

Machine Learning for Design

Lecture 3

Machine Learning for Images. *Part 1*

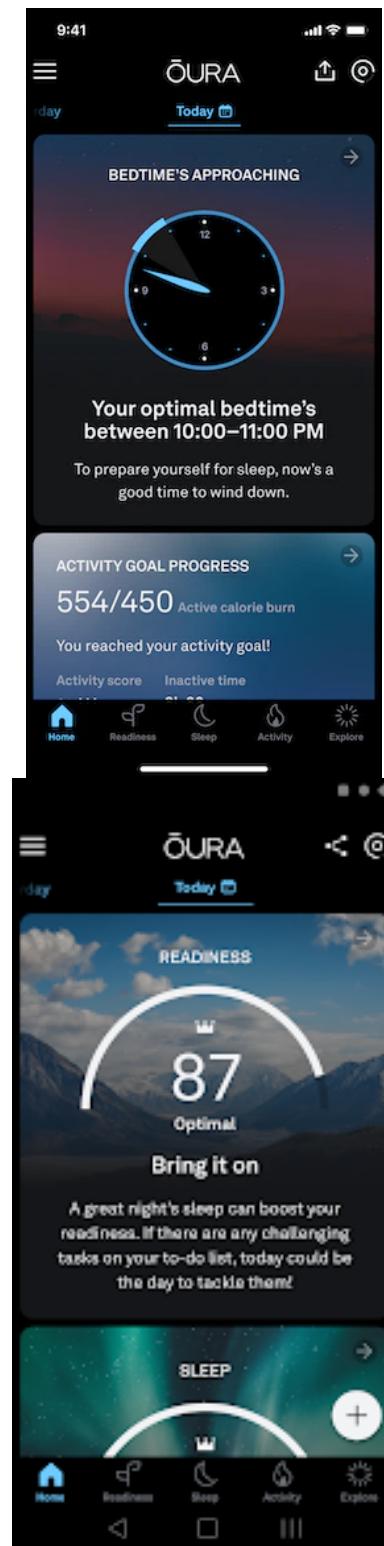
Administration

- Group formation
 - 48 students
 - 3 group of 3 → get together?
 - Or reach out to me, and I will re-distribute
- Forum in Brightspace
 - Collaboration > Discussions
 - For questions and discussions about lectures, assignments, etc.

(Reverse) Design Exercise Oura Ring 3



<https://ouraring.com/oura-experience>



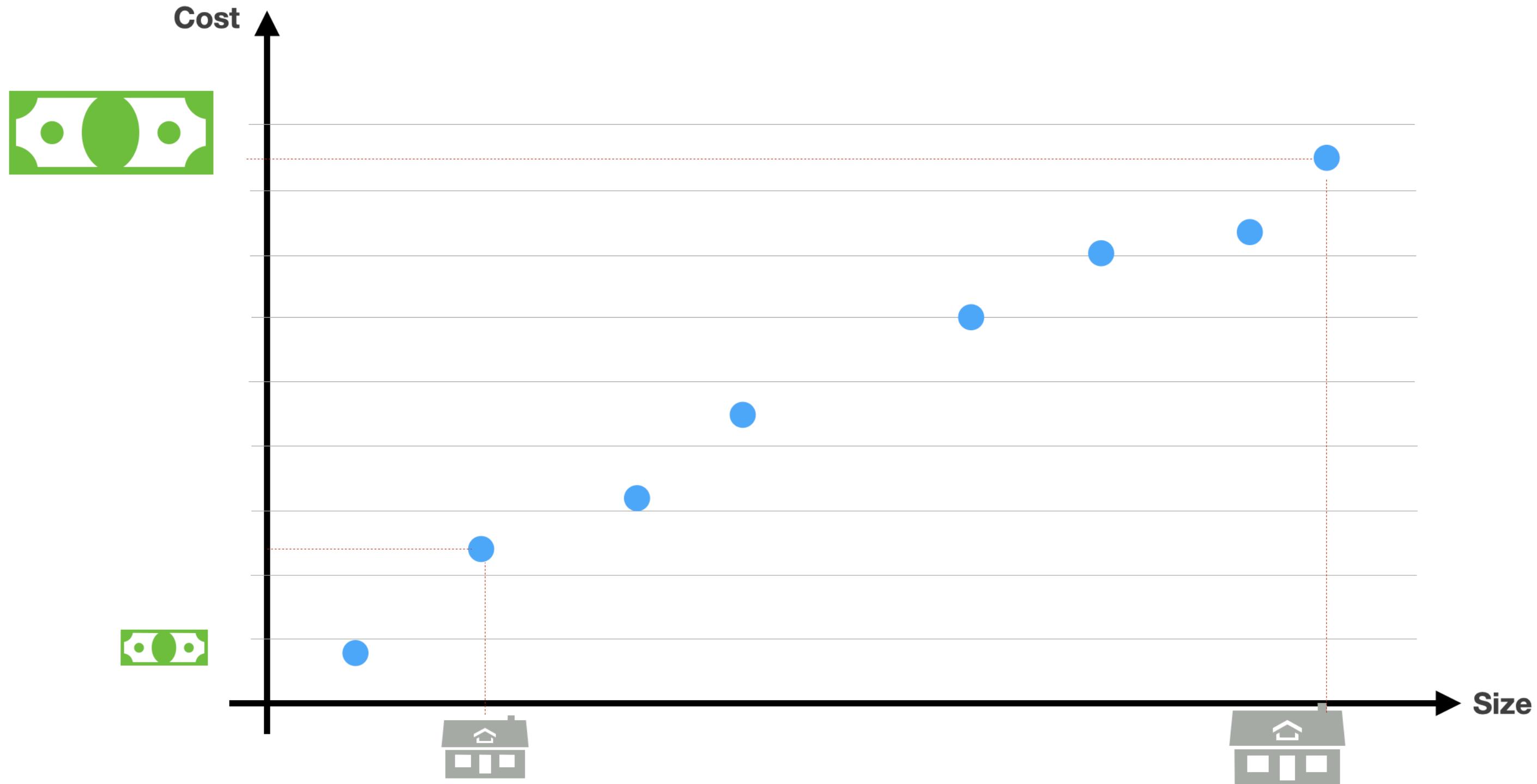
Formative Assignment

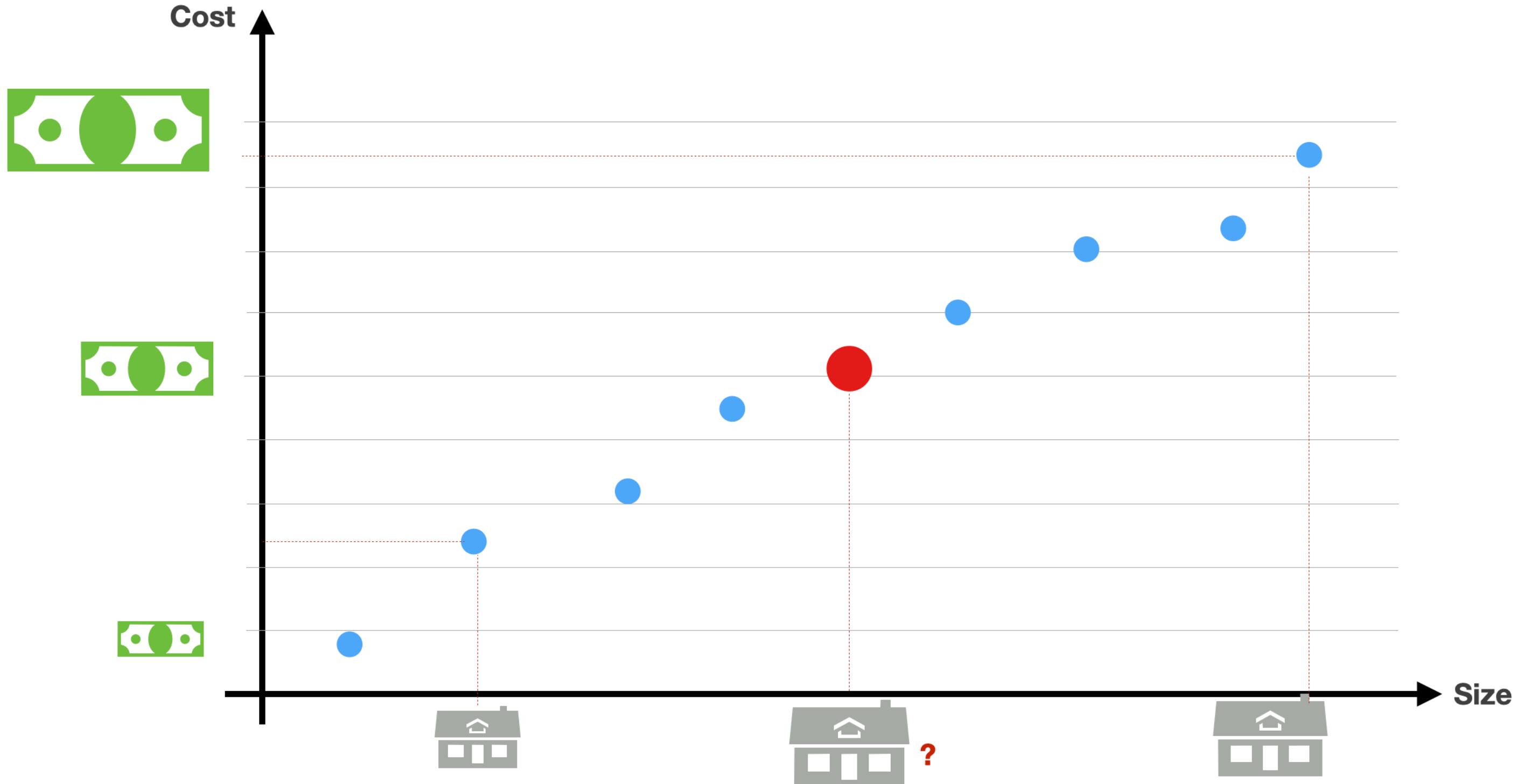
- Fill in the questionnaire in Bright Space
 - Materials > Week 3 > ML Design
- Questionnaire: OUR A
- I will analyze your answers, and we will discuss them together next week

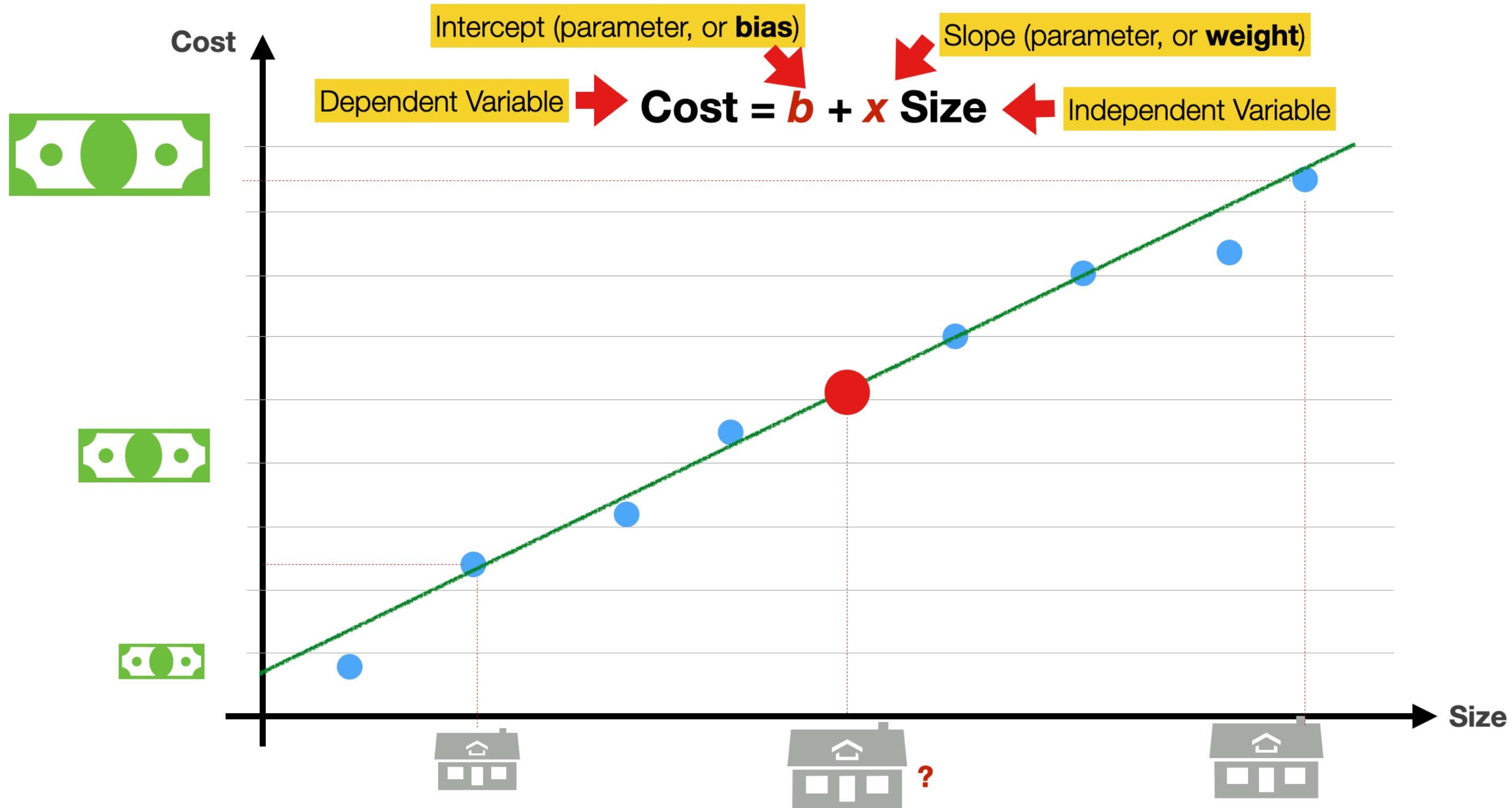
A bit more on regression and classification

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

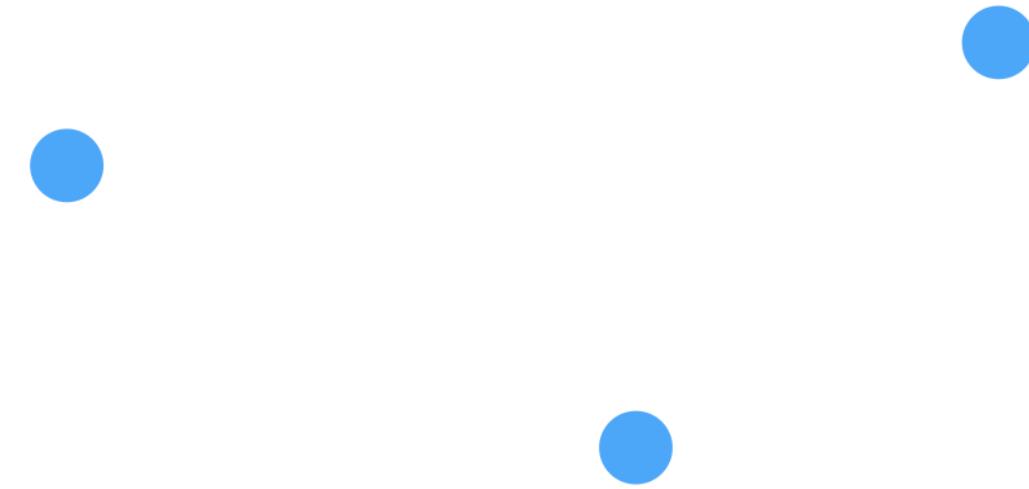
Linear Regression



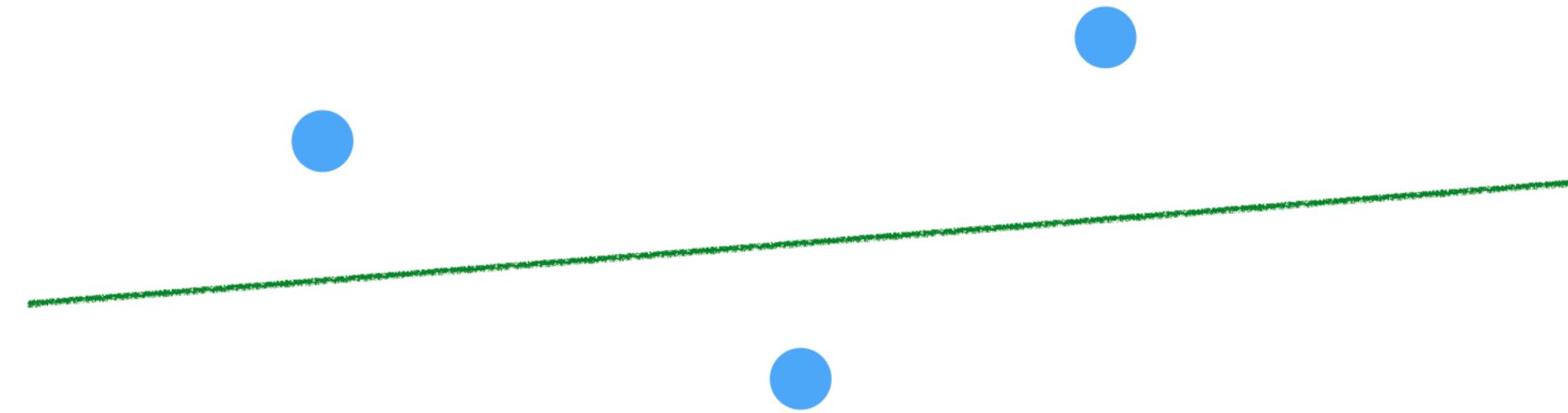




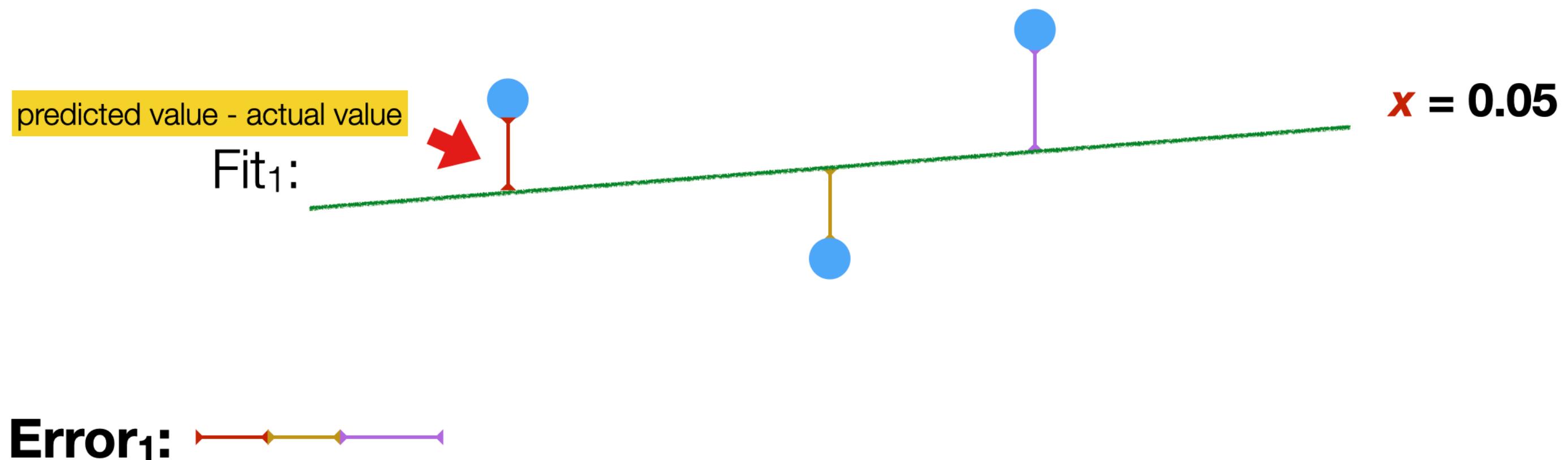
Cost = \times Size



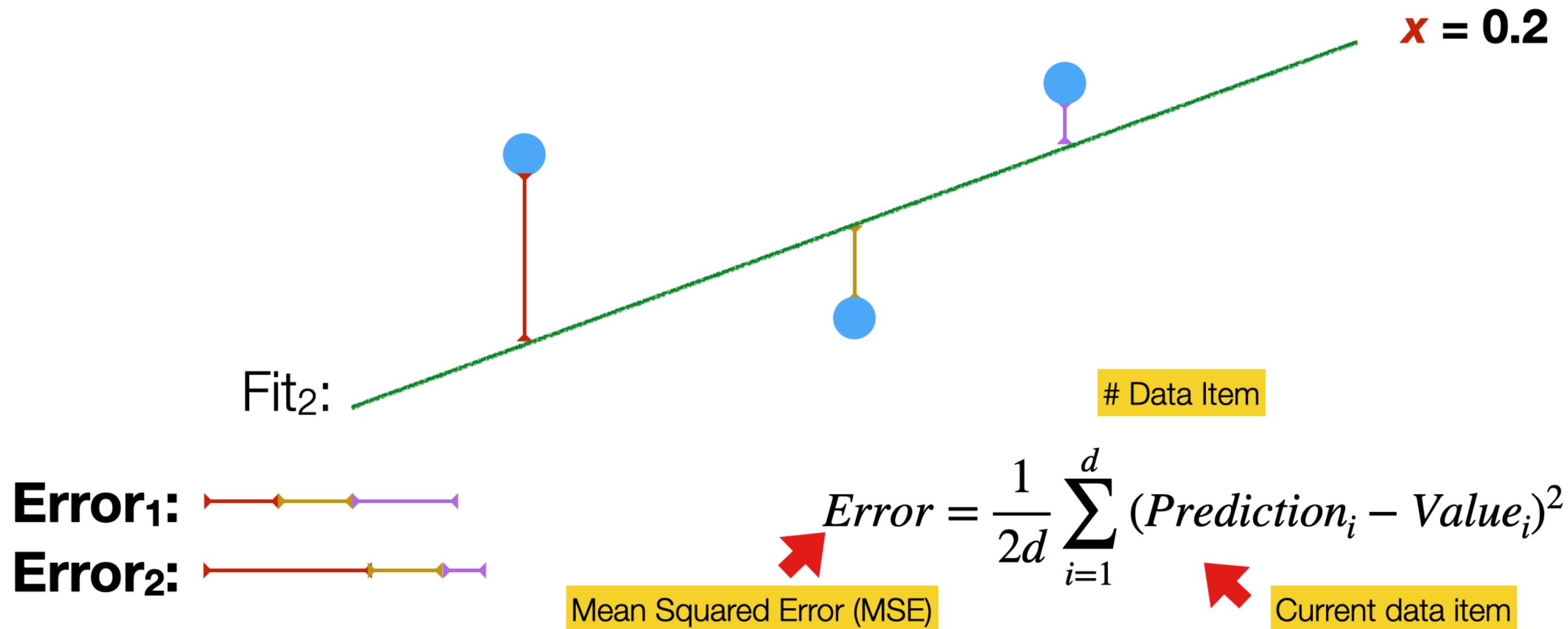
Cost = \times Size



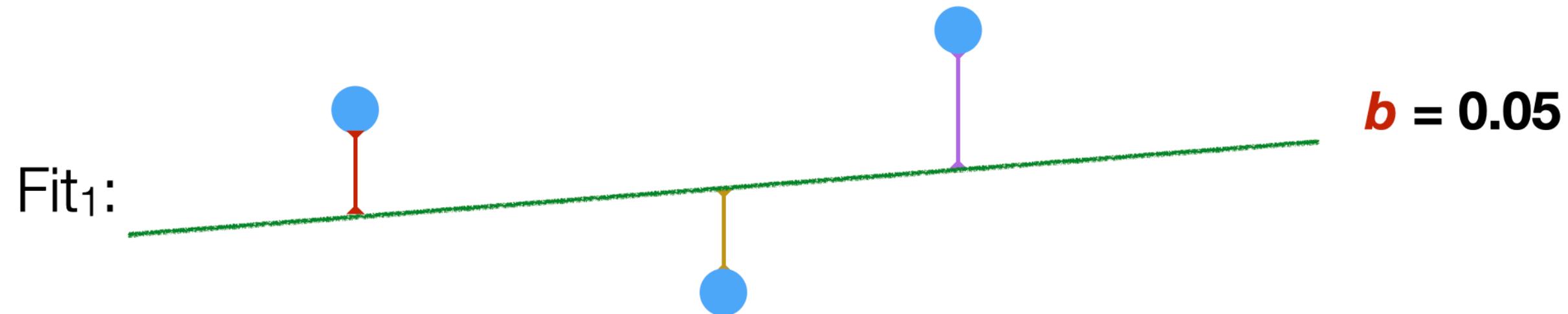
Cost = \times Size



Cost = x Size



Cost = x Size



Error₁:

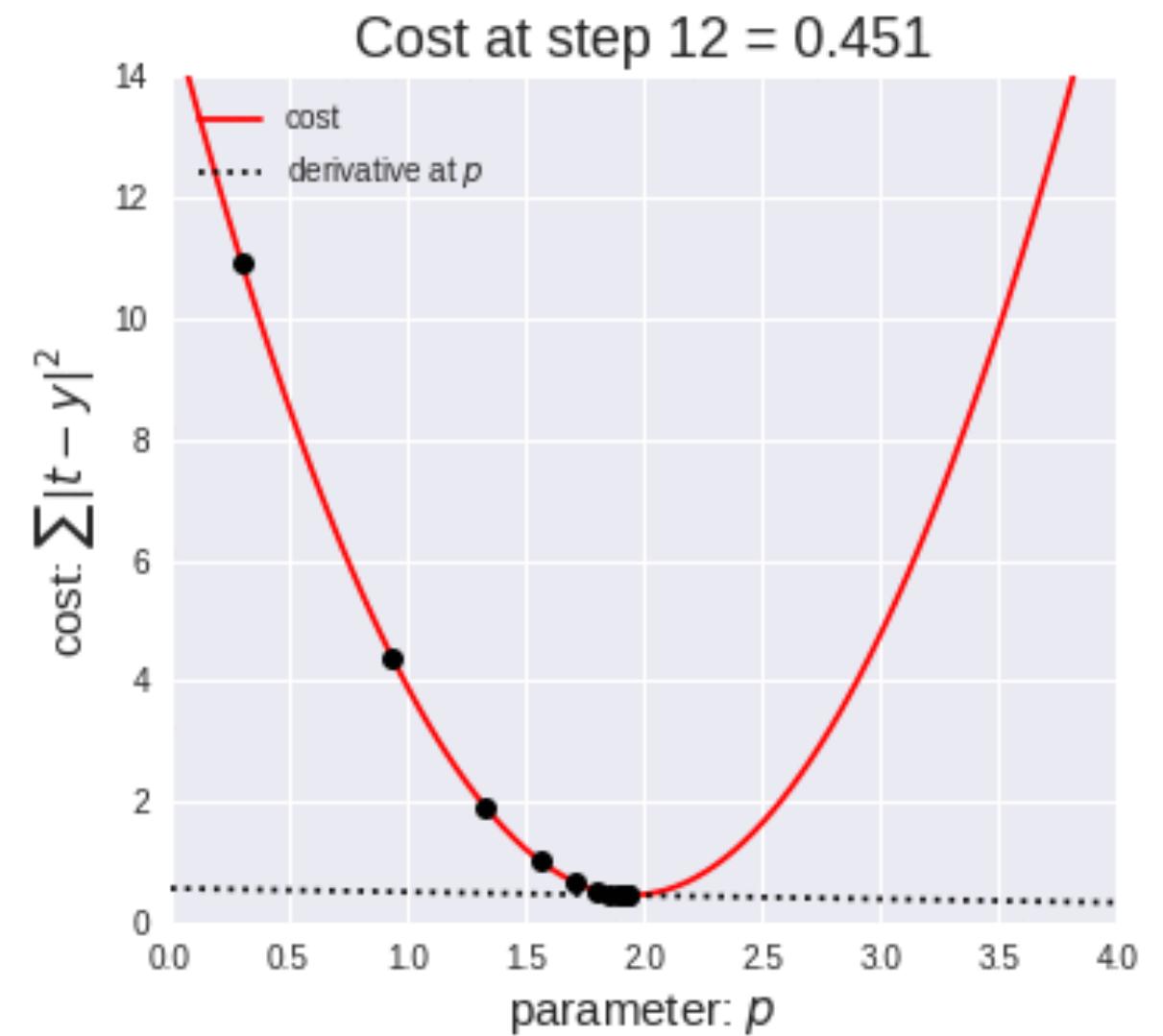
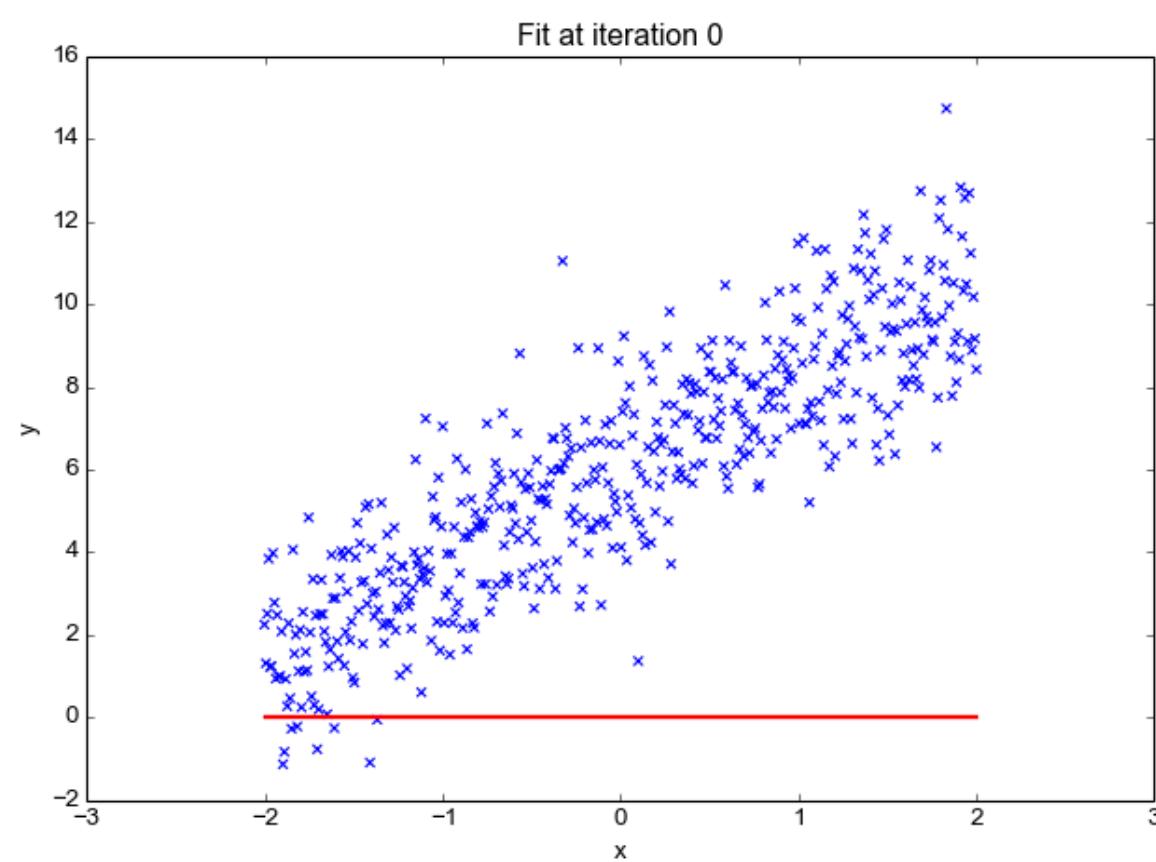
Error₂:

Fit_1 is a better fit on the training data than Fit_2

We select $b = 0.05$

Finding the best parameter values

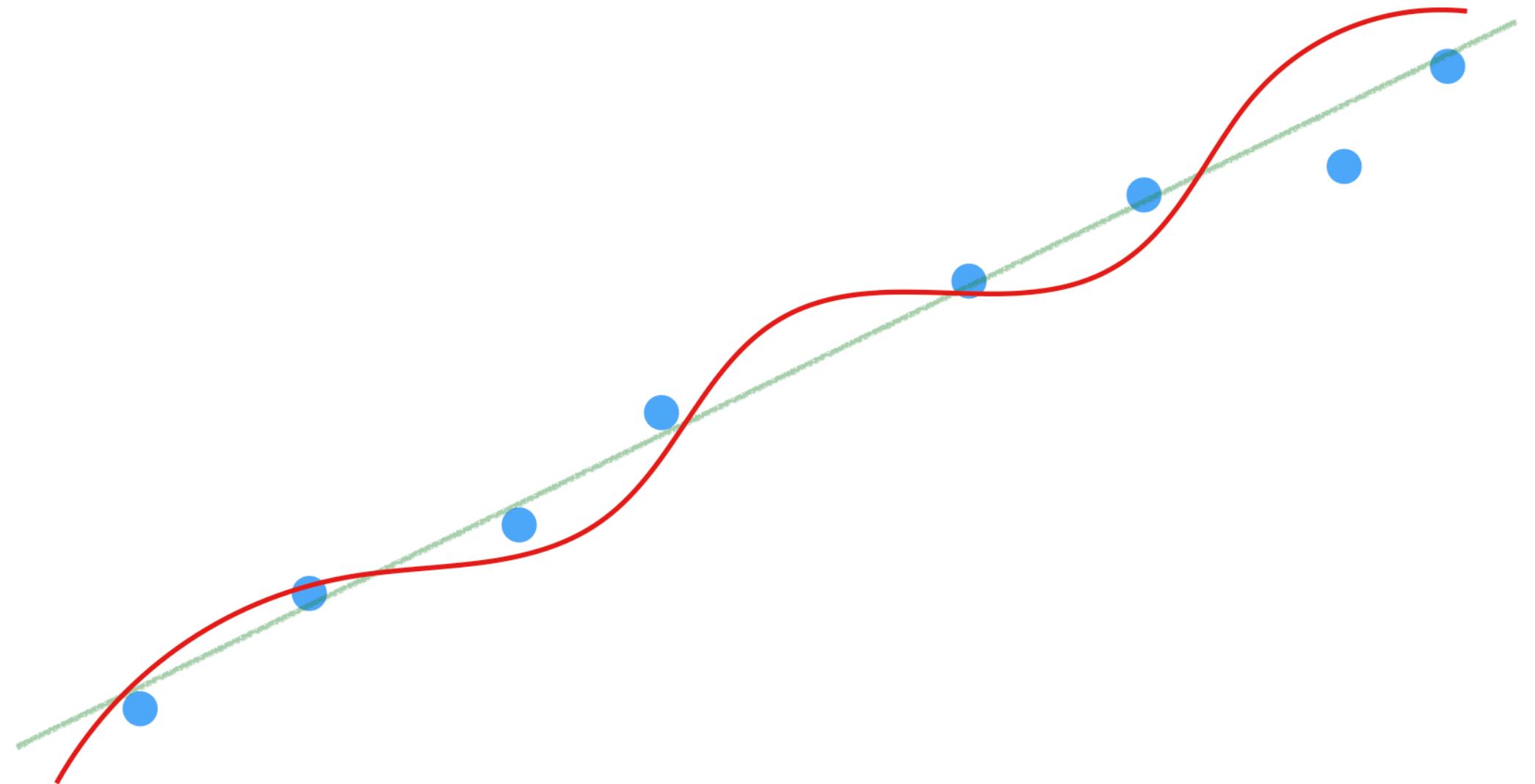
- Training the model
- **Gradient Descent:** an algorithm to find the minimum point of a function
- **Hyper parameters:** parameters of the Gradient Descent
 - *Learning Rate:* speed of descent
 - *Epochs:* max number of steps



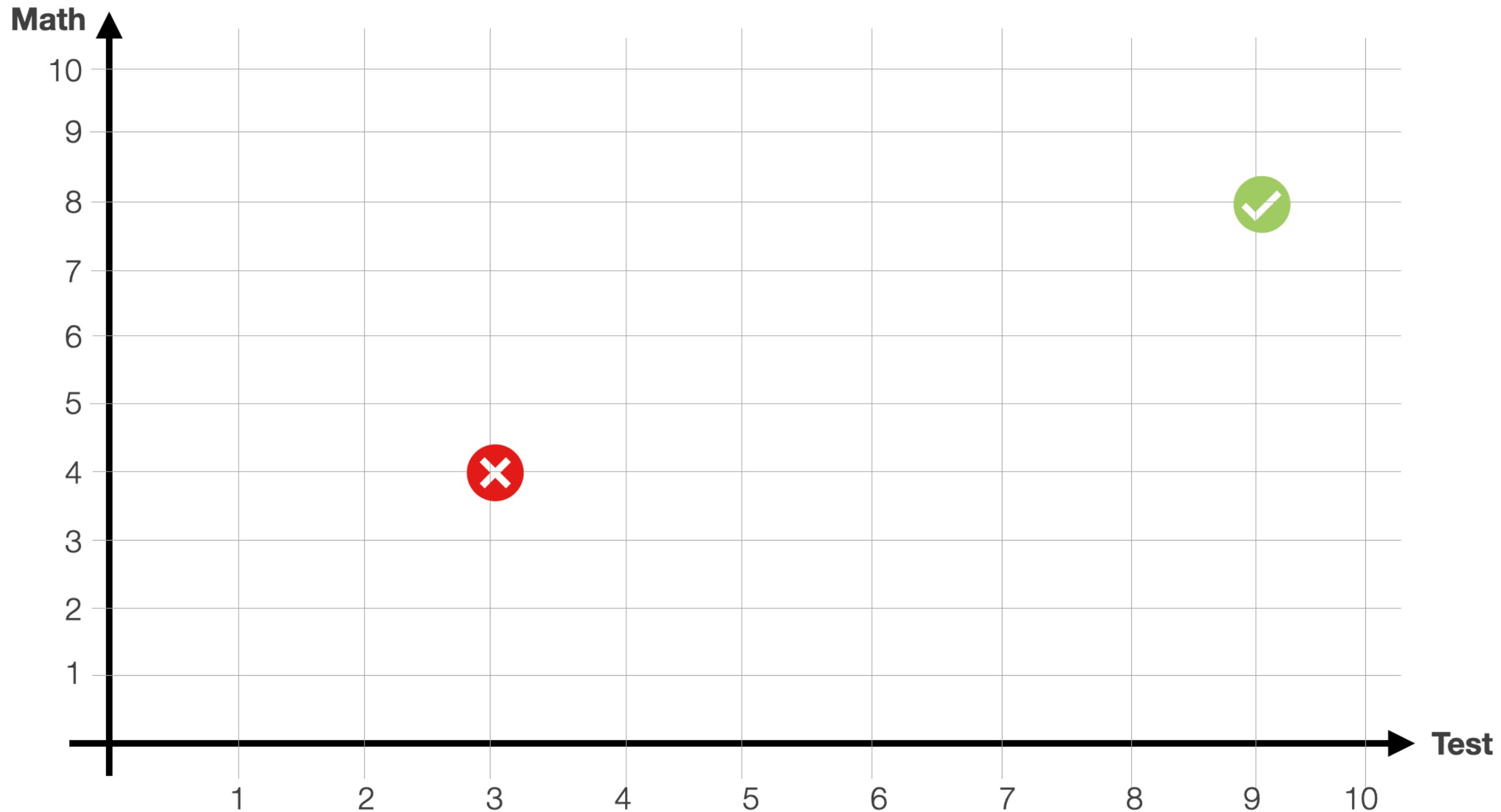
Nth degree polynomial

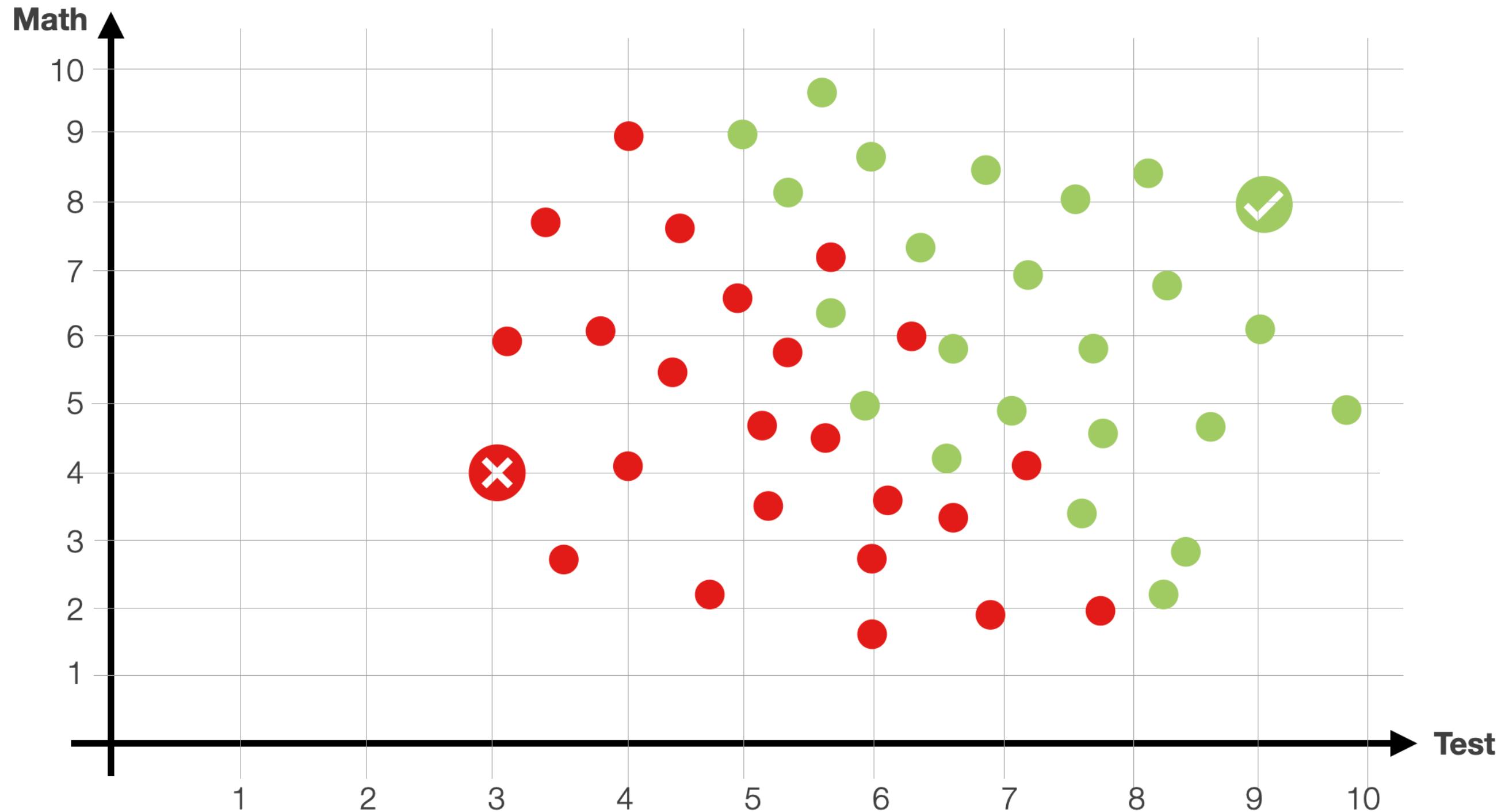


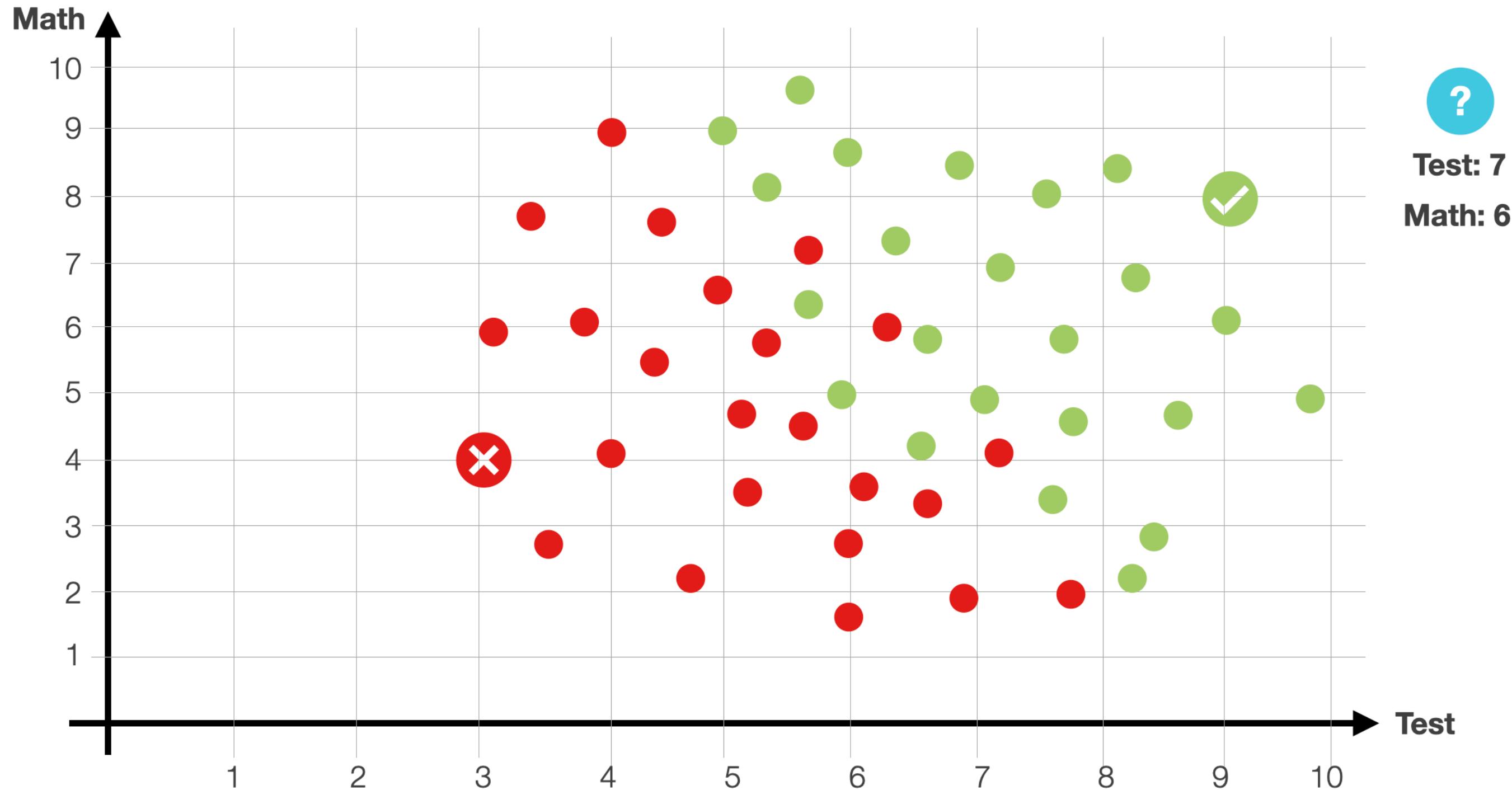
$$\text{Cost} = b + x_1 \text{ Size} + x_2 \text{ Size}^2 + \dots + x_n \text{ Size}^n$$

