

# Demystifying Machine Learning

*Uncloaking the Math in the Black Box*

# Agenda

What's in an Algorithm?

Machine Learning Applications

Math for Machine Learning

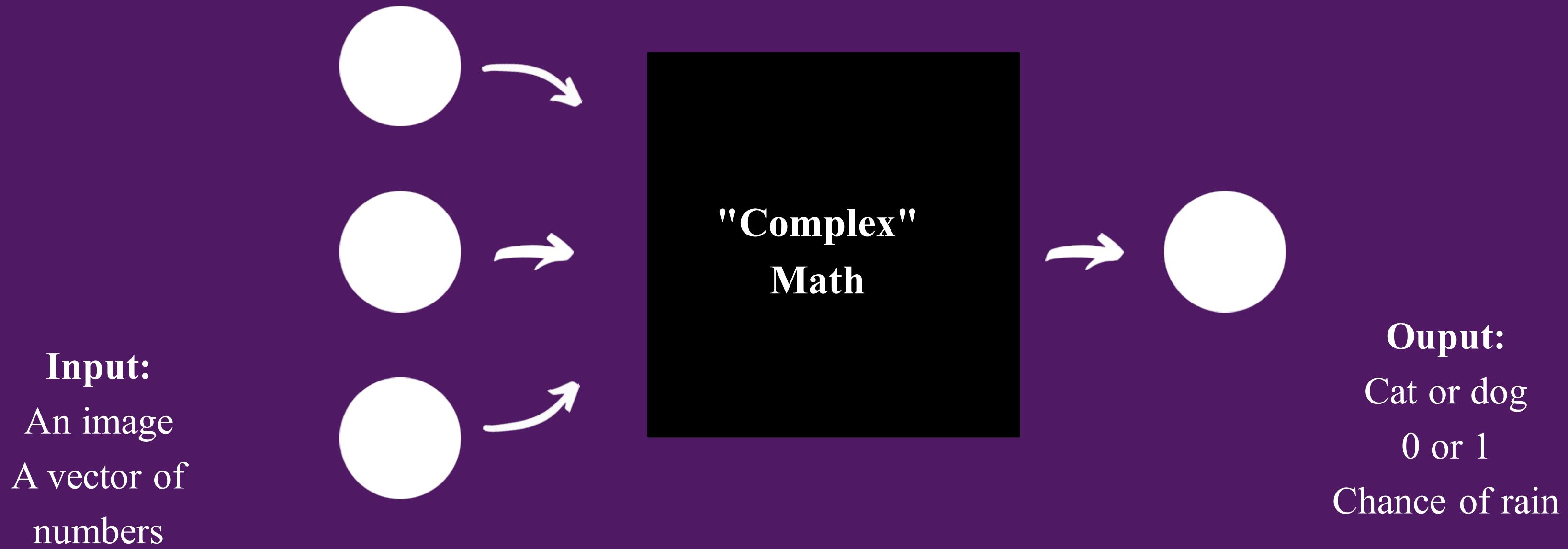
Types of ML



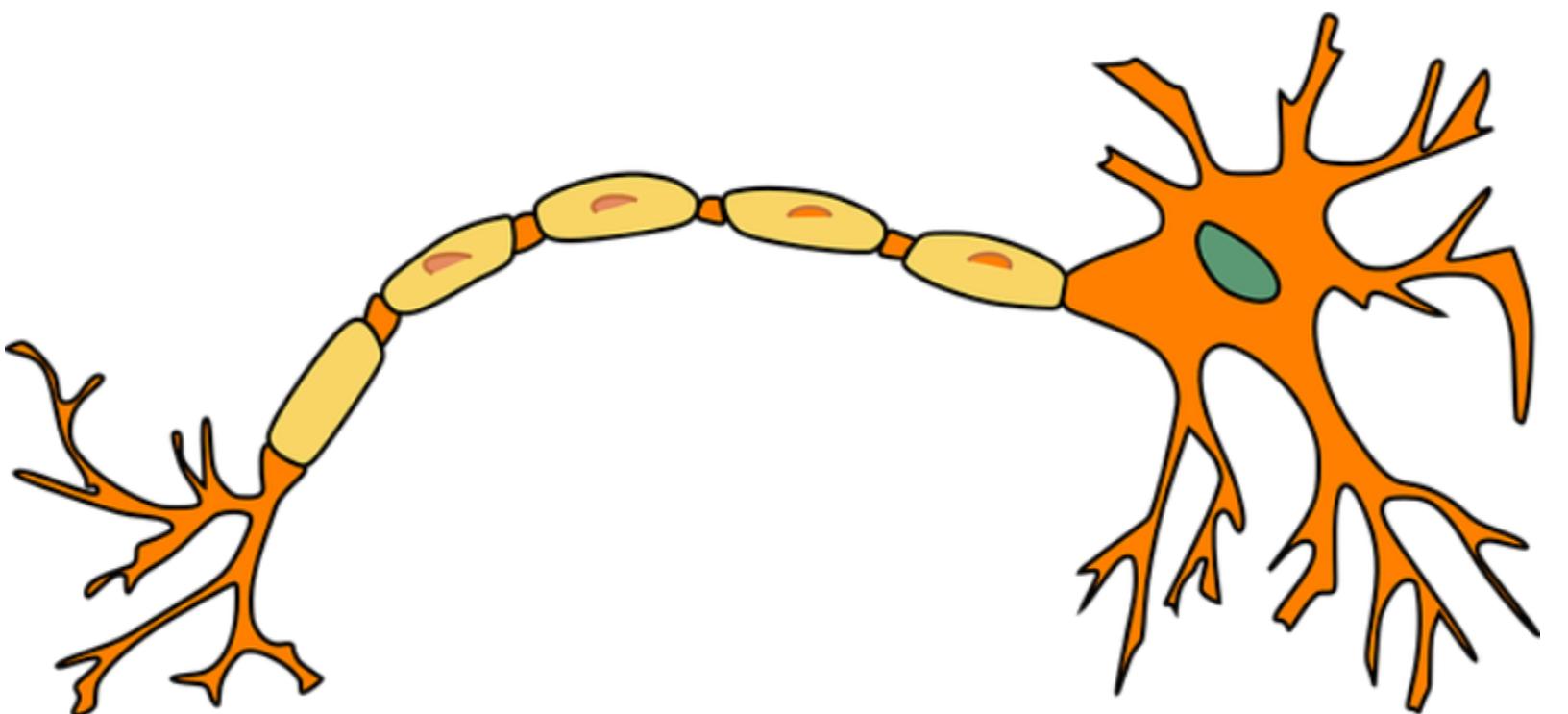


**WHAT'S IN THE BLACK  
BOX?**

# WHAT DOES AN ALGORITHM DO?



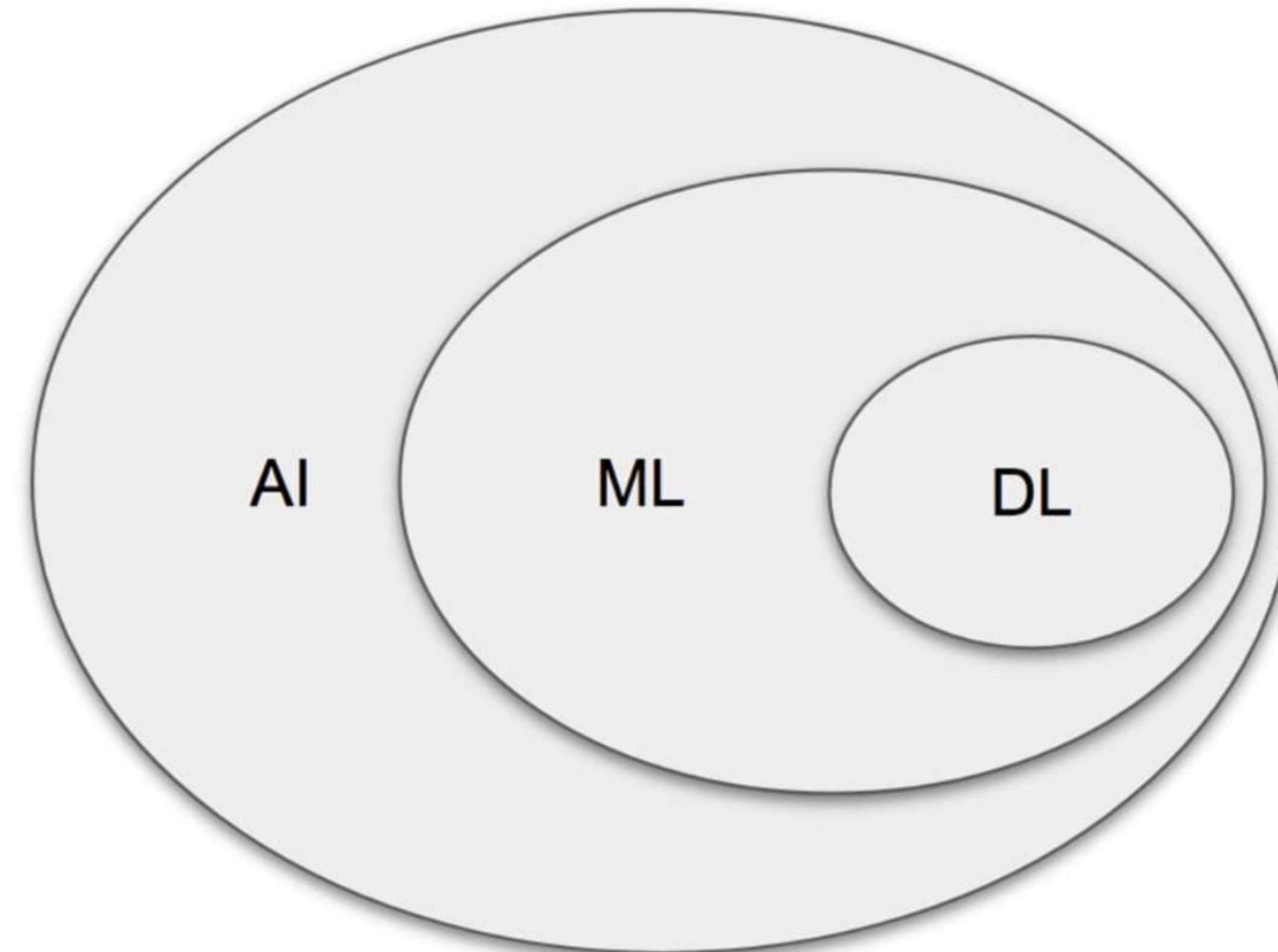
# ML takes cues from neuroscience



## □ Training

- Show a program lots of data
- Teach it to recognize patterns
- Check if it learned well

