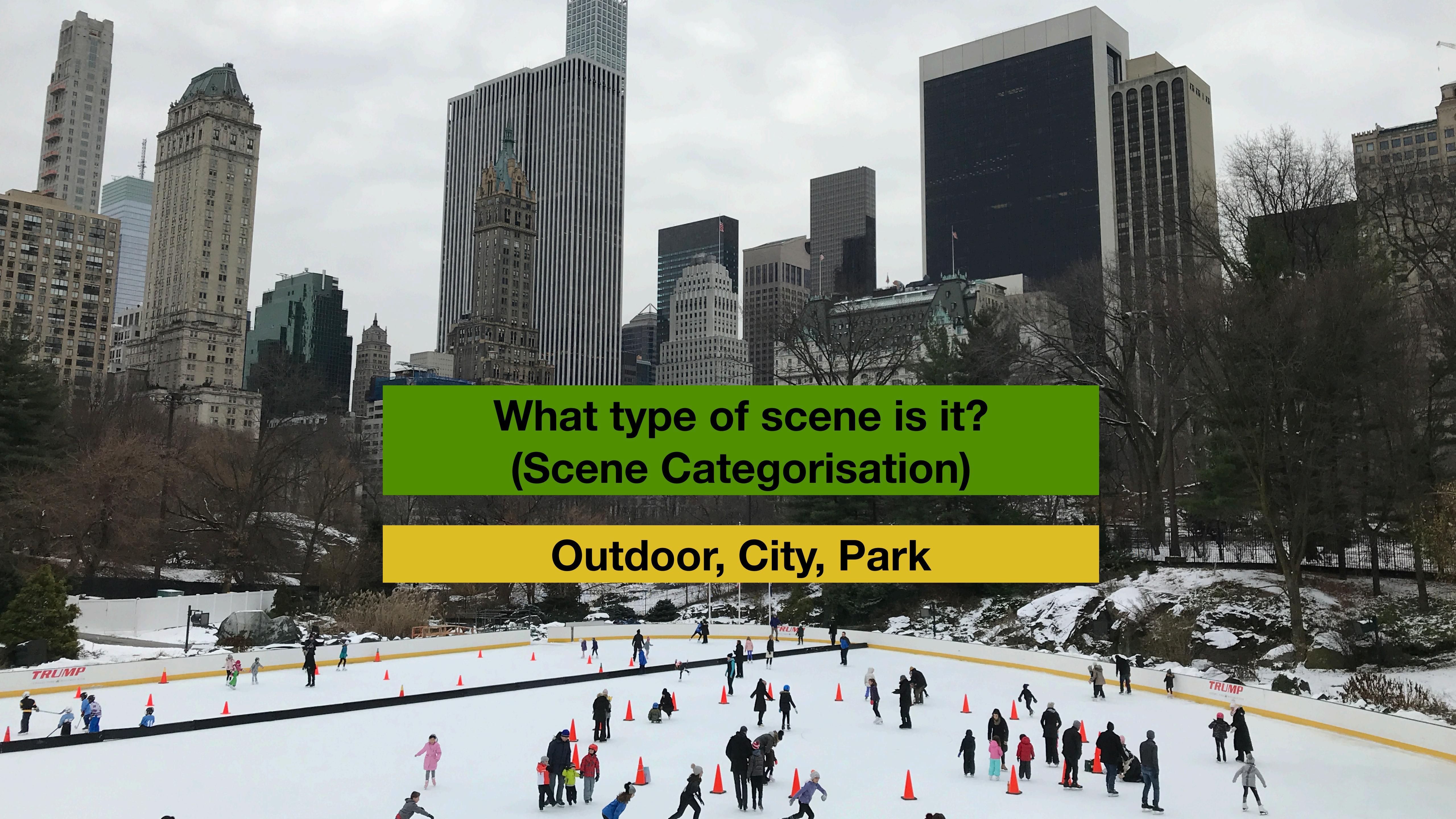
Data explorer

Solar 101

FAQ

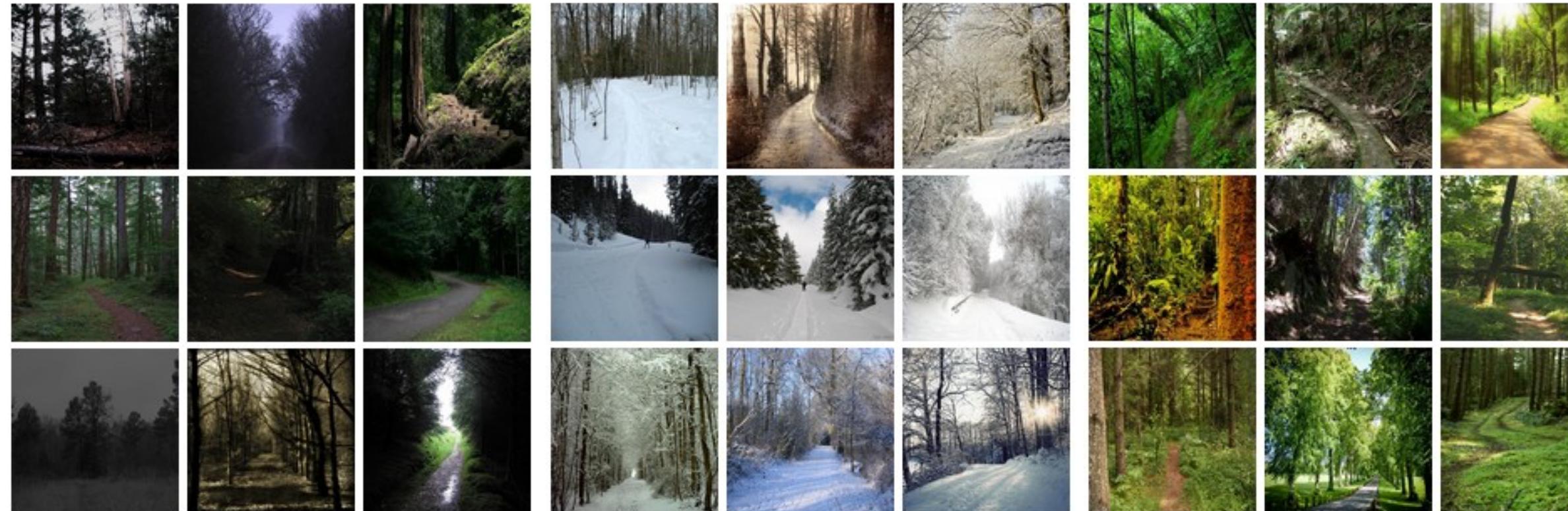


<https://www.google.com/get/sunroof>



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

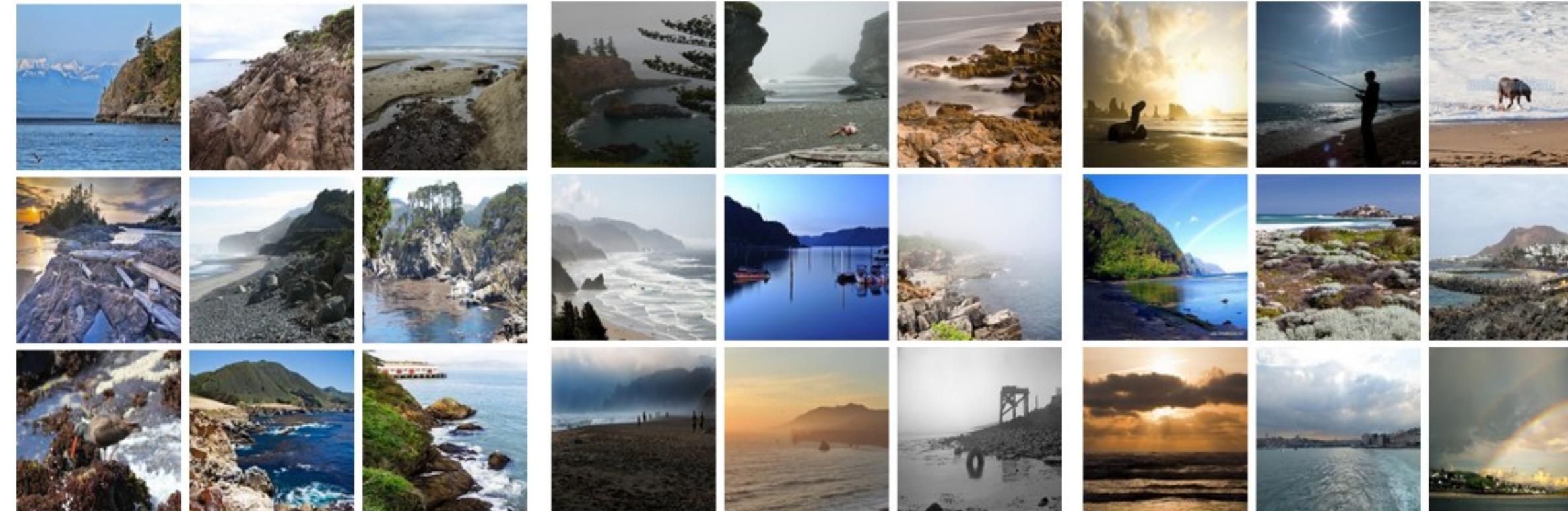