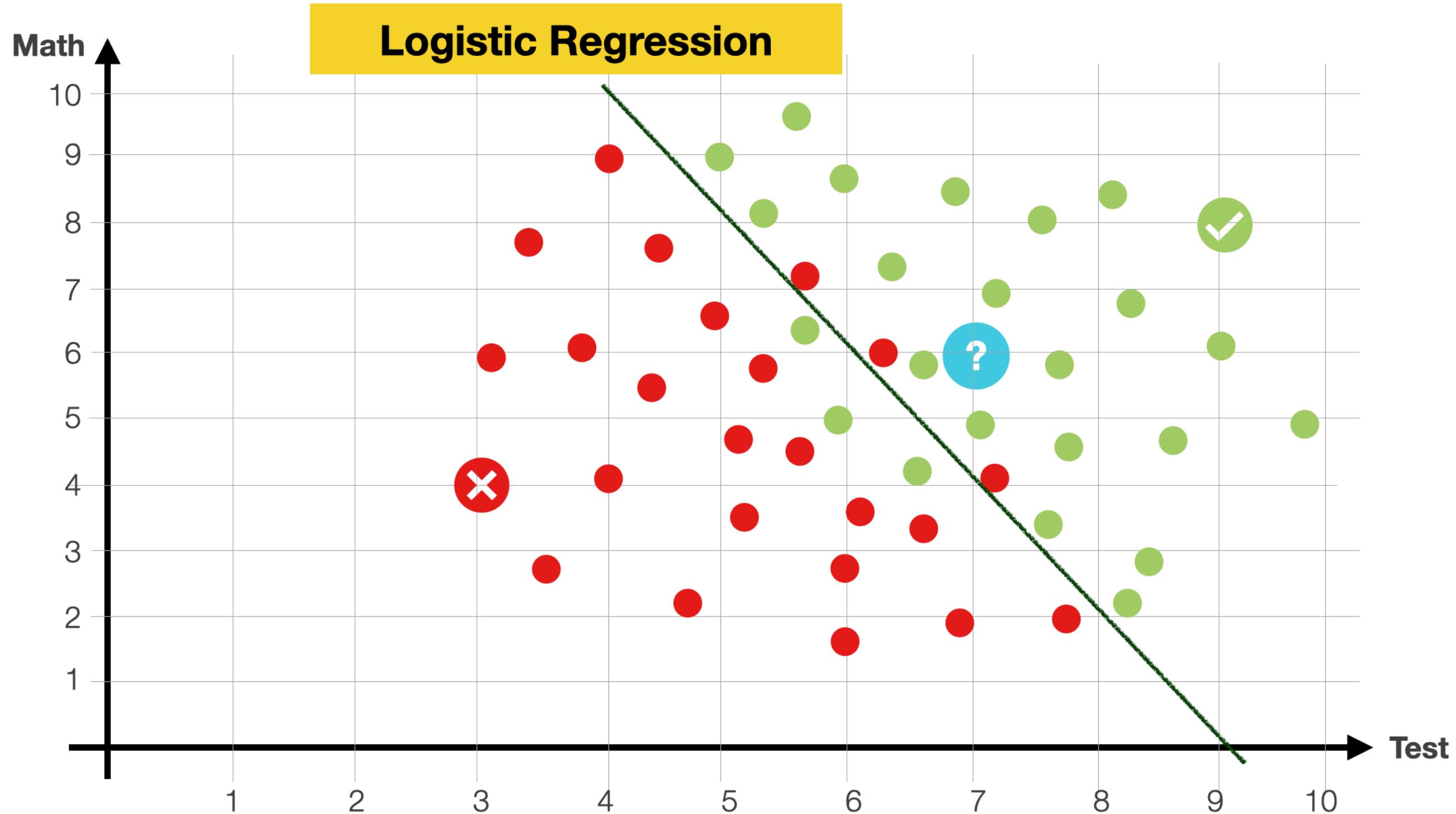


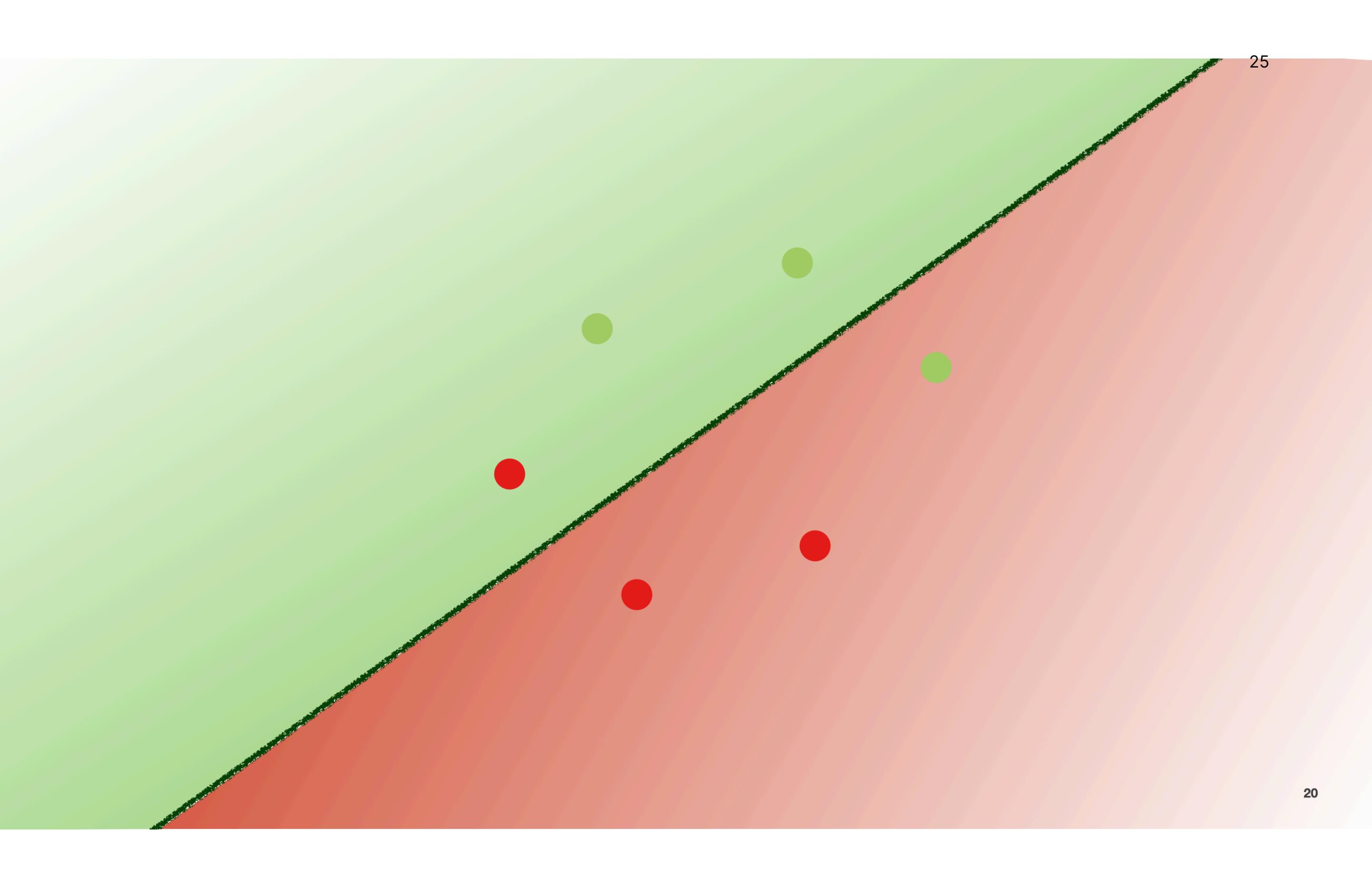
Classification











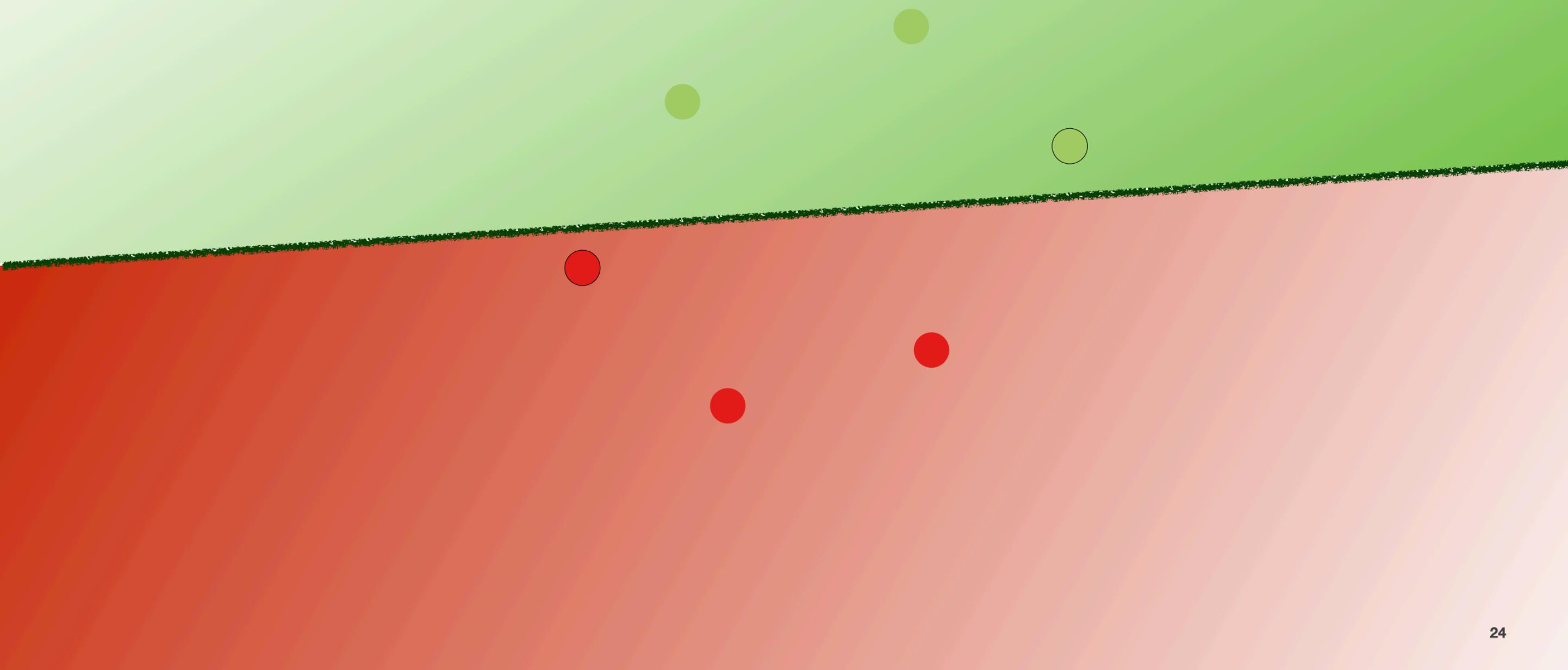
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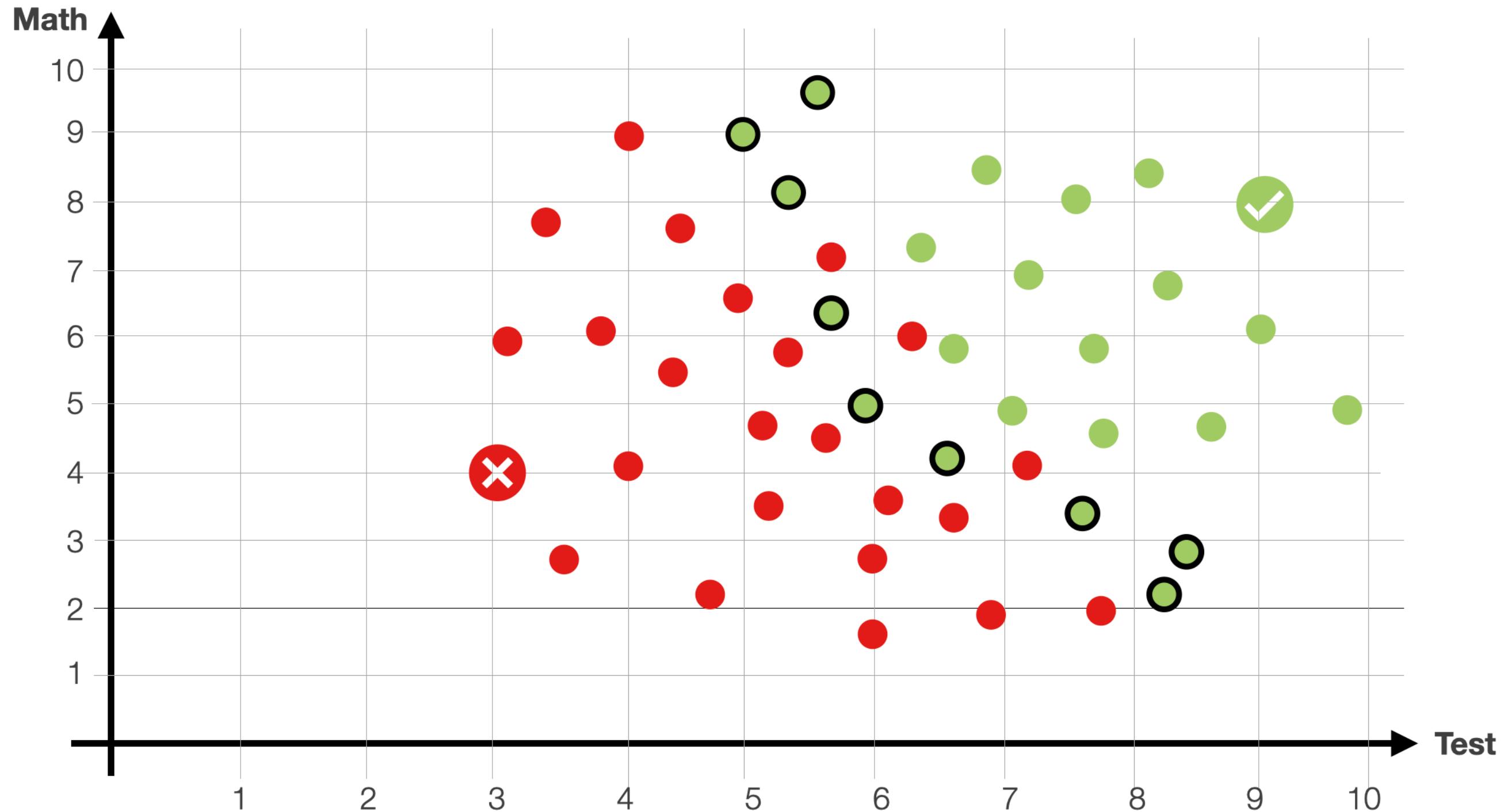
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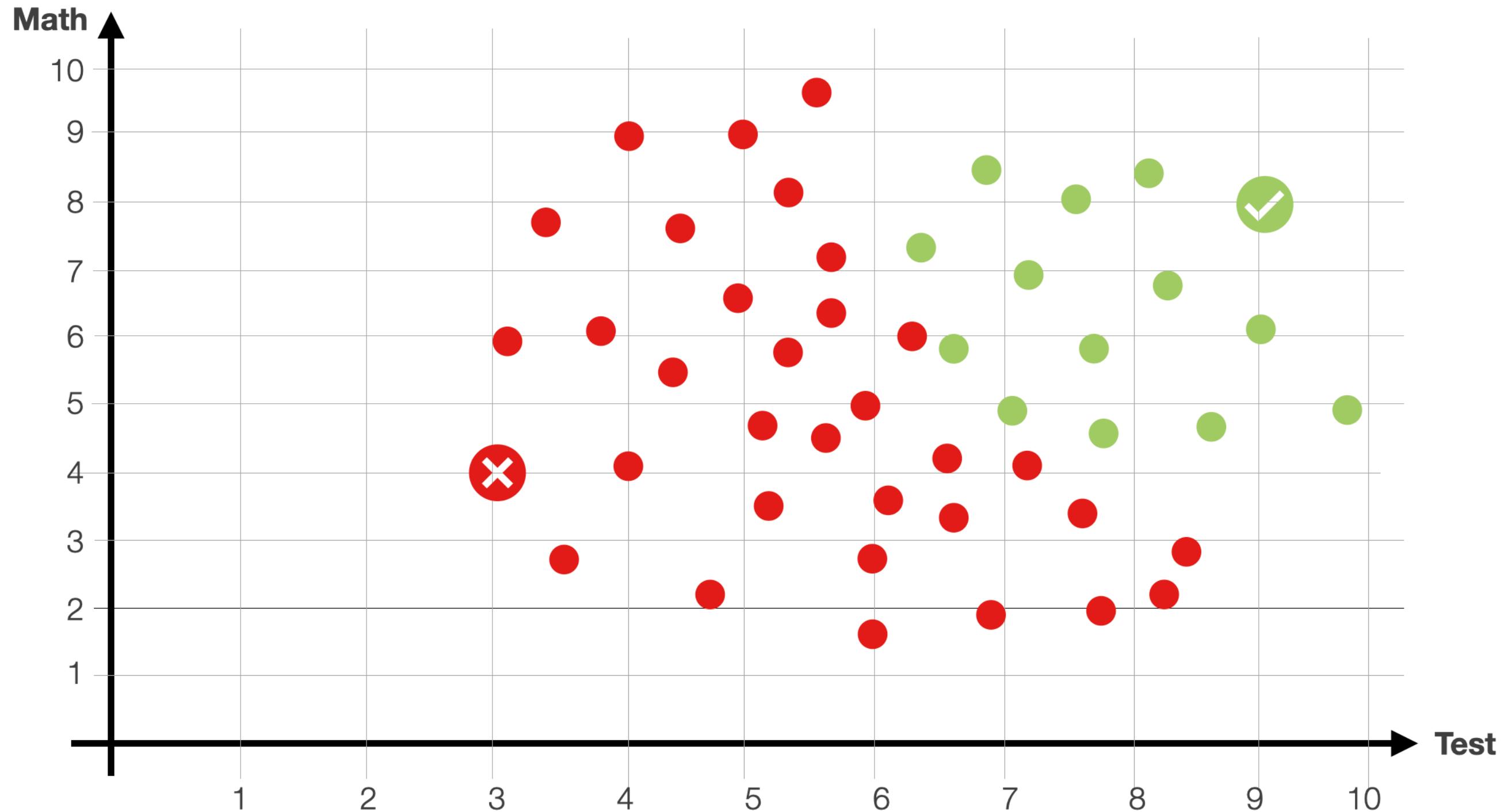
Error: 2

