



The image is a 3D visualization of a network graph. It features a central circular area with a light gray background, containing the text "K-NEAREST NEIGHHORS" in a bold, black, sans-serif font. Surrounding this central area is a complex network of nodes and edges. The nodes are represented by spheres of various sizes and colors, including white, orange, red, and blue. The edges are thin, white lines connecting the nodes, creating a dense web of connections. The background is a gradient of light pink and blue, with the pink area on the left and the blue area on the right. The overall effect is a modern, high-tech aesthetic.

K-NEAREST NEIGHHORS

KNN Classification

Theory and Intuition

KNN

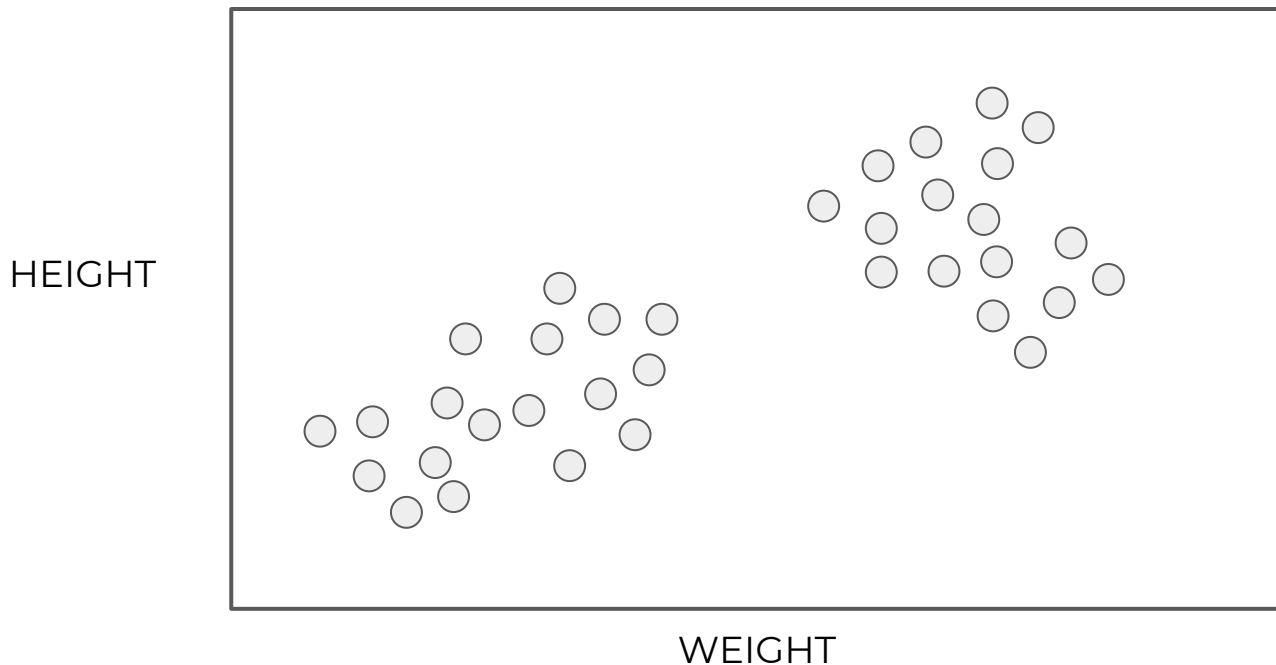
- K nearest neighbors is one of the simplest machine learning algorithms.
- It simply assigns a label to new data based on the **distance** between the old data and new data.
- Let's go through the intuition with an example use case...

KNN

- Sexing chicks is still a very manual process:
 - en.wikipedia.org/wiki/Chick_sexing
- Let's imagine we gathered a dataset of baby chick heights and weights.
- How could we train an algorithm to identify the sex of a new baby chick based on historical features?