# What's the landscape?



# MACHINE LEARNING IN THE WILD

## ROUTE OPTIMIZATION



## MOVIE RECOMMENDATION S



## VIDEO GAME ENEMIES



## PERSONAL ASSISTANTS



# WILD MACHINE LEARNING

RECIDIVISM



HR  
SCREENINGS



POLICING



CREDIT  
SCORING



# Math Foundations

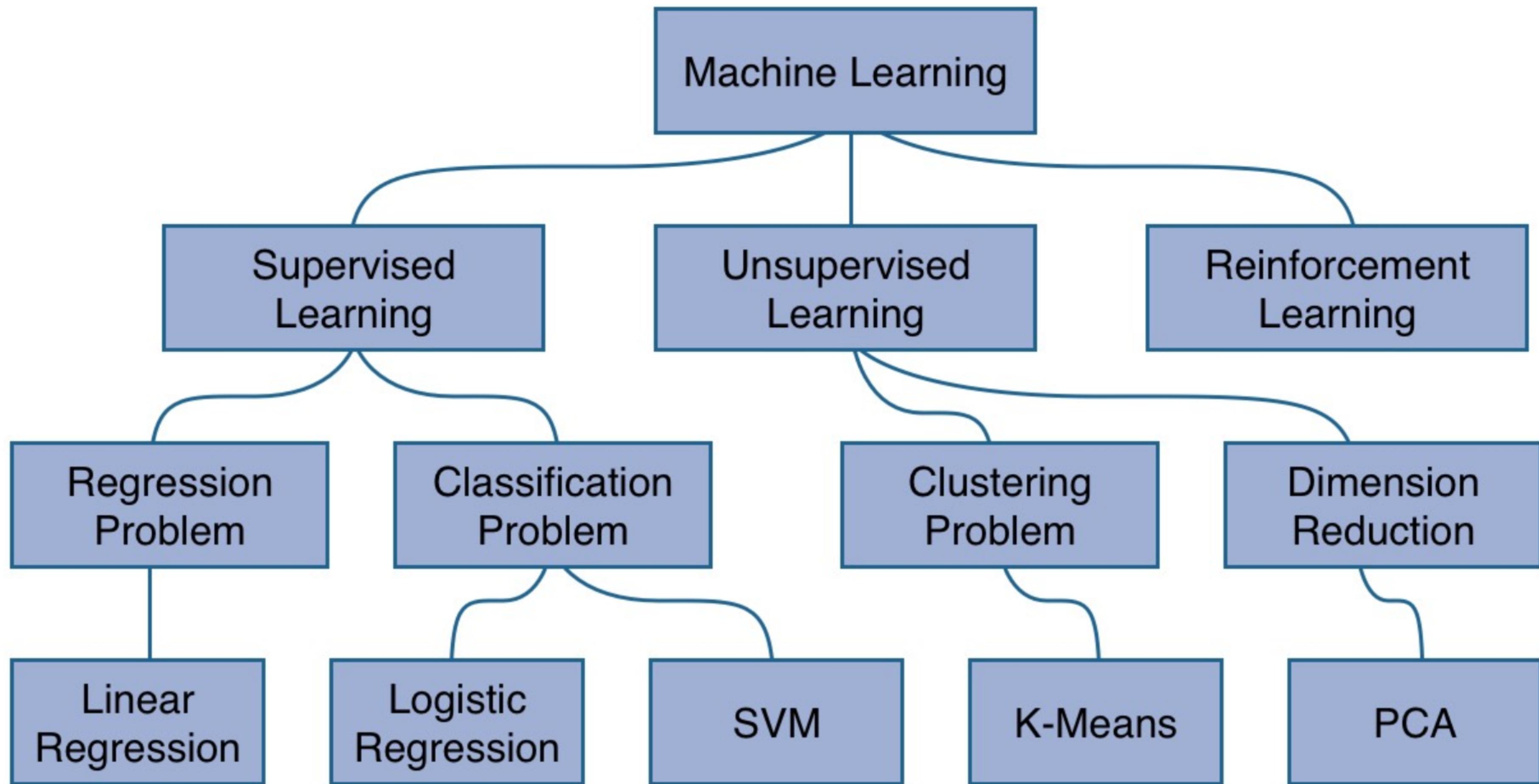
BIGGEST BARRIER TO ENTRY

WHAT THEY SAY YOU NEED

- Linear Algebra
- Calculus
- Statistics
- Discrete Math

WHAT'S USED FREQUENTLY

- SOME Linear Algebra
- Probability
- Statistics



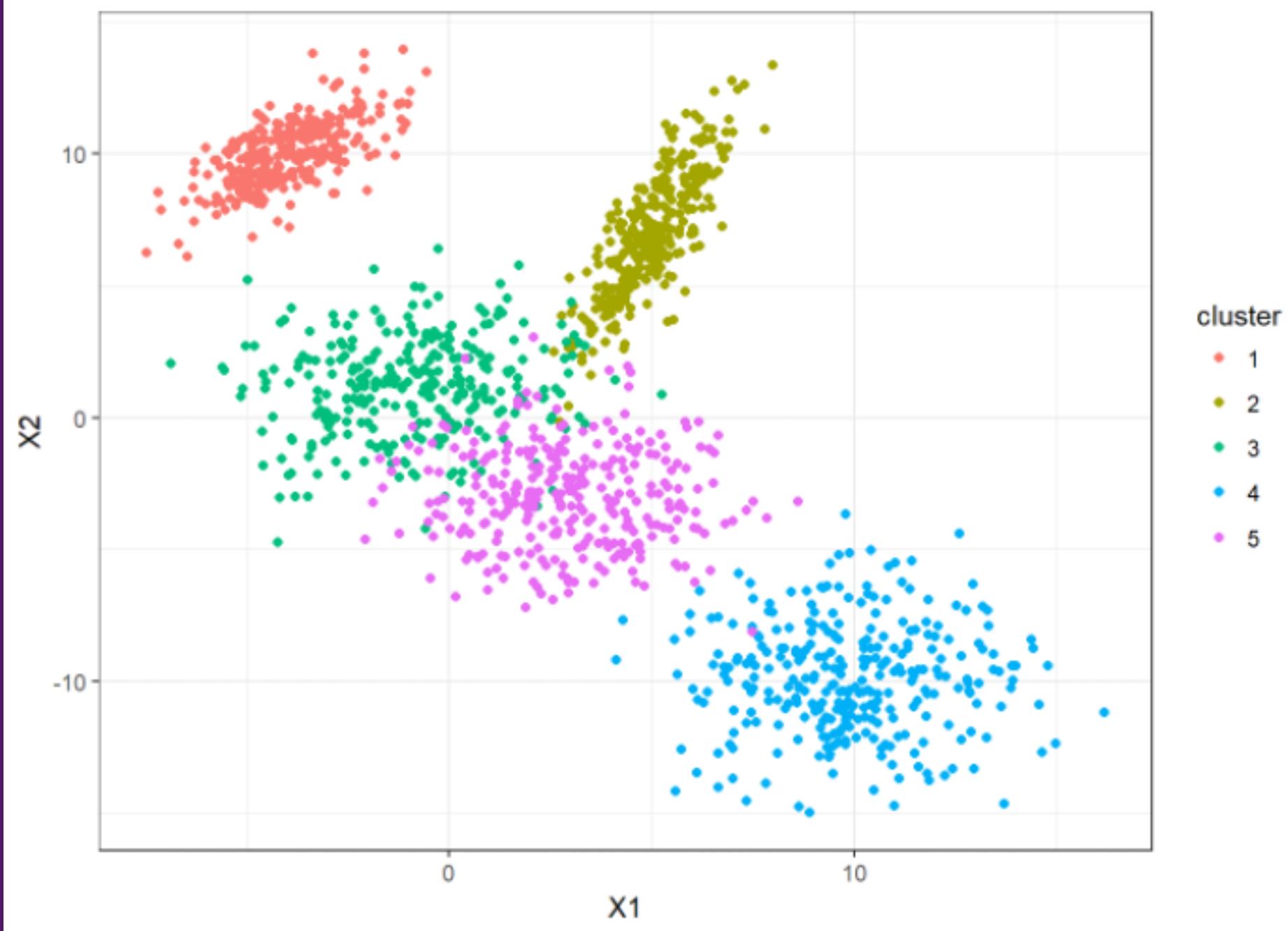
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1	0.505828	0.870442	9.949194	0.000552	1
2	-0.871557	0.868241	9.924039	0.000358	1
3	-0.871557	0.868241	9.924039	0.000729	1
4	-0.871557	0.868241	9.924039	0.005013	1
5	-0.046967	0.871548	9.961837	0.013645	1
6	0.871557	0.868241	9.924039	0.011765	1
7	0.871557	0.382027	9.954619	0.009366	1
8	0.871557	0.868241	9.924039	0.010479	1
9	0.871557	-0.316847	9.956907	0.013448	1
10	0.603430	-0.301062	9.977236	0.014846	1
11	0.555697	-0.295141	9.980185	0.015184	1
12	0.871557	-0.157112	9.960708	0.017696	1
13	0.871557	0.473205	9.950702	0.016759	1
14	0.698513	0.869429	9.937614	0.012372	1
15	-0.871557	-0.868241	9.924039	0.010986	1

# SUPERVISED

LABELED OR NAH?