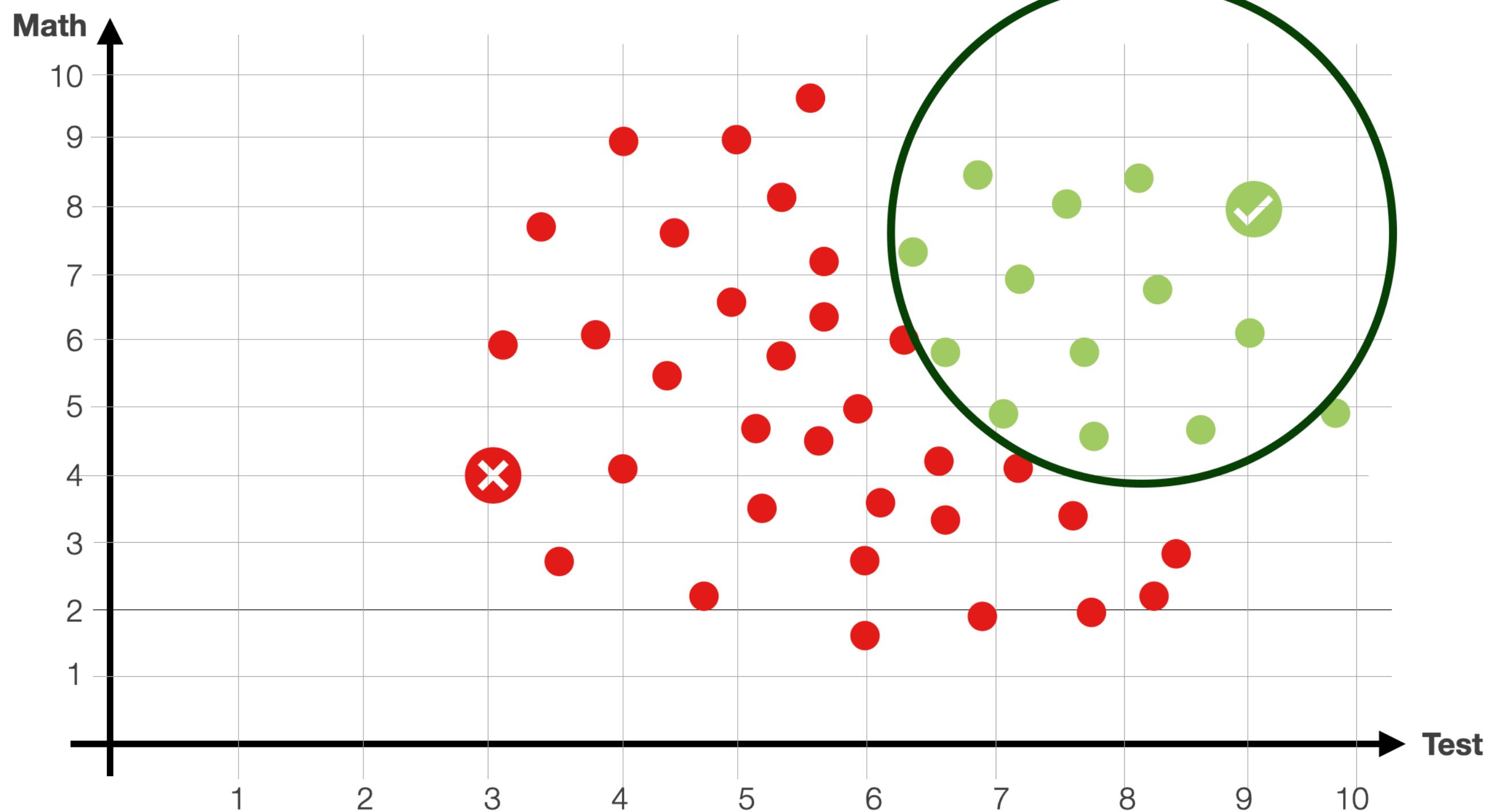
Error: 1

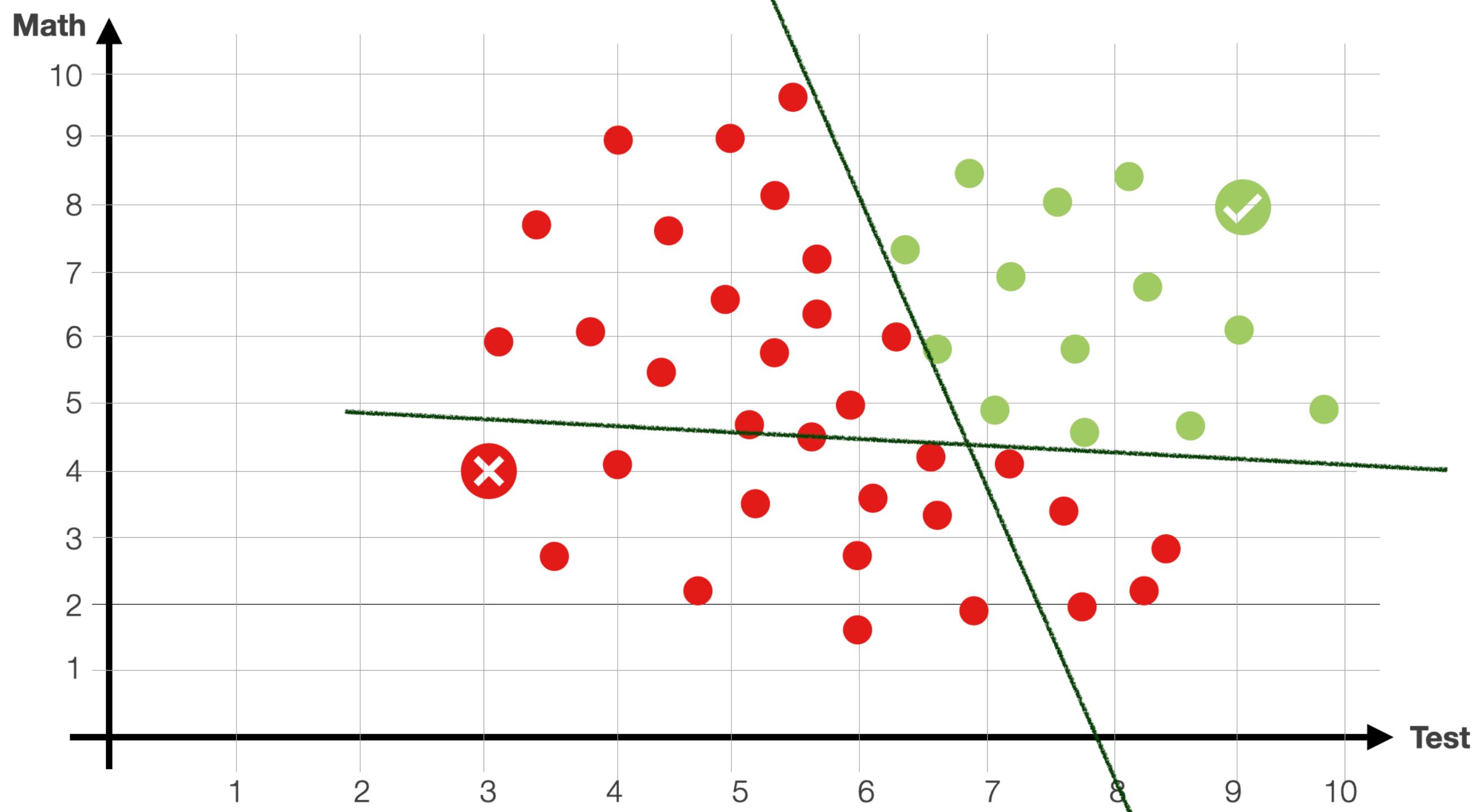
Error: 0

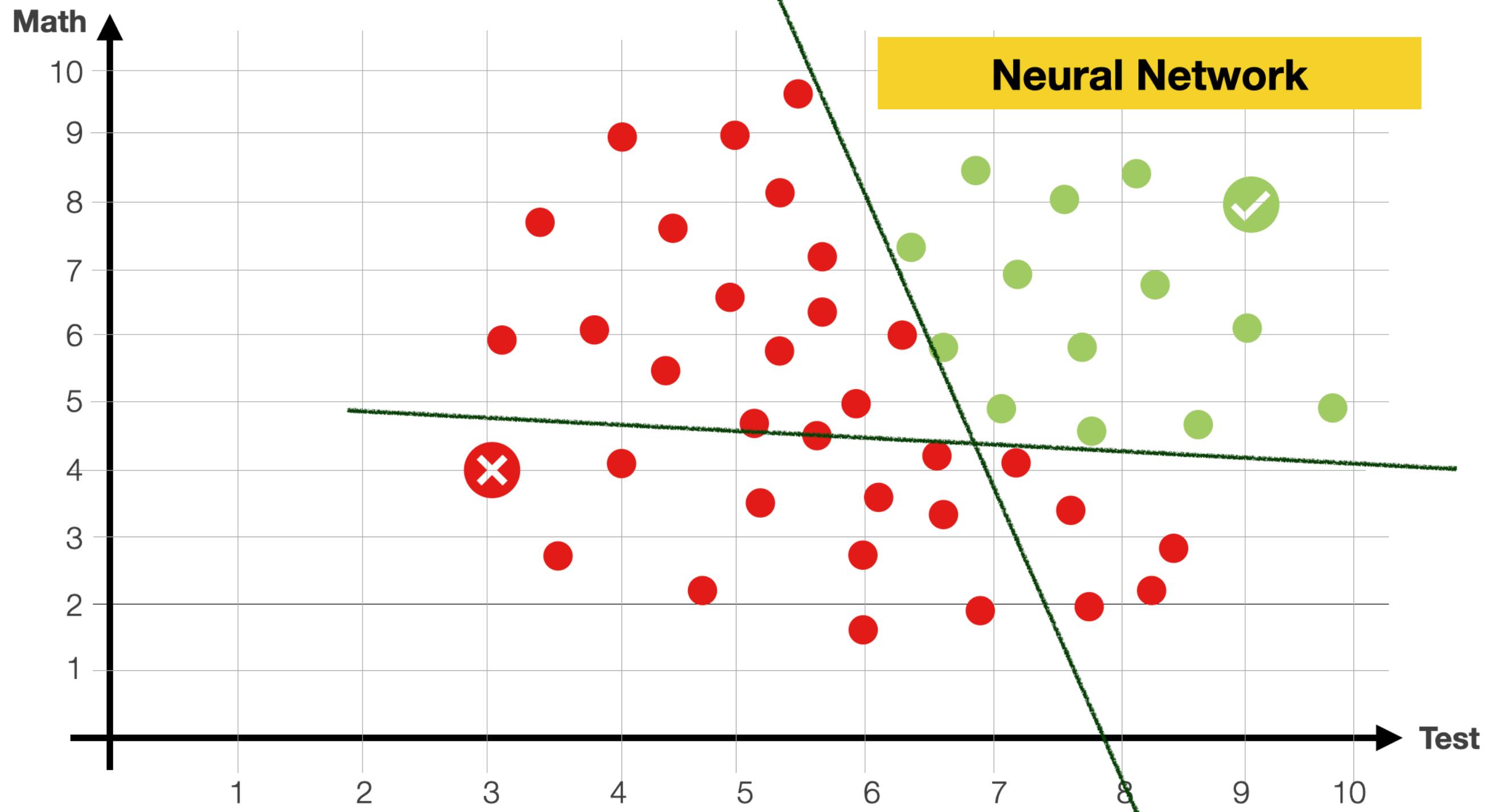


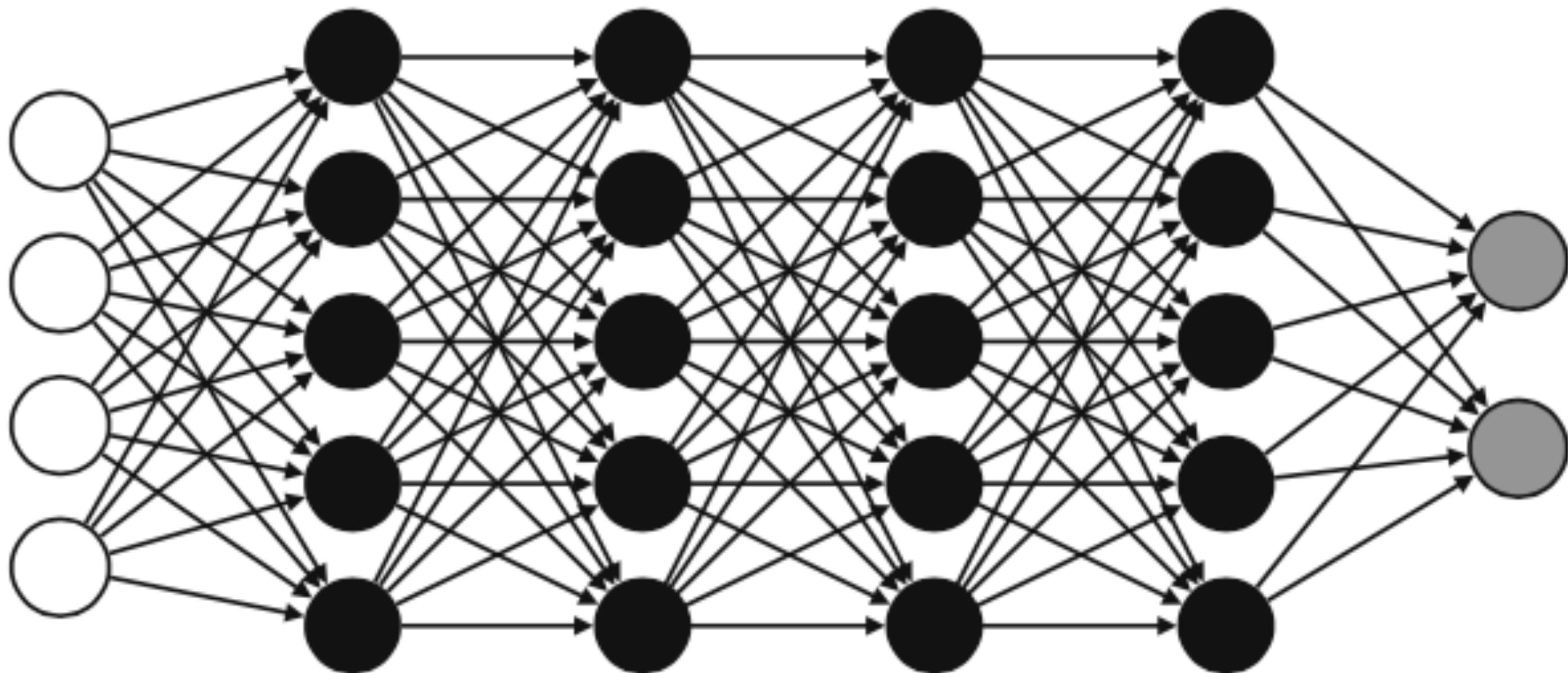


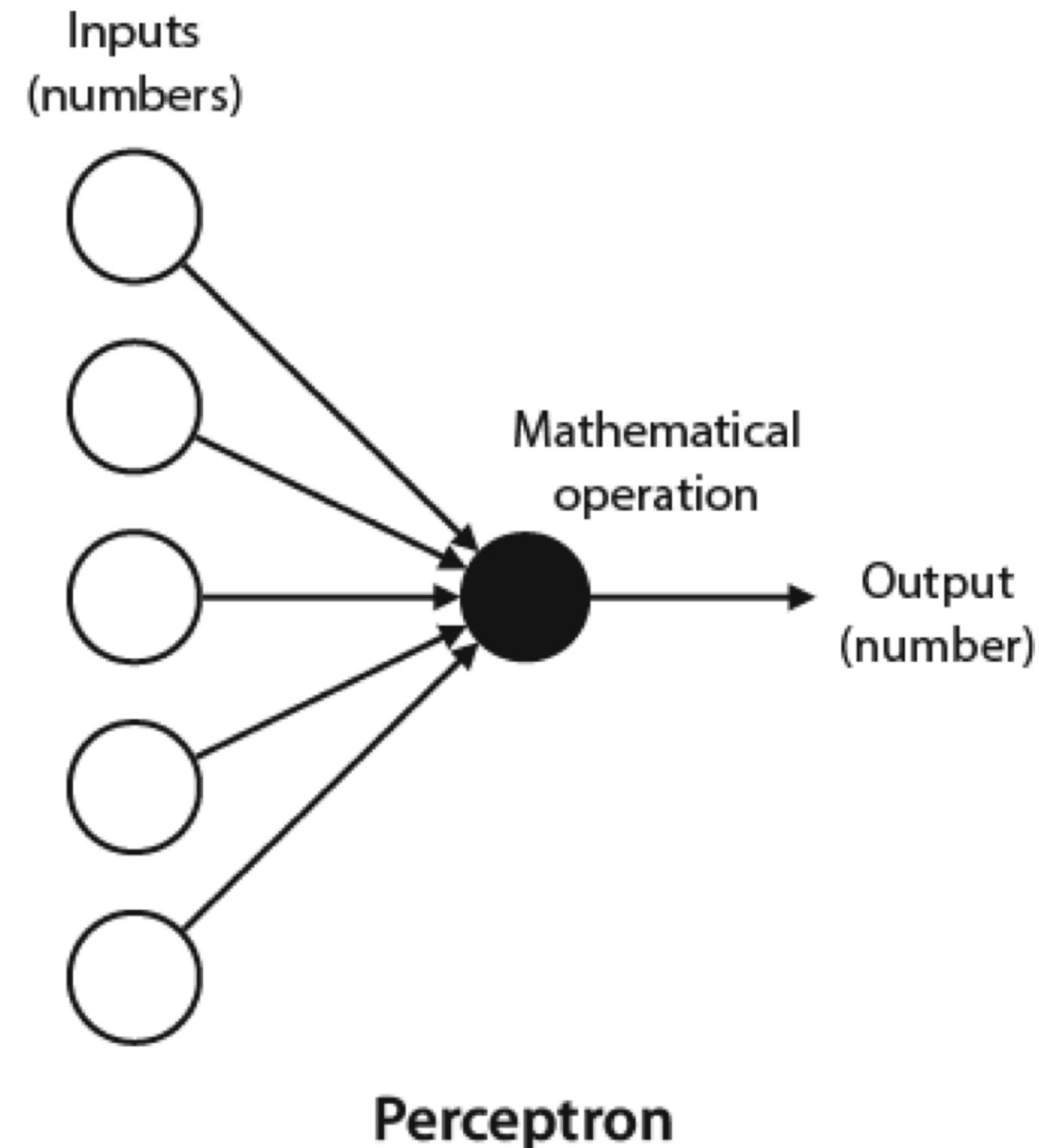
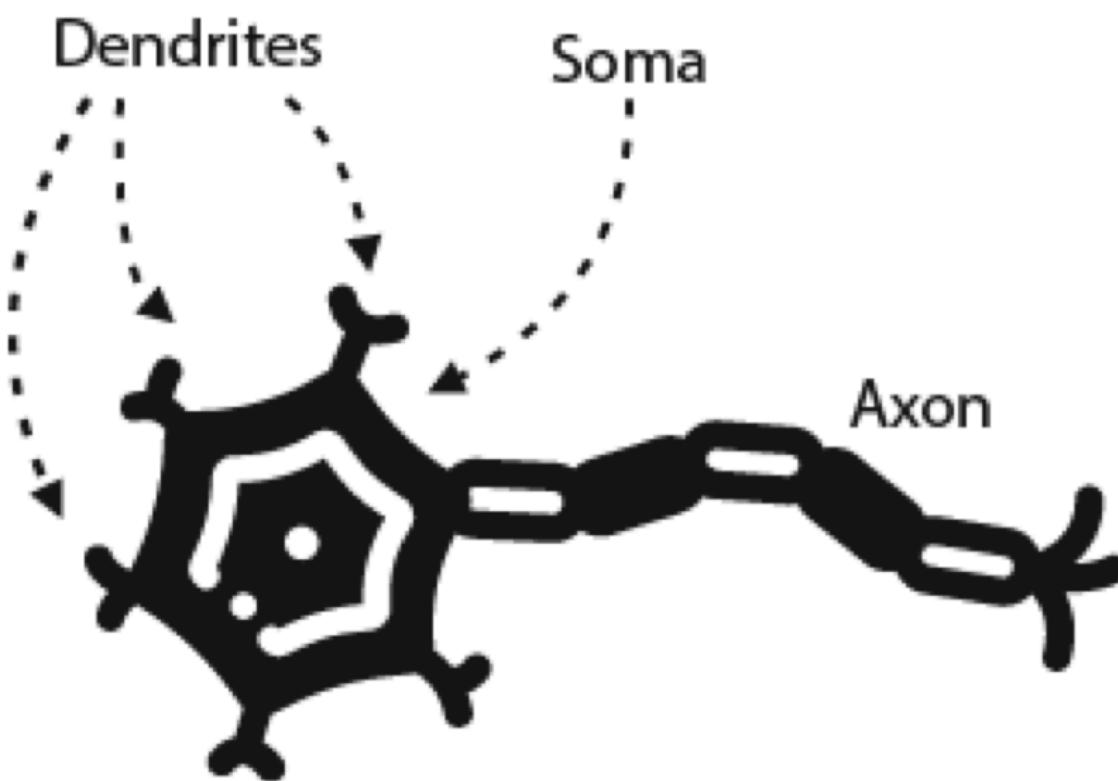