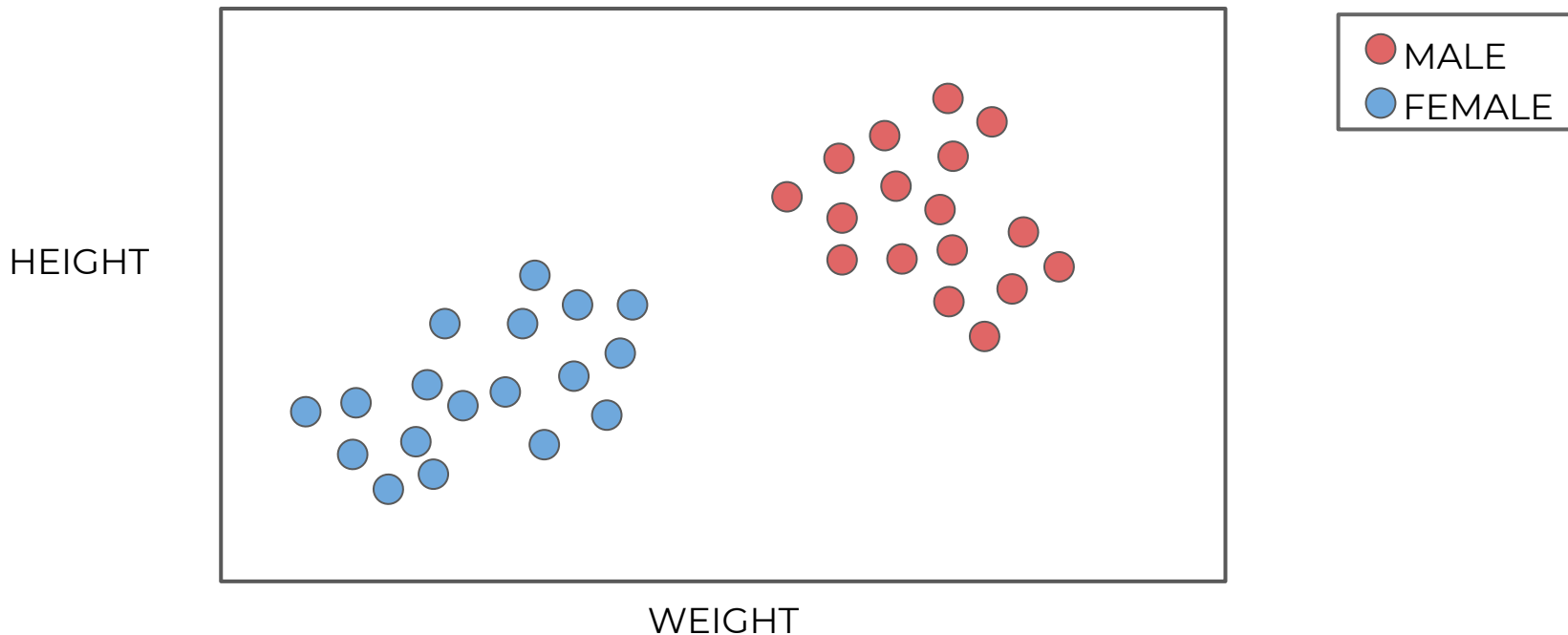
KNN

- Imagine a height and weight data set



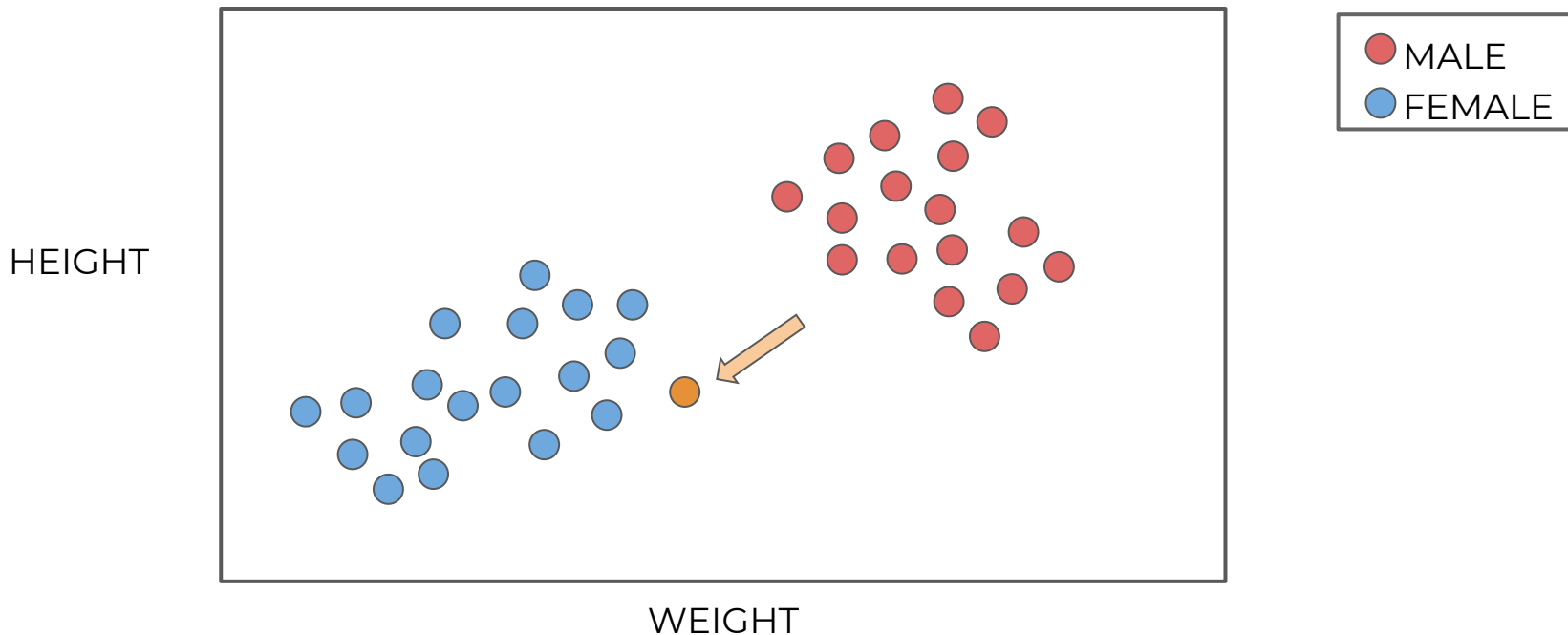
KNN

- We historically know the sex of the chicks:



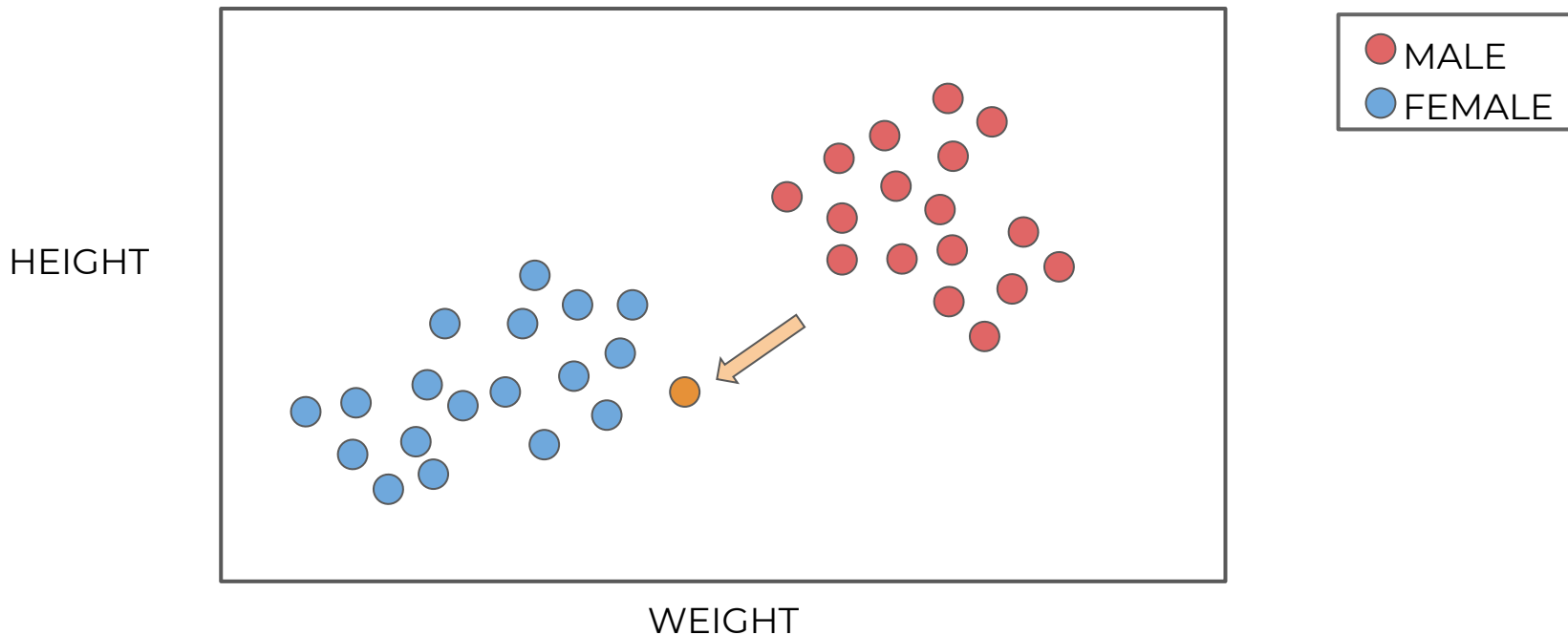
KNN

- How would we assign sex to a new point?



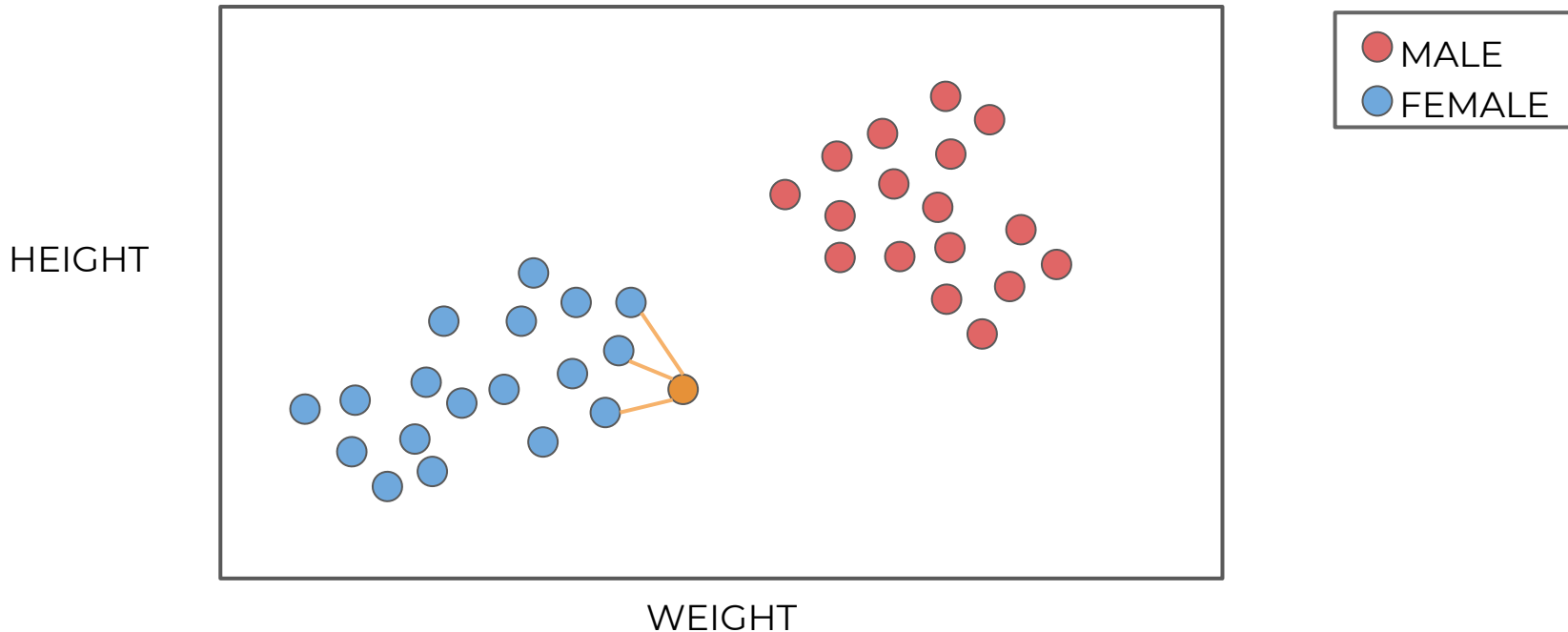
KNN

- We intuitively “know” this is likely female.