# UNSUPERVISED

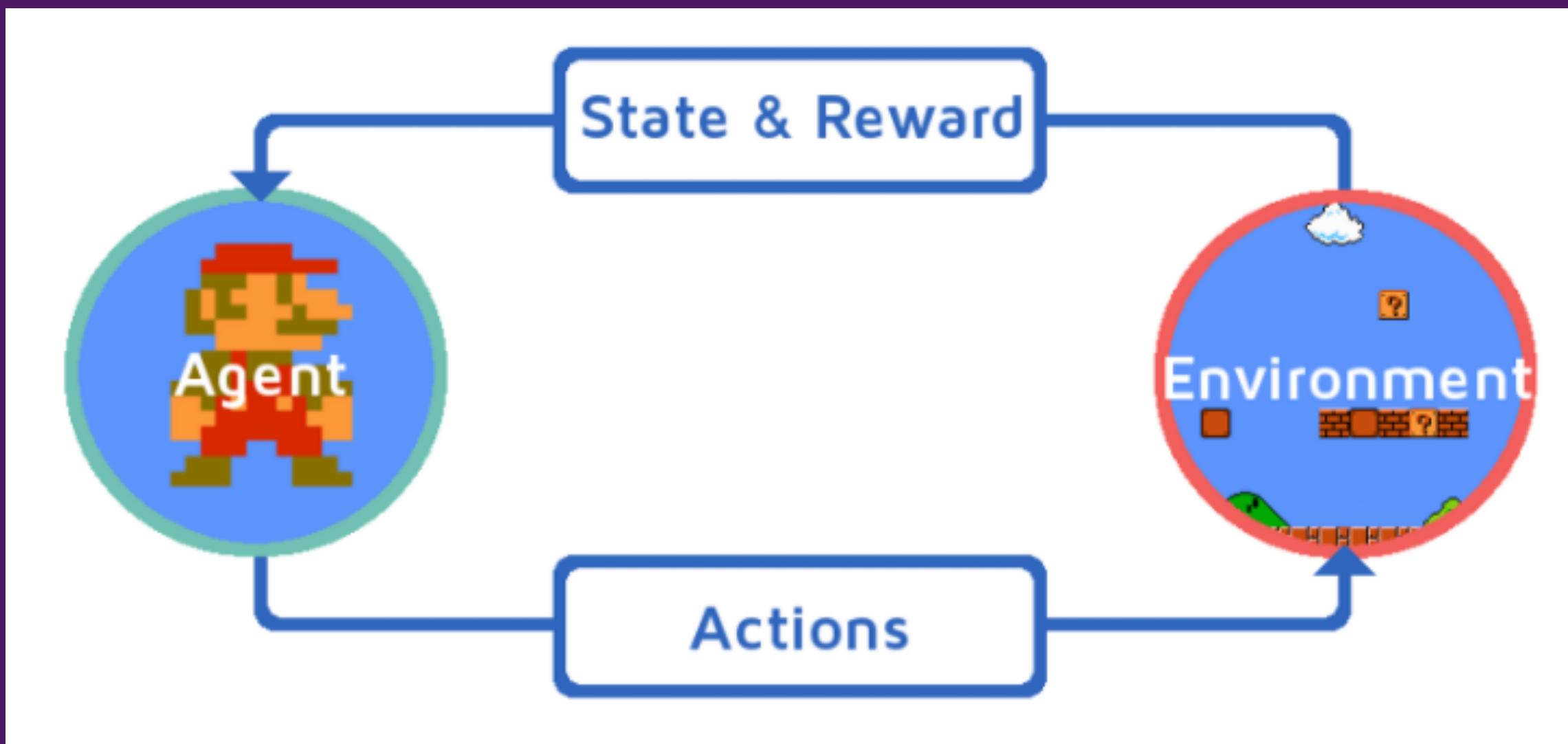
CREATES SIMILAR GROUPINGS

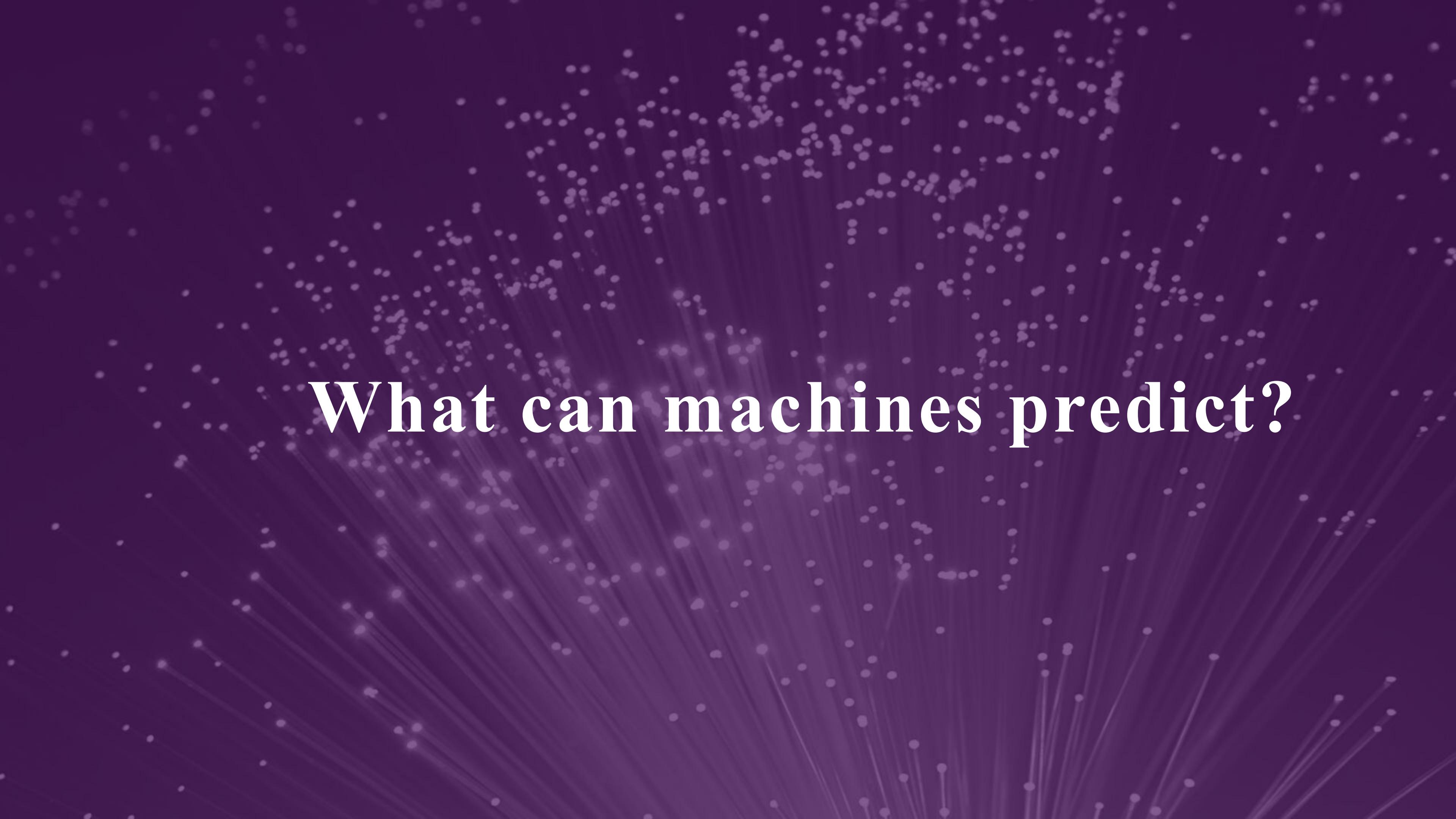


# REINFORCEMENT

## T

WHAT MANY CONSIDER AI



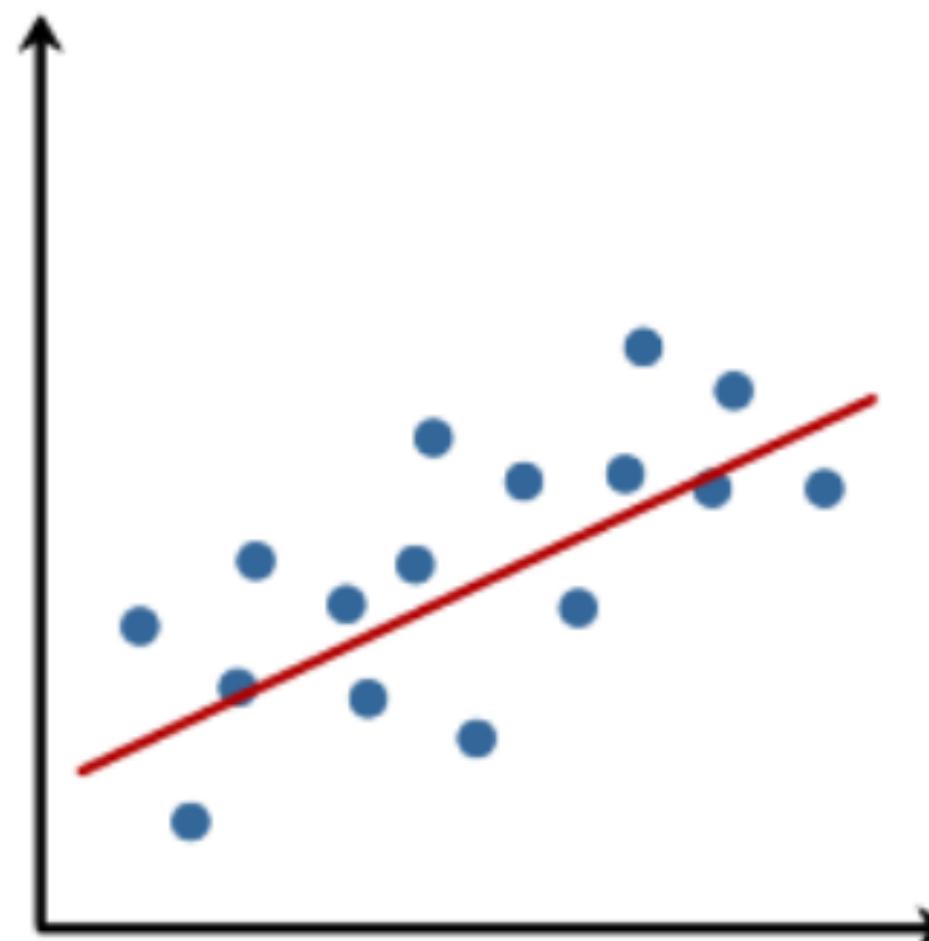


# What can machines predict?

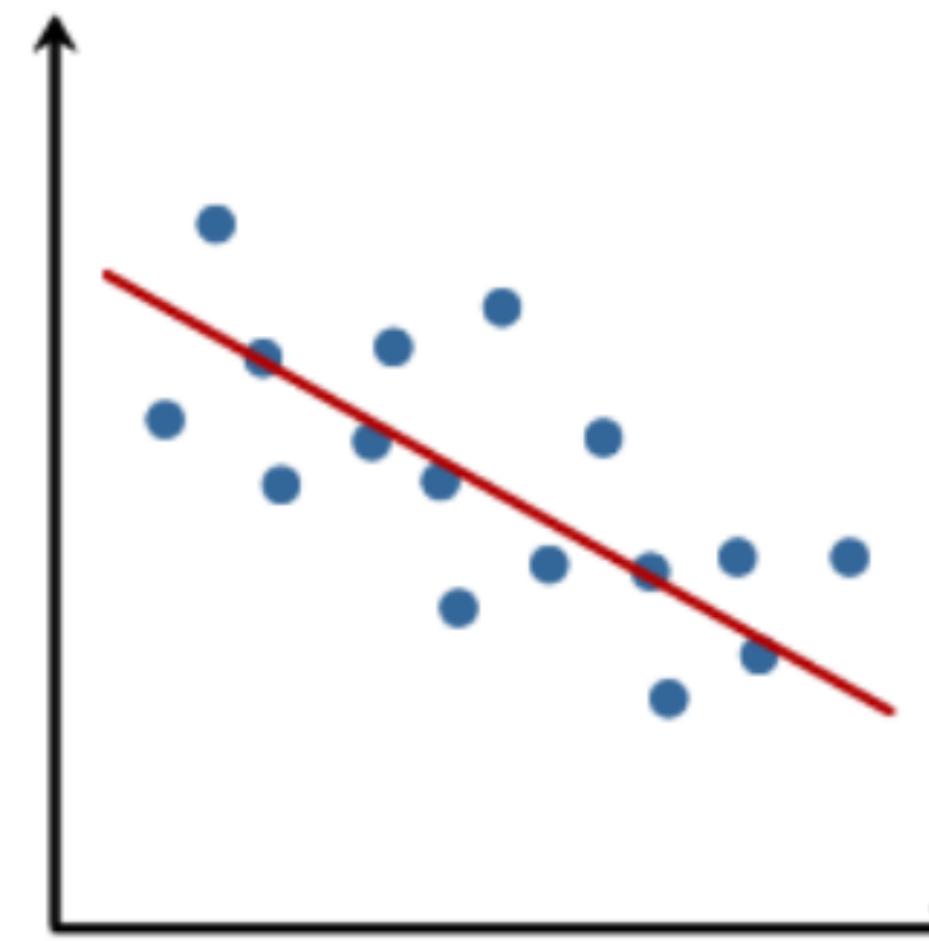
# REGRESSION

REAL/CONTINUOUS NUMBERS

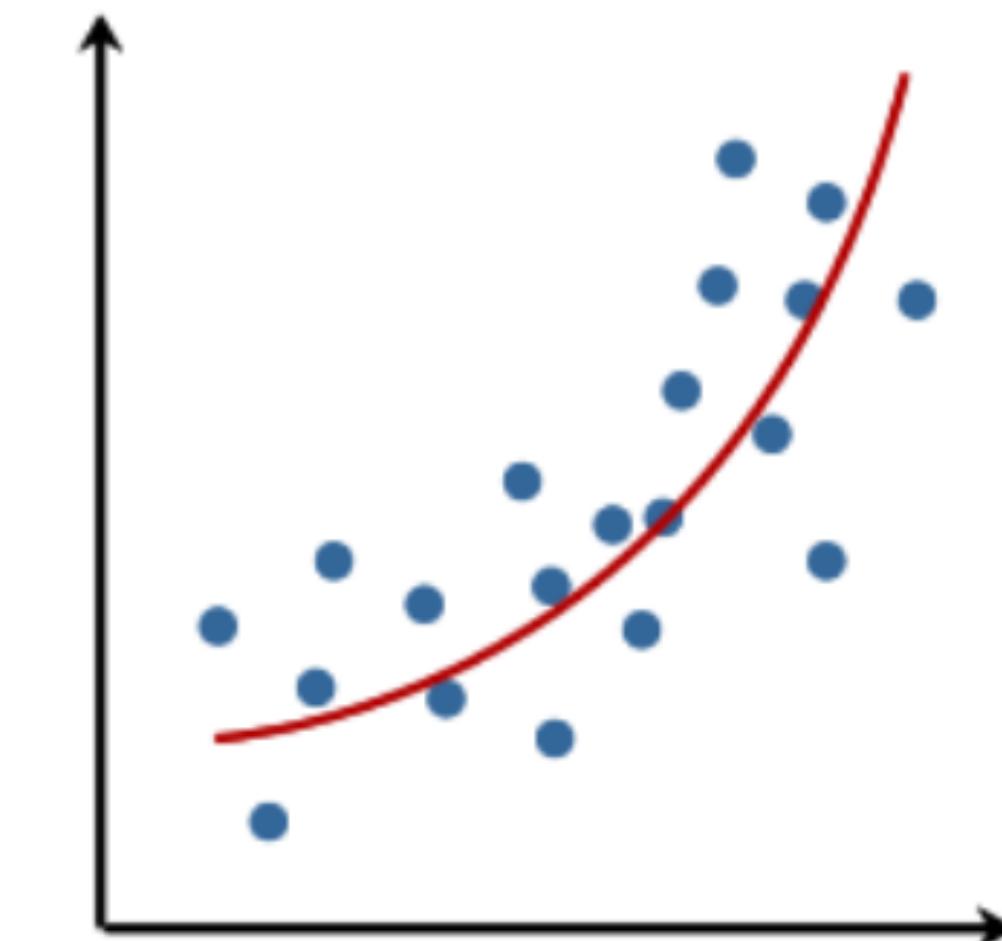
Linear



Linear

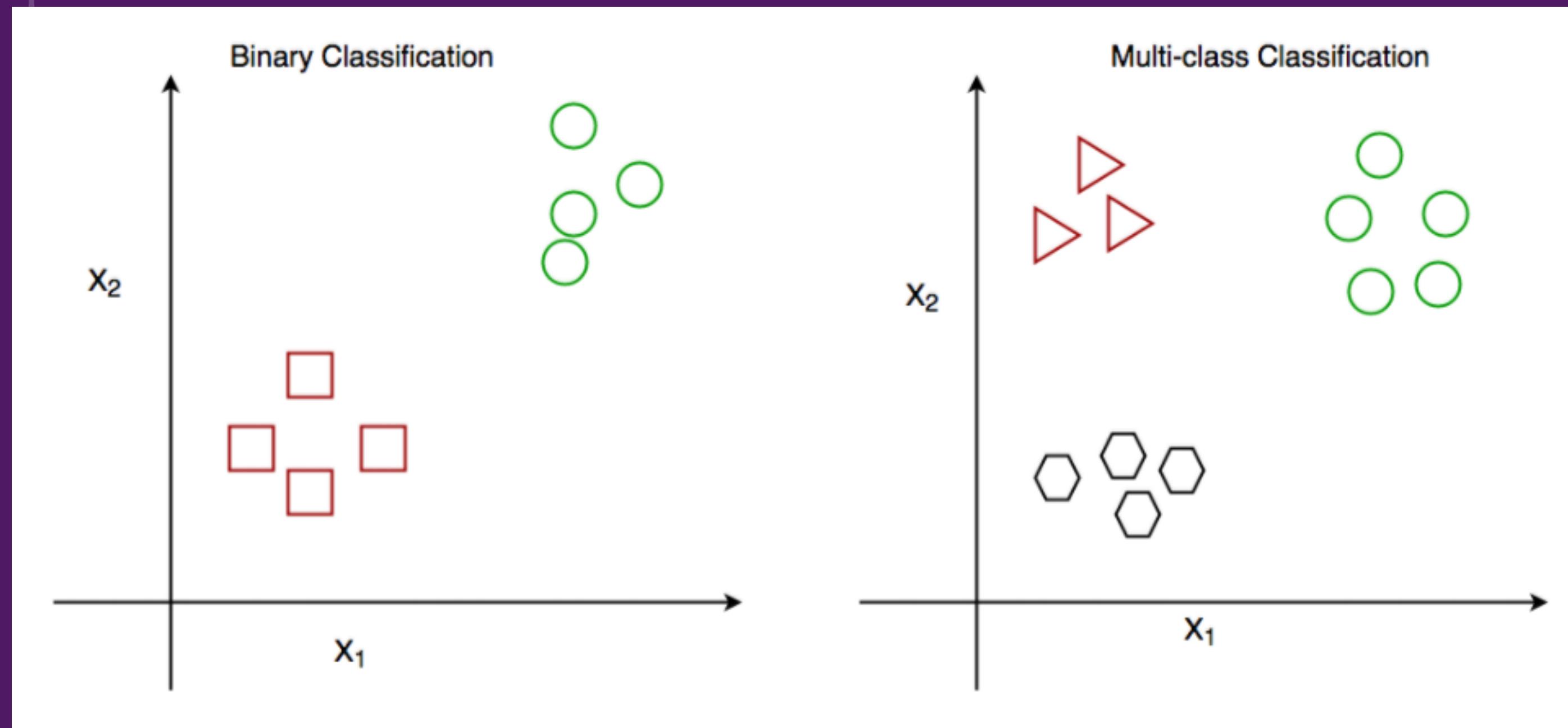


No linear relationship

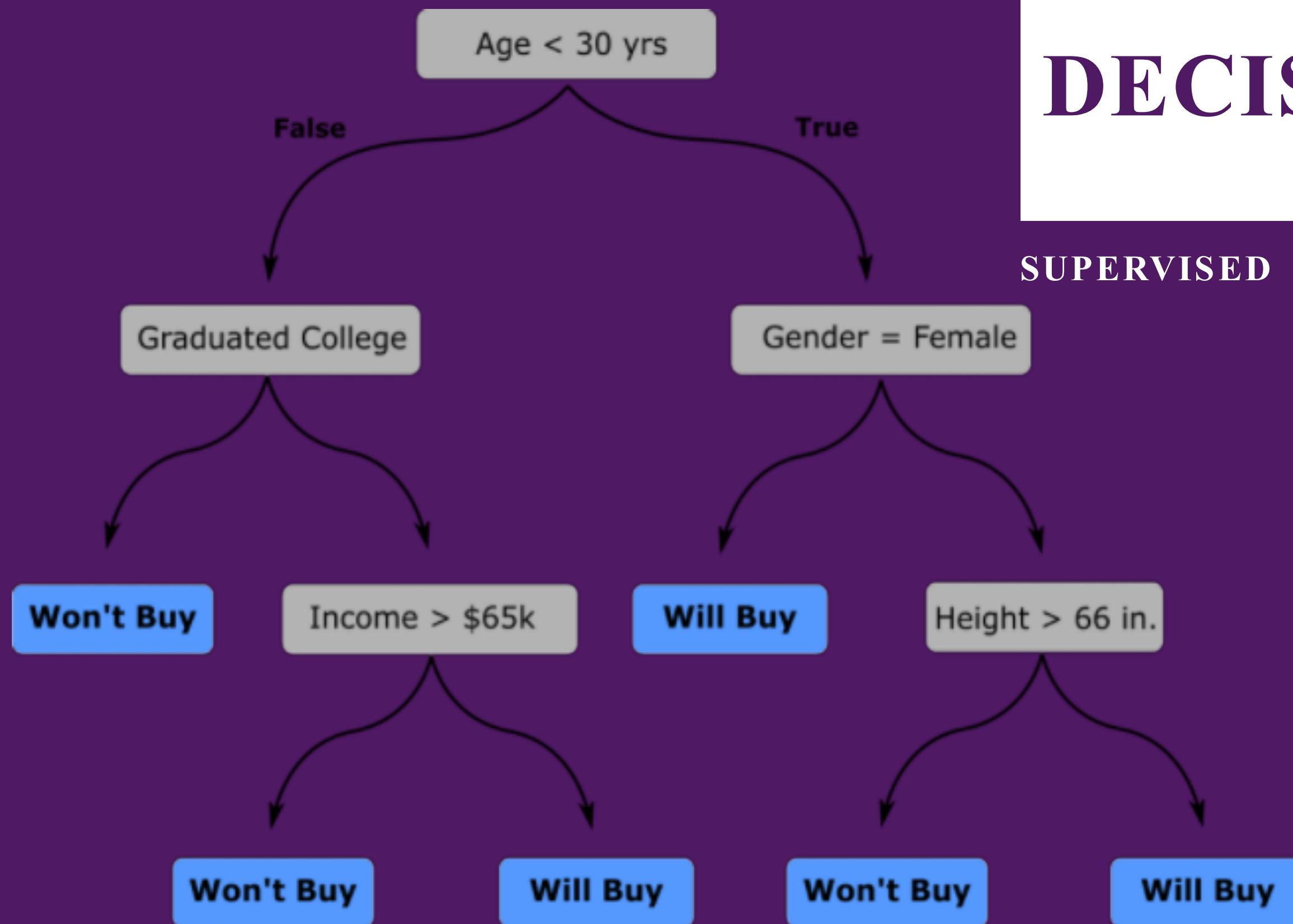


# CLASSIFICATION

DISCRETE/CATEGORICAL VARIABLE

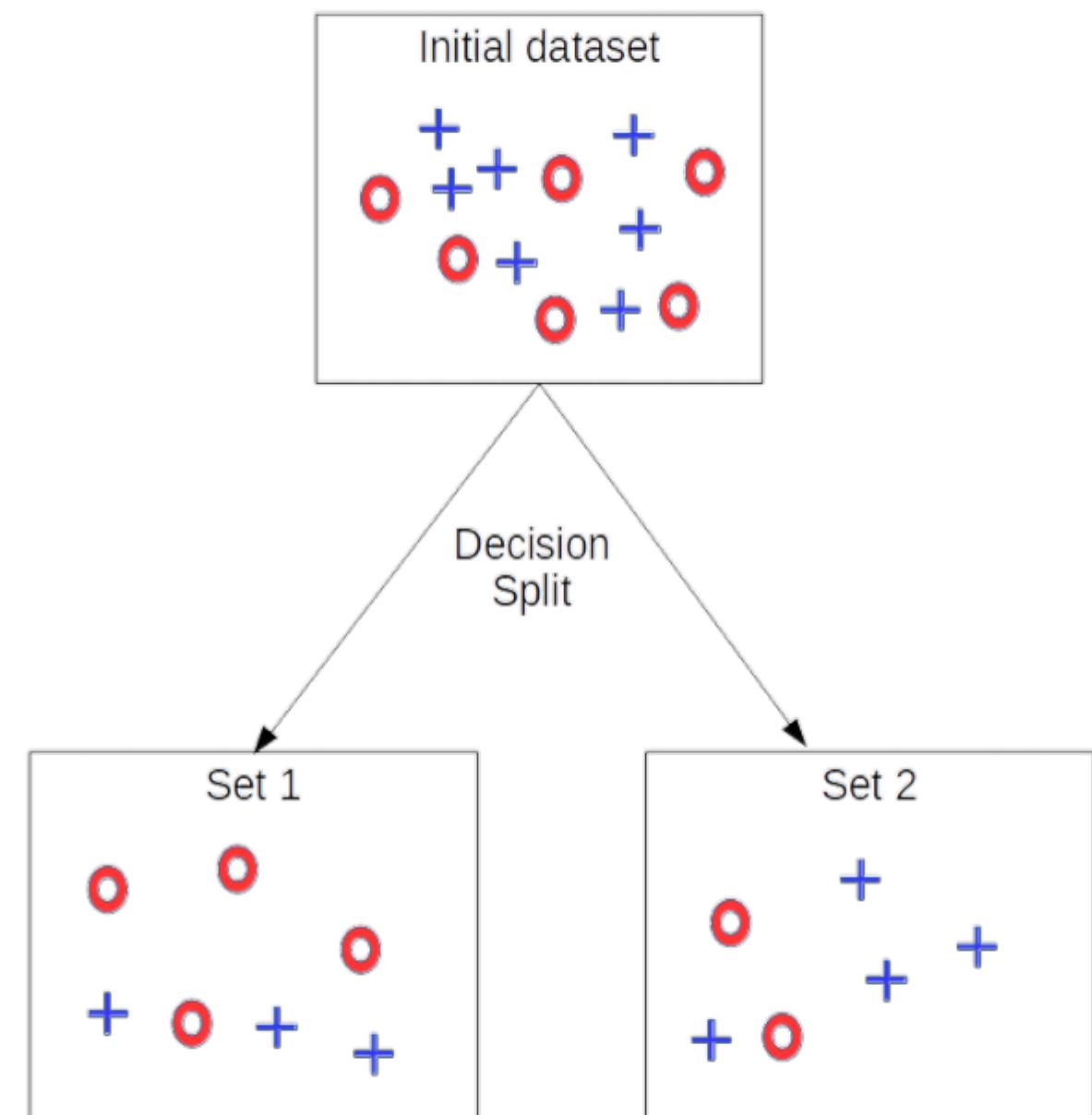


# DECISION TREES



# Entropy

- A measure of the degree of randomness in a variable
- "Good" Decision Trees have homogenous leaf nodes
- The higher the entropy, the harder to draw conclusions



# Information Gain

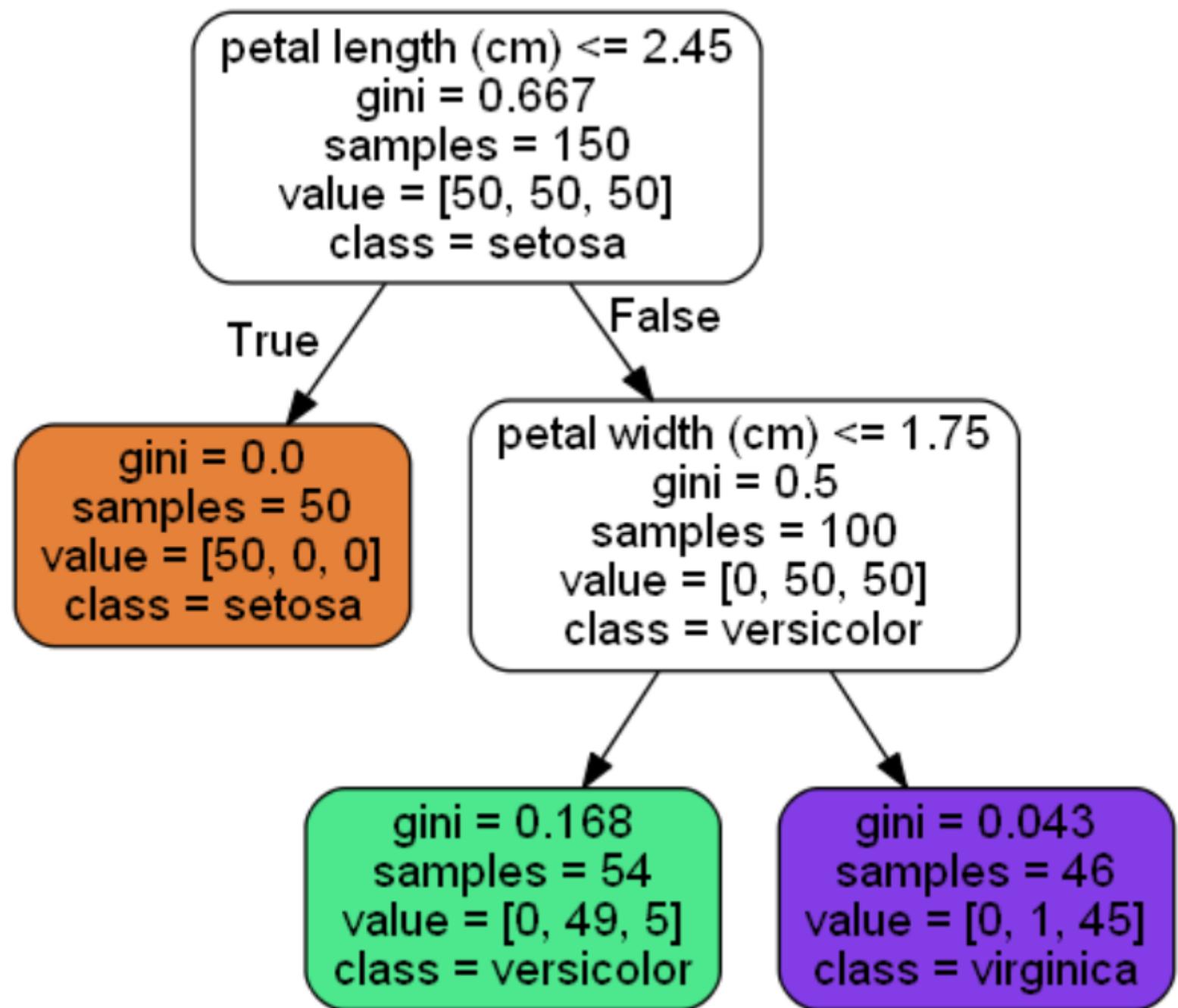
- Used to decide which of the attributes are most relevant
- The purpose is to find the attribute that returns the most information gain
- Expected information gain = decrease in entropy
- The less random the variables, the more information is gained

# Gini Index

- Measures how impure a node is
- Calculated per node

Gini index is used in CART  
(Classification and Regression Trees)

ID3 search algorithm uses entropy and information gain



# Pre-pruning

Involves setting the tree parameters before building it so it stops early without completely being built.

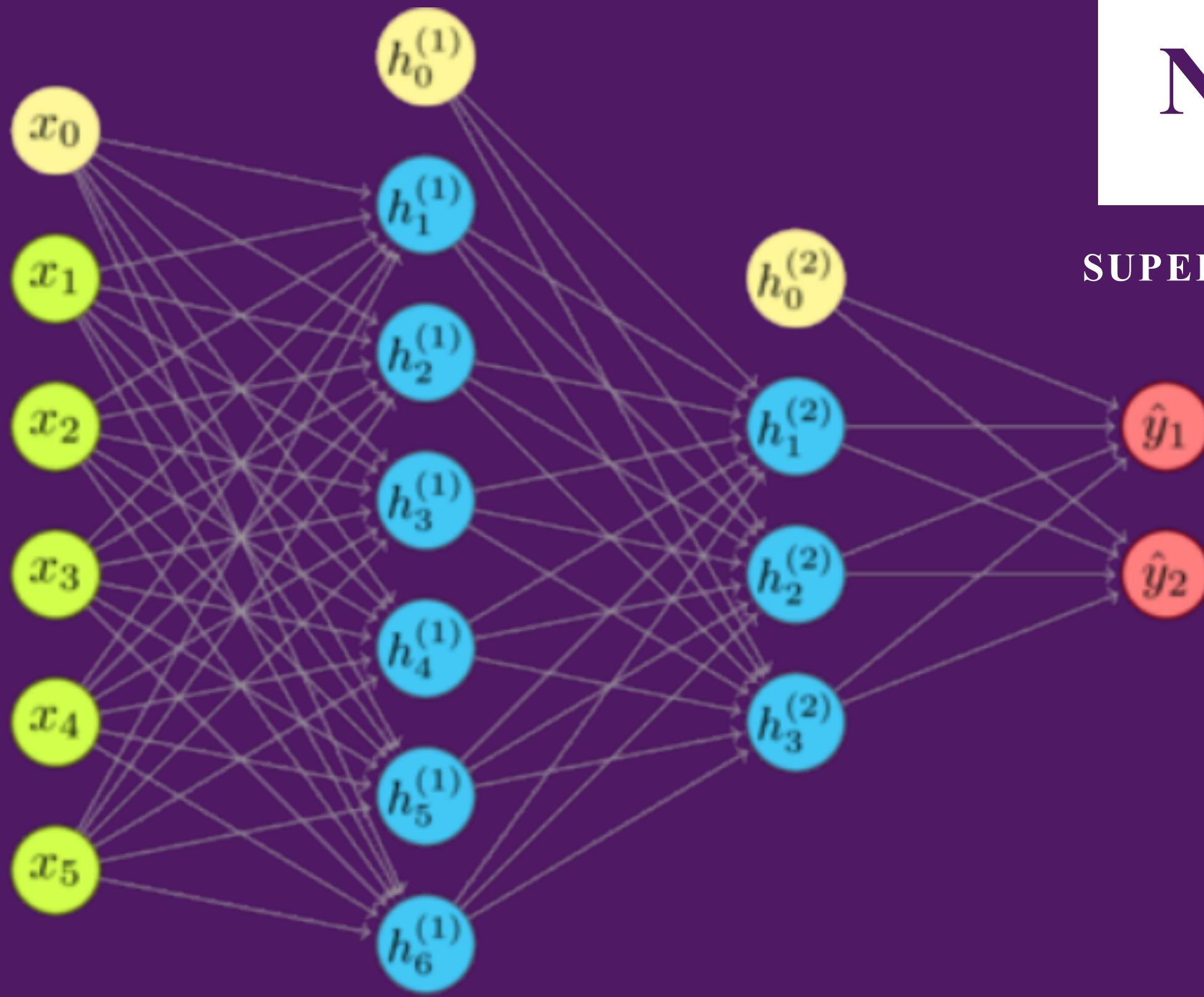
Variables to tune:

- Set max tree depth
- Set max terminal nodes
- Set max number of features
- Set max samples for a node split
  - controls the size of terminal nodes

# Post-pruning

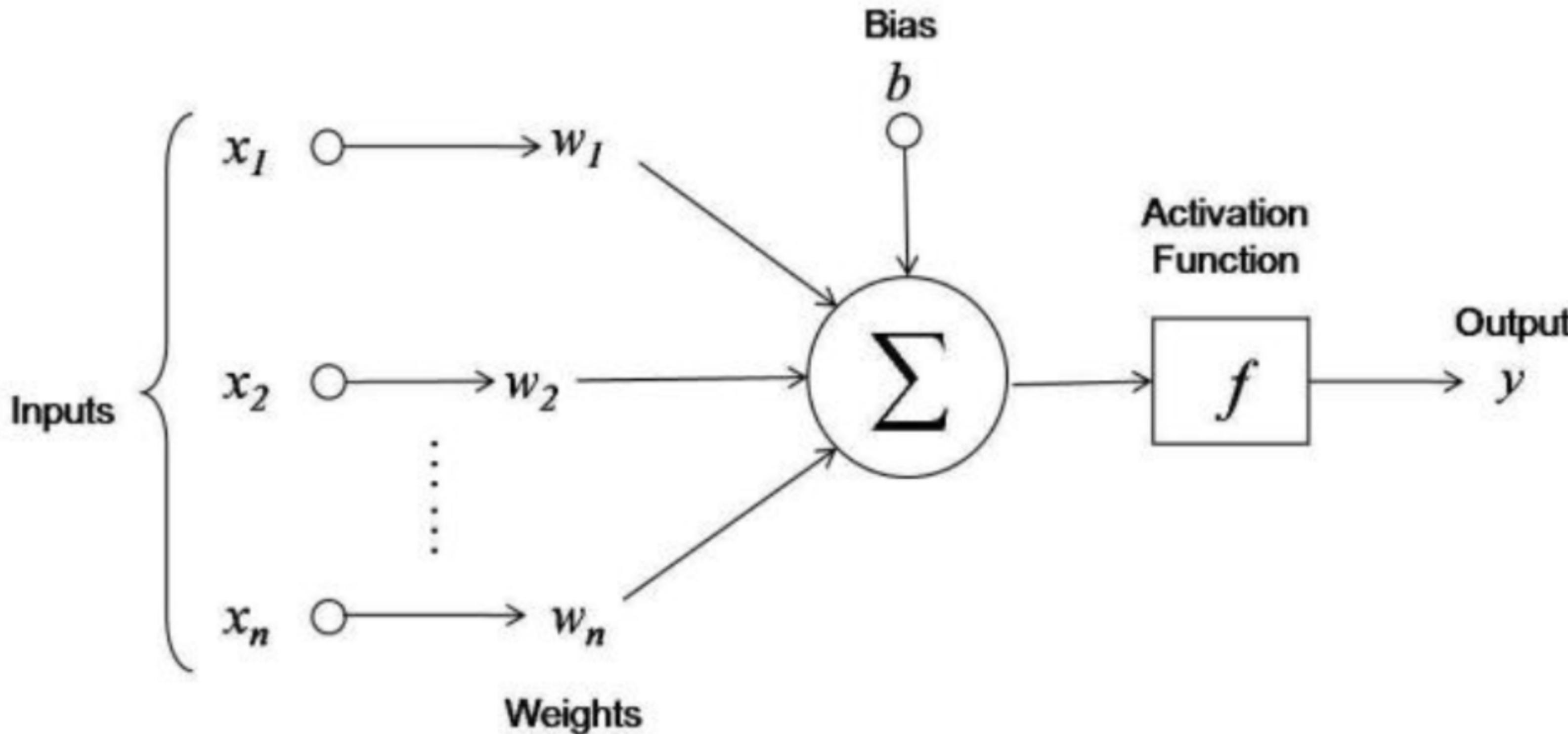
- Validate the performance of the model on a test
- Cut back splits that seem to overfit the noise in the training set
- Removes a branch from a fully grown tree

# NEURAL NETWORKS

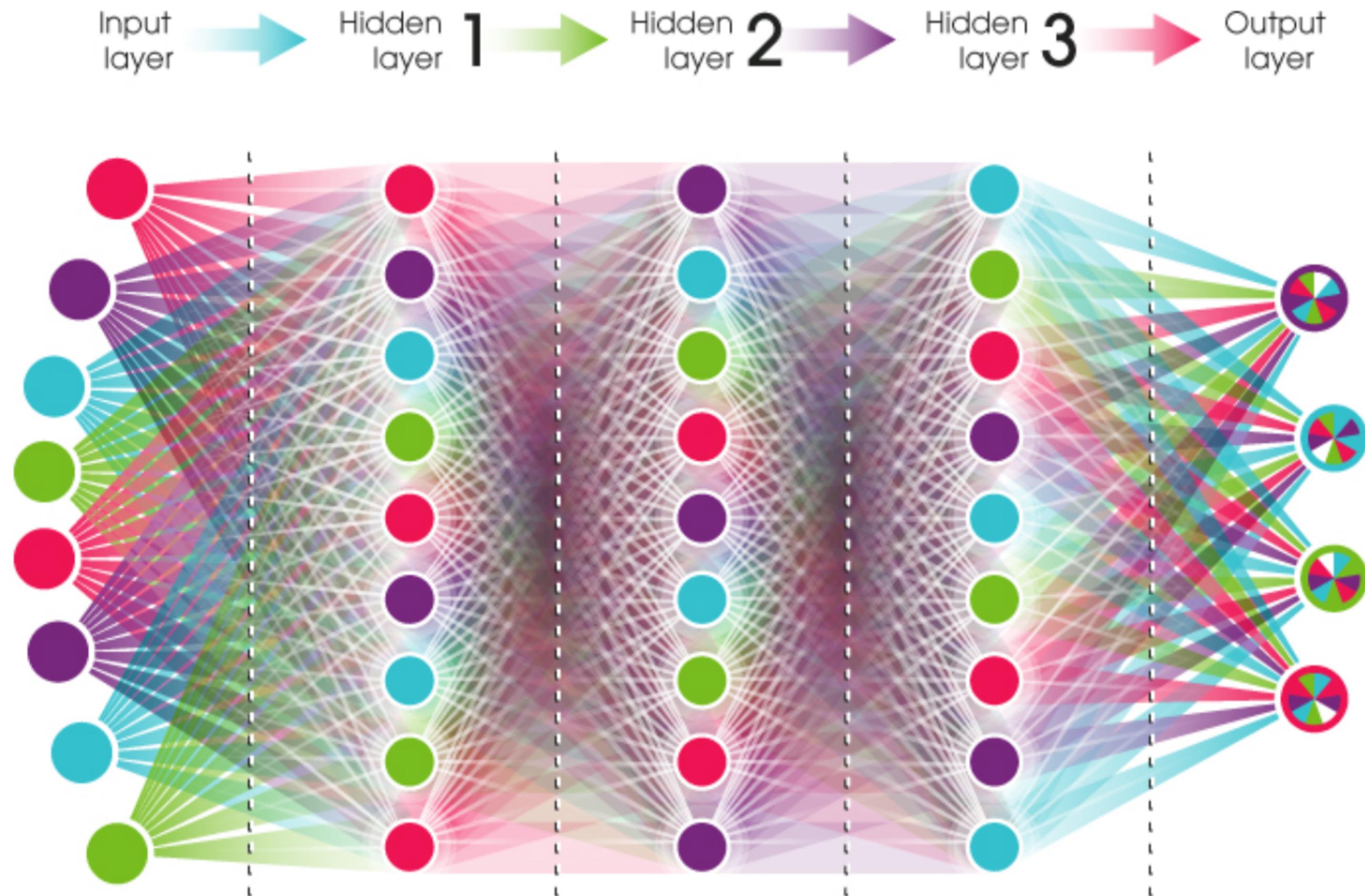


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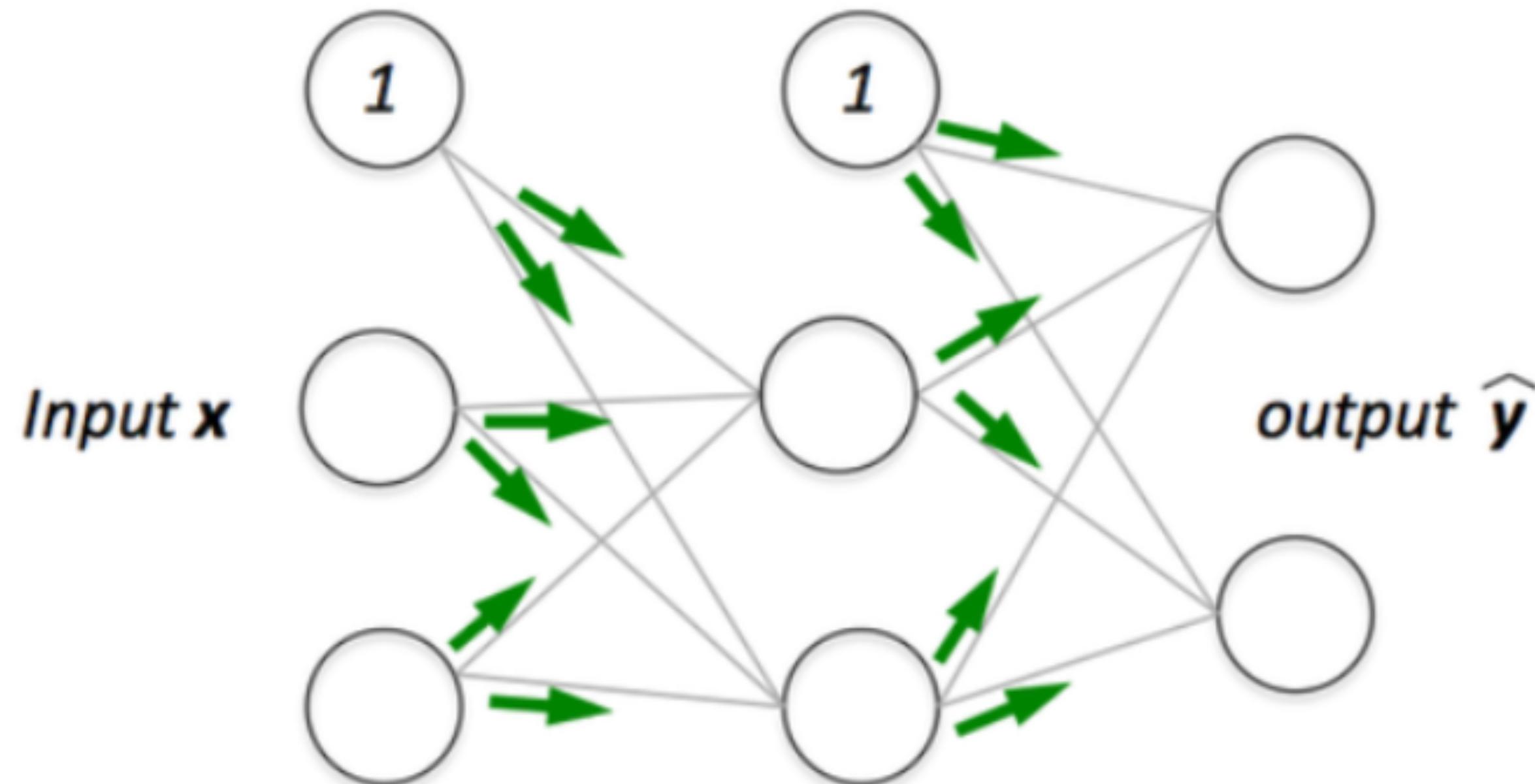
# Neuron Structure



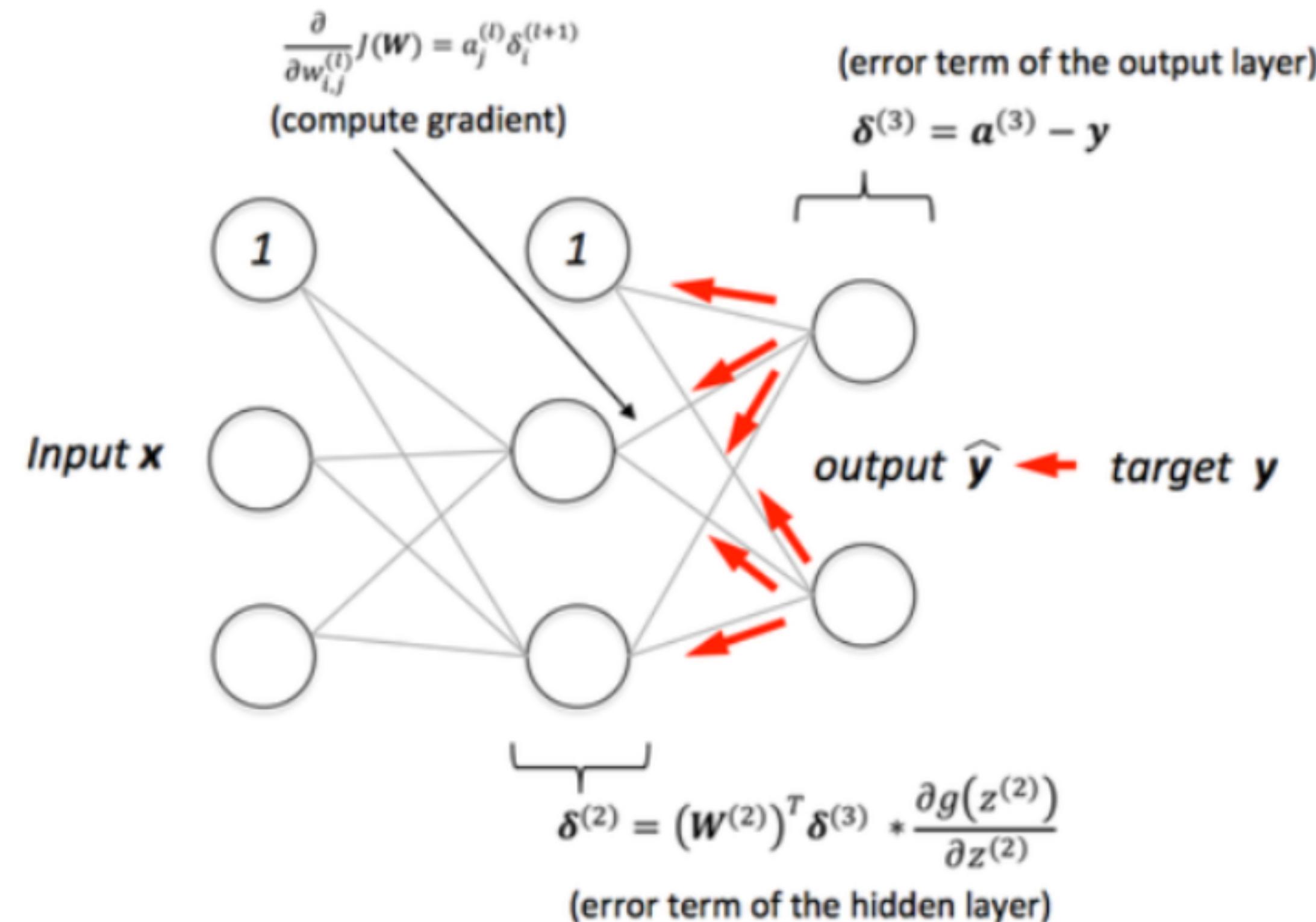
# DEEP NEURAL NETWORK



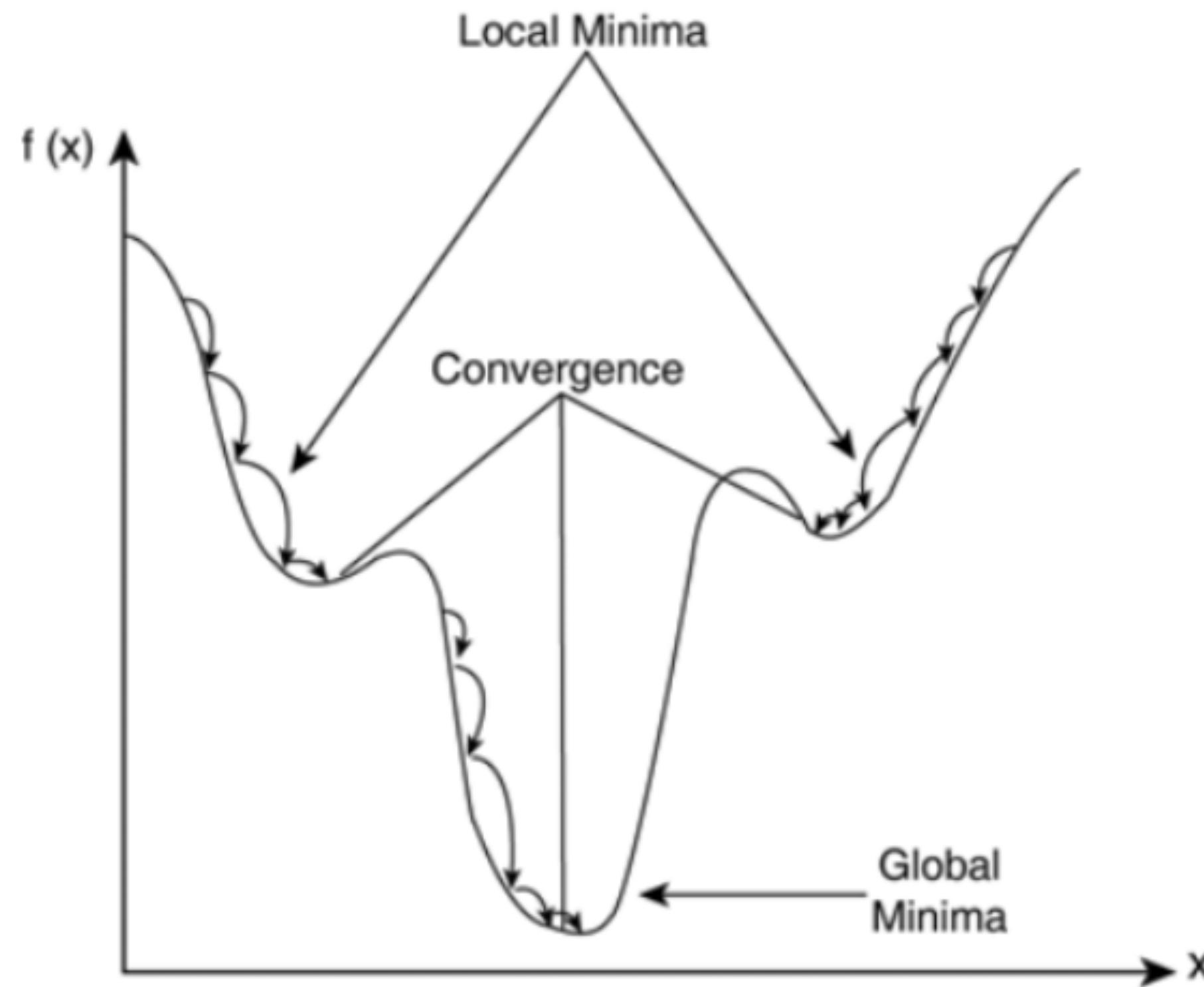
# Forward Propogation



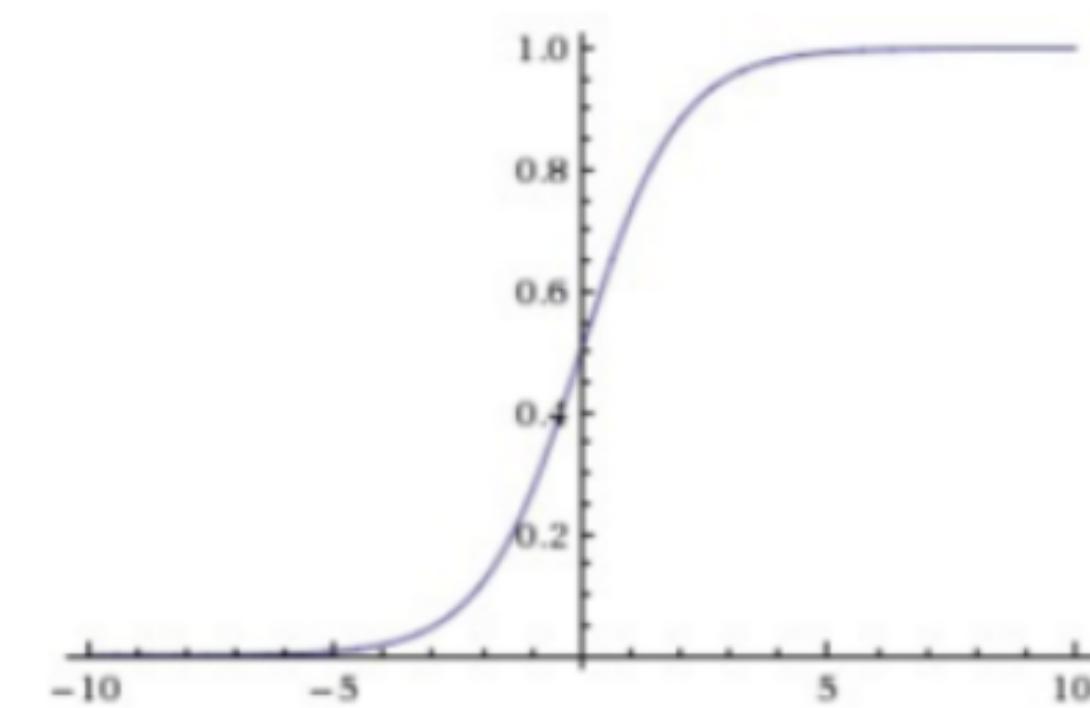
# Back Propogation



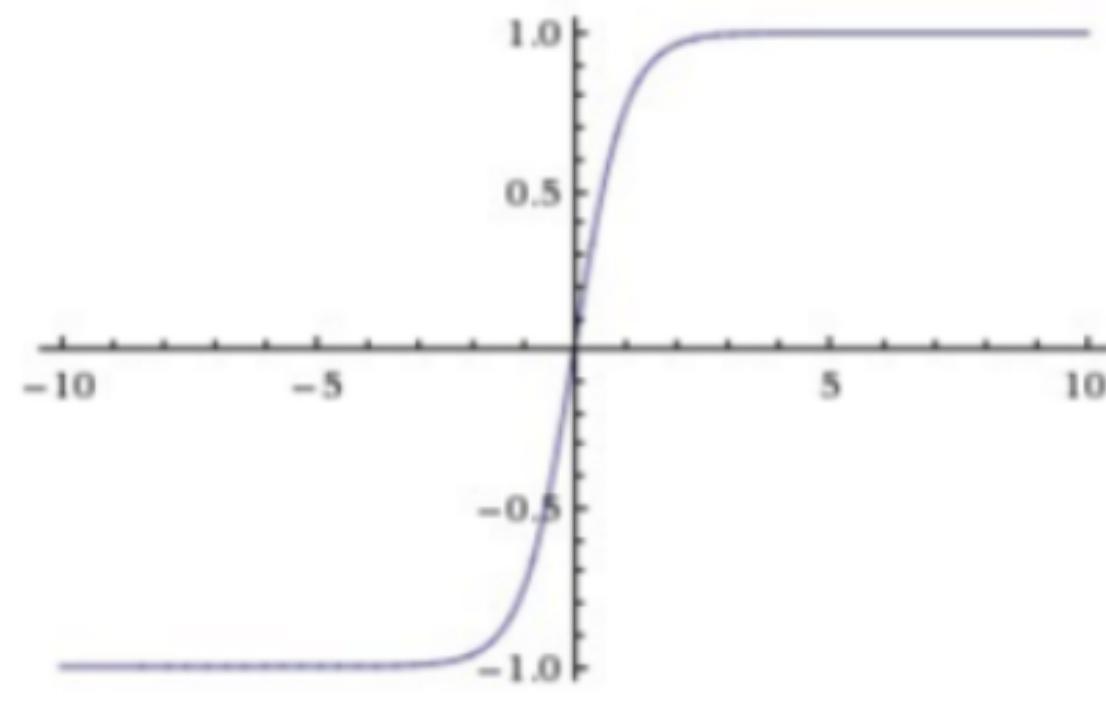
# Learning Rate



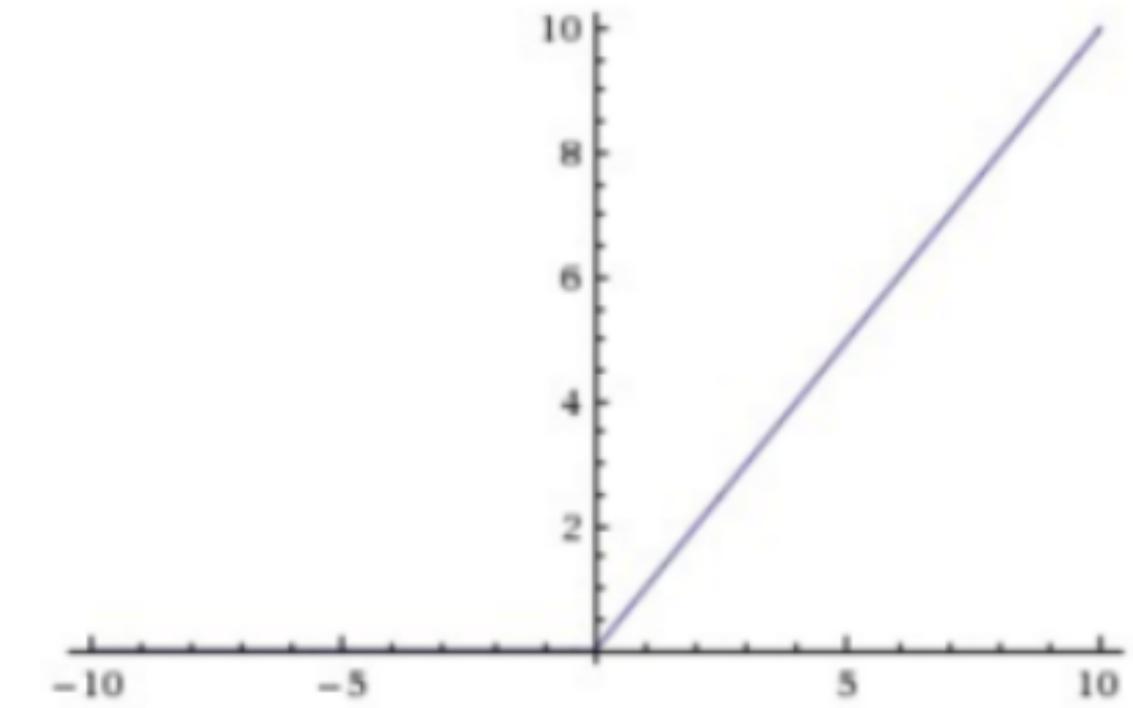
# Activation Functions



Sigmoid

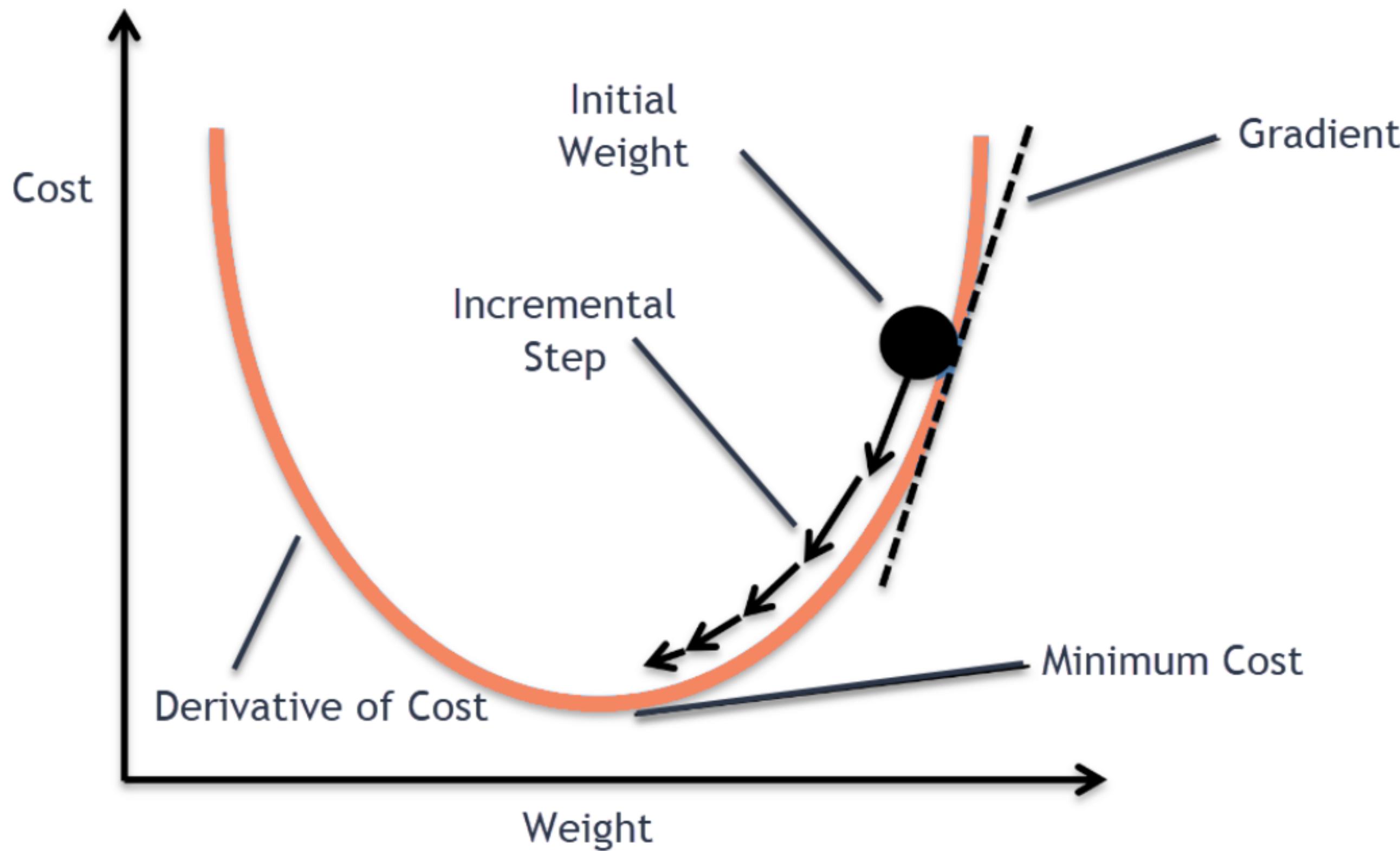


tanh



ReLU

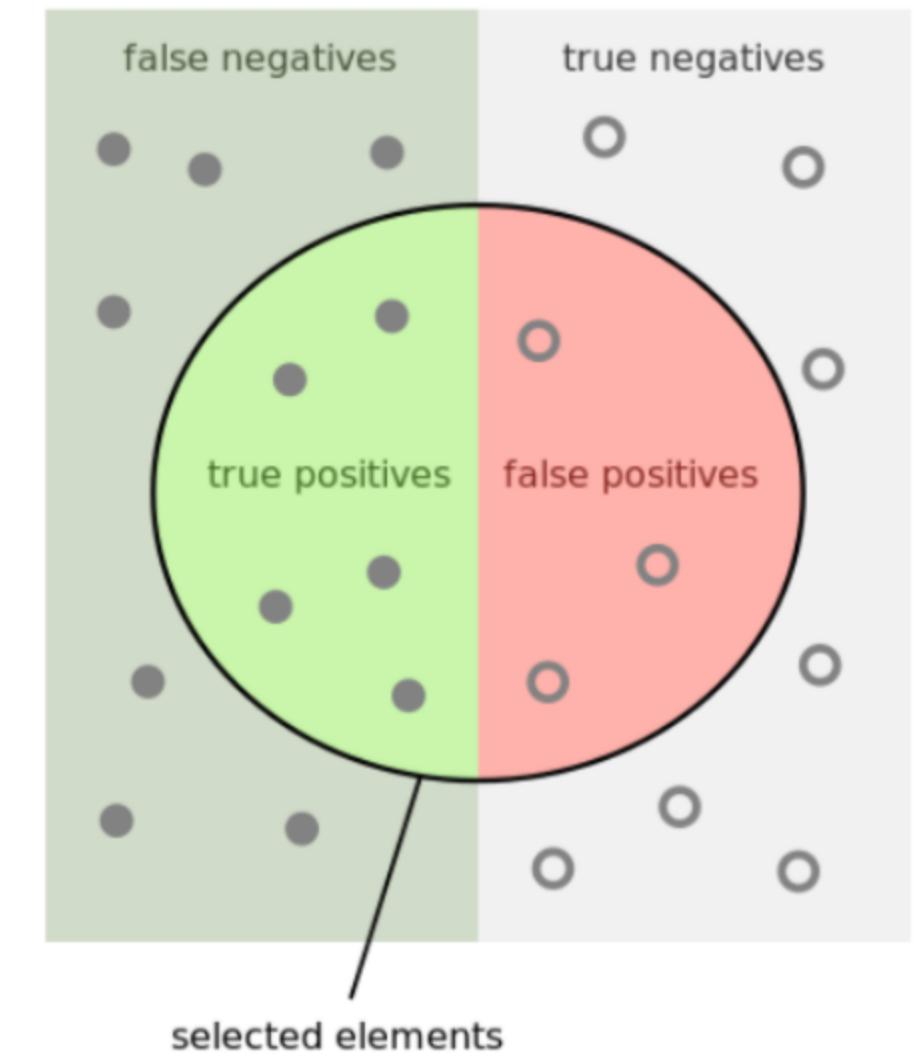
# Gradient Descent



# What Makes A Successful Model?

# Evaluating Models

- Classification Accuracy
- Confusion matrix
- Logarithmic Loss
- Area under curve (AUC)
- F-Measure



How many selected items are relevant?	$\text{Precision} = \frac{\text{green}}{\text{green} + \text{red}}$	How many relevant items are selected?	$\text{Recall} = \frac{\text{green}}{\text{green} + \text{black}}$
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# Classification Accuracy

precision: TP/cancer diagnoses

True state

		Diagnosis	
		No cancer	Cancer
True state	No cancer	$TN$	$FP$
	Cancer	$FN$	$TP$

recall: TP/cancer true states

The diagram illustrates a 2x2 confusion matrix for cancer diagnosis:

		Diagnosis	
		No cancer	Cancer
True state	No cancer	$TN$	$FP$
	Cancer	$FN$	$TP$

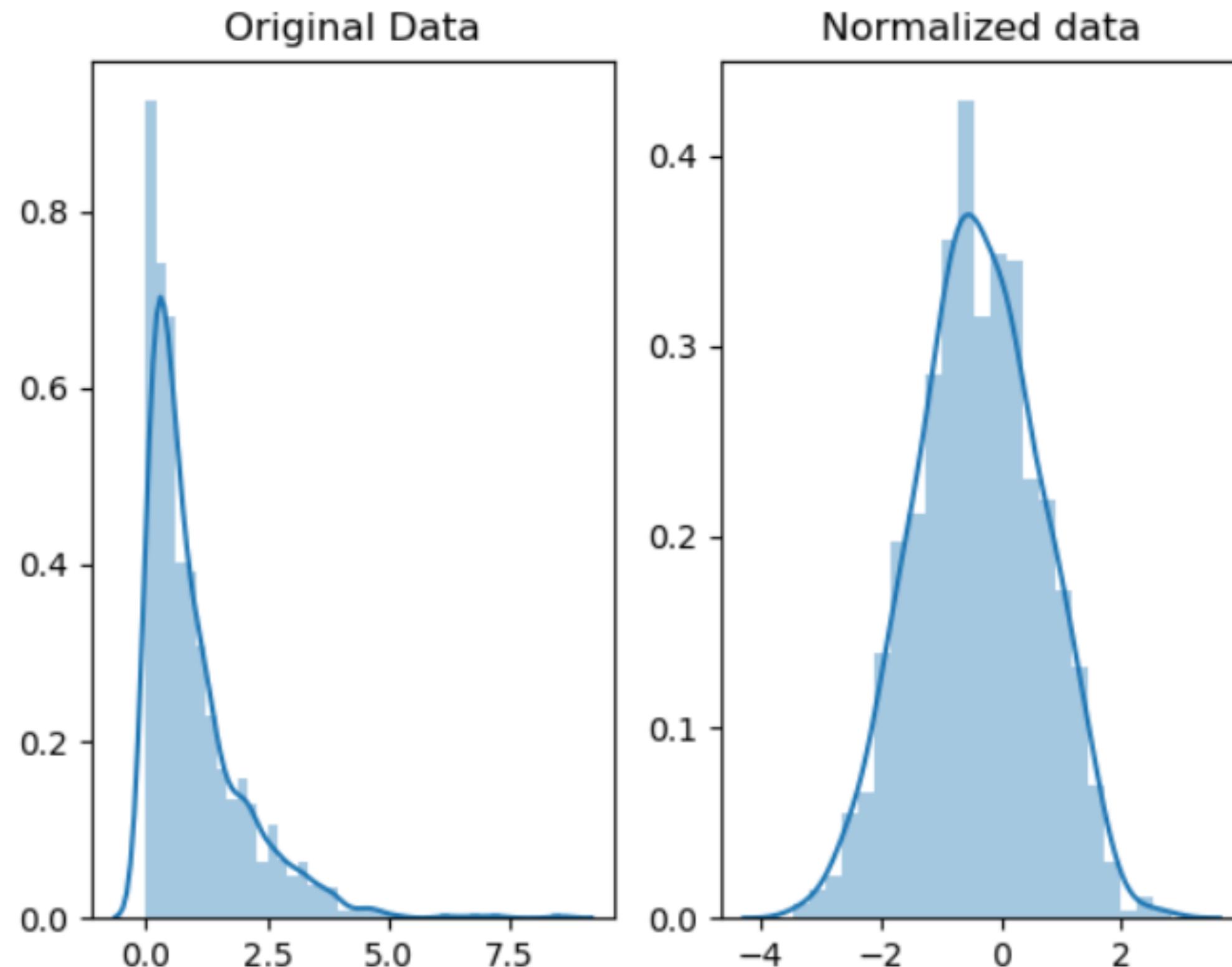
Annotations:

- A red oval highlights the bottom-left cell ( $FN$ ) and the entire bottom row.
- A green oval highlights the bottom-right cell ( $TP$ ) and the entire right column.

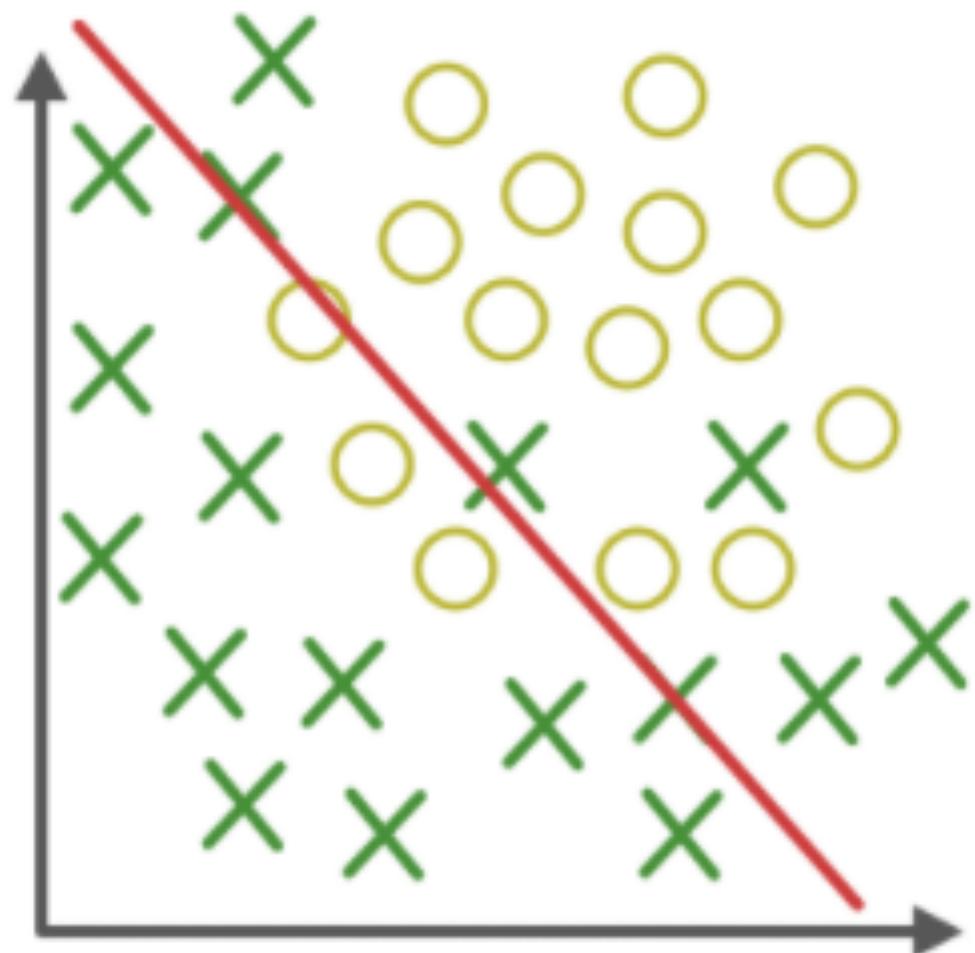
# Confusion Matrices

		Predicted Class		
		Positive	Negative	
Actual Class	Positive	True Positive (TP)	False Negative (FN) <b>Type II Error</b>	<b>Sensitivity</b> $\frac{TP}{(TP + FN)}$
	Negative	False Positive (FP) <b>Type I Error</b>	True Negative (TN)	<b>Specificity</b> $\frac{TN}{(TN + FP)}$
Precision	$\frac{TP}{(TP + FP)}$	Negative Predictive Value	$\frac{TN}{(TN + FN)}$	Accuracy $\frac{TP + TN}{(TP + TN + FP + FN)}$

# Normalization

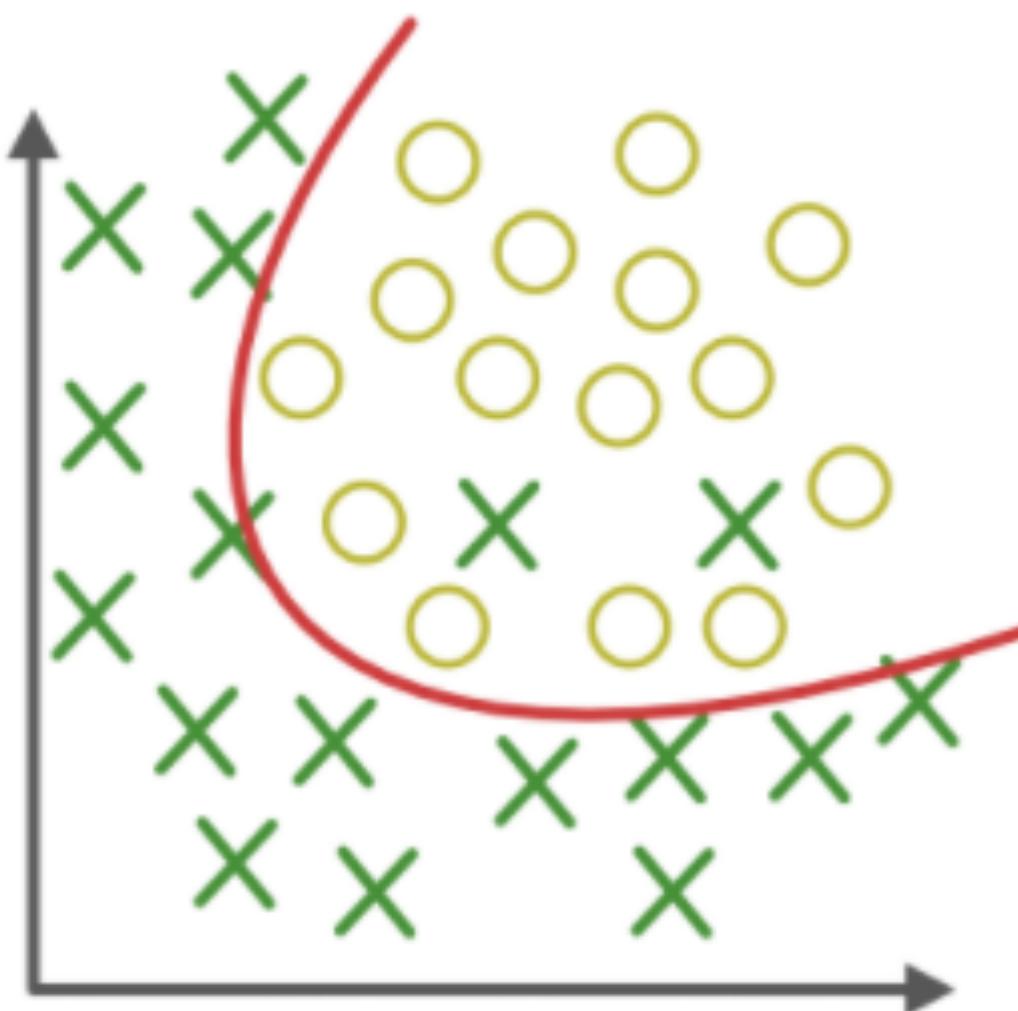


# Overfitting

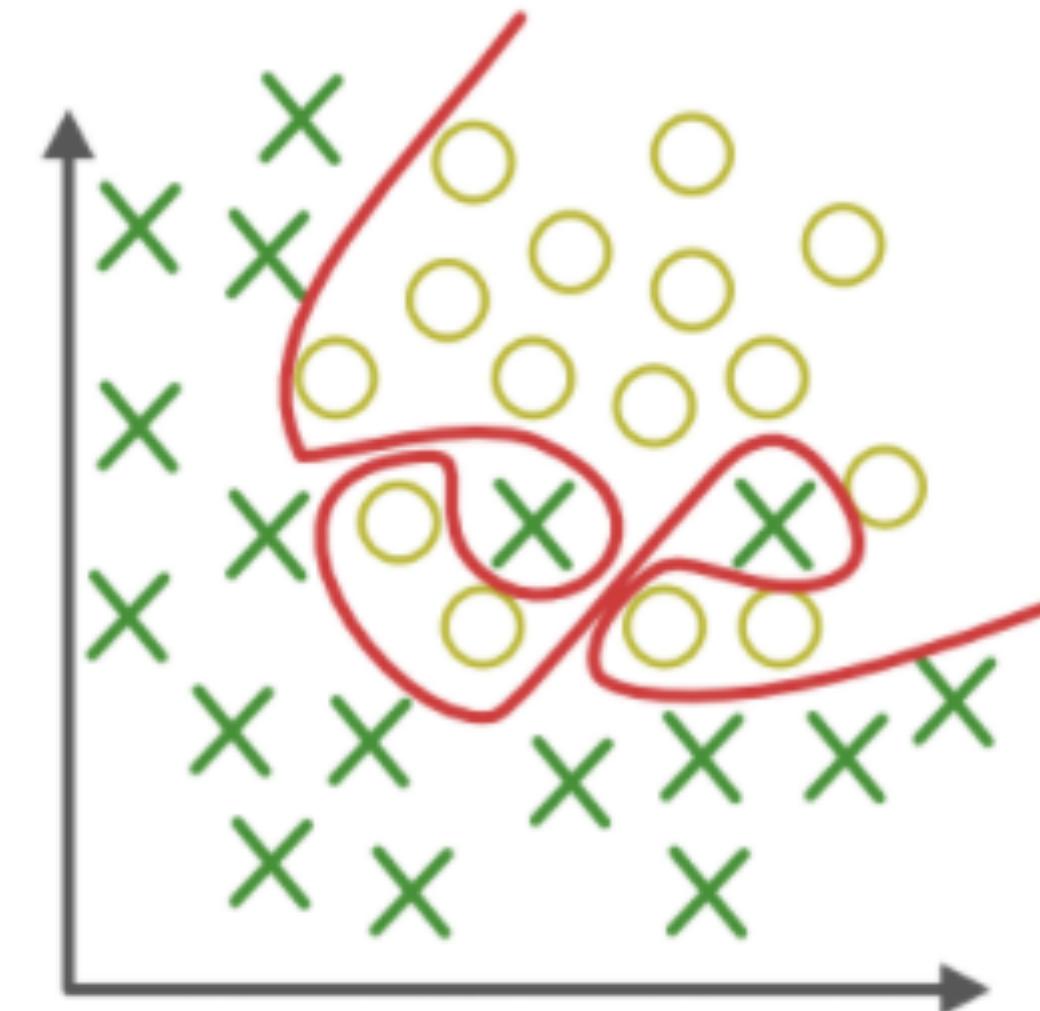


**Under-fitting**

(too simple to explain the variance)



**Appropriate-fitting**



**Over-fitting**

(forcefitting--too good to be true)

DG

# Regularizarion

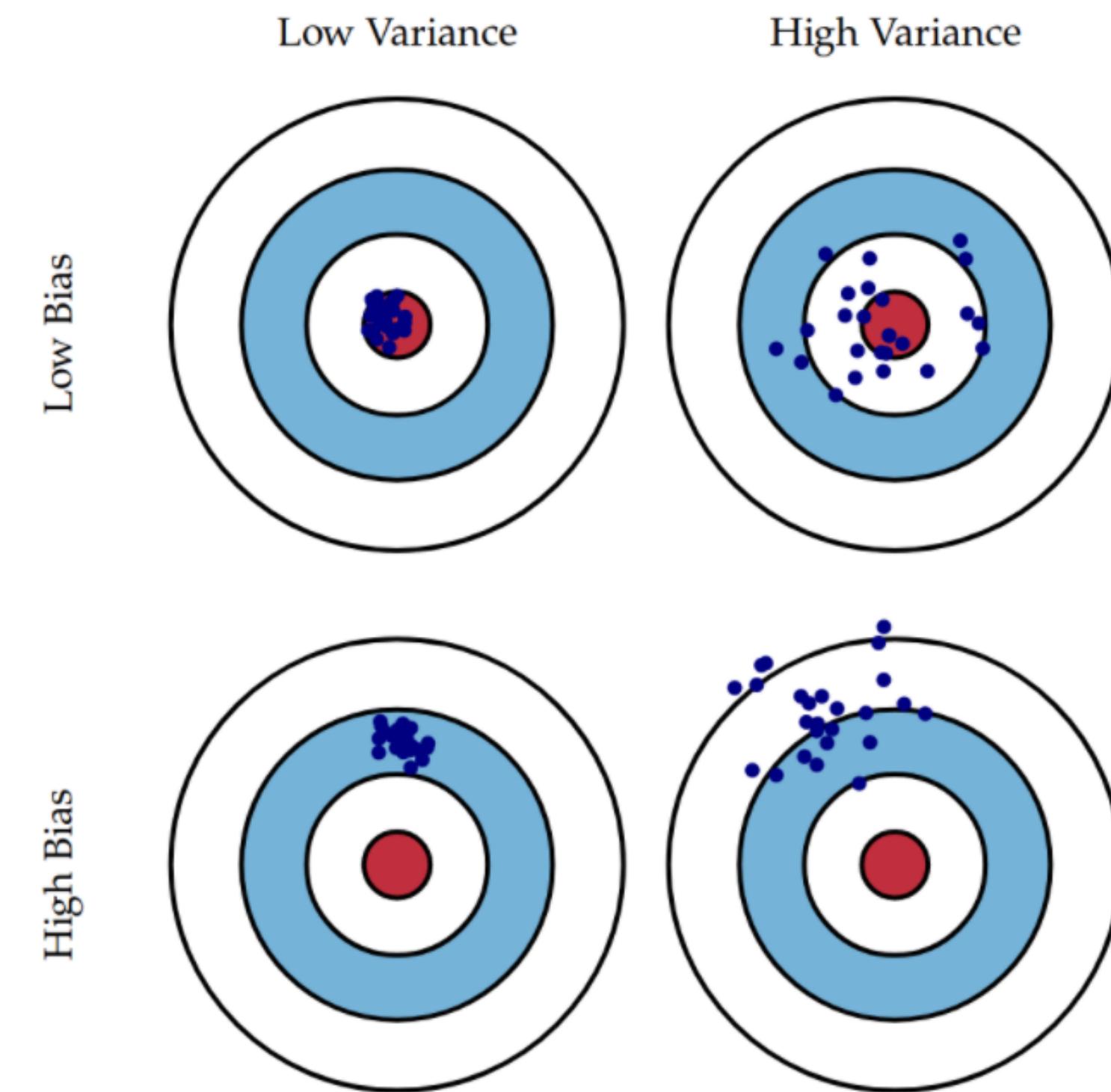
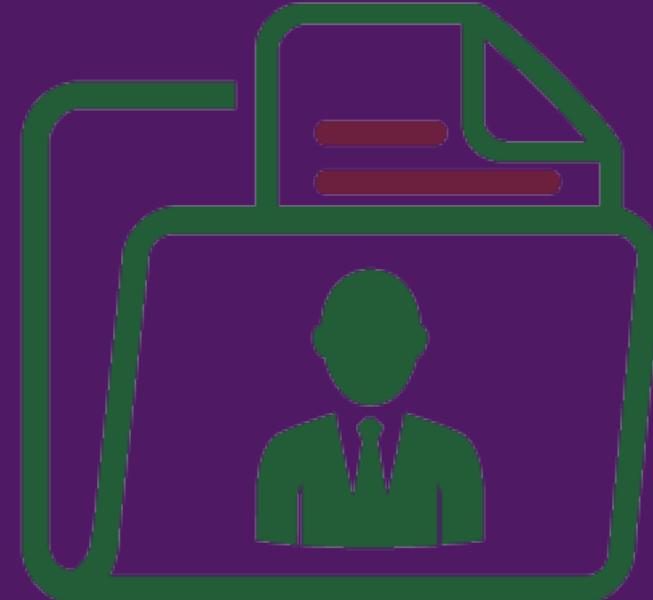


Fig. 1 Graphical illustration of bias and variance.

# DATA IS VALUABLE

WHO  
COLLECTS IT



WHO OWNS IT



IF IT'S  
PROTECTED



WHO'S USING  
IT



# Critical Questions

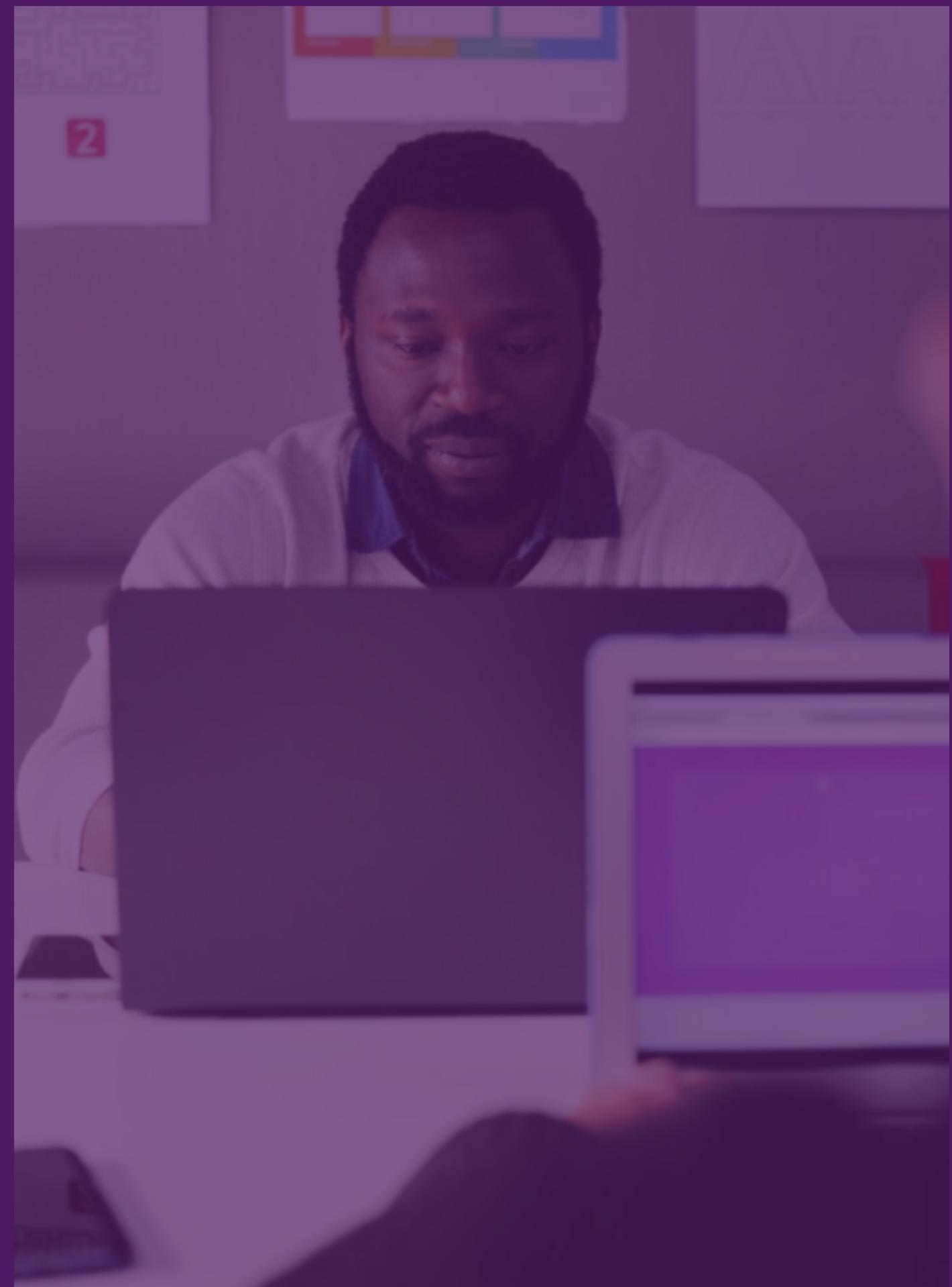
# What does the prediction data look like?



If I don't have  
the data, can I  
get it?



# When will models make predictions?



# What do we do about protected classes?



# How often will this model get feedback?

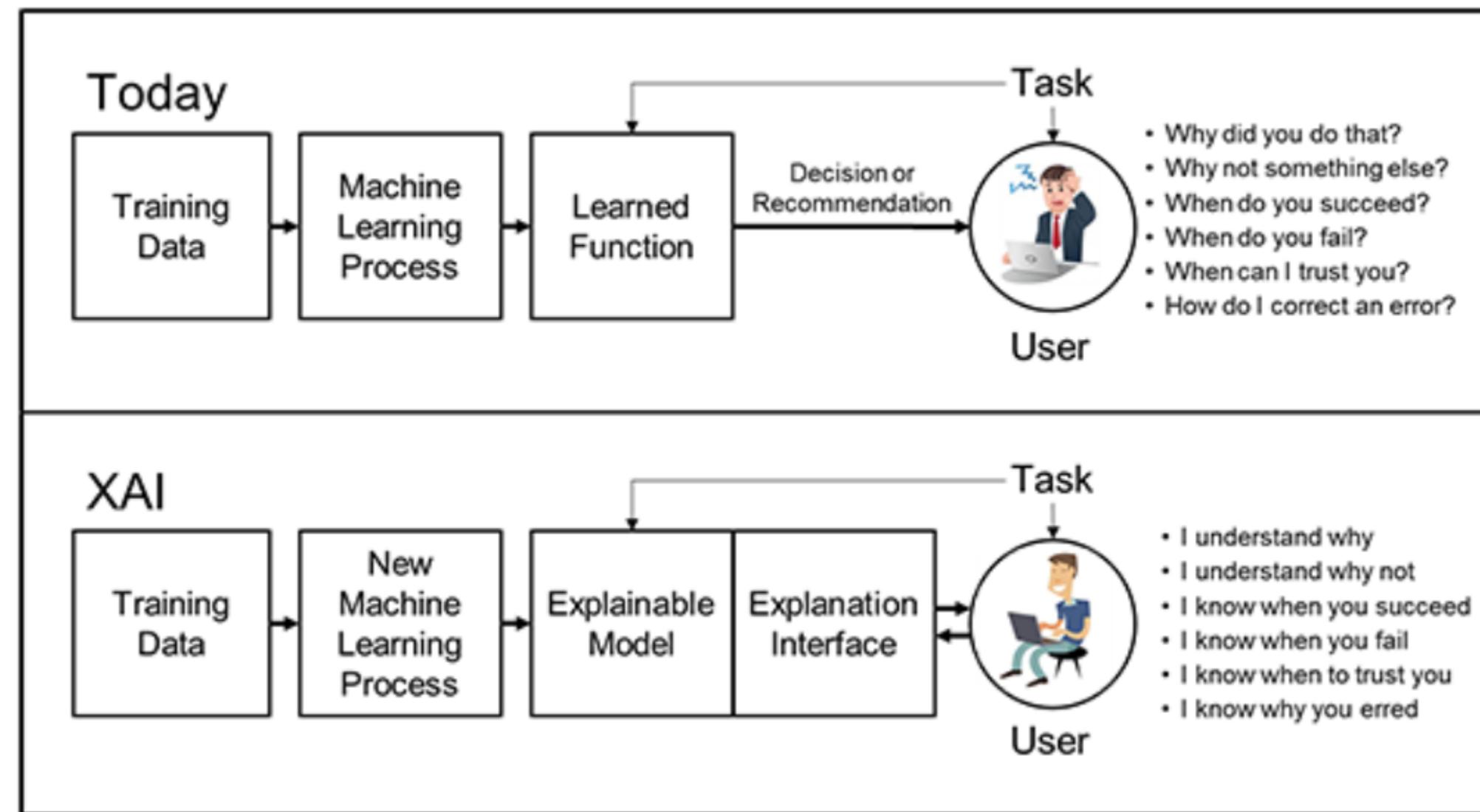


# PICKING A MODEL

## NO FREE LUNCH

- What type of data are my outputs?
- What am I trying to predict
- What is wrong with the data?
- What data cleaning did I do?
- Will this be a problem when running the model in the real world?

# The Future: XAI



# Thank you!



@DataSciBae



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/ayodeleodubela