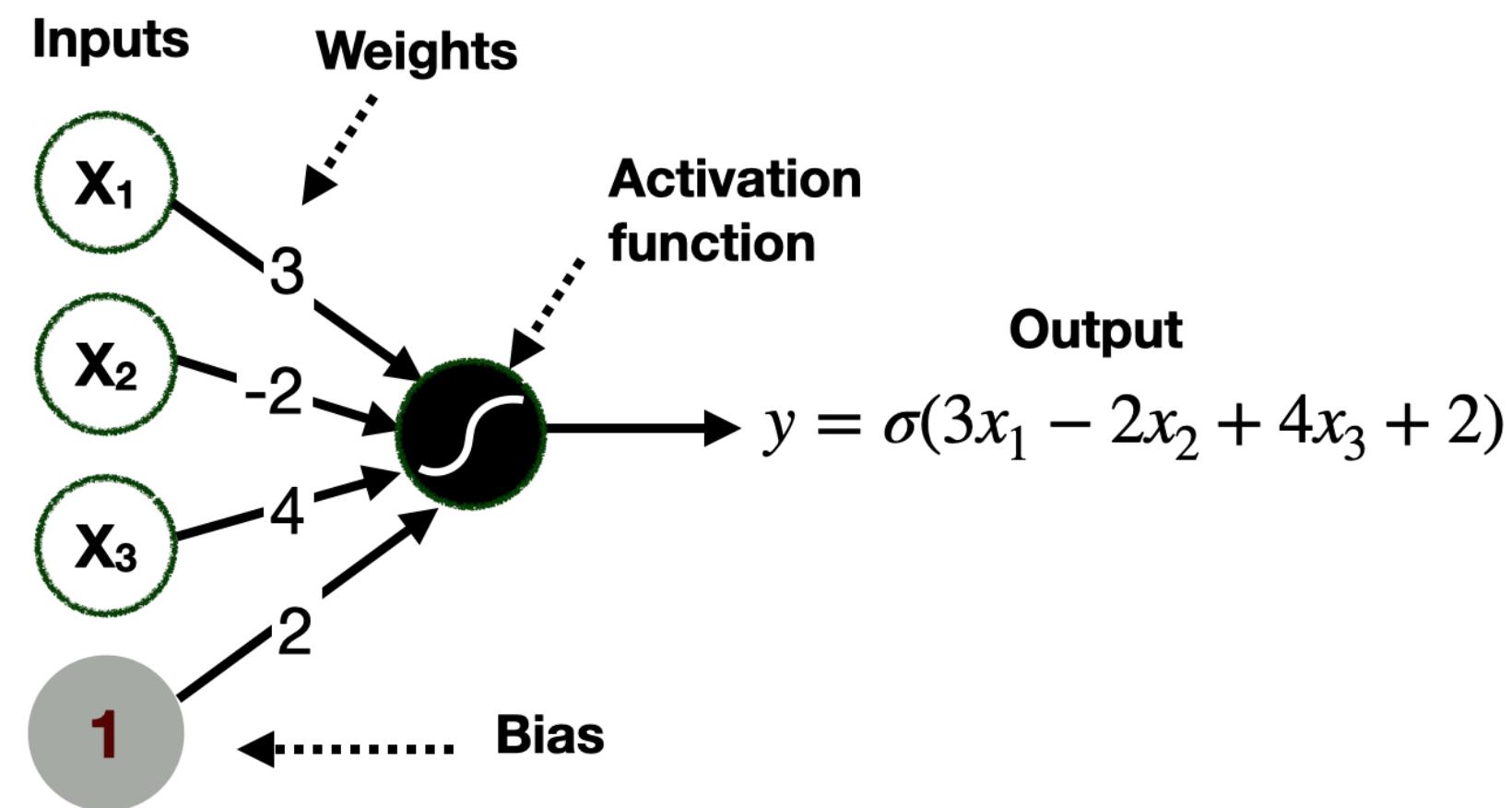






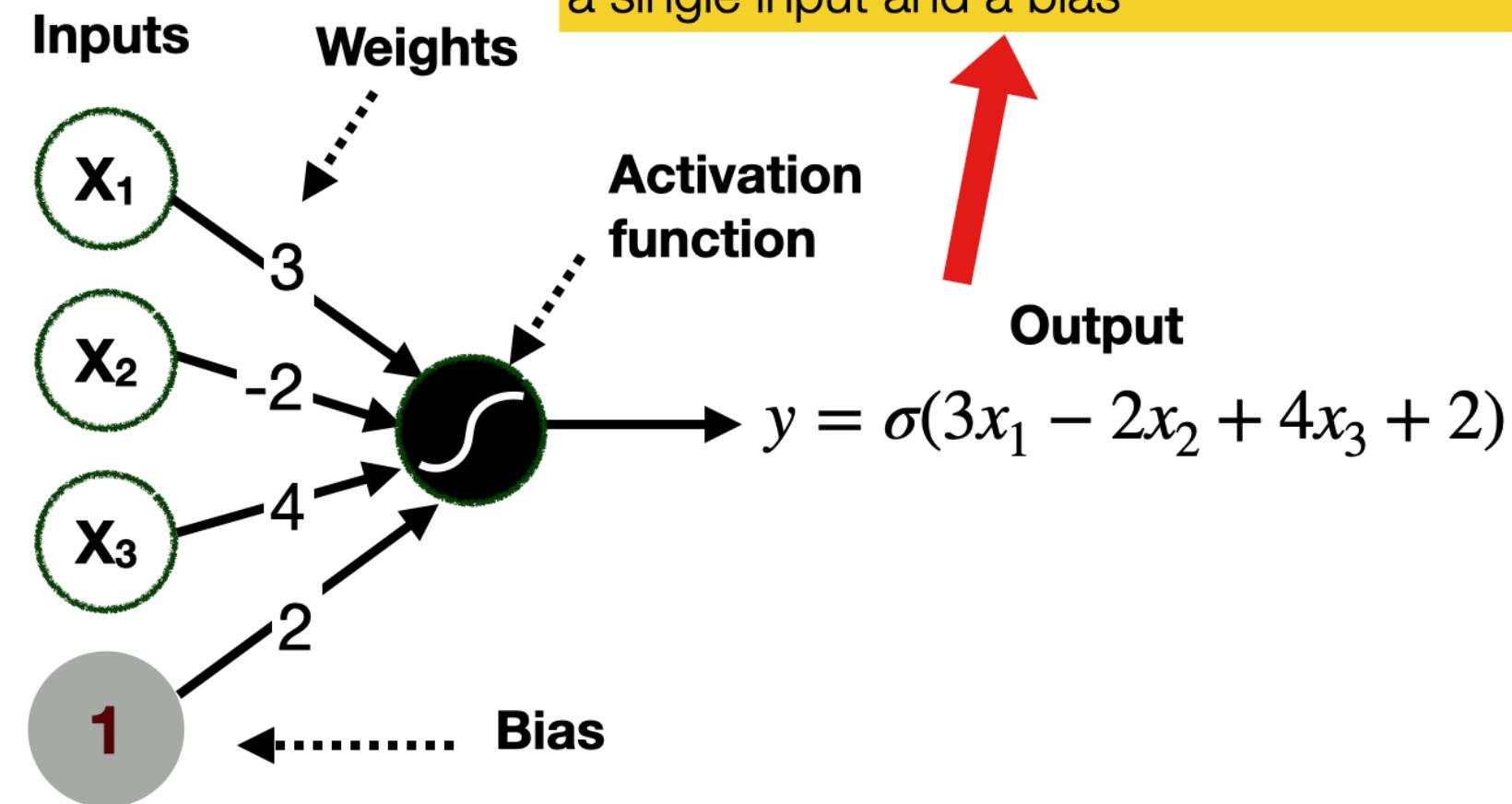






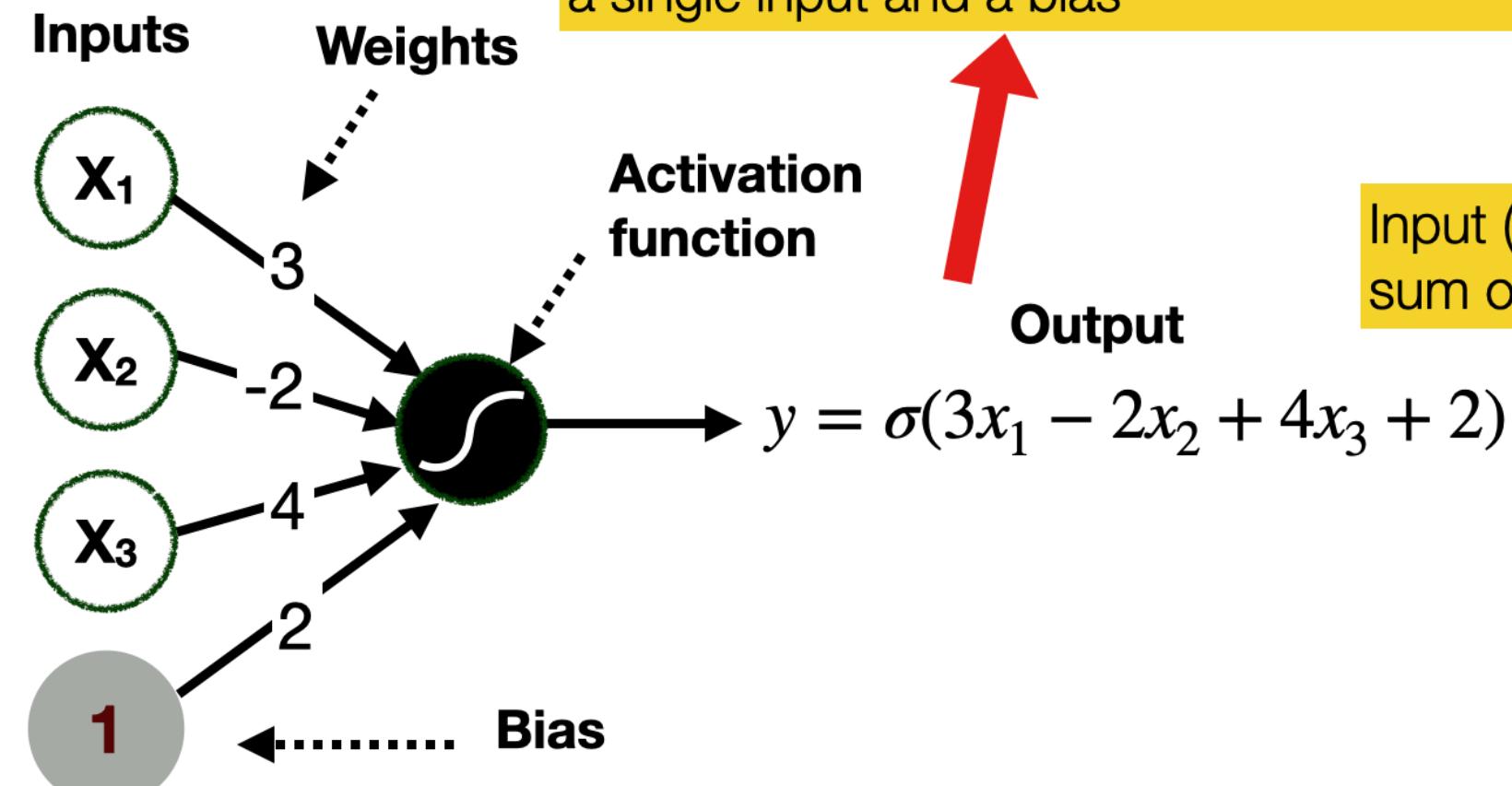
Cost = $a + b$ Size

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

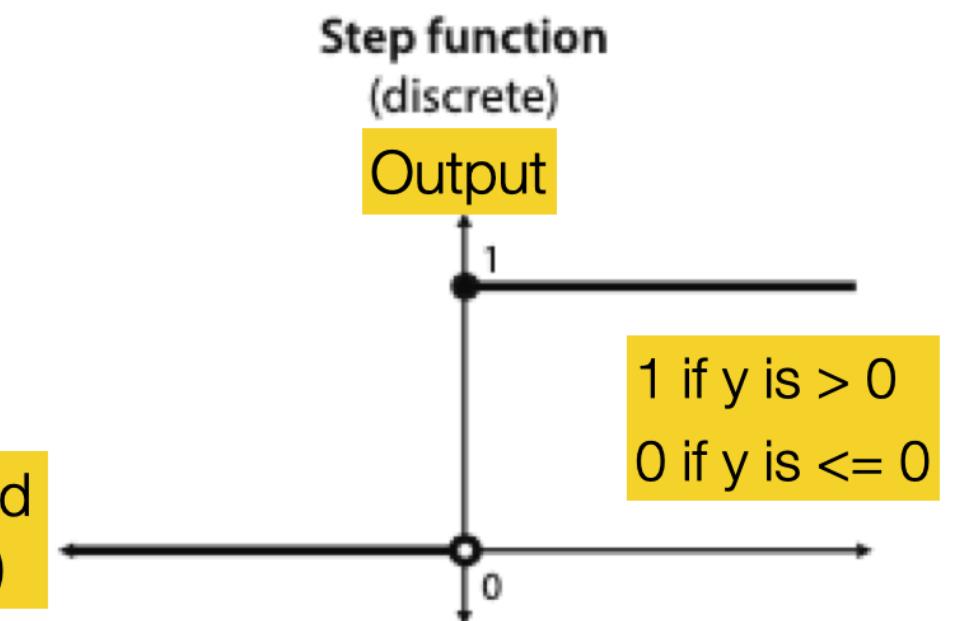


Cost = $a + b$ Size

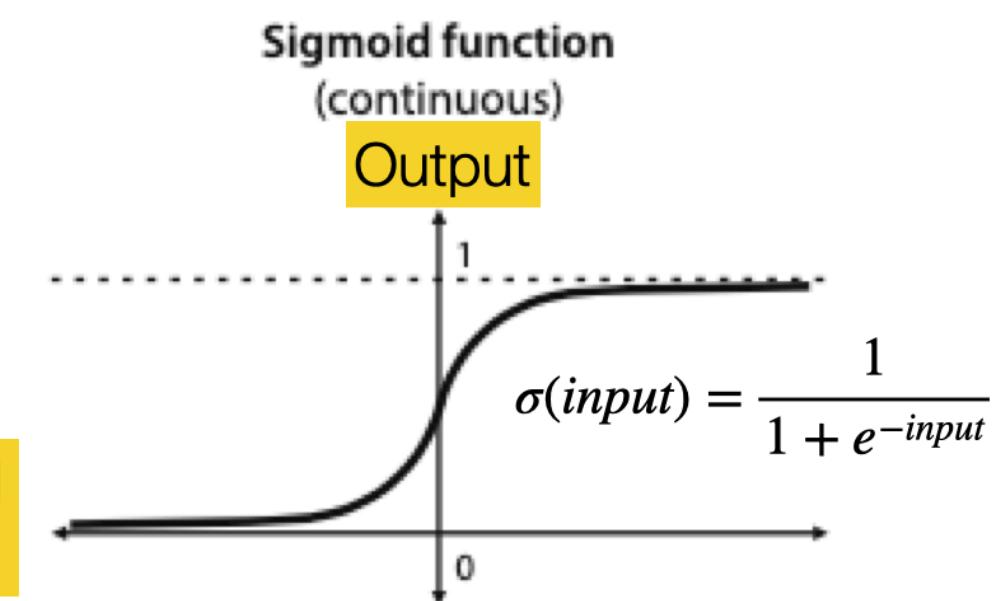
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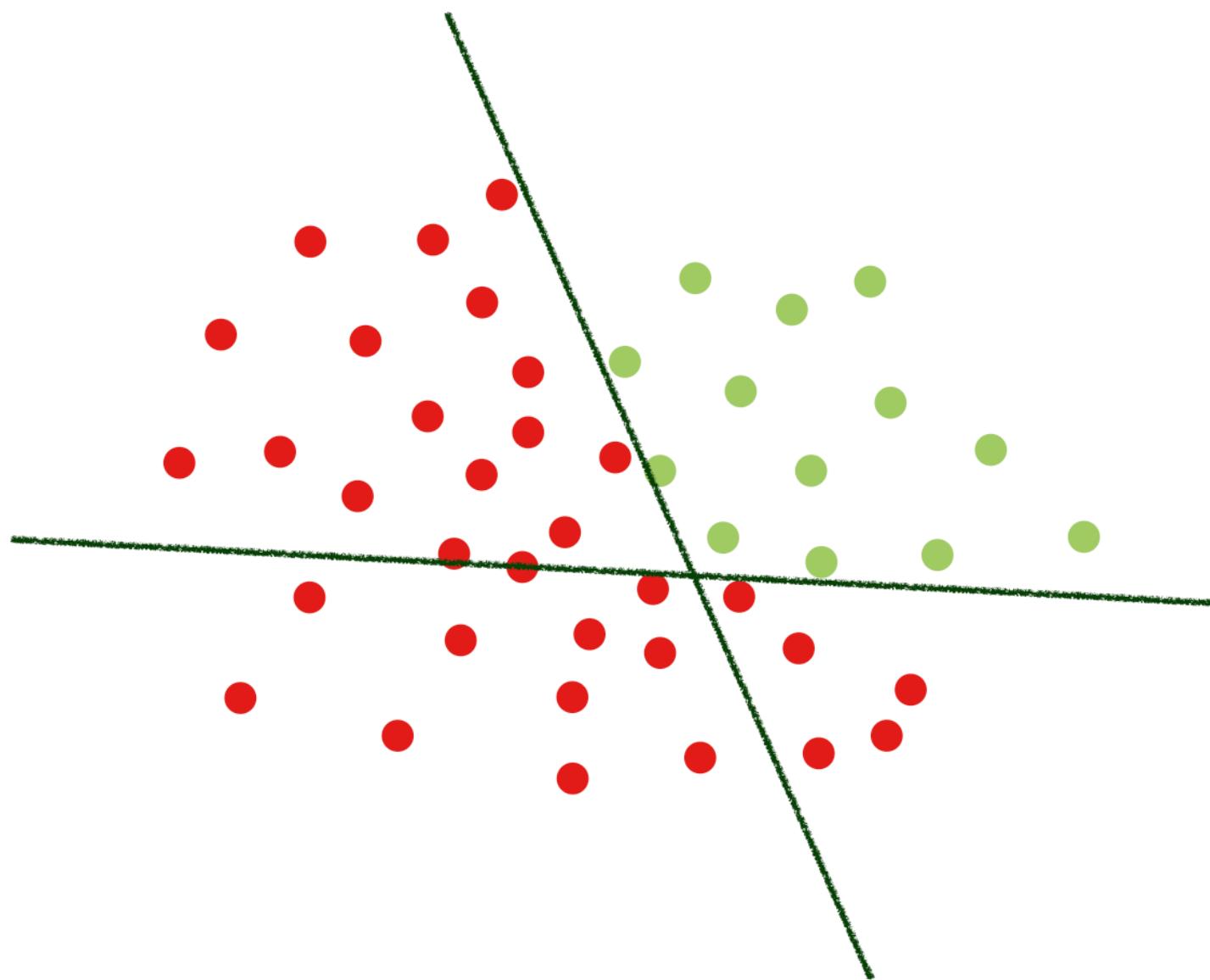


Input (weighted sum of values)

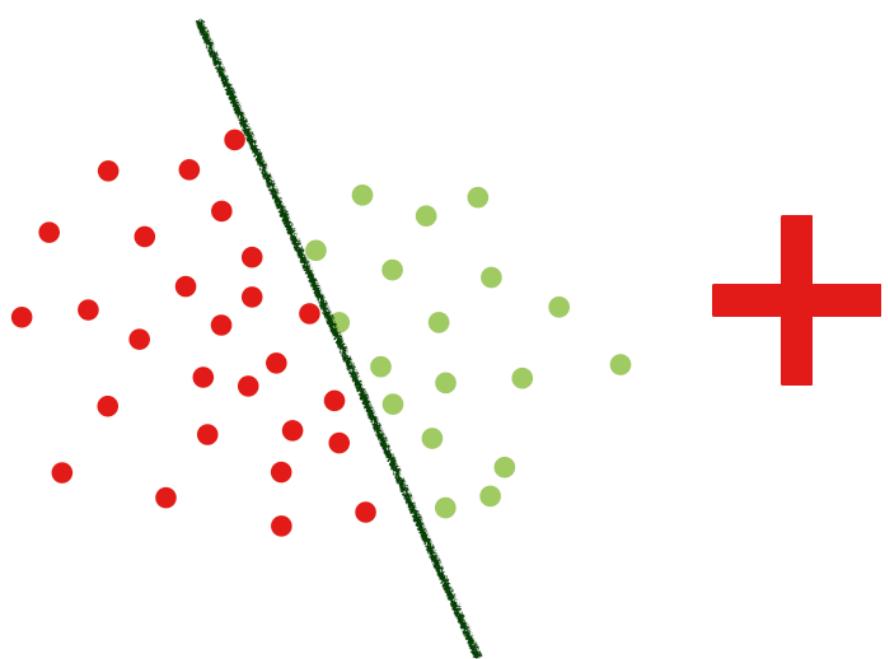


Input (weighted sum of values)

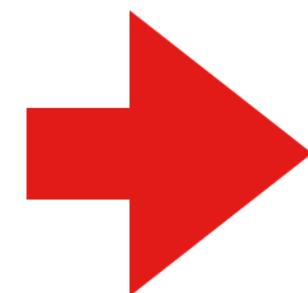
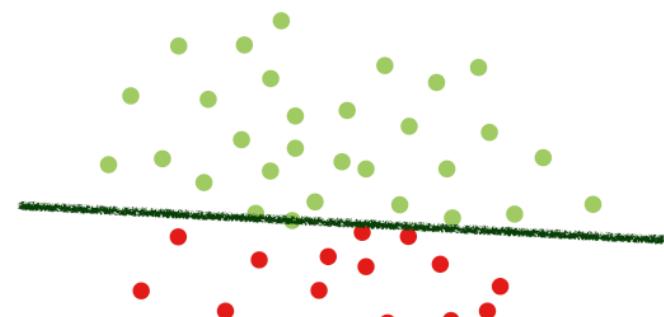