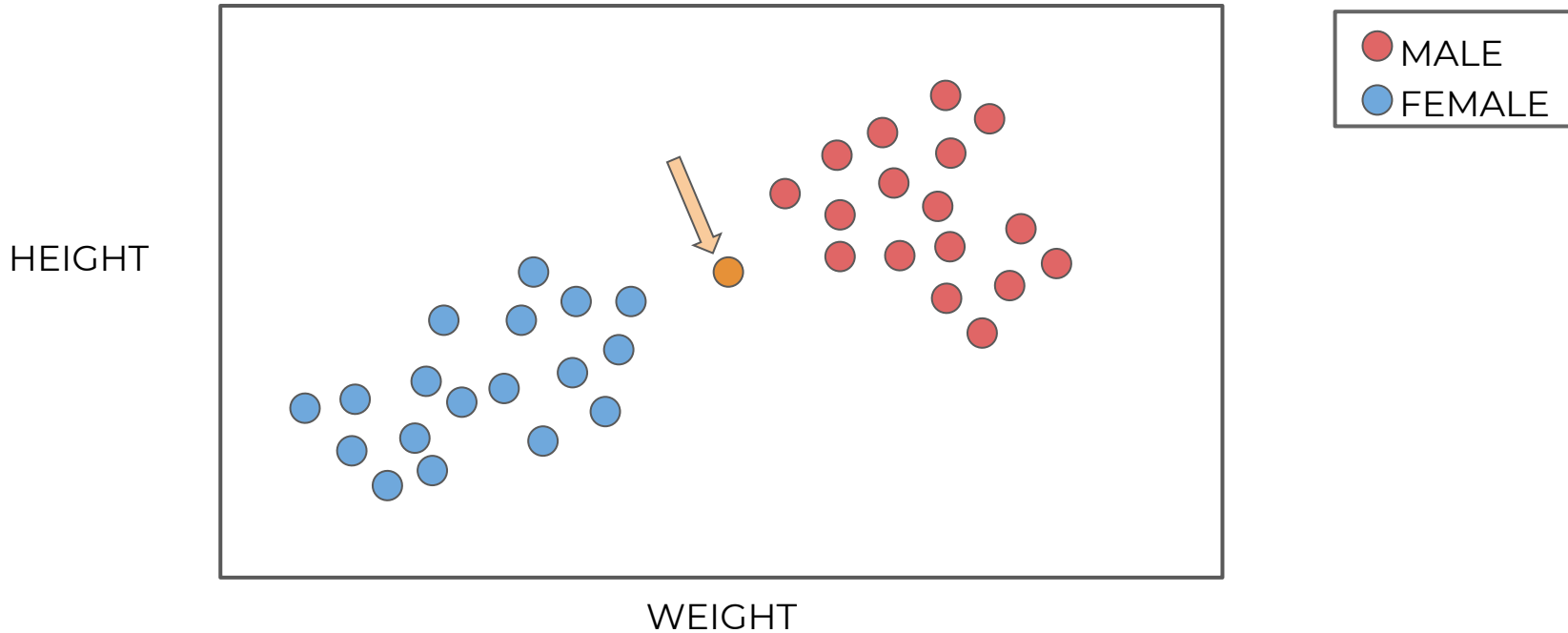
KNN

- Intuition comes from **distance** to points!



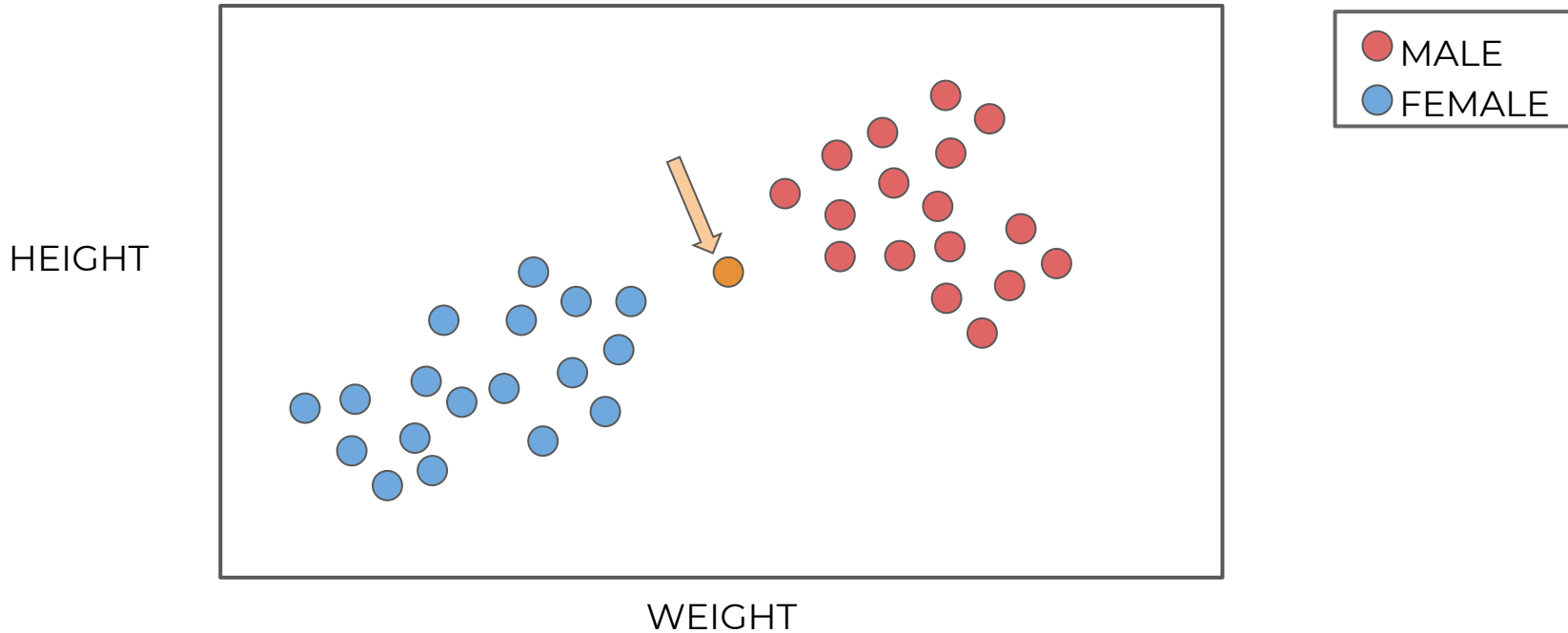
KNN

- What about a less obvious point?