**Outdoor, City, Park**



**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

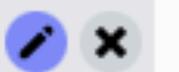
Running

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

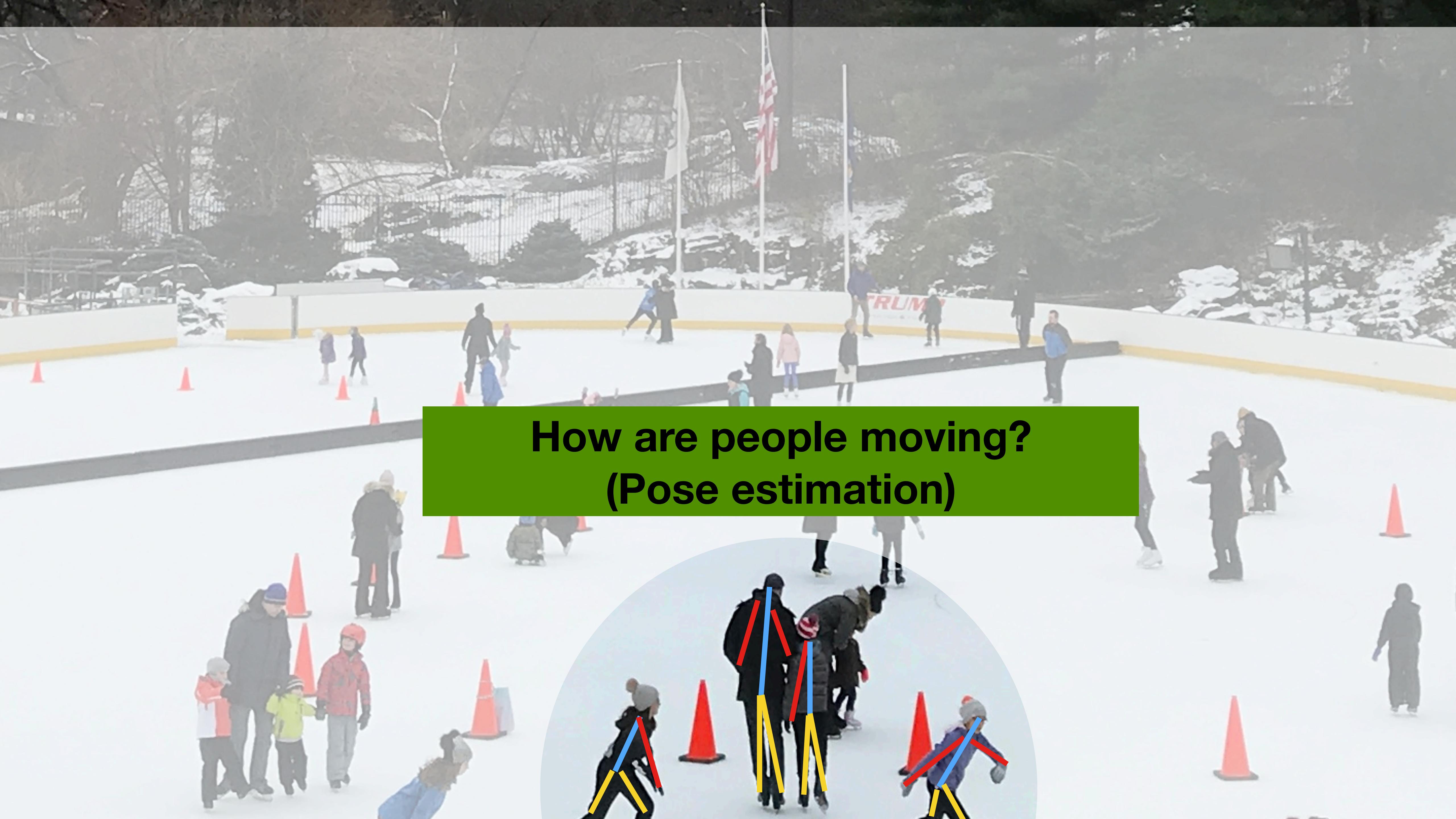
Submit

The background image shows a wide outdoor ice skating rink in a park. Numerous