

CS500-Data Science Tools and Technique

K-Nearest Neighbor (KNN)

A Supervised Machine Learning Algorithm

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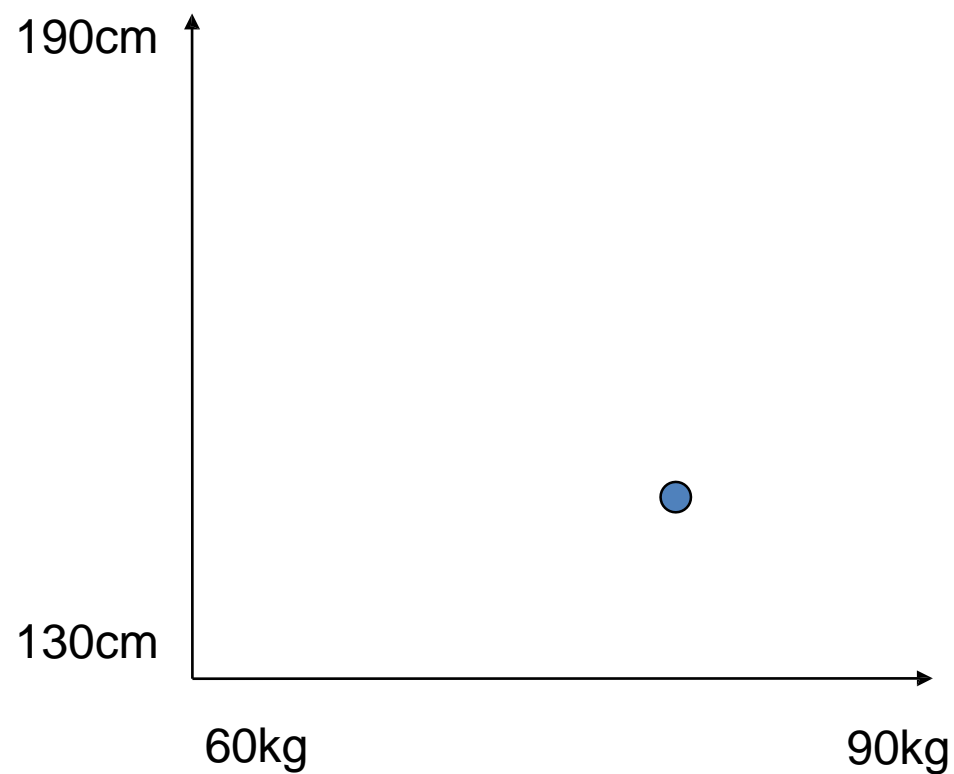
K Nearest Neighbor & Neural Networks

Can we LEARN to recognise a rugby player?