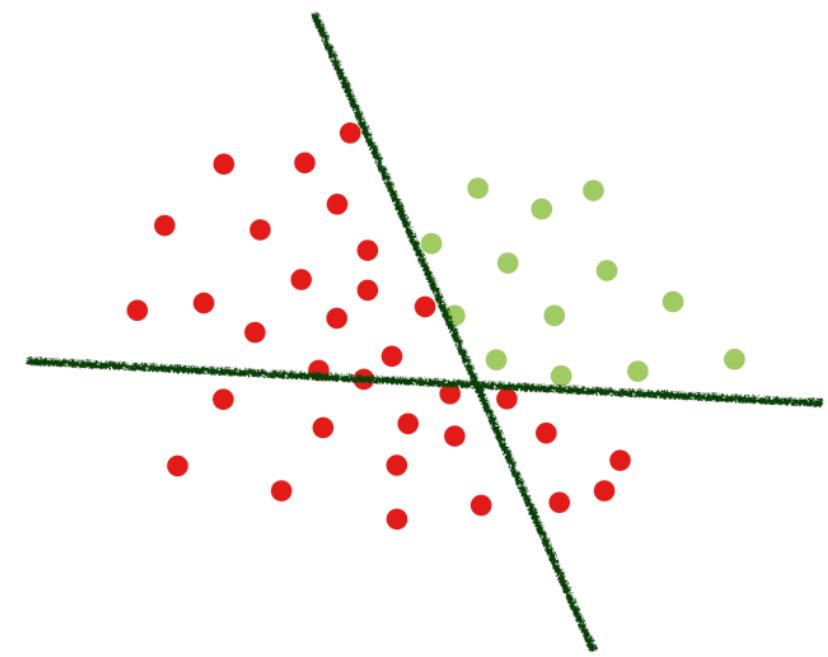
**Test Grade
Classifier**

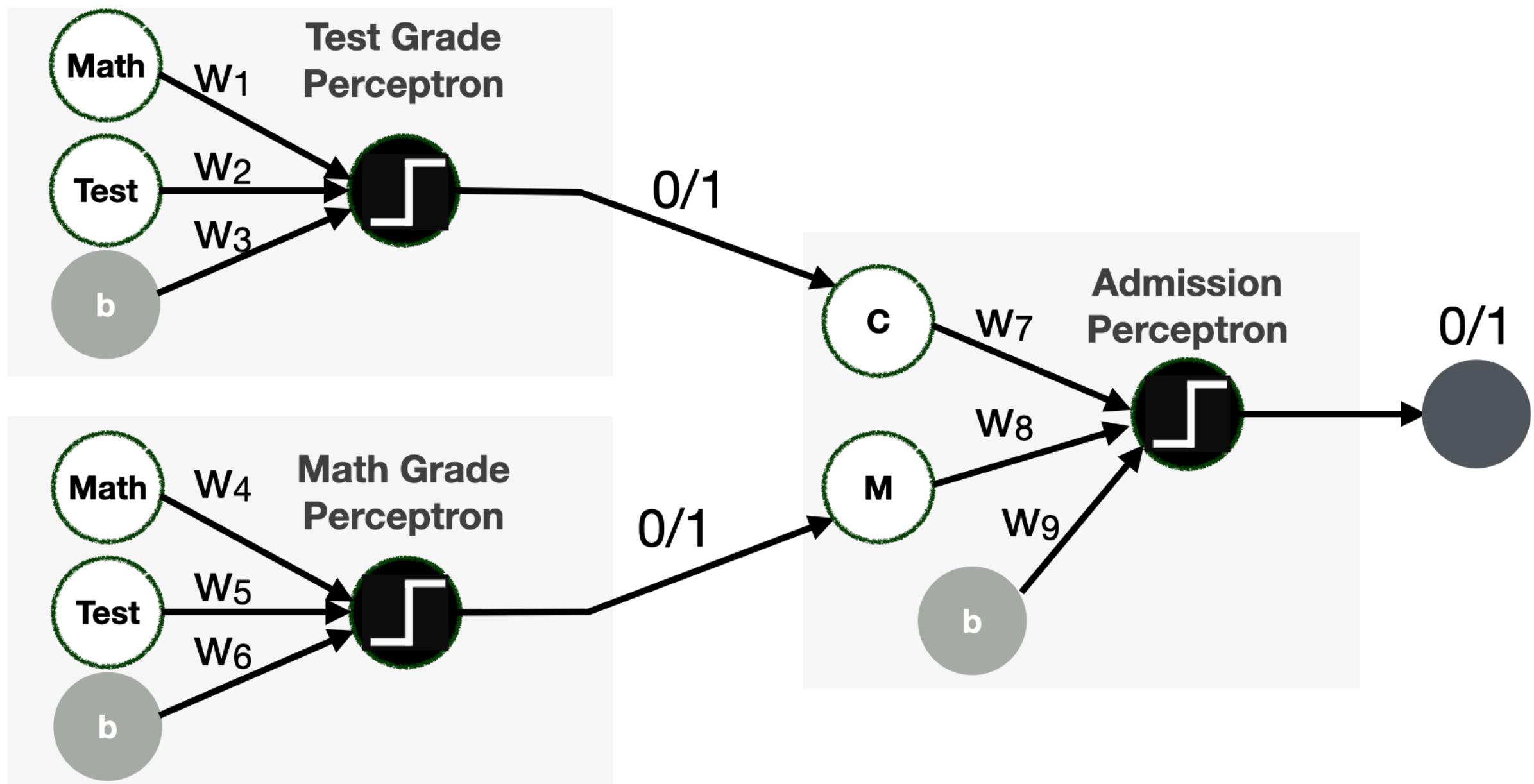


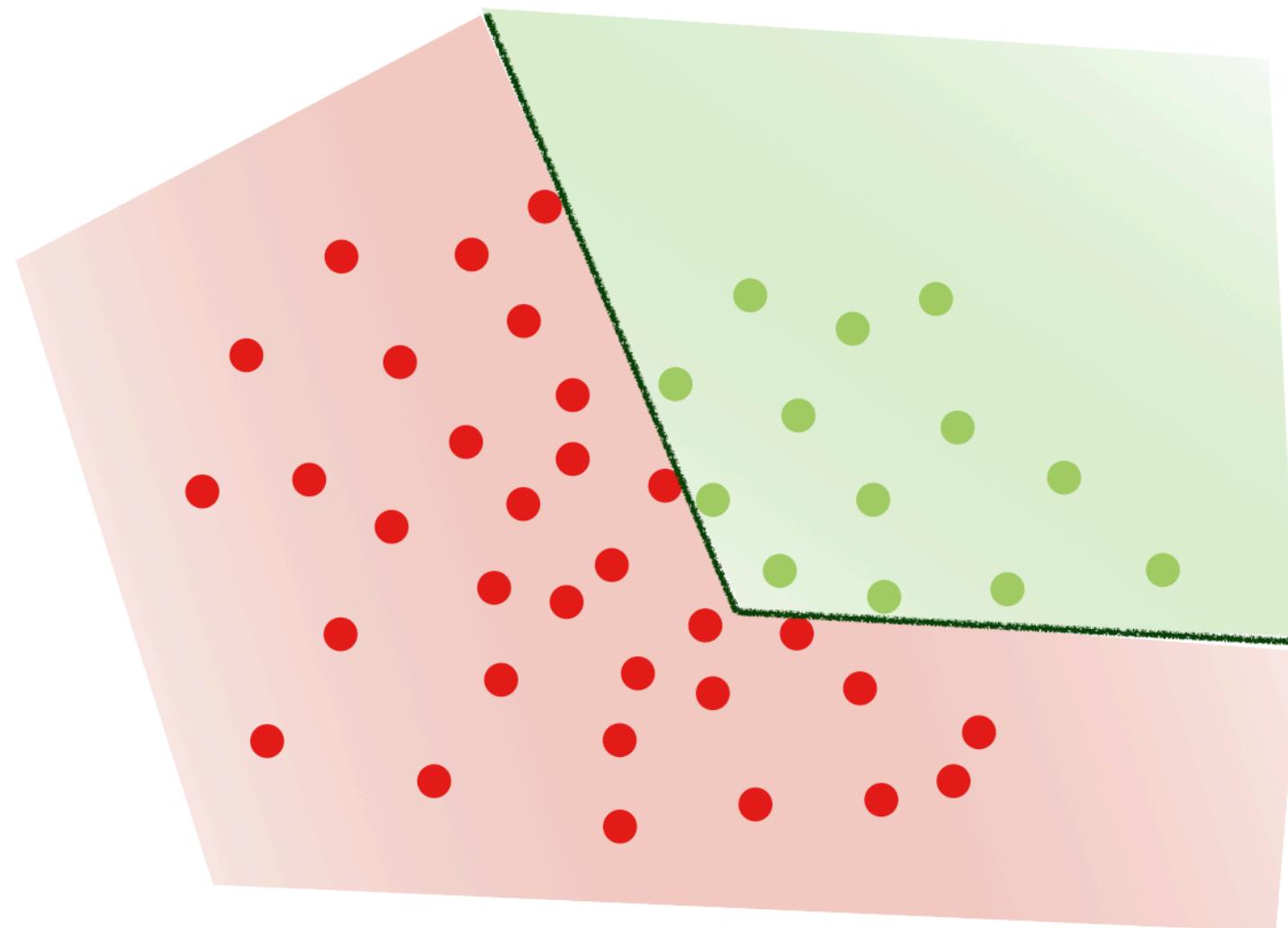
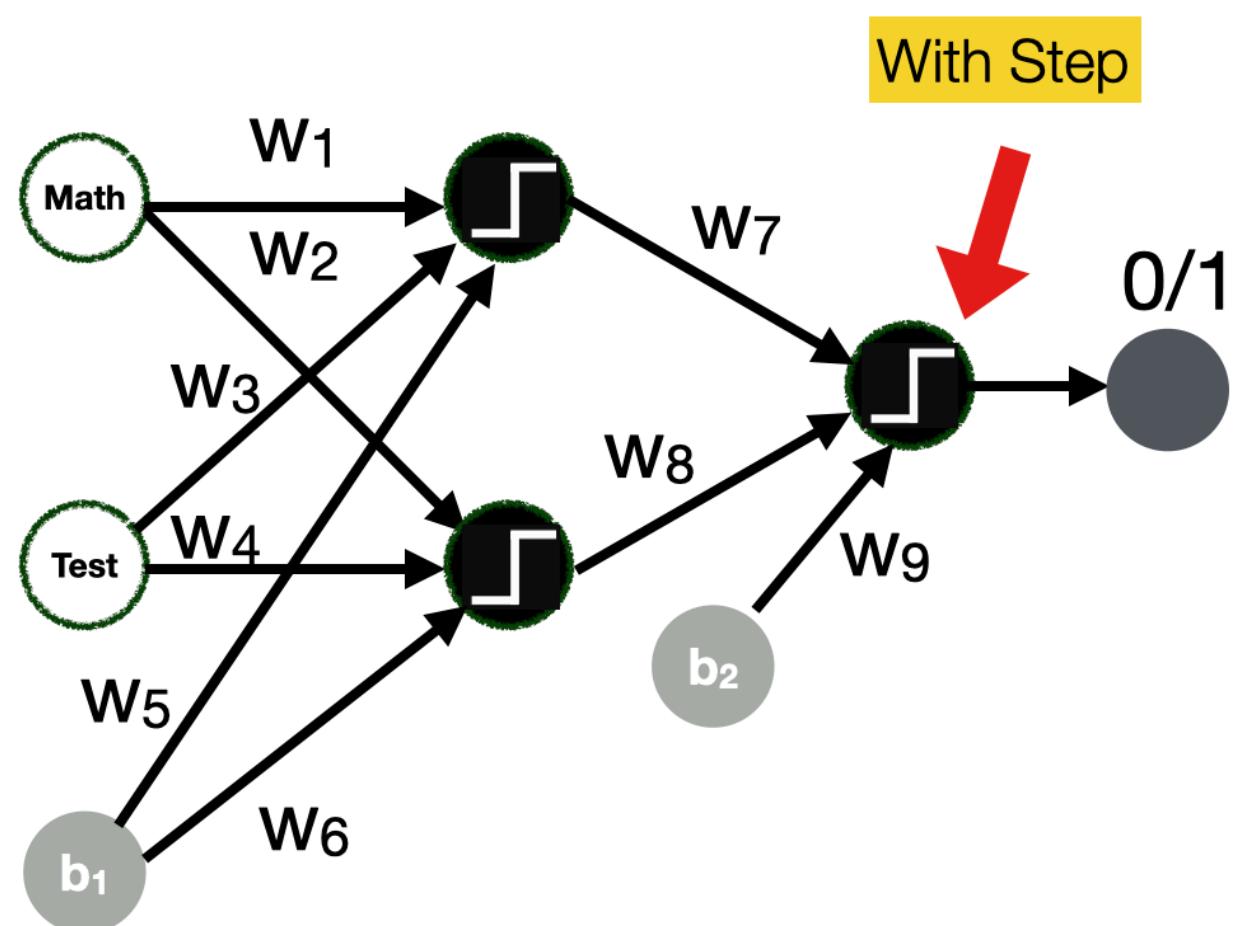
**Math Grade
Classifier**

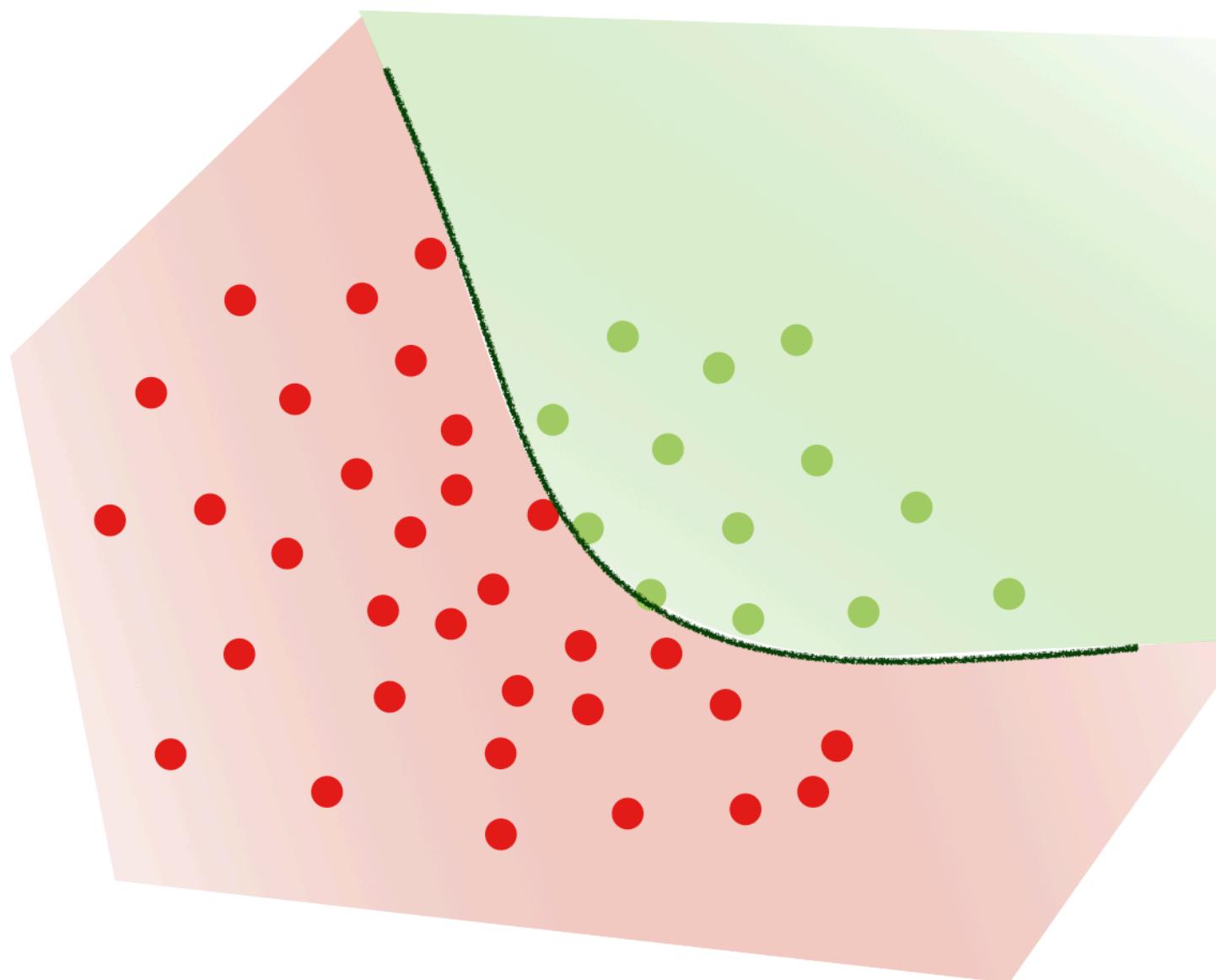
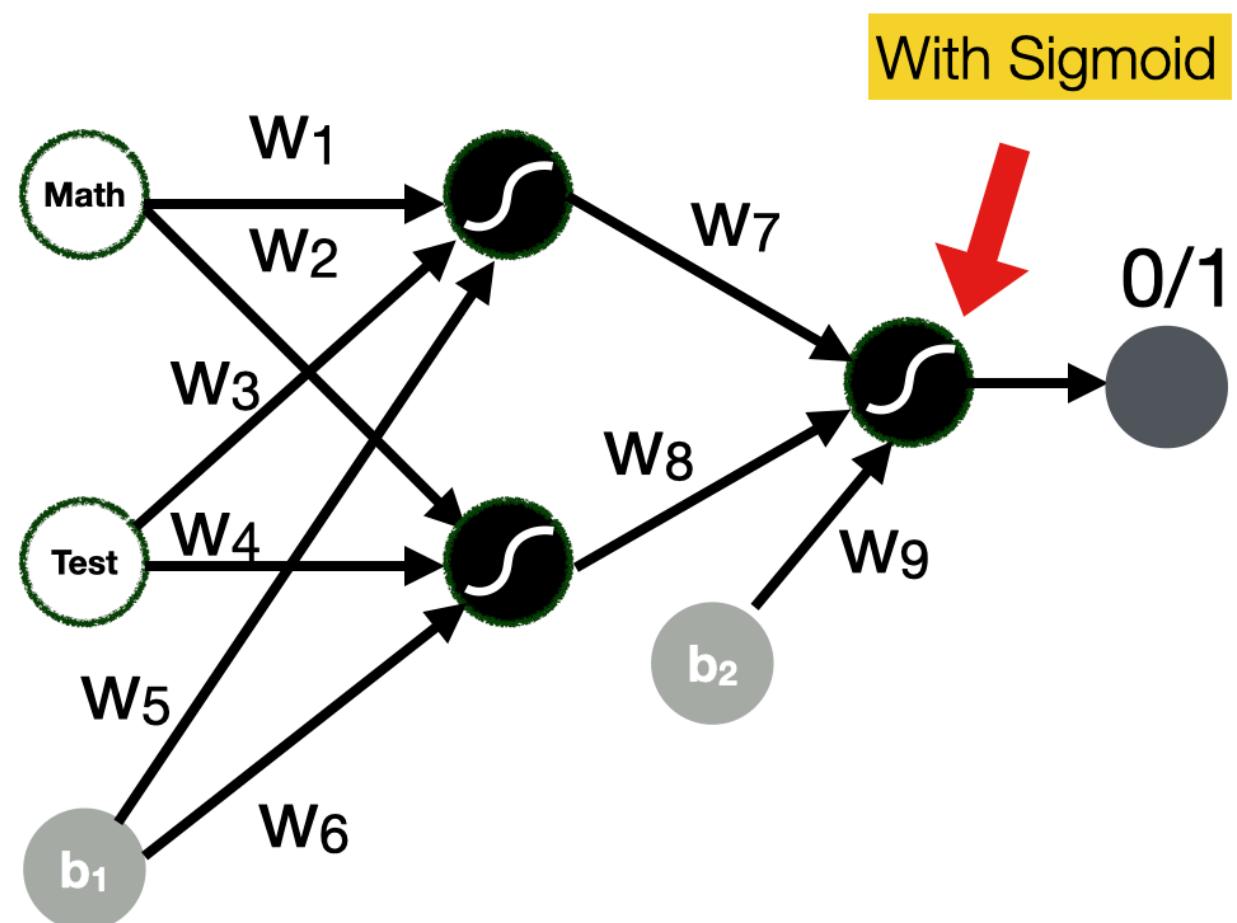


**Admission
Classifier**

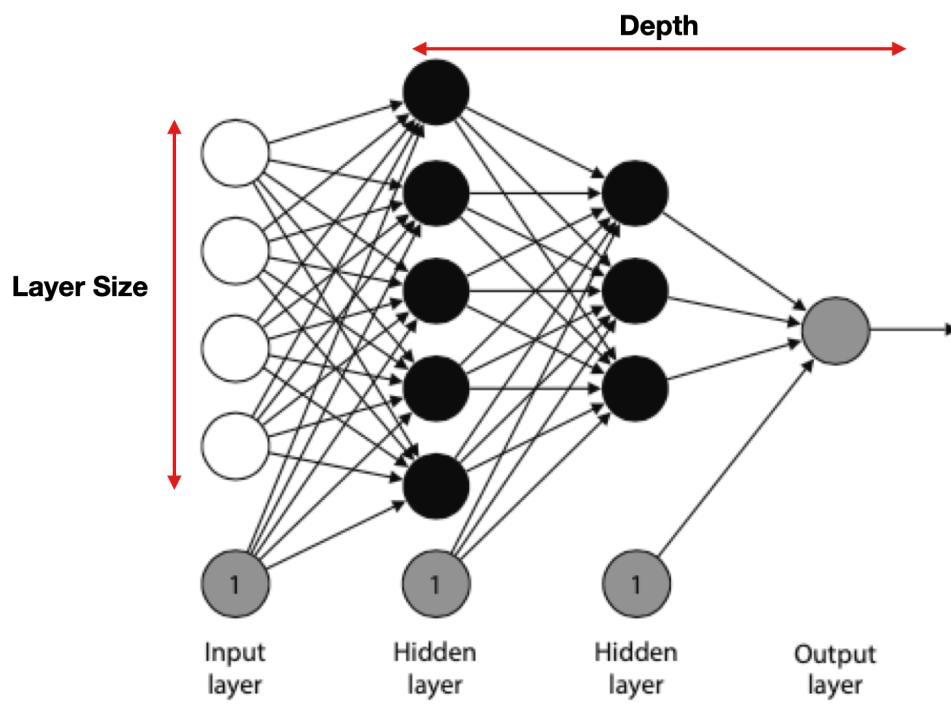






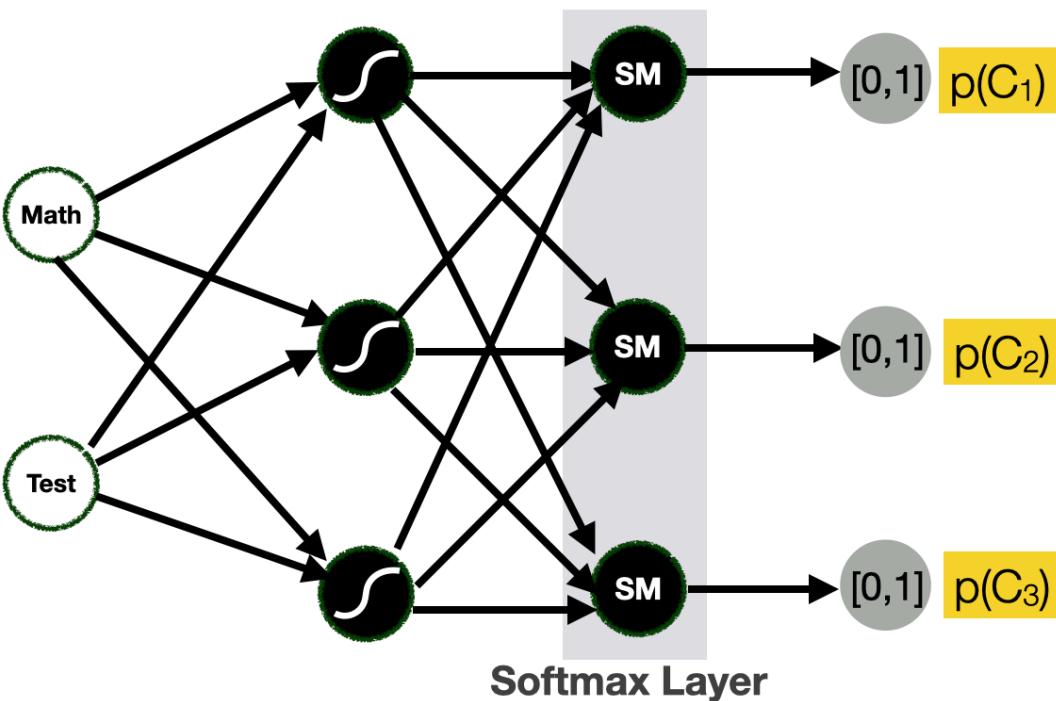


Fully connected Neural Network



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

Classifying into multiple classes - *Softmax* function



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

Value of class i

Normalisation term on K classes

- Return a probability for each class
 - example C1= ADMITTED, C2 = NOT ADMITTED, C3 = 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

Tensorflow Playground

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

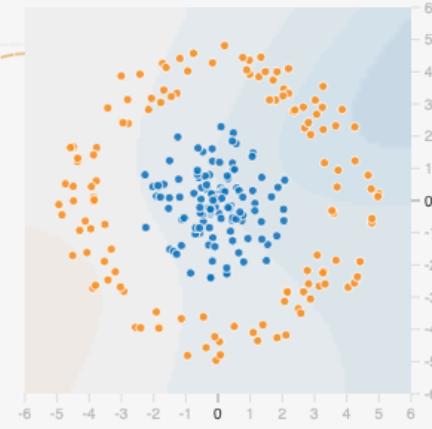
Epoch 000,000 Learning rate 0.03 Activation Tanh Regularization None Regularization rate 0 Problem type Classification

DATA
Which dataset do you want to use?

Ratio of training to test data: 50%
Noise: 0
Batch size: 10
REGENERATE

FEATURES
Which properties do you want to feed in?
 x_1 , x_2 , x_1^2 , x_2^2 , x_1x_2 , $\sin(x_1)$, $\sin(x_2)$

2 HIDDEN LAYERS
4 neurons 2 neurons
This is the output from one neuron. Hover to see it larger.
The outputs are mixed with varying weights, shown by the thickness of the lines.

OUTPUT
Test loss 0.497
Training loss 0.502

Colors show data, neuron and weight values.
 Show test data Discretize output

Machine Learning and Images

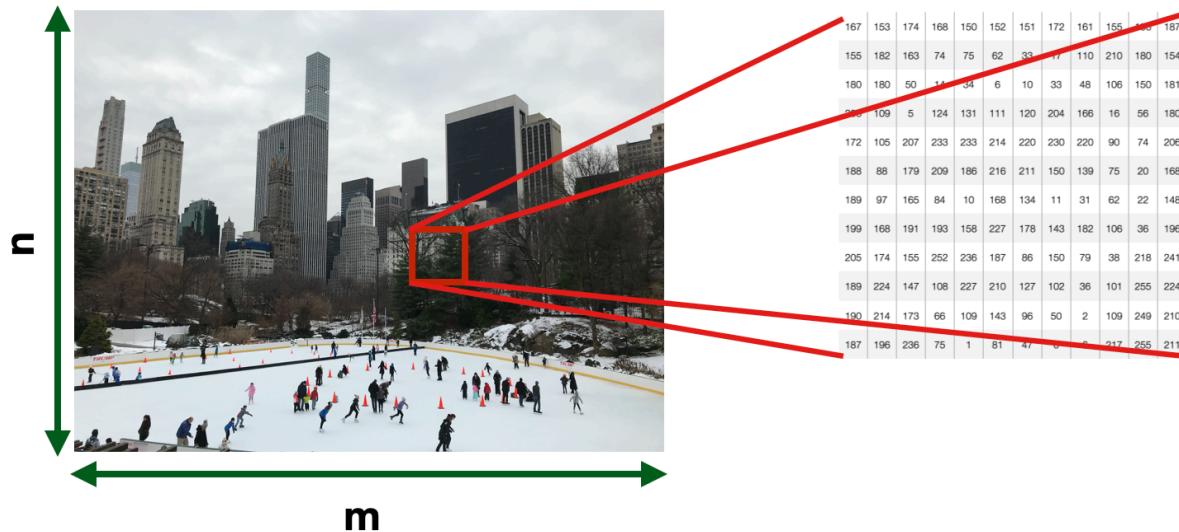
What do you see?



This is what a computer “sees”

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	155	182	163	74	75	62	33	17	110	210	180	154	180	180	50	14	34	6	10	33	17	110	210	180	154							
180	180	50	14	34	6	10	33	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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187	196	236	75	1	81	47	0	6	217	255	211	187	196	236	75	1	81	47	0	6	217	255	211	187	196	236	75	1	81	47	0	6	217														

Images



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

Computer Vision

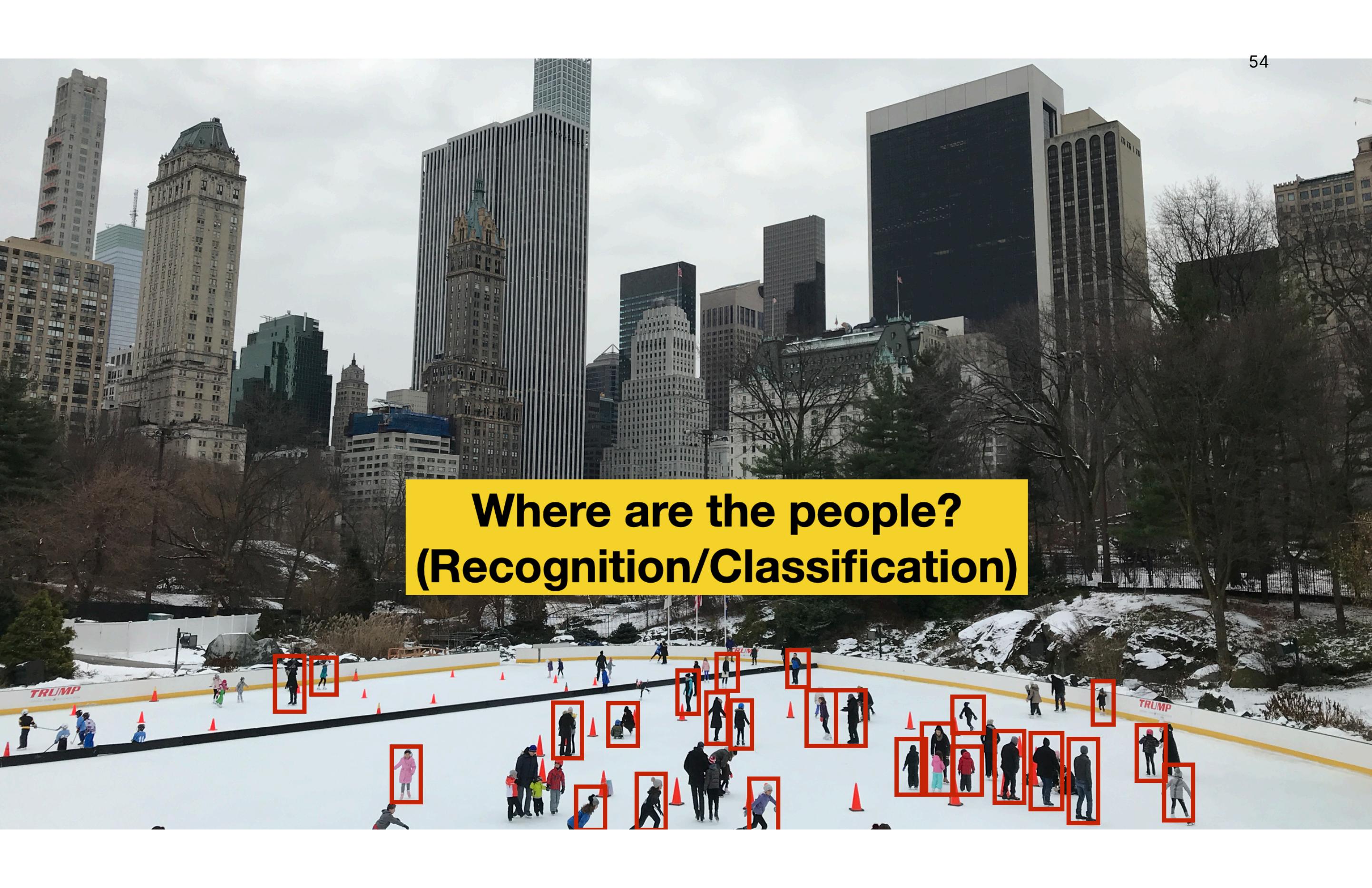
- Building algorithms that can “understand” the content of images and use it for other applications
- It is a “Strong AI” problem
 - signal-to-symbol conversion
 - The **semantic gap**
- A general-purpose vision system **requires**
 - Flexible, robust visual representation
 - Updated and maintained
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

**What specific tasks can
we train a CV system to
perform?**

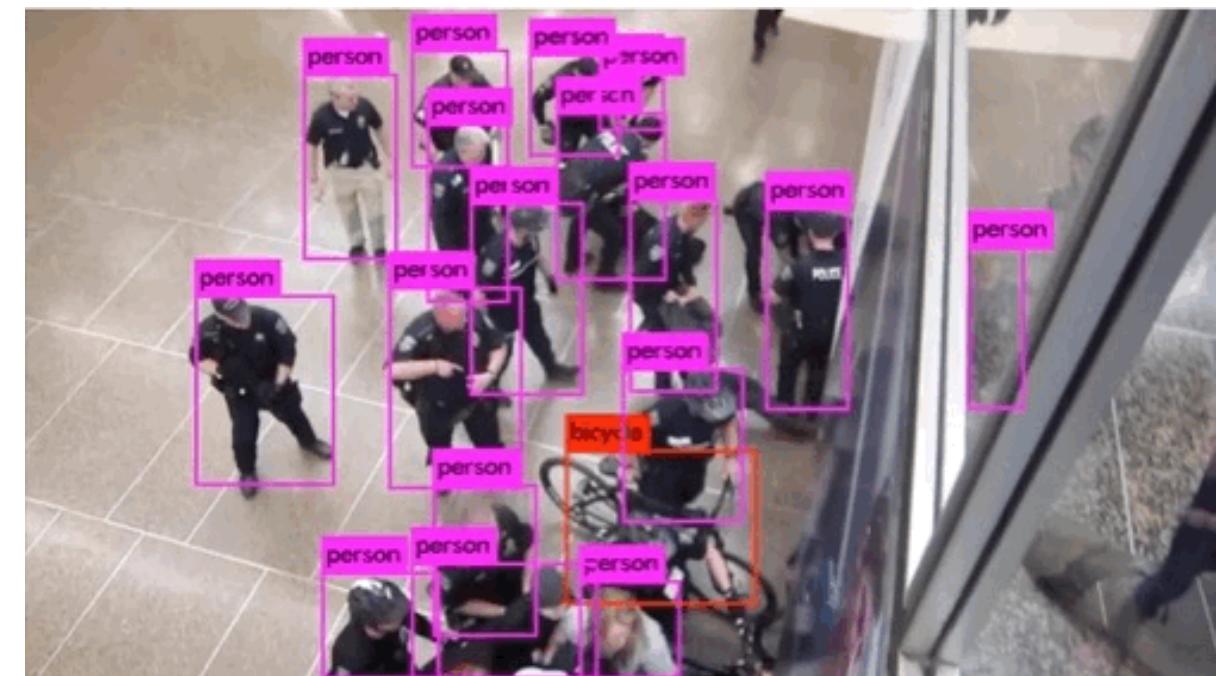
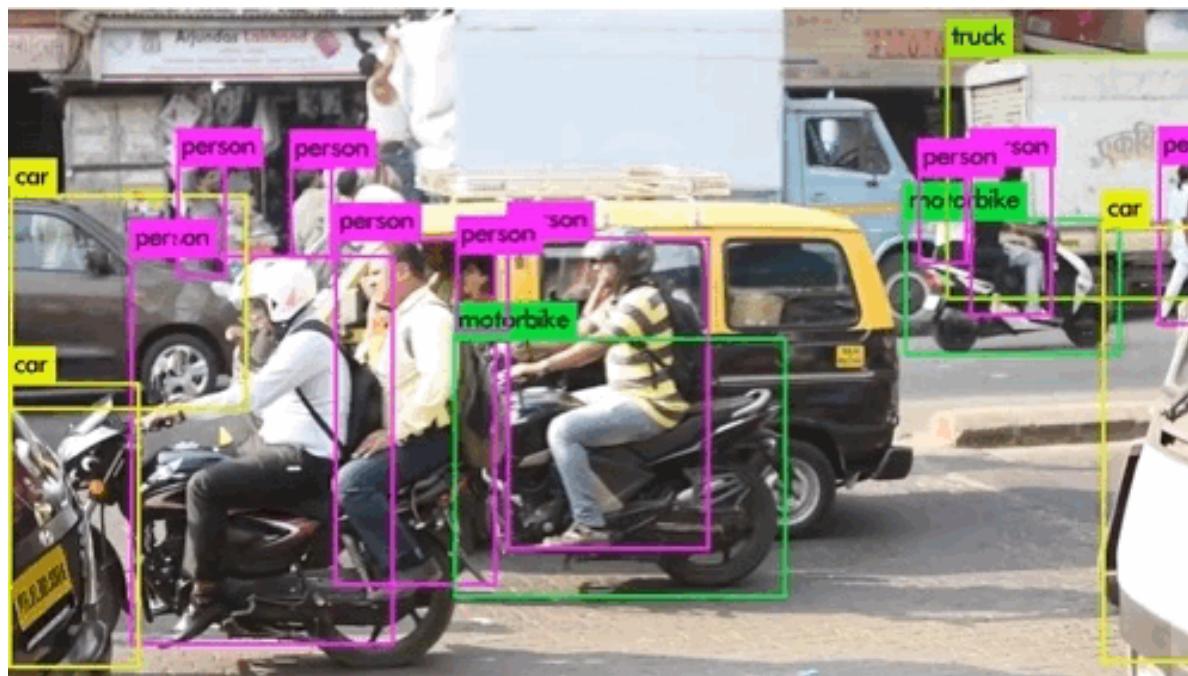
A wide-angle photograph of an outdoor ice skating rink in a park. In the foreground, many people are skating on the white ice. Orange traffic cones are scattered across the rink, some marking paths or boundaries. The rink is surrounded by a fence. In the background, a dense city skyline with numerous skyscrapers rises against a cloudy sky. Bare trees are visible in the middle ground between the rink and the buildings.

**Is this a flag?
(Recognition / Classification)**



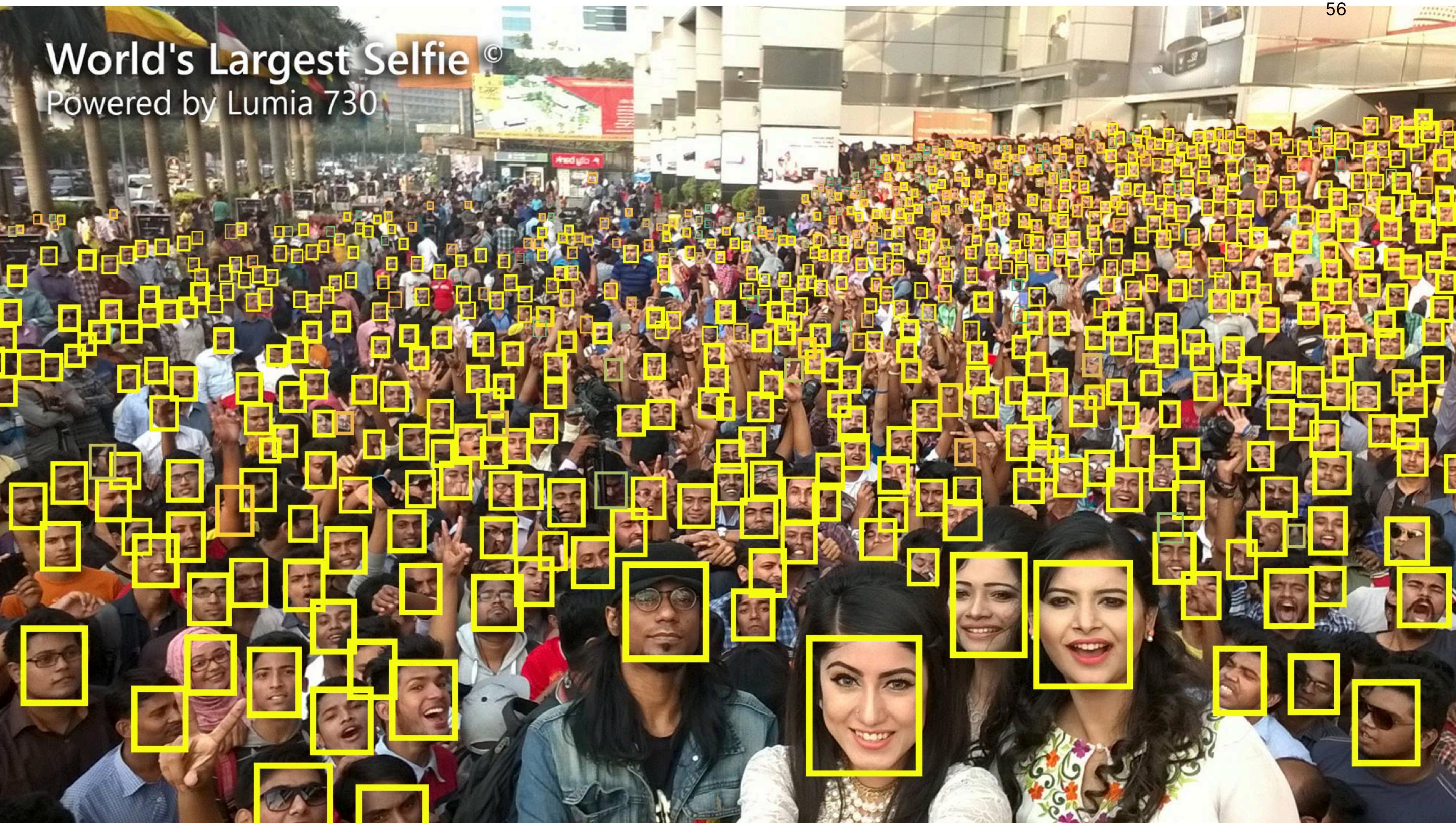


**Where are the people?
(Recognition/Classification)**

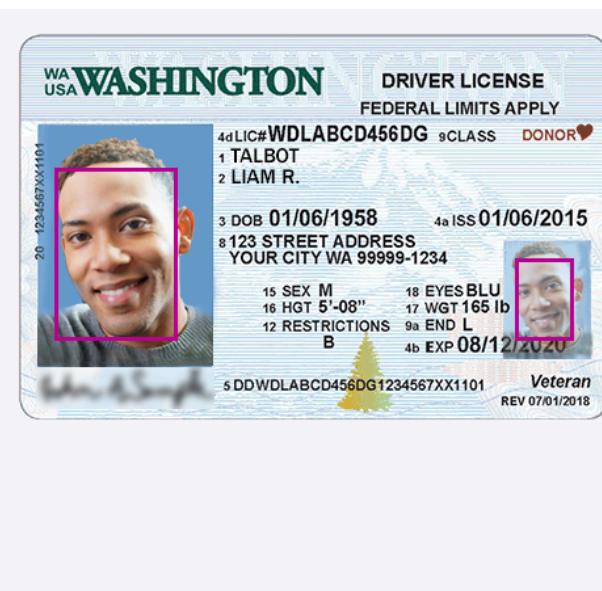
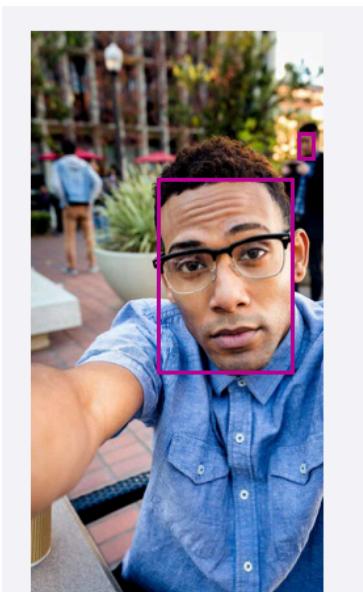