people are skating on the ice. In the distance, a dense skyline of tall buildings, including recognizable landmarks like the Woolworth Building and the Trump International Hotel and Tower, rises against a cloudy sky. Bare trees are scattered throughout the scene.

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

**Skating**





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



X: -1.11702489838885962072754  
Z: 25.3908080944445801  
Work station



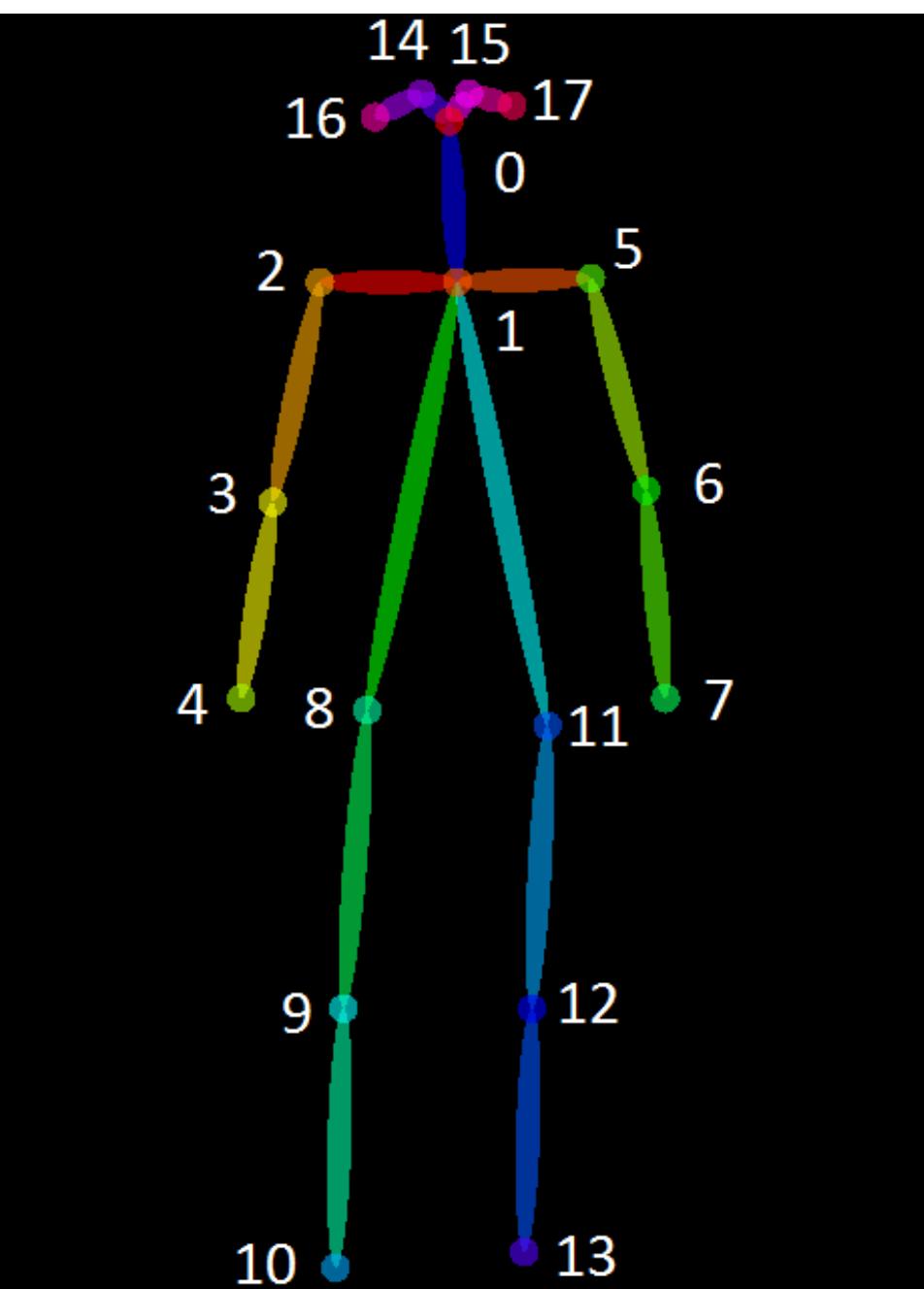
Ethical task tracking of operators in agile manufacturing

<http://resolver.tudelft.nl/>

uuid:3408e8c3-809b-436d-94eb-efb4f0532b17

# Stereolabs ZED Camera

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





**Are these images of the same person?  
(Image / Face Similarity)**

**Bonus if you guess  
the movie!**

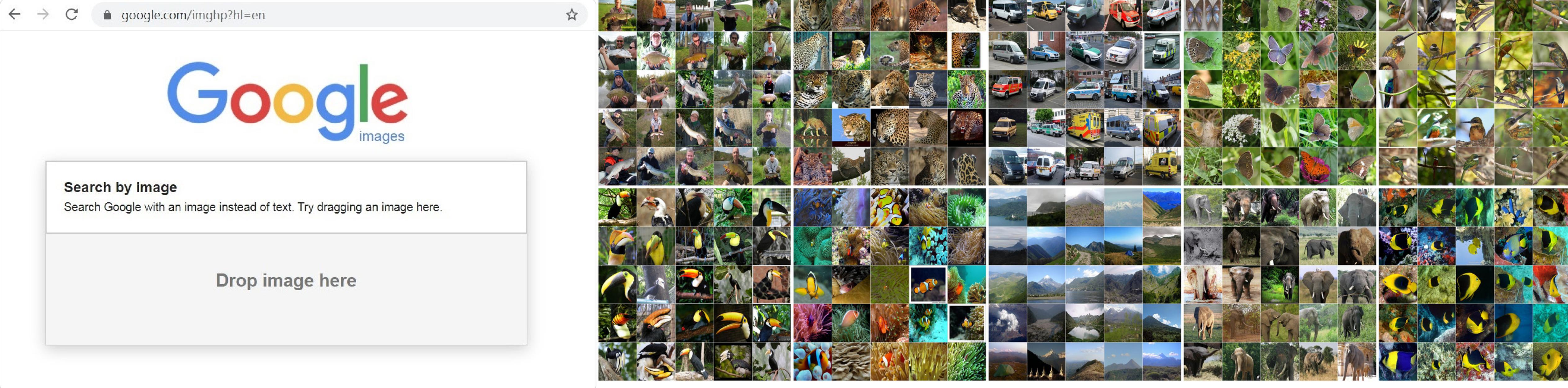
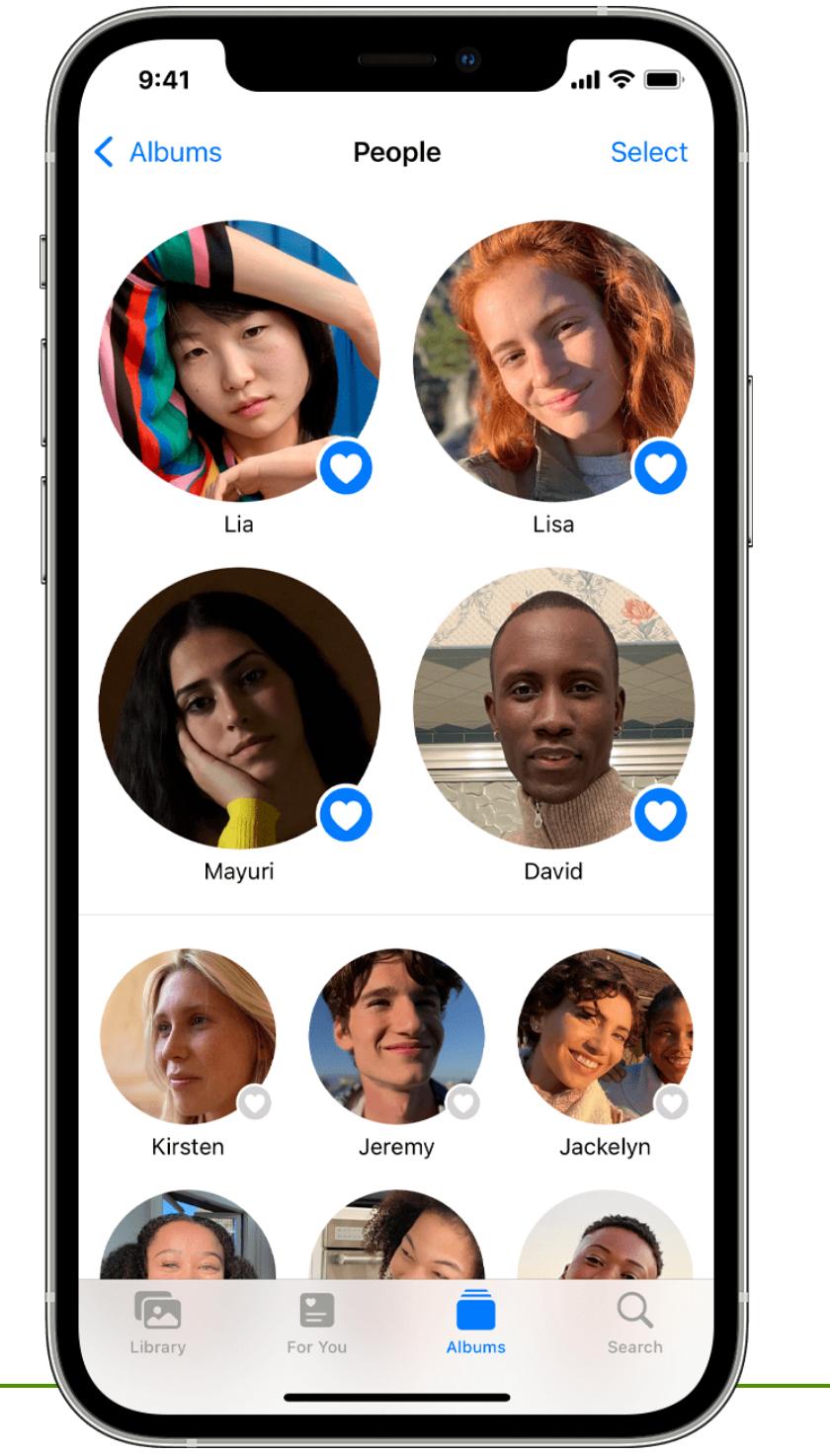
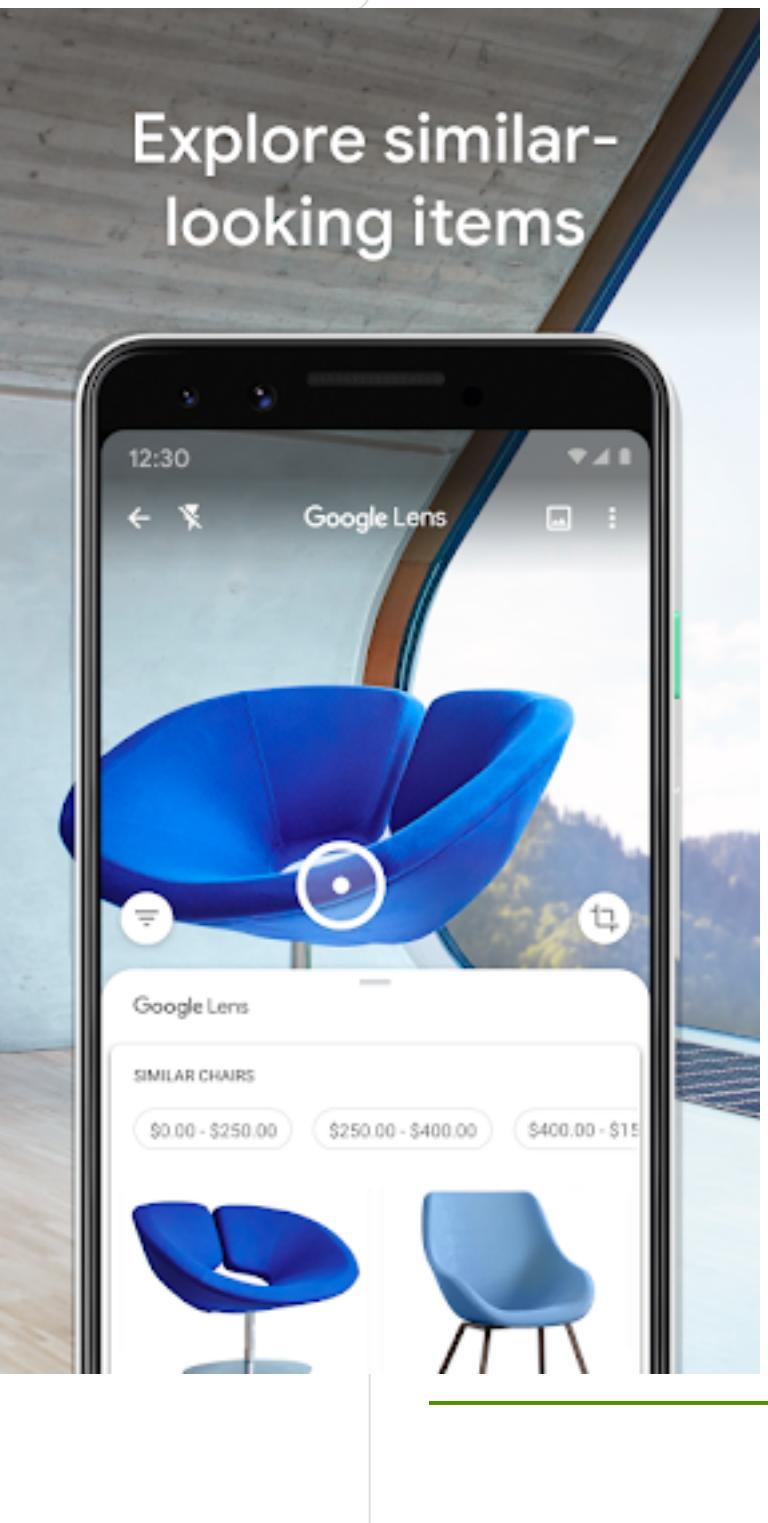


Image search results

88432.jpg X Lamp

Visually similar images

A screenshot of the Google Images search results for the term "Lamp". The search bar at the top shows the file name "88432.jpg" and the word "Lamp". Below the search bar, the heading "Visually similar images" is displayed. A grid of images shows various desk lamps, mostly red and black, arranged in a 5x5 pattern. At the bottom left, there is a link "Report images".

# Machine Learning For Design

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Lecture 3 - Machine Learning for Images

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16/02/2022

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[www.ml4design.com](http://www.ml4design.com)

# Credits

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- CMU Computer Vision course - Matthew O'Toole. <http://16385.courses.cs.cmu.edu/spring2022/>