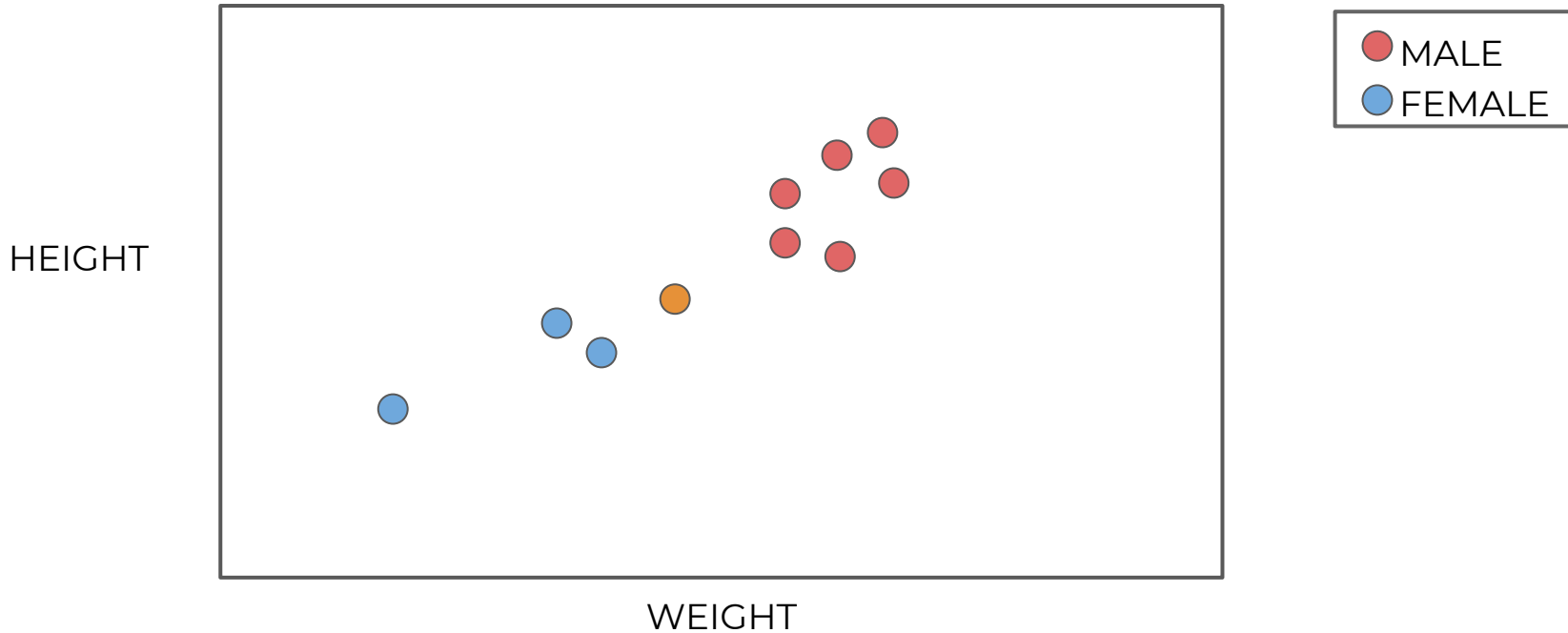
KNN

- How many points to we consider?



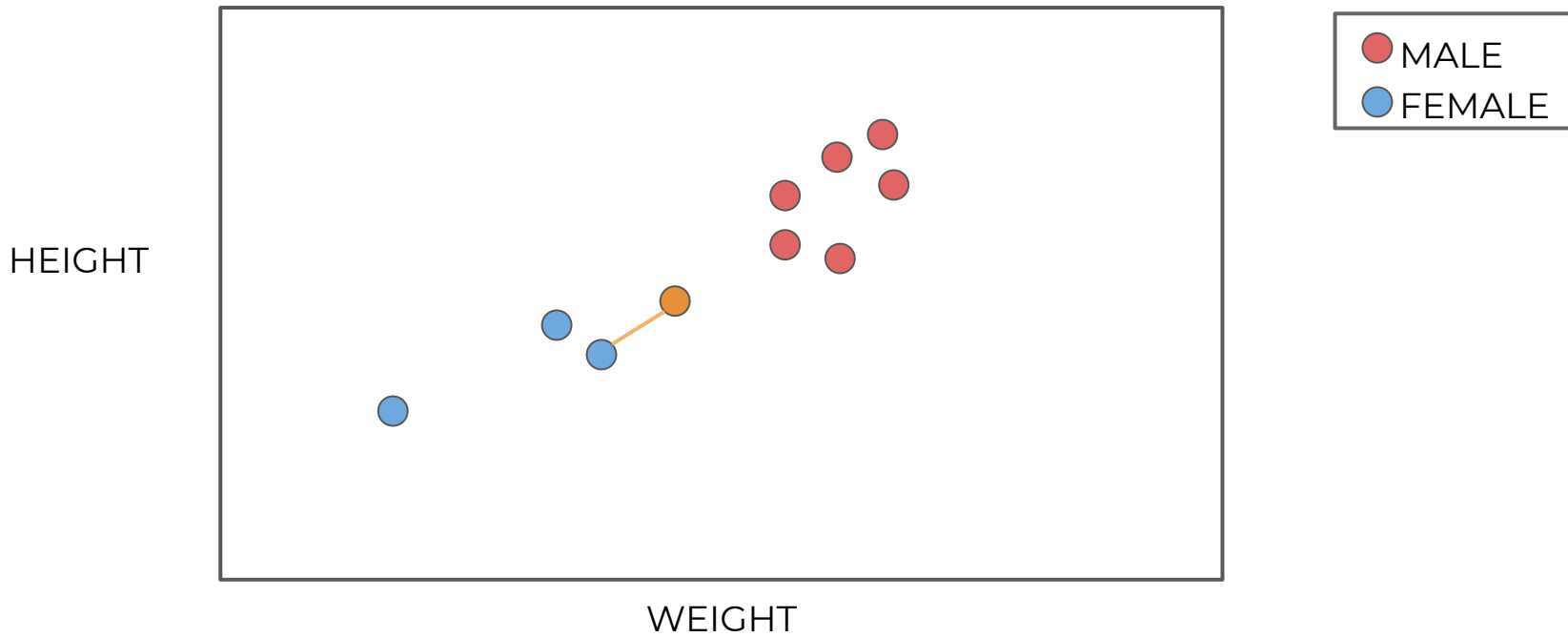
KNN

- Let's imagine a situation like this:



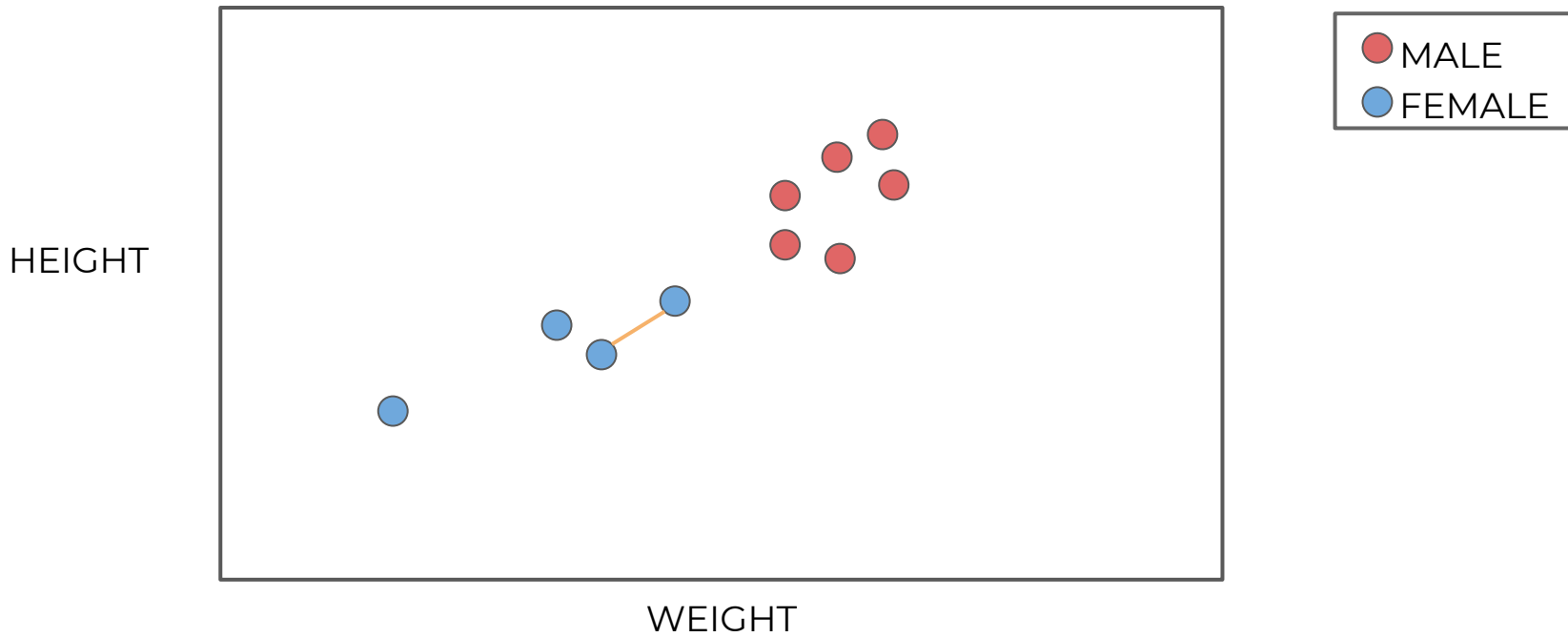
KNN

- $K=1$



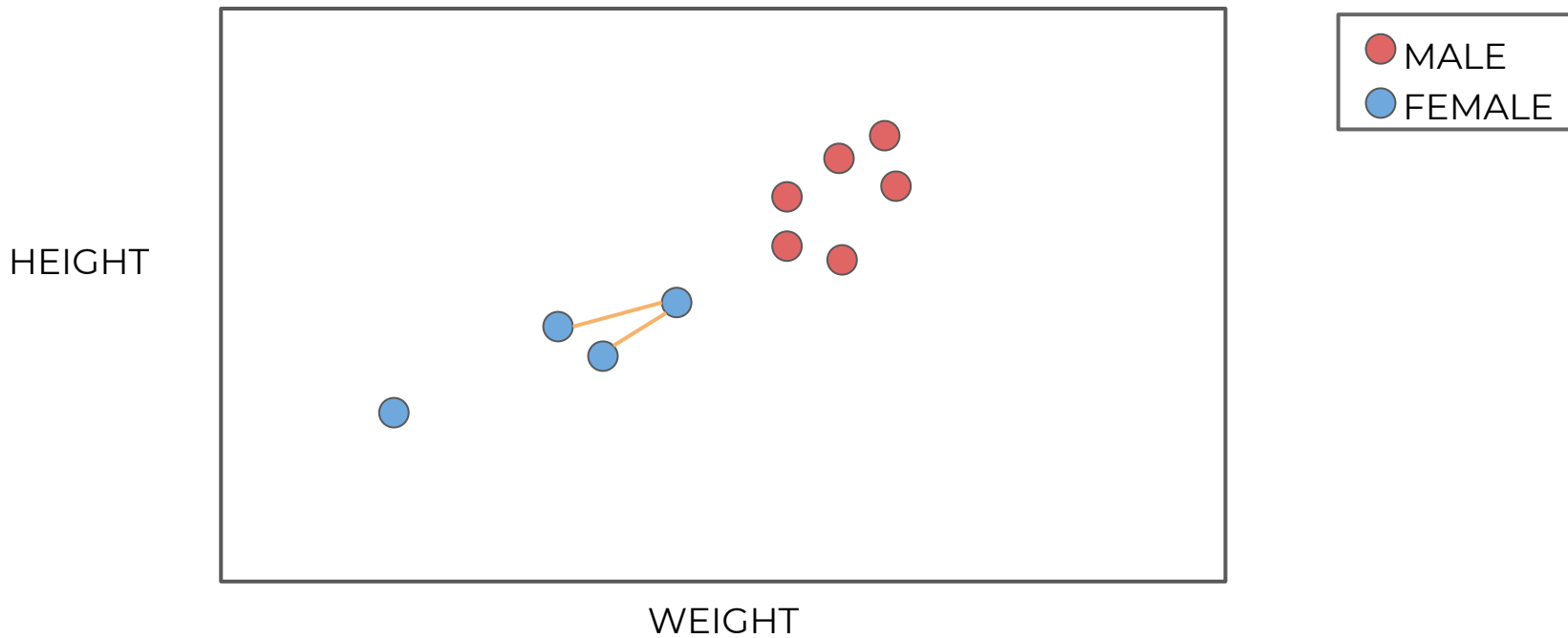
KNN

- $K=1$



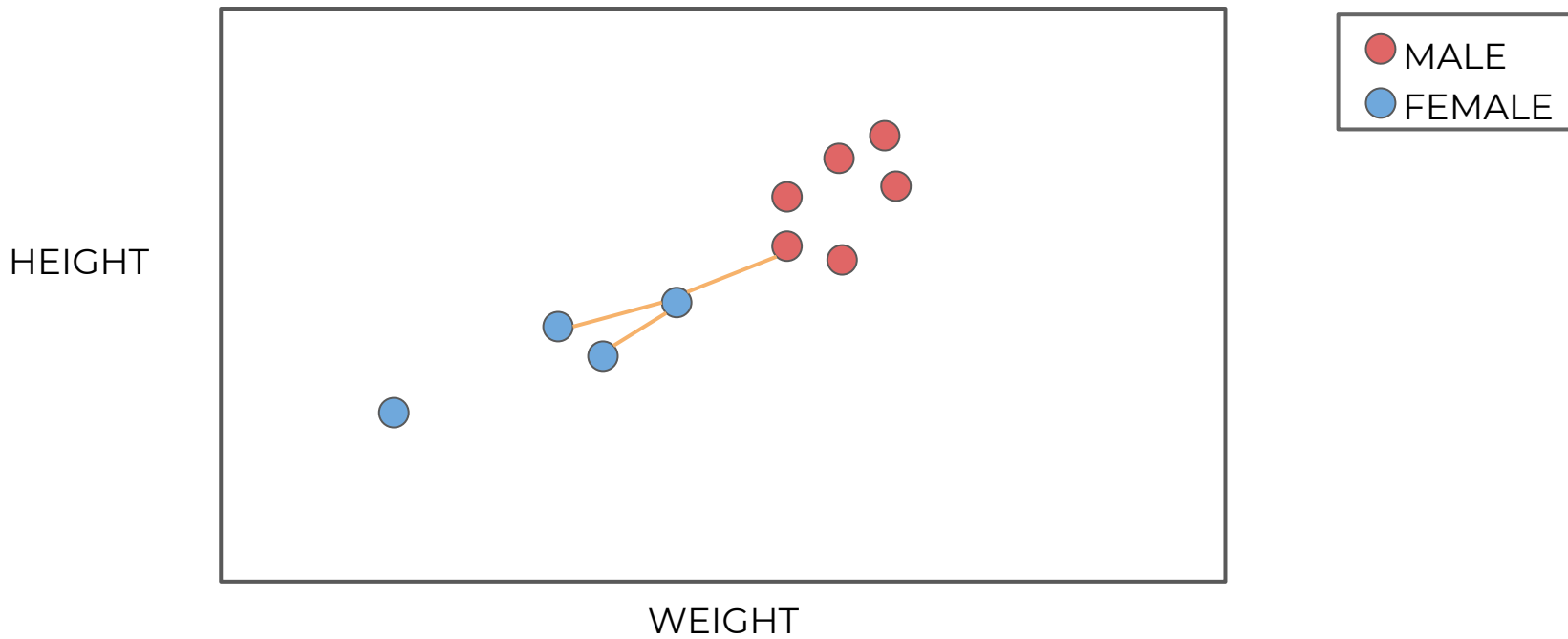
KNN

- $K=2$



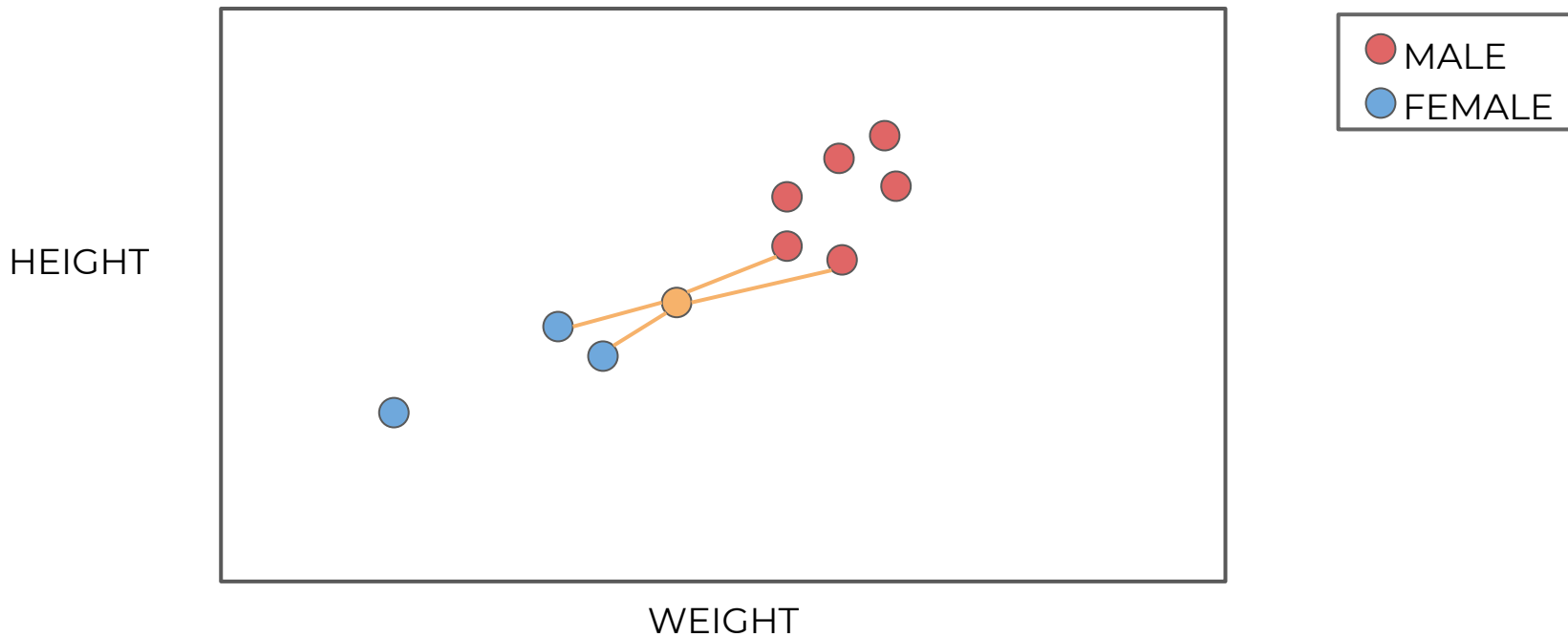
KNN

- $K=3$



KNN

- K=4 leads to a tie!



KNN

- Tie considerations and options:
 - Always choose an odd K .
 - In case of tie, simply reduce K by 1 until tie is broken.
 - Randomly break tie.
 - Choose nearest class point.

KNN

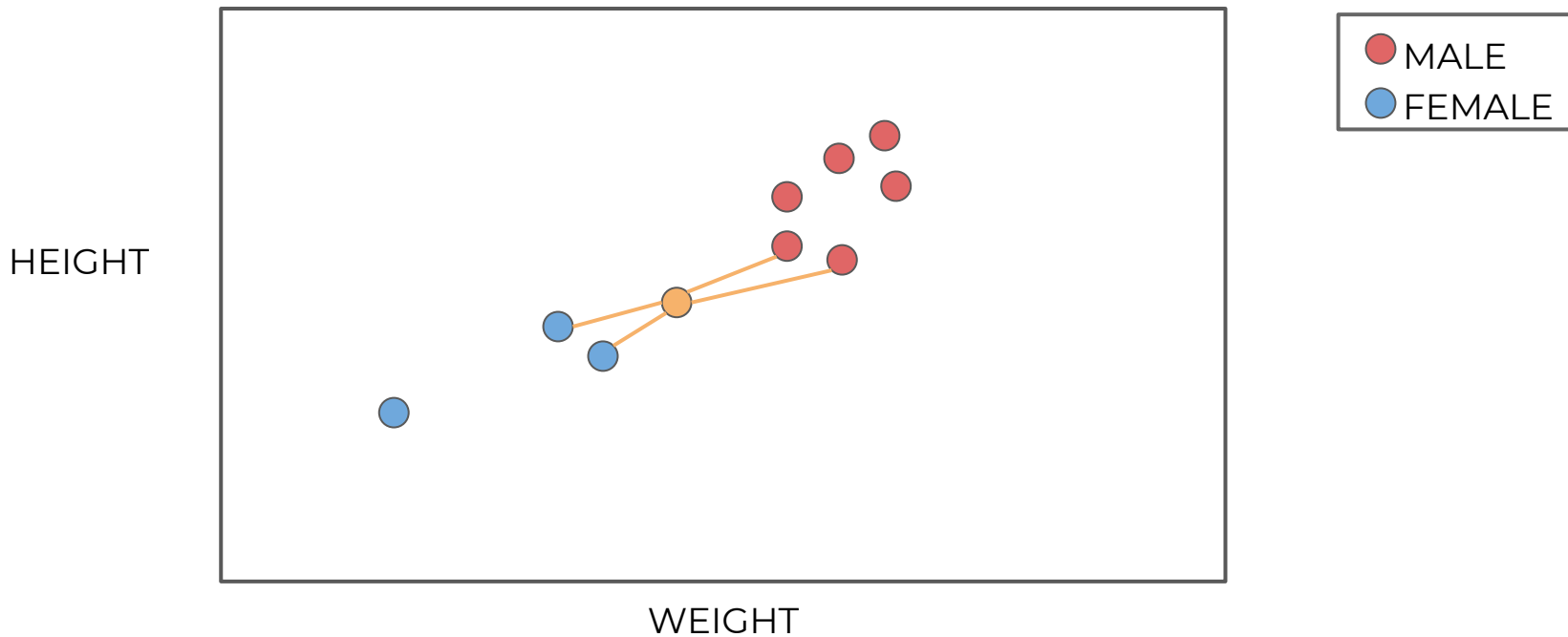
- What does Scikit-Learn do in case of tie?
 - *Warning: Regarding the Nearest Neighbors algorithms, if it is found that two neighbors, neighbor $k+1$ and k , have identical distances but different labels, the results will depend on the ordering of the training data.*

KNN

- What does Scikit-Learn do in case of tie?
 - *In the case of ties, the answer will be the class that happens to appear first in the set of neighbors.*
 - *Results are ordered by distance, so it chooses the class of the closest point.*