World's Largest Selfie ©

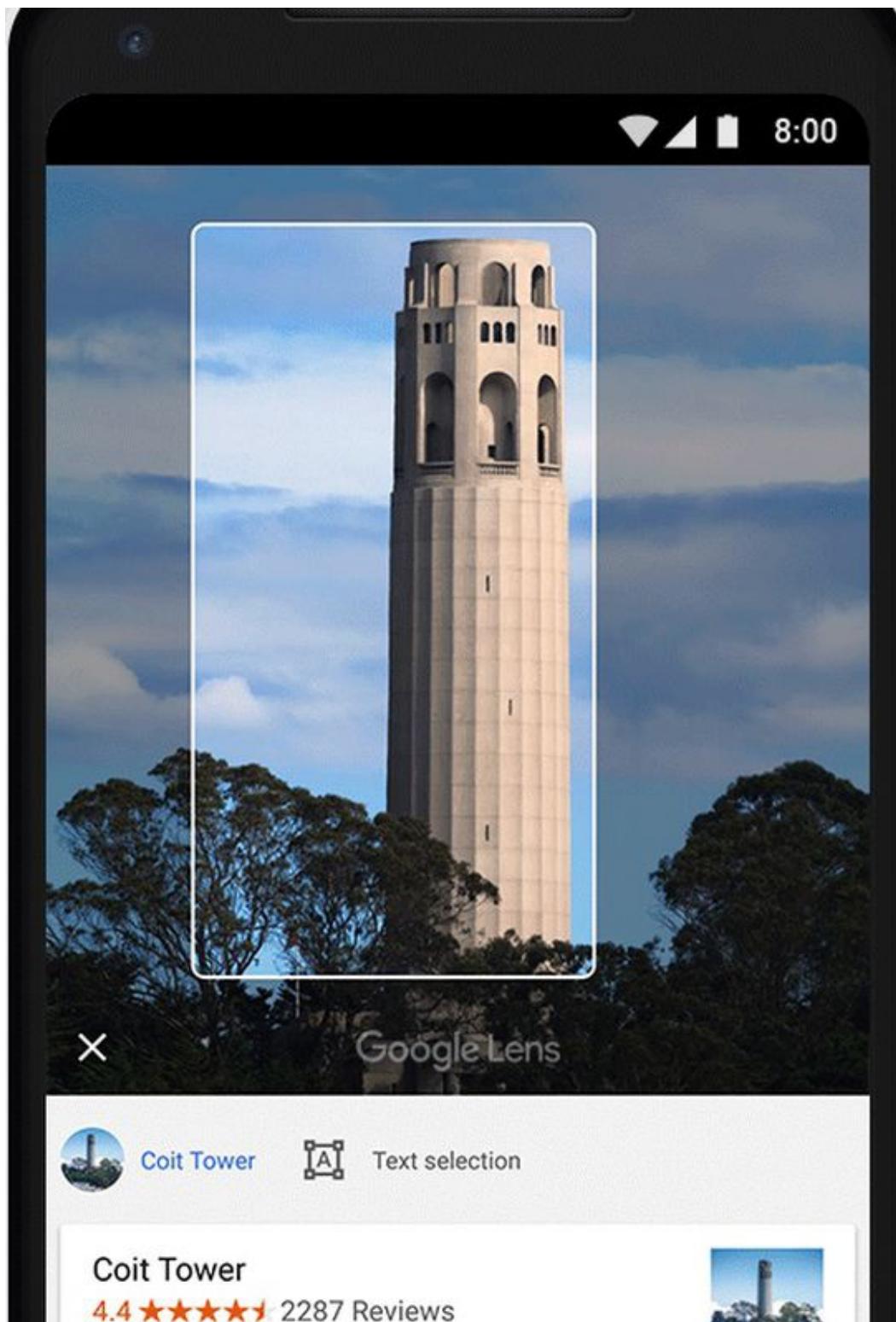
Powered by Lumia 730

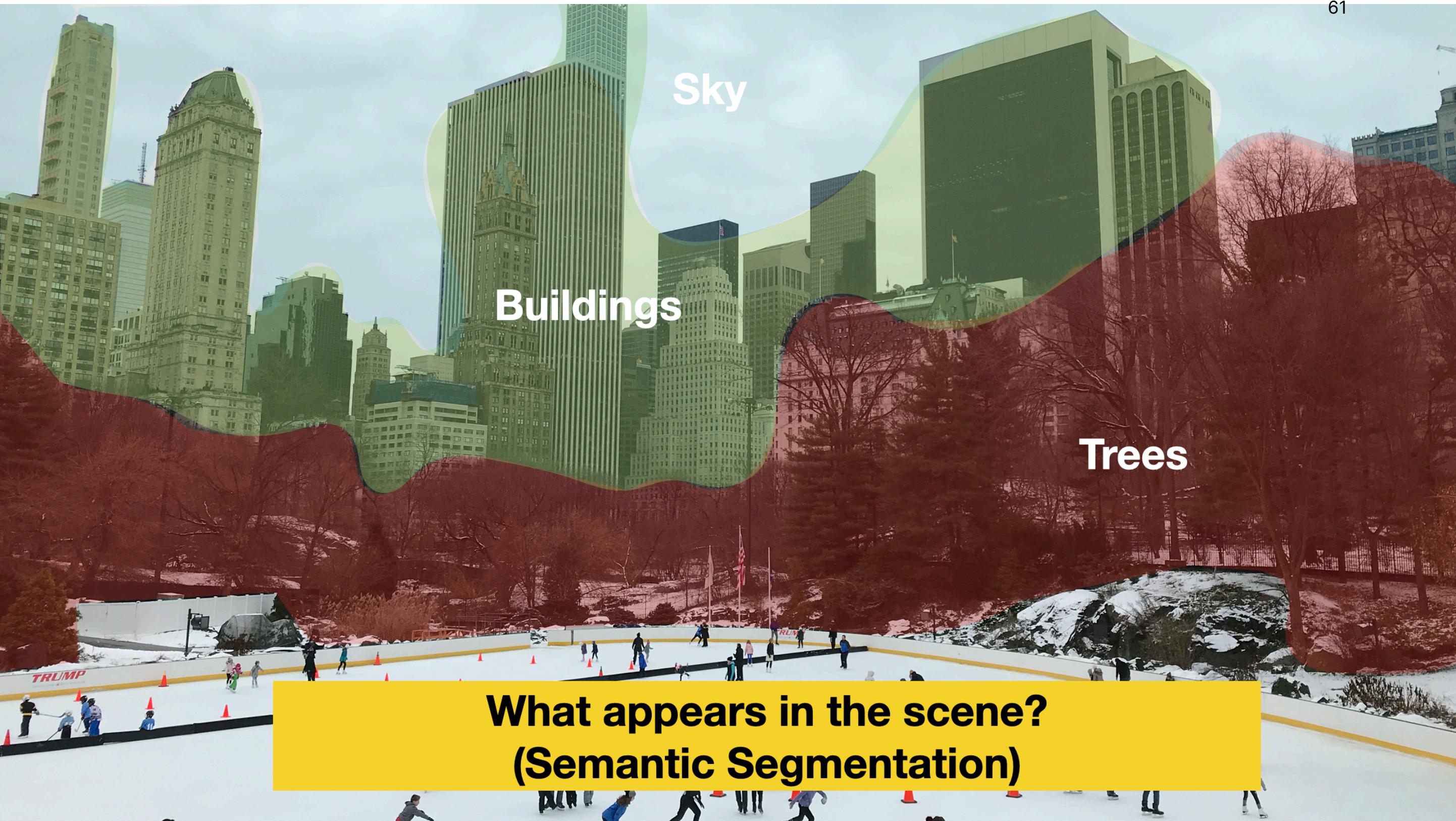


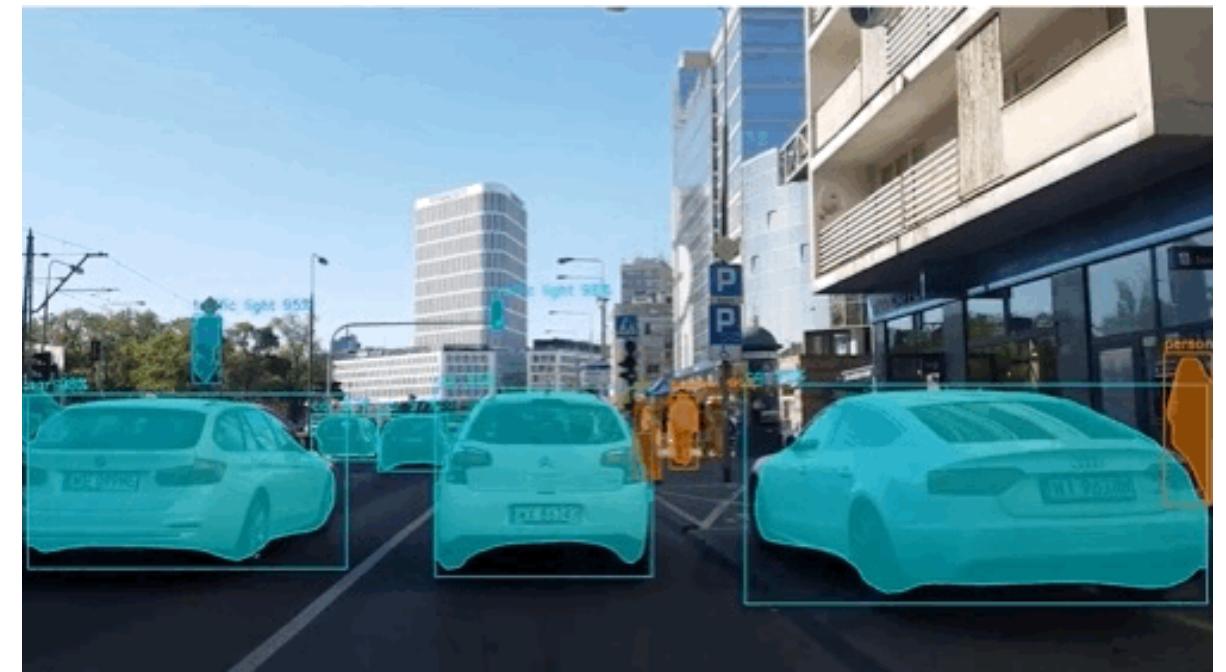
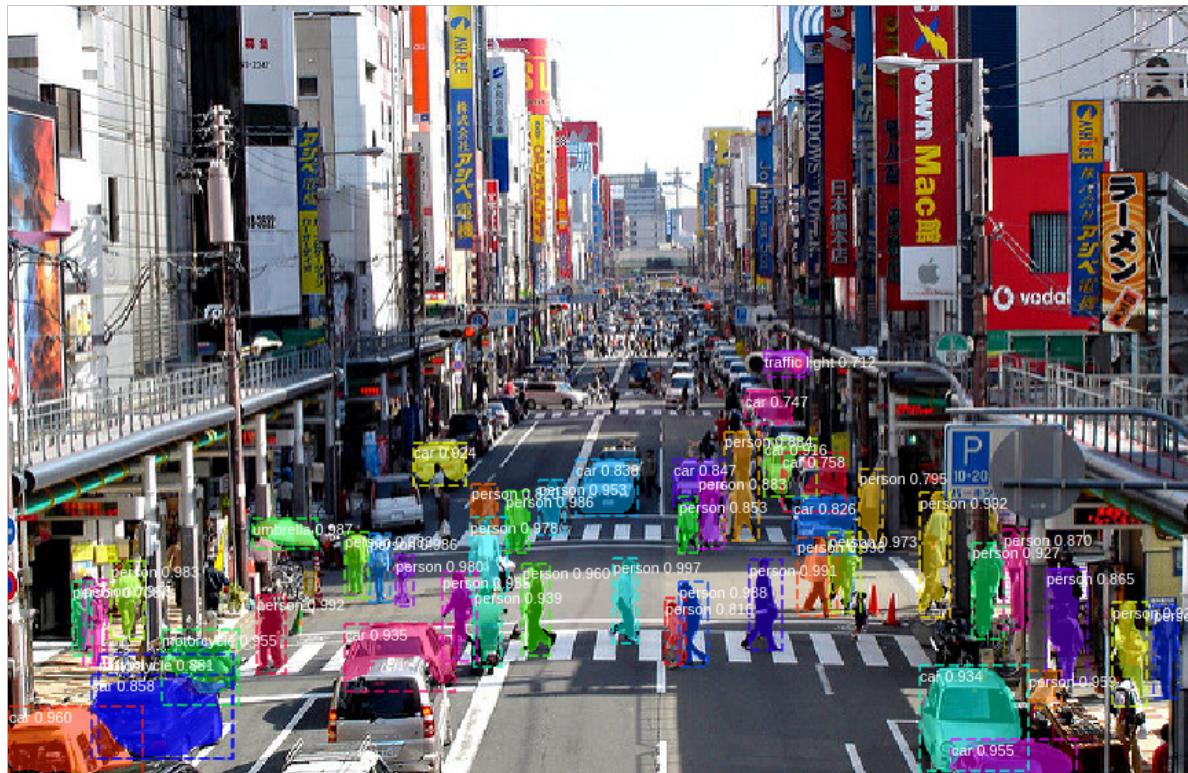












Project Sunroof

≡ Google Project Sunroof

22314 Cupertino Rd, Cupertino, CA 95014, USA GO

Analysis complete. Your roof has:

- 1,910 hours of usable sunlight per year
Based on day-to-day analysis of weather patterns
- 863 sq feet available for solar panels
Based on 3D modeling of your roof and nearby trees

\$13,000 savings
Estimated net savings for your roof over 20 years

Wrong building? Click another roof to view details.

Sunny

Shady

Map data ©2019 Google | Terms of Use

Savings estimator Data explorer Solar 101 FAQ



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

Outdoor, City, Park



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:**



Hugging Face

Models Datasets Spaces Docs Solutions Pricing

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

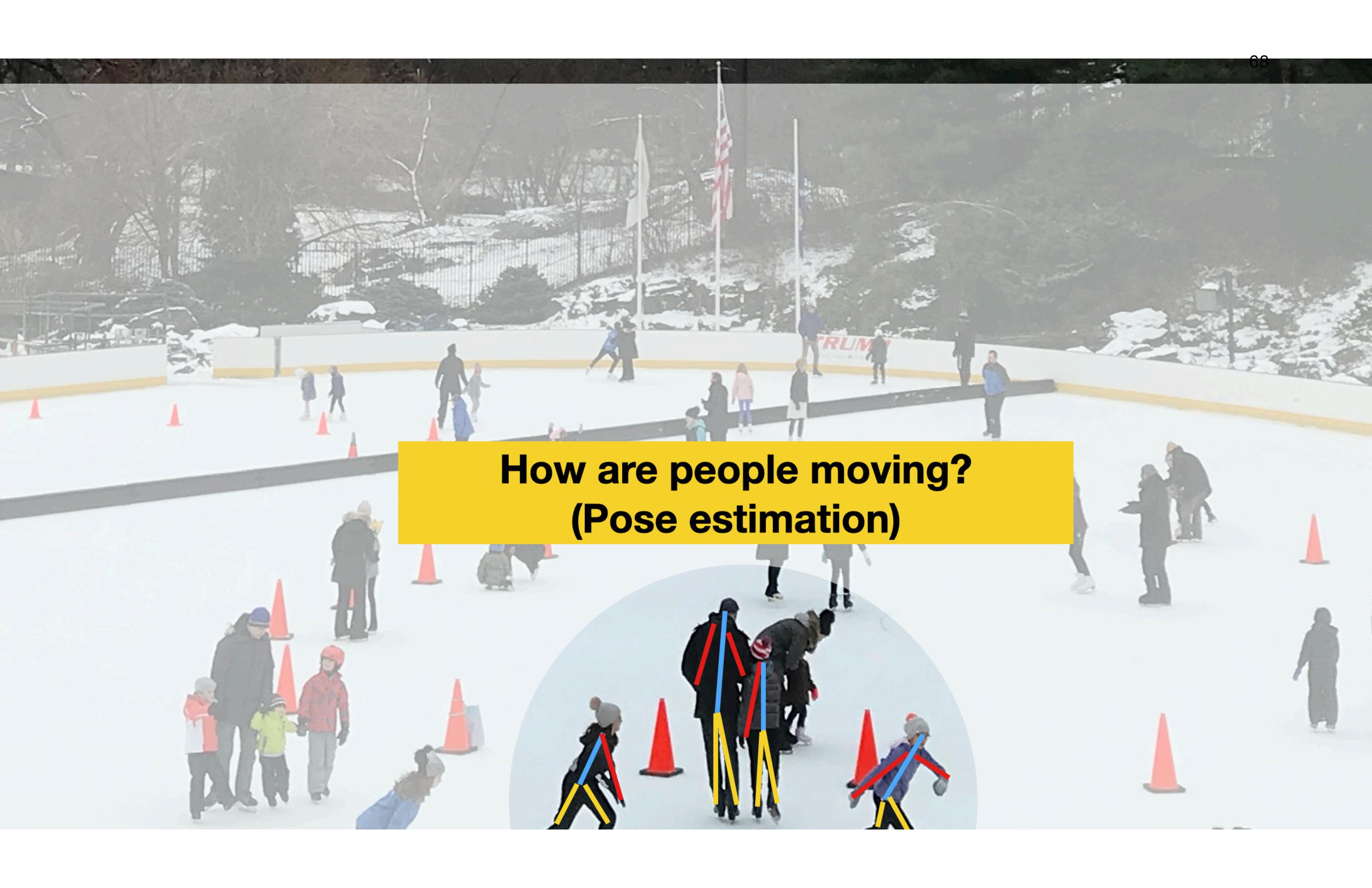
people skating on a rink in a city park with skyscrapers

Clear Submit



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

Skating



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

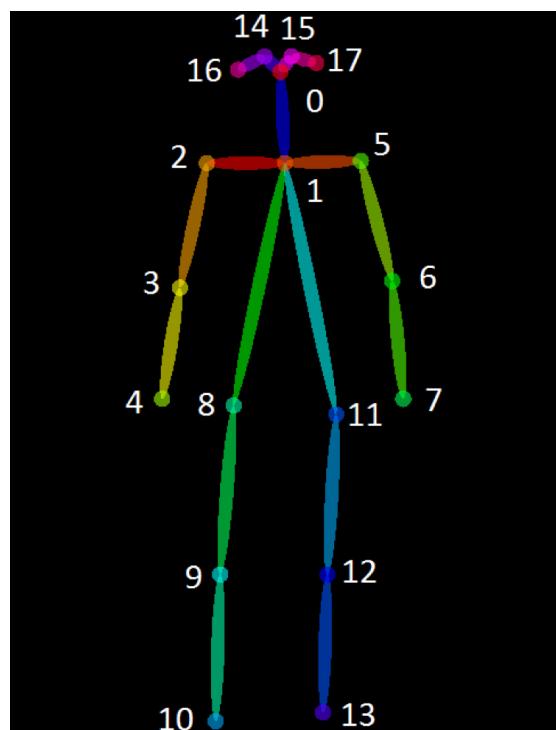


Stereolabs ZED Camera

3D Object Detection

Body tracking

Positional tracking



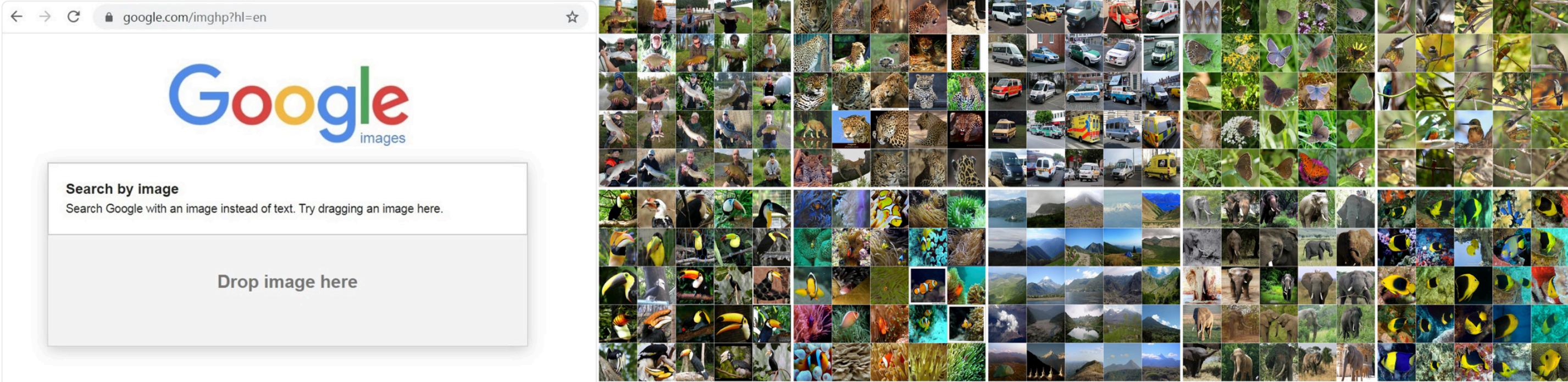


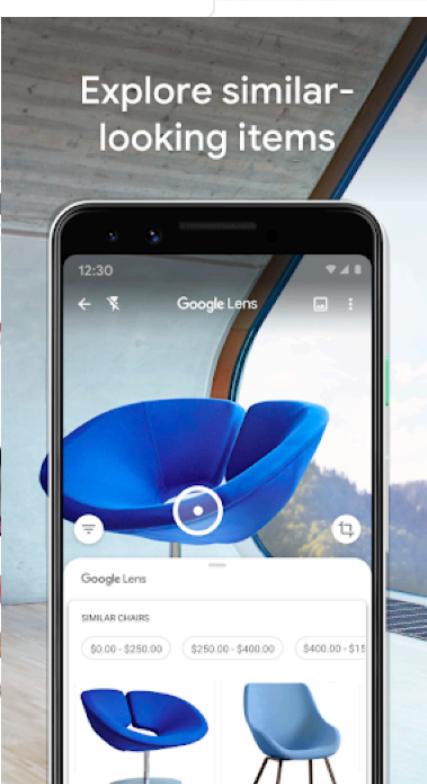
Image search results

88432.jpg Lamp

Visually similar images

Report images

This section shows the results of a visual search for a lamp. It displays a grid of numerous red desk lamps with black flexible necks and shades. The search interface includes a file name "88432.jpg" and a category "Lamp". A "Report images" button is located at the bottom.



Machine Learning for Design

Lecture 3

Machine Learning for Images. *Part 1*

Credits

CMU Computer Vision course - Matthew O'Toole.

Grokking Machine Learning. Luis G. Serrano. Manning, 2021