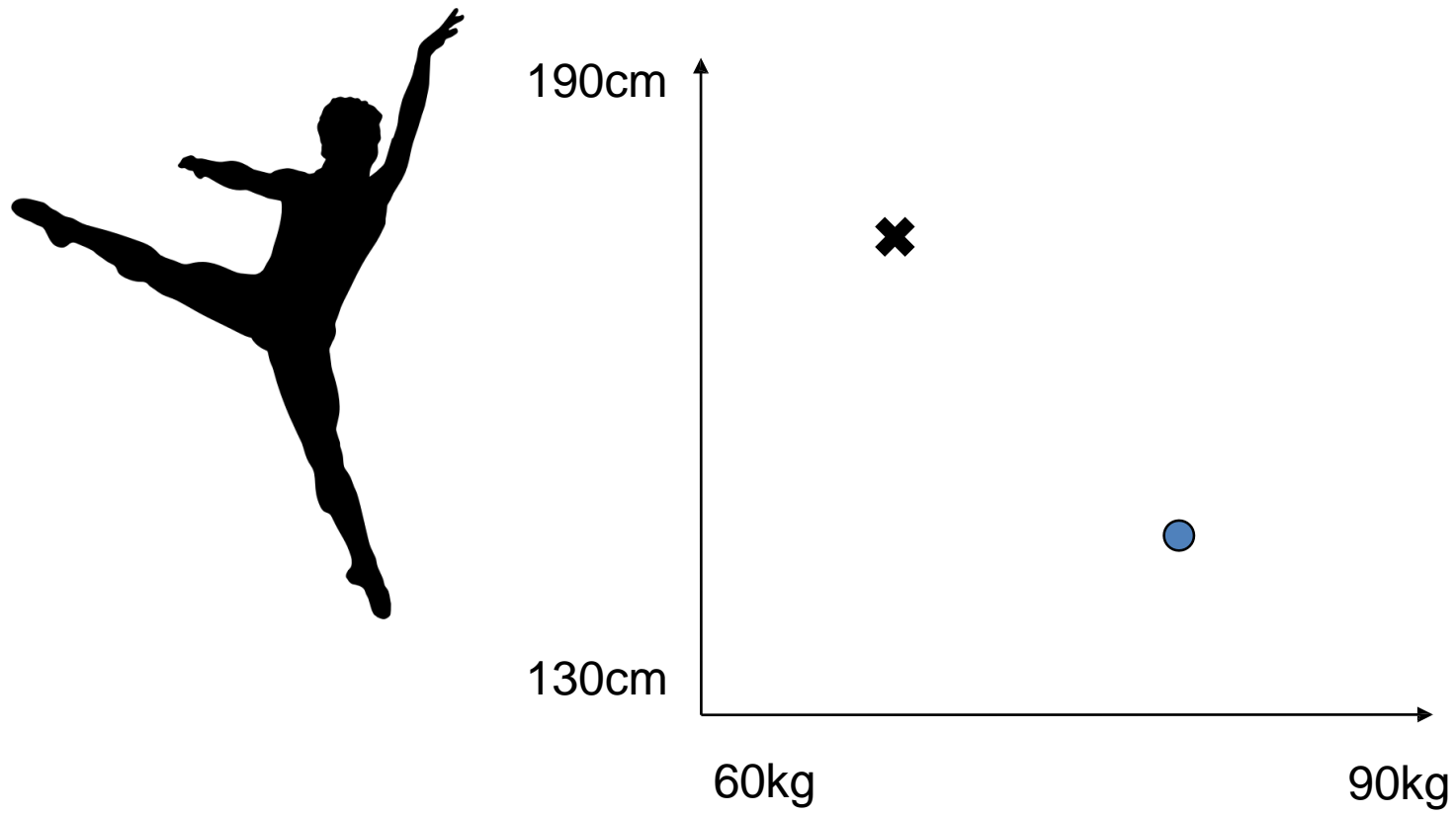


What are the “features” of a rugby player?

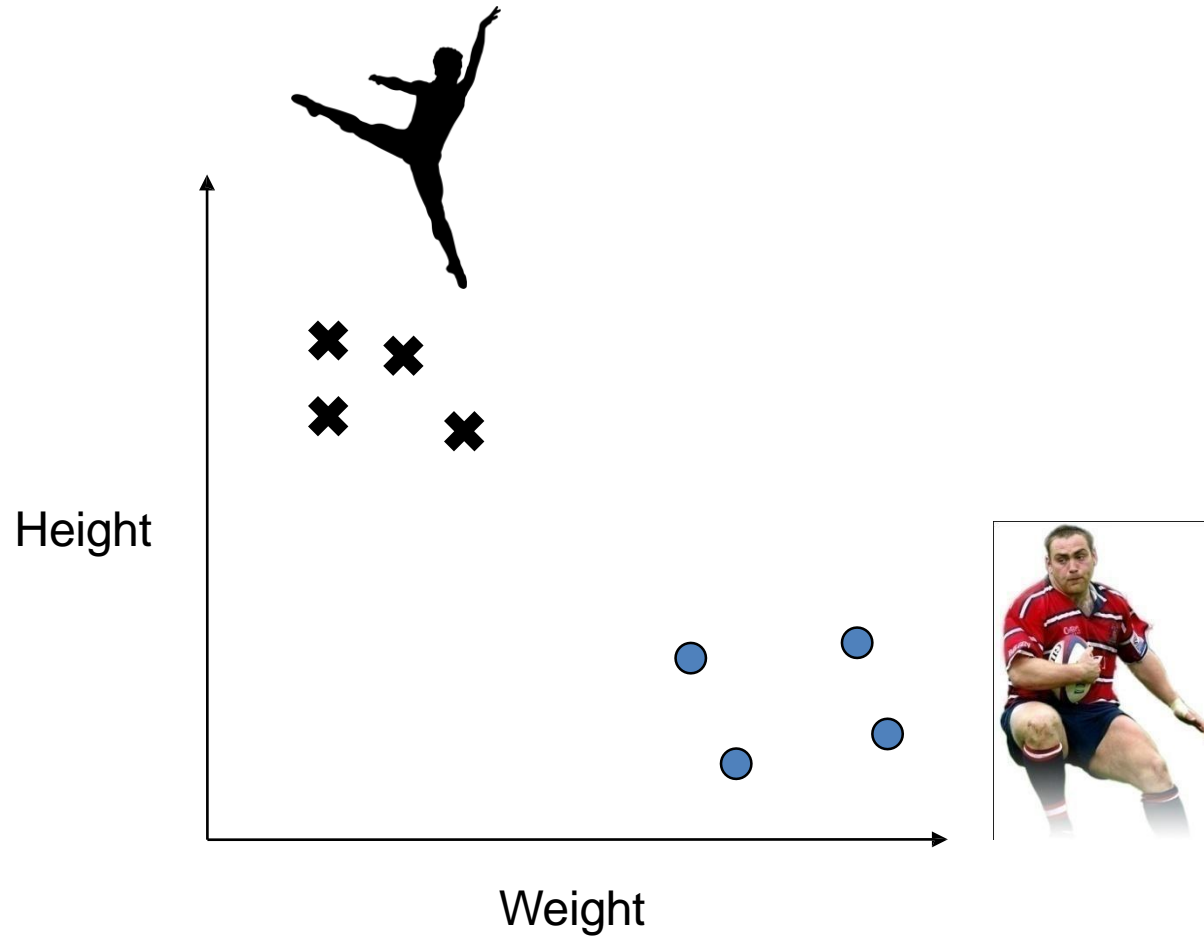
Rugby players = short + heavy?



Ballet dancers = tall + skinny?

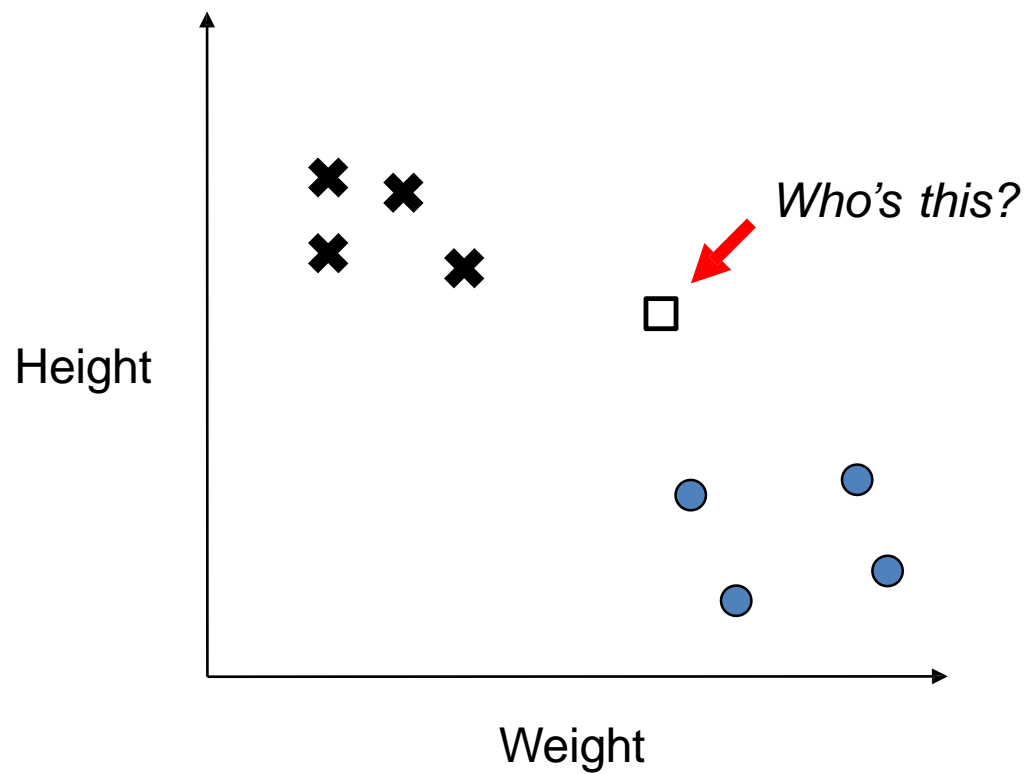


Rugby players “cluster” separately in the space.



K Nearest Neighbors

The K-Nearest Neighbour Algorithm



The K-Nearest Neighbour Algorithm

Step 1: Determine the value for K

Step 2: Calculate the distances between the new input (test data) and all the training data. The most commonly used metrics for calculating distance are Euclidean, Manhattan and Minkowski

Step 3: Sort the distance and determine k nearest neighbors based on minimum distance values

Step 4: Analyze the category of those neighbors and assign the category for the test data based on majority vote

Step 5: Return the predicted class

Distance functions

Euclidean

$$\sqrt{\sum_{i=1}^k (x_i - y_i)^2}$$

Manhattan

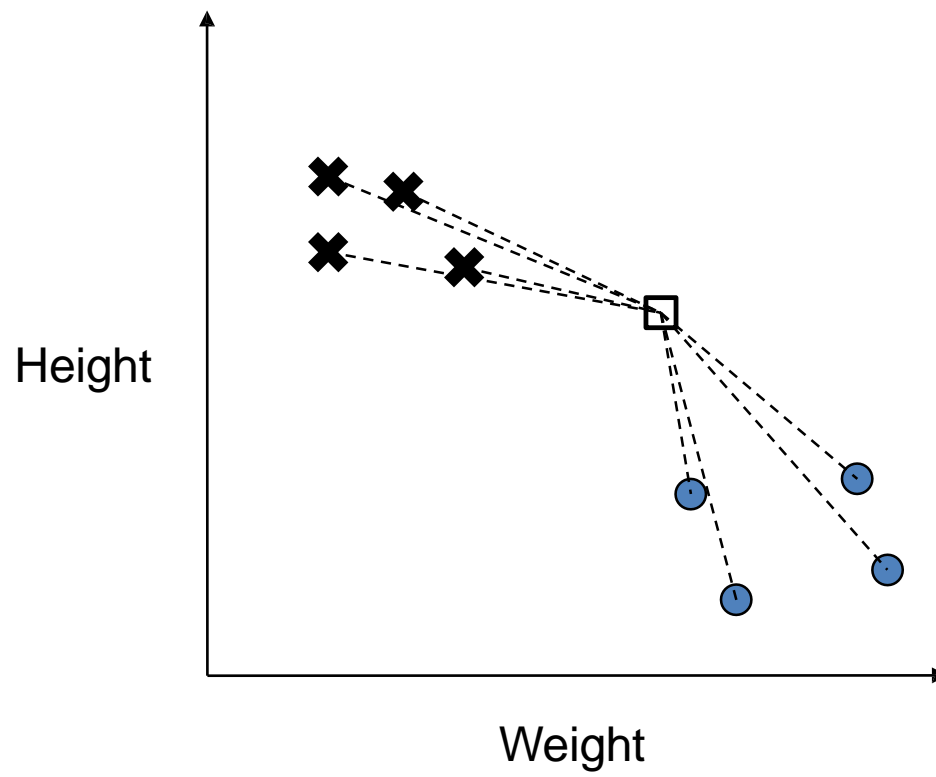
$$\sum_{i=1}^k |x_i - y_i|$$

Minkowski

$$\left(\sum_{i=1}^k (|x_i - y_i|)^q \right)^{1/q}$$

The K-Nearest Neighbour Algorithm

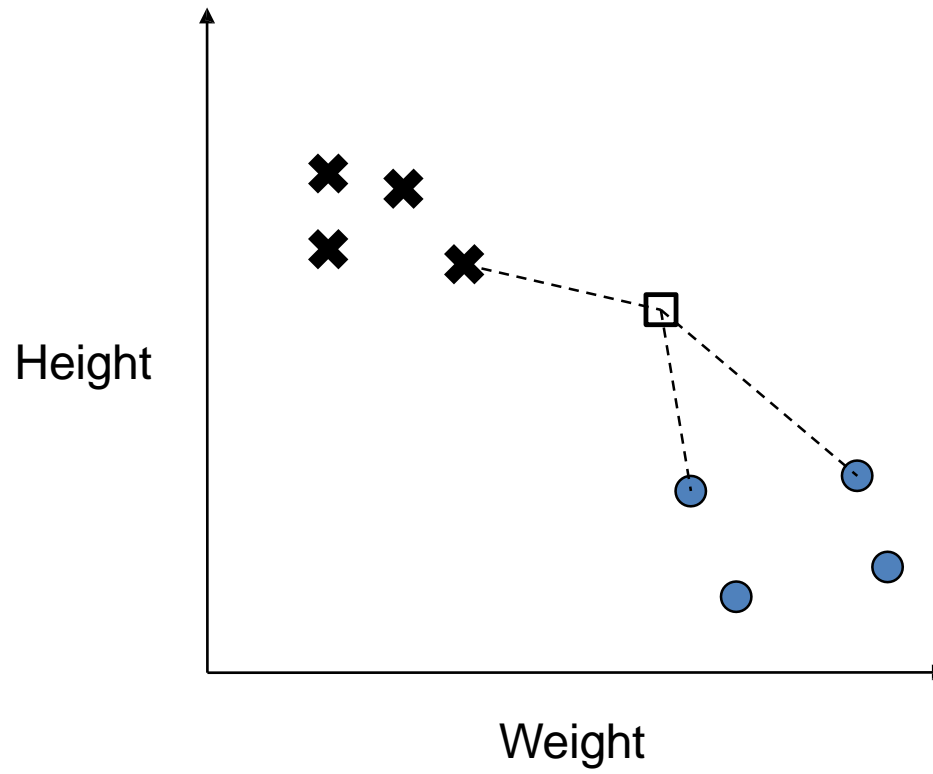
1. *Measure distance to all points*



The K-Nearest Neighbour Algorithm

1. *Measure distance to all points*
2. *Find closest “k” points*

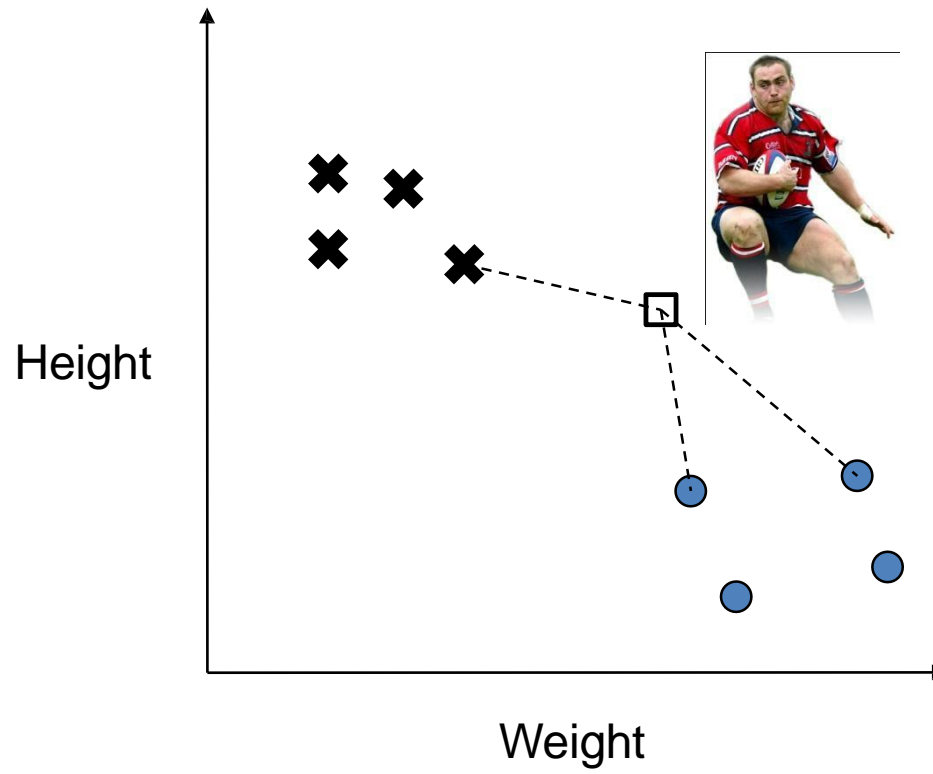
← (here $k=3$, but it could be more)



The K-Nearest Neighbour Algorithm

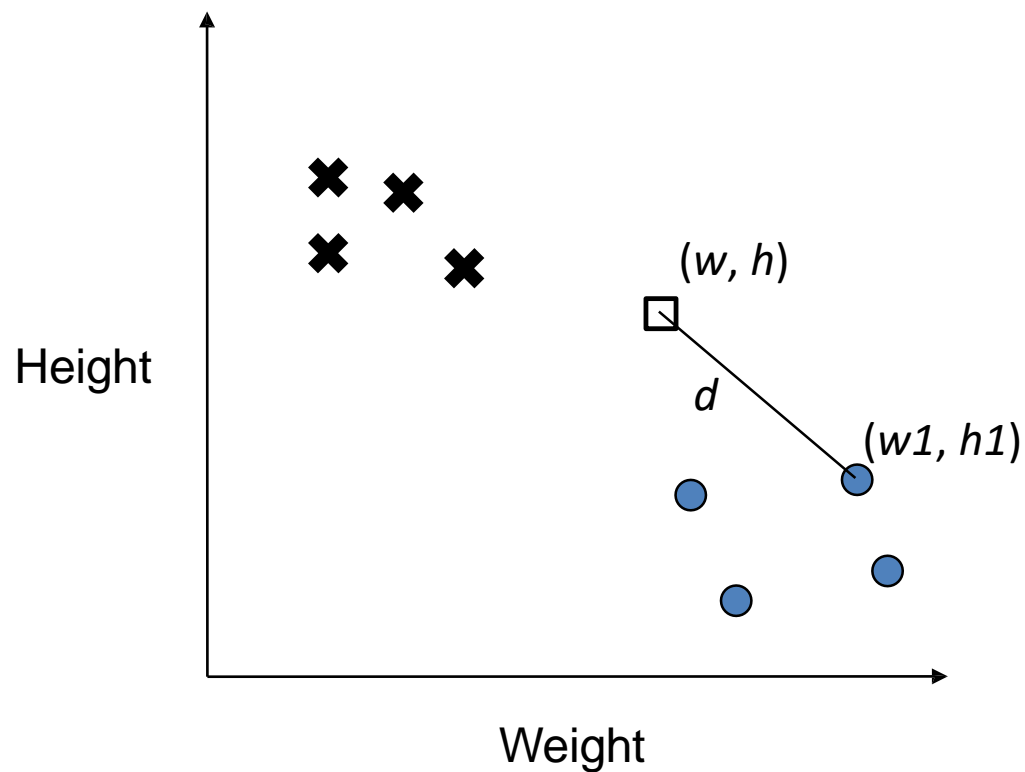
1. *Measure distance to all points*
2. *Find closest “k” points*
3. *Assign majority class*

← (here $k=3$, but it could be more)



“Euclidean distance”

$$d = \sqrt{(w - w_1)^2 + (h - h_1)^2}$$



Euclidean distance still works in 3-d, 4-d, 5-d, etc....

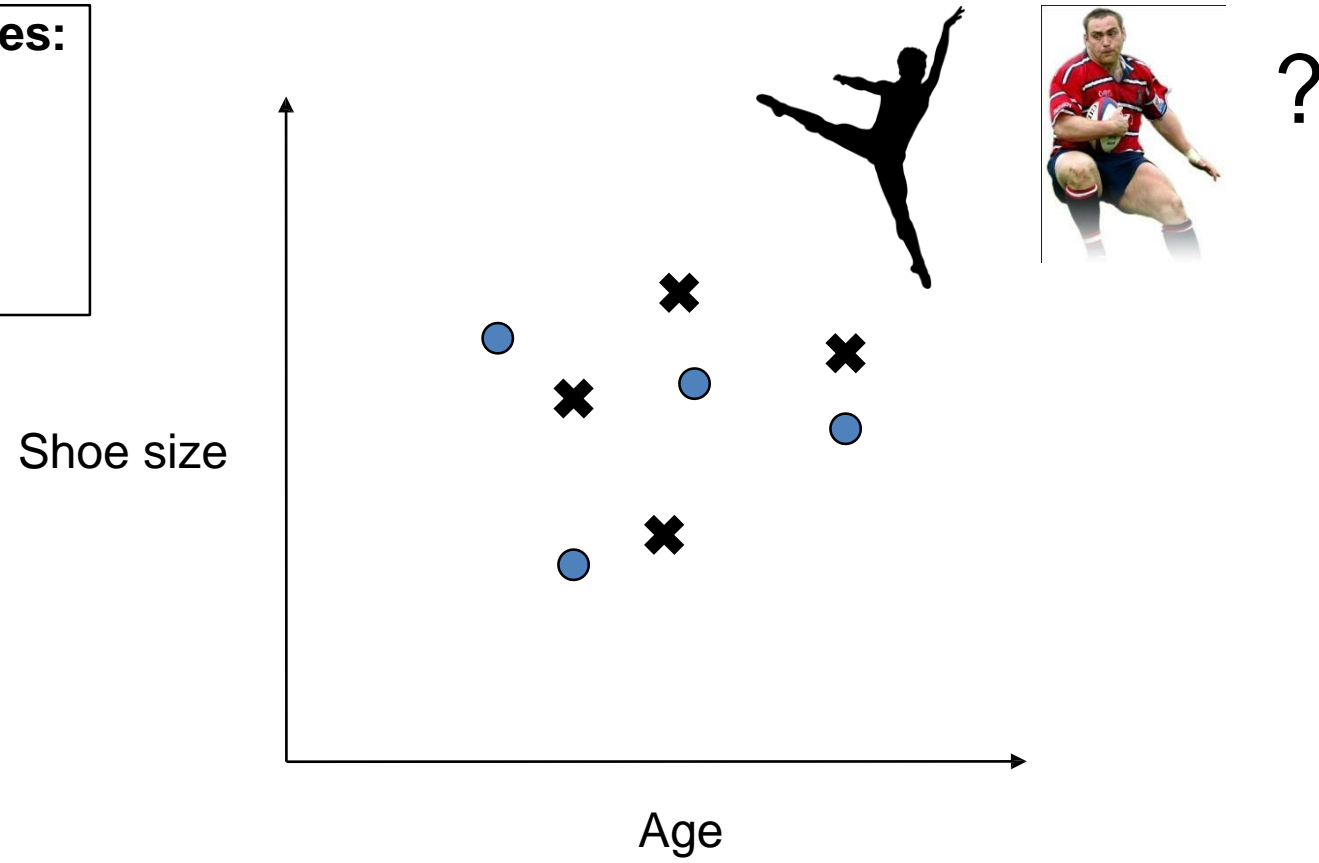
$$d = \sqrt{(x - x_1)^2 + (y - y_1)^2 + (z - z_1)^2}$$

$x = \text{Height}$
 $y = \text{Weight}$
 $z = \text{Shoe size}$

Choosing the wrong features makes it difficult,
too many and it's computationally intensive.

Possible features:

- Shoe size ✓
- Height
- Age ✓
- Weight



K-Nearest Neighbour Model

- Example : Classify whether a customer will respond to a survey question using a 3-Nearest Neighbor classifier

Customer	Age	Income	No. credit cards	Response
John	35	35K	3	No
Rachel	22	50K	2	Yes
Hannah	63	200K	1	No
Tom	59	170K	1	No
Nellie	25	40K	4	Yes
David	37	50K	2	?

K-Nearest Neighbour Model

- Example : 3-Nearest Neighbors

Customer	Age	Income	No. credit cards	Response
John	35	35K	3	No
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Hannah	63	200K	1	No
Tom	59	170K	1	No
Nellie	25	40K	4	Yes
David	37	50K	2	?

Distances from David to his 3 nearest neighbors:

- Nellie: 15.74
- Rachel: 122
- Hannah: 152.23

Distance from David to John: 15.16

Distance from David to Tom: 15

Distance from David to David: 15

K-Nearest Neighbour Model

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Distances from David to other customers:

- David to Nellie: 15.74
- David to Tom: 122
- David to Hannah: 152.23
- David to Rachel: 15.16

Three nearest ones to David are: No, Yes, Yes

K-Nearest Neighbour Model

- Example : 3-Nearest Neighbors

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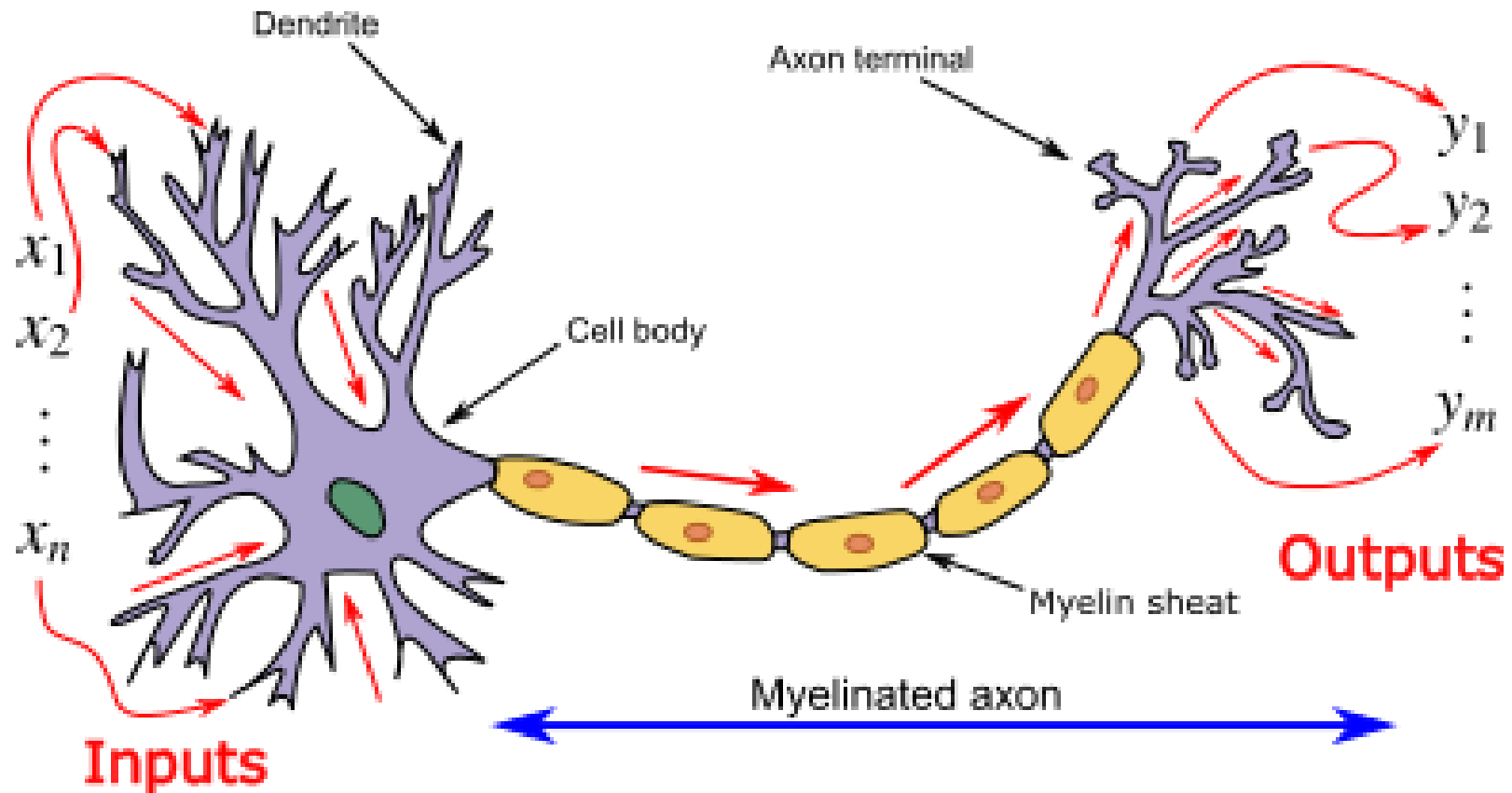
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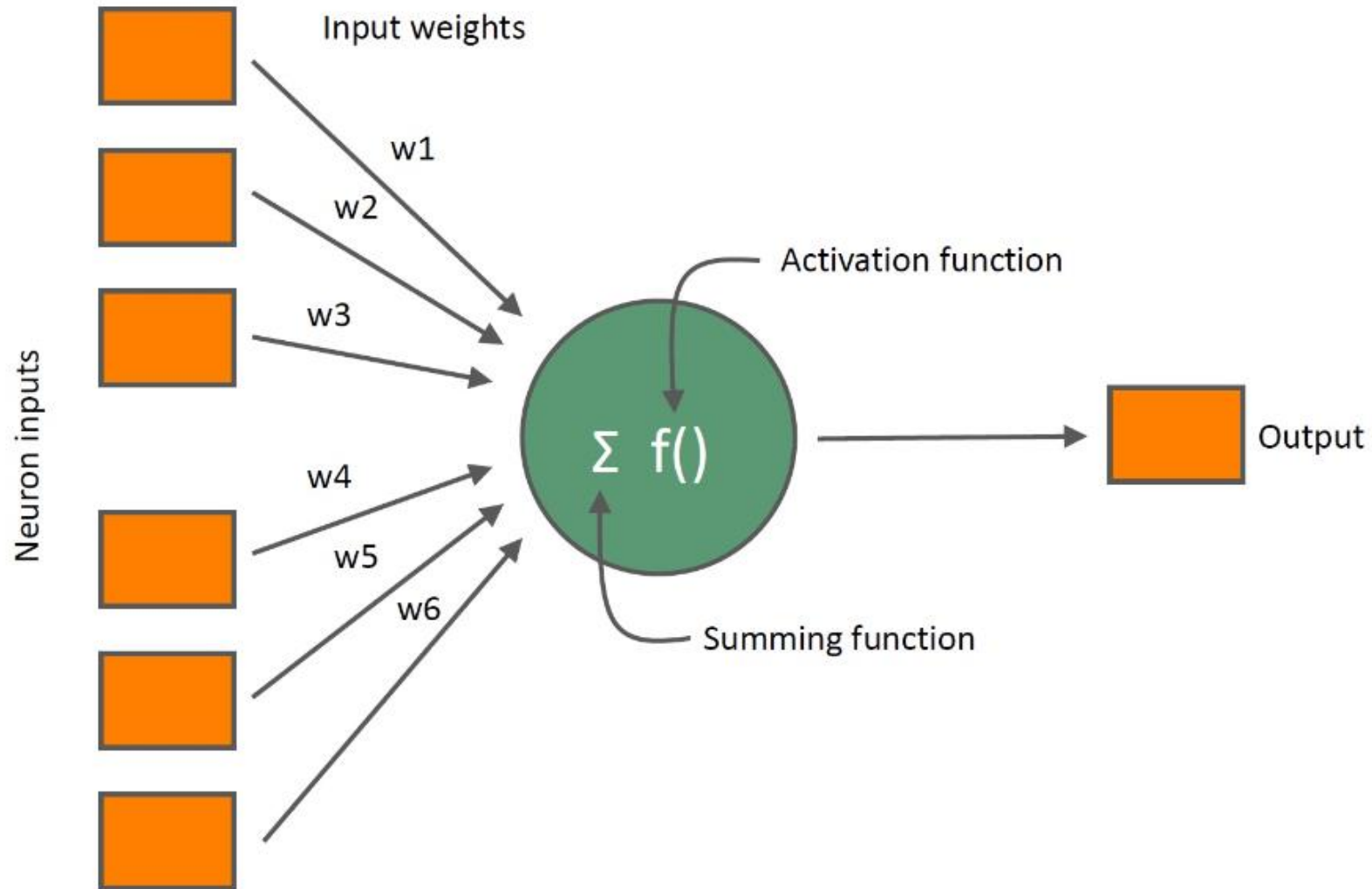
Three nearest ones to David are: No, Yes, Yes

Artificial Neural Network

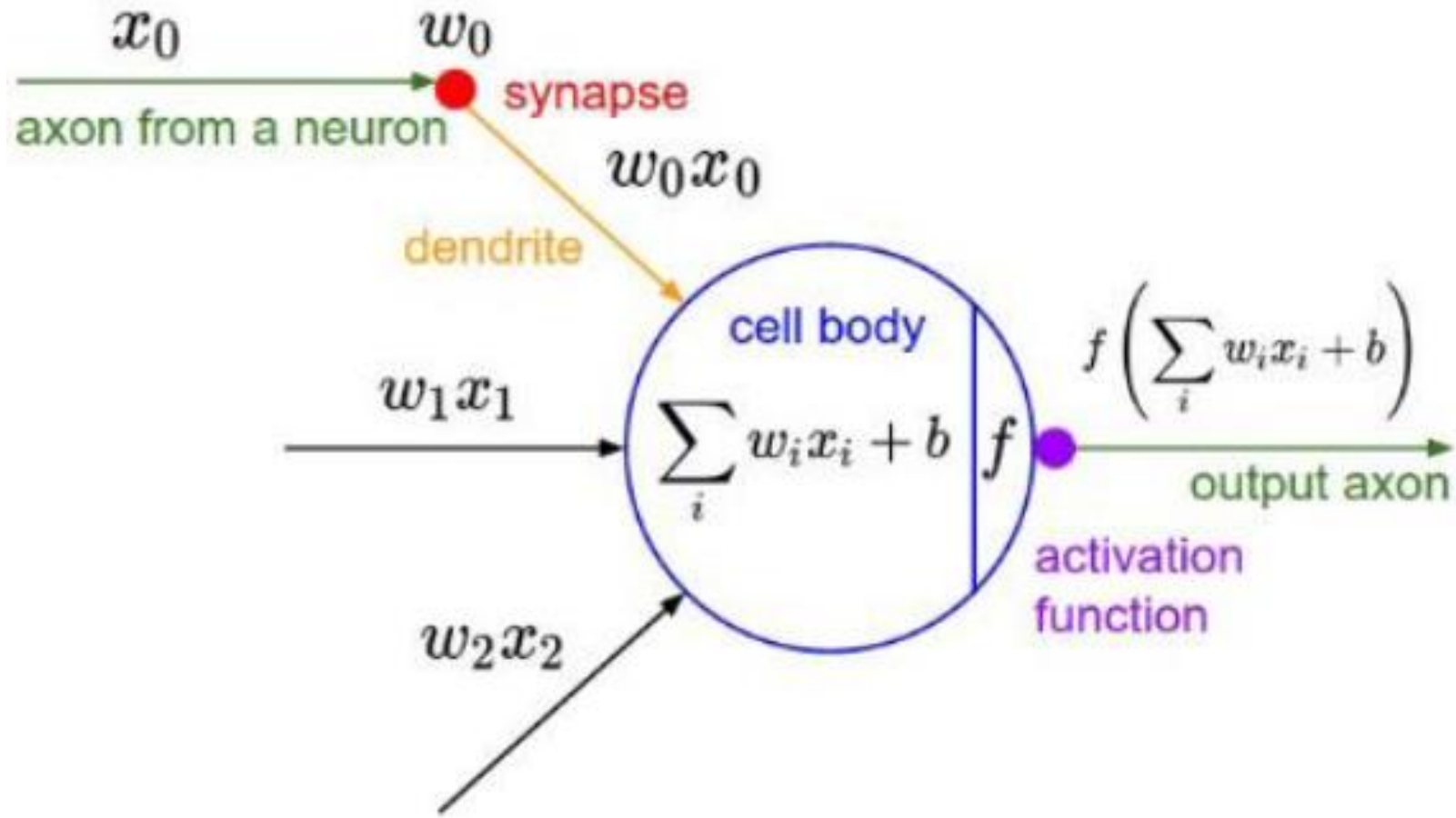
Biological Neuron/ Biological Neural Network



Artificial Neuron/ Artificial Neural Network



Artificial Neuron/ Artificial Neural Network



Artificial Neural Network Layers