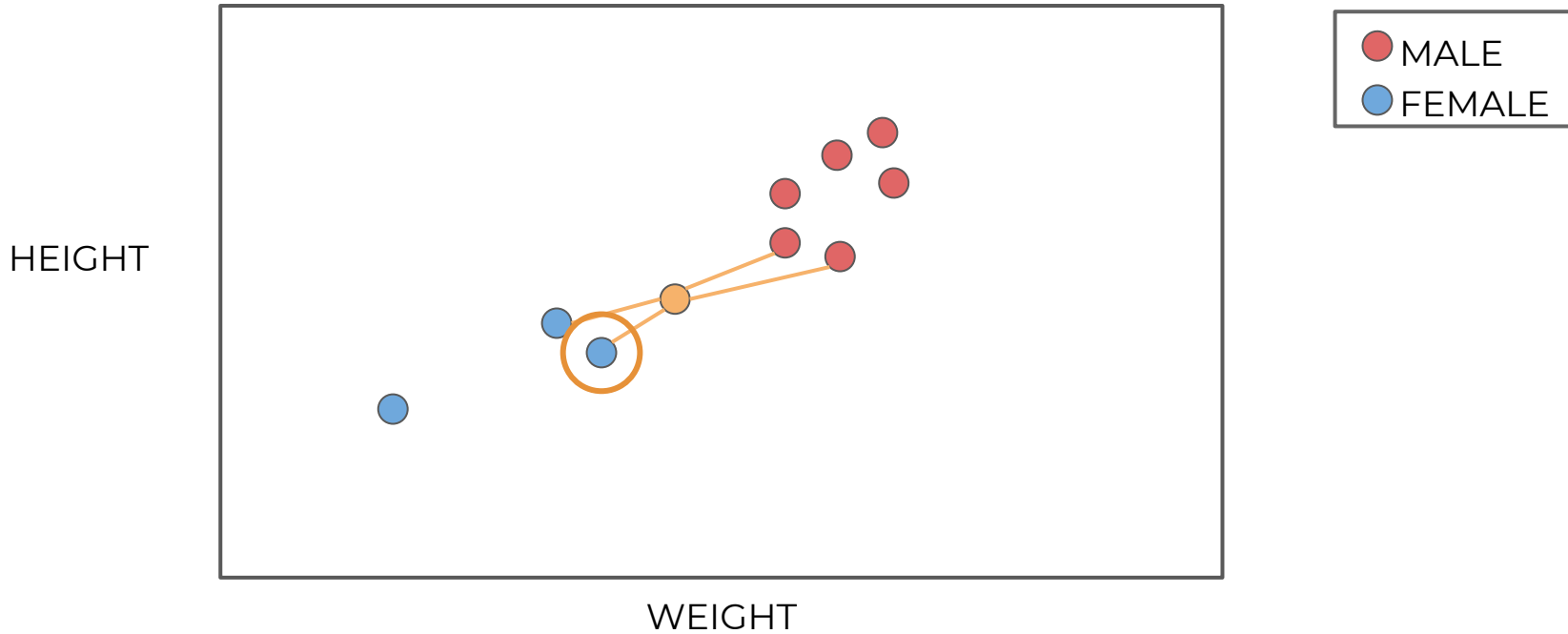
KNN

- K=4 leads to a tie!



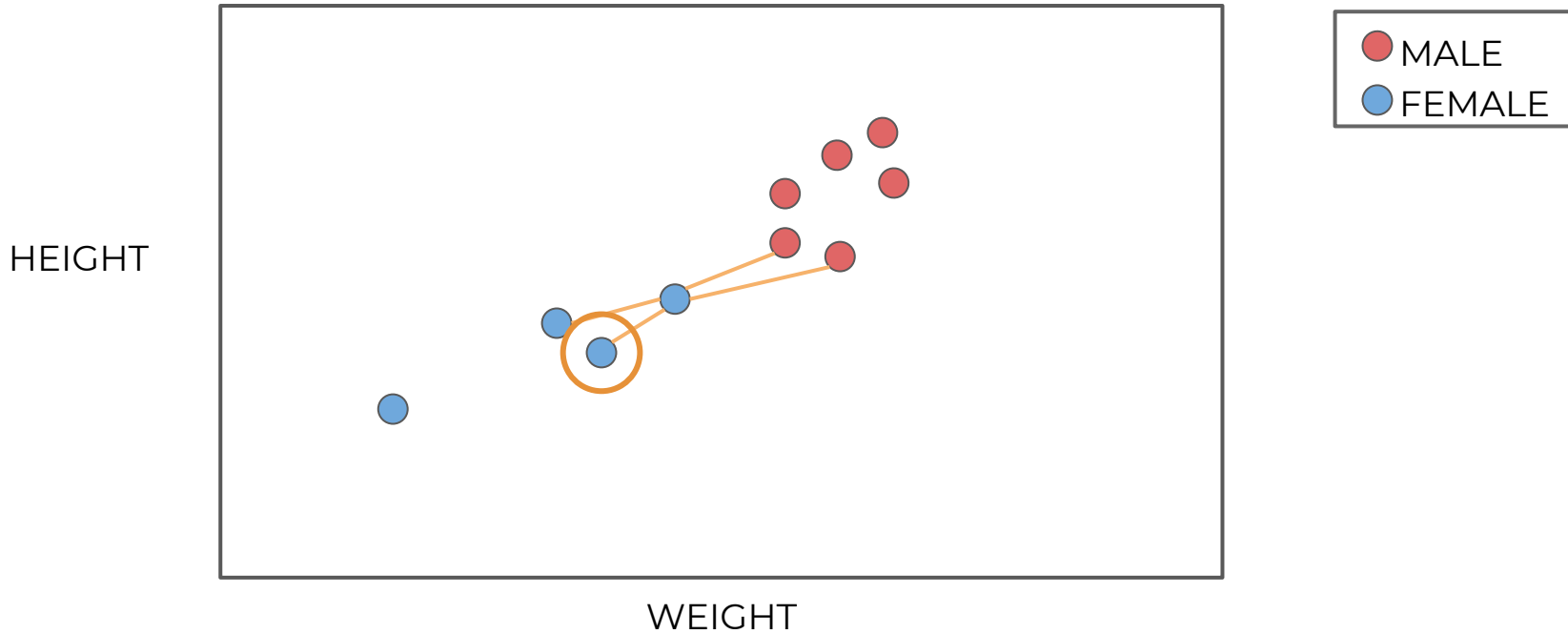
KNN

- Choose closest K



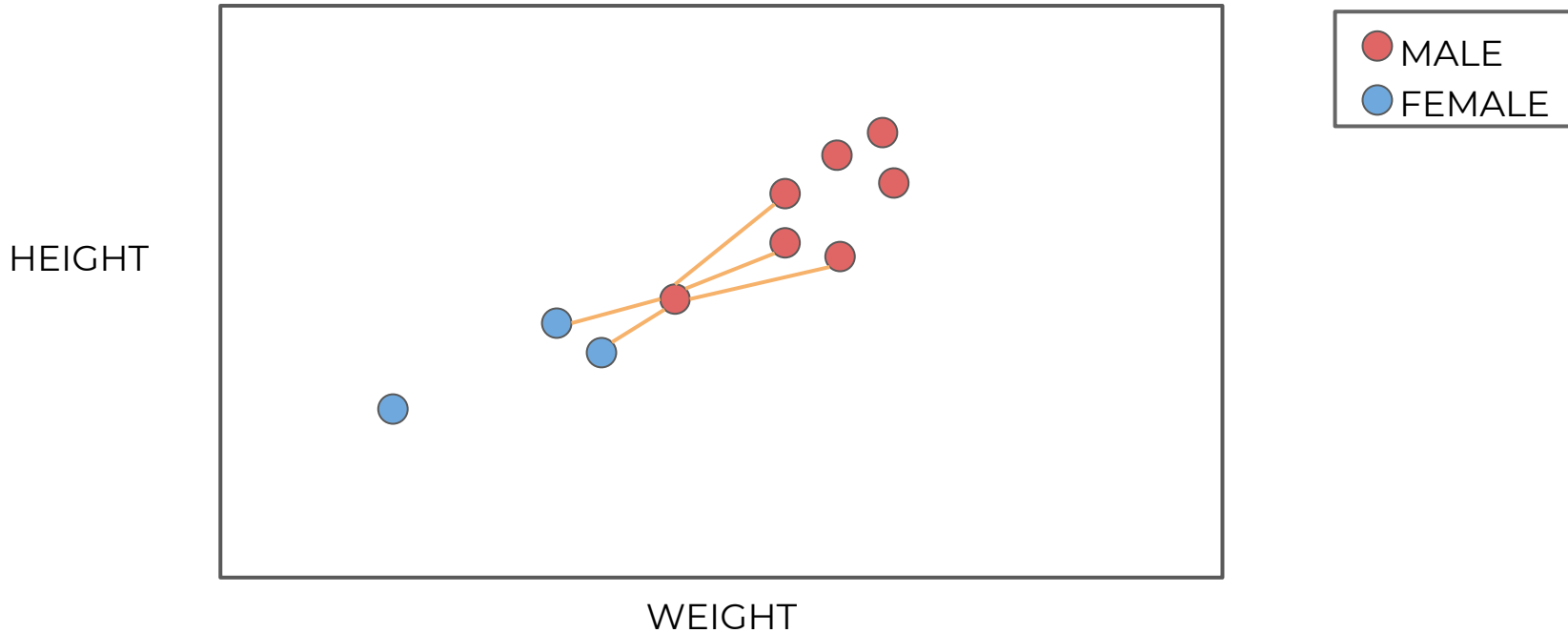
KNN

- Choose closest K



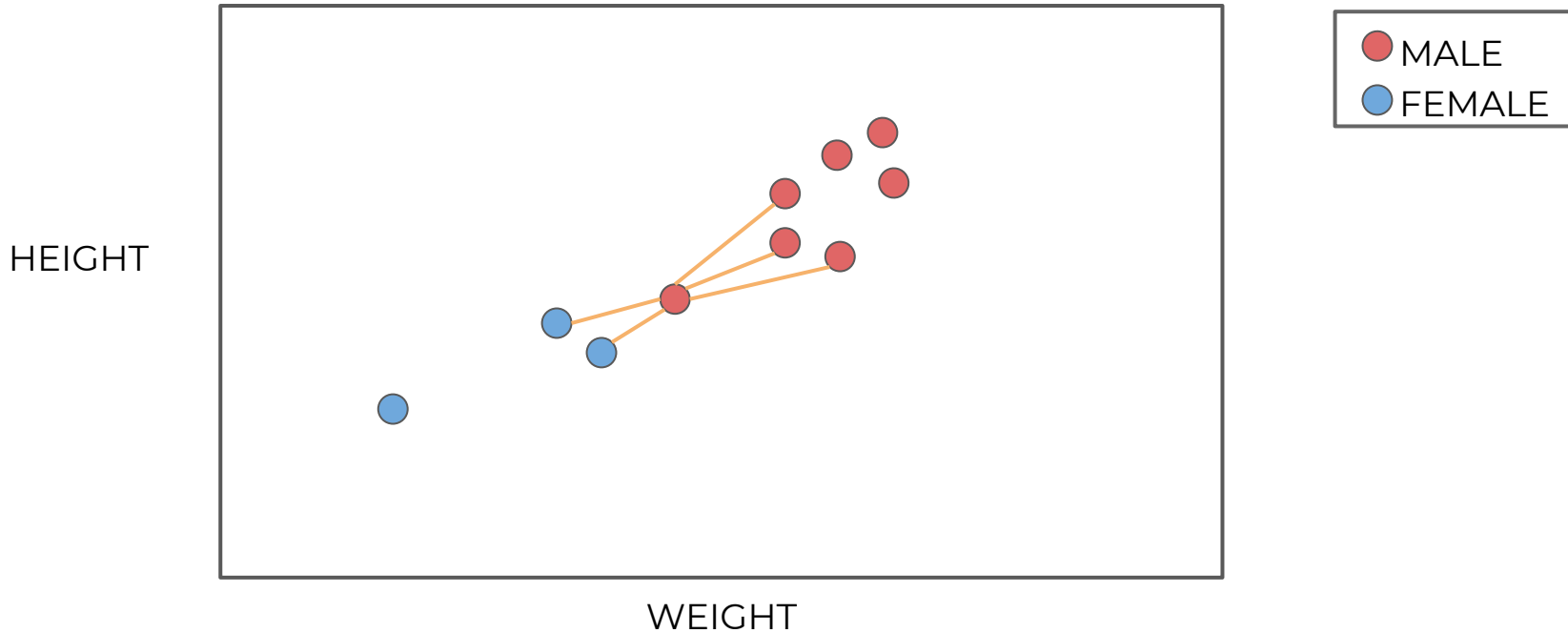
KNN

- K=5 causes a switch from previous K values.



KNN

- How to choose best K value?

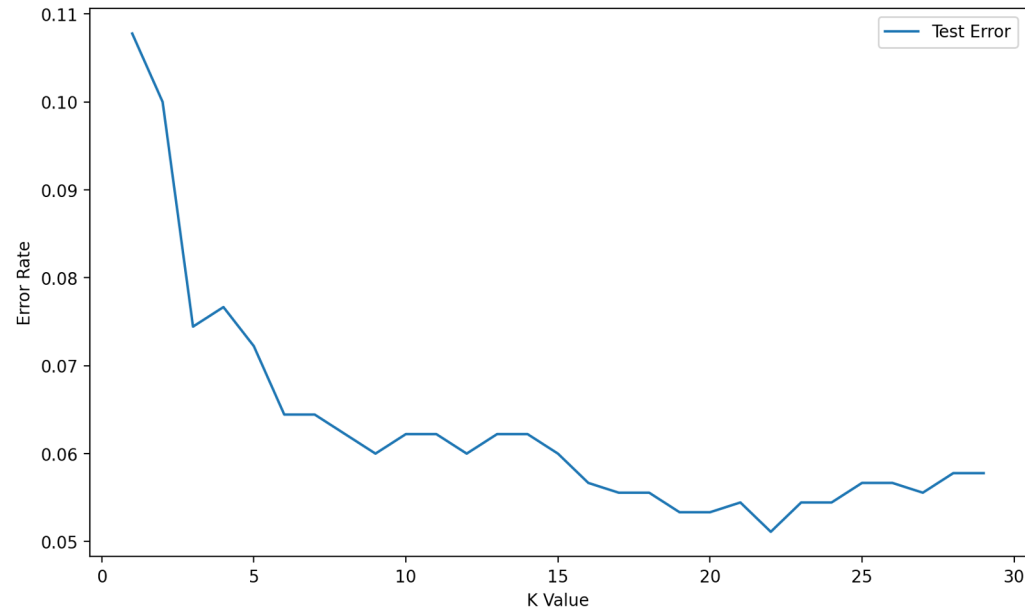


KNN

- We want a K value that **minimizes** error:
 - $\text{Error} = 1 - \text{Accuracy}$
- Two methods:
 - Elbow method.
 - Cross validate a grid search of multiple K values and choose K that results in lowest error or highest accuracy.

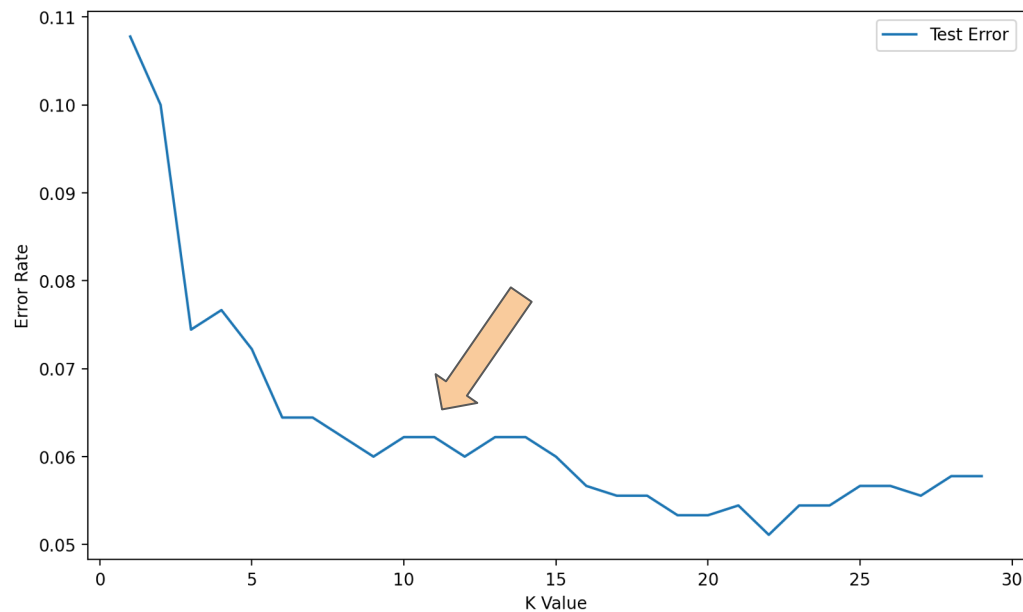
KNN

- Elbow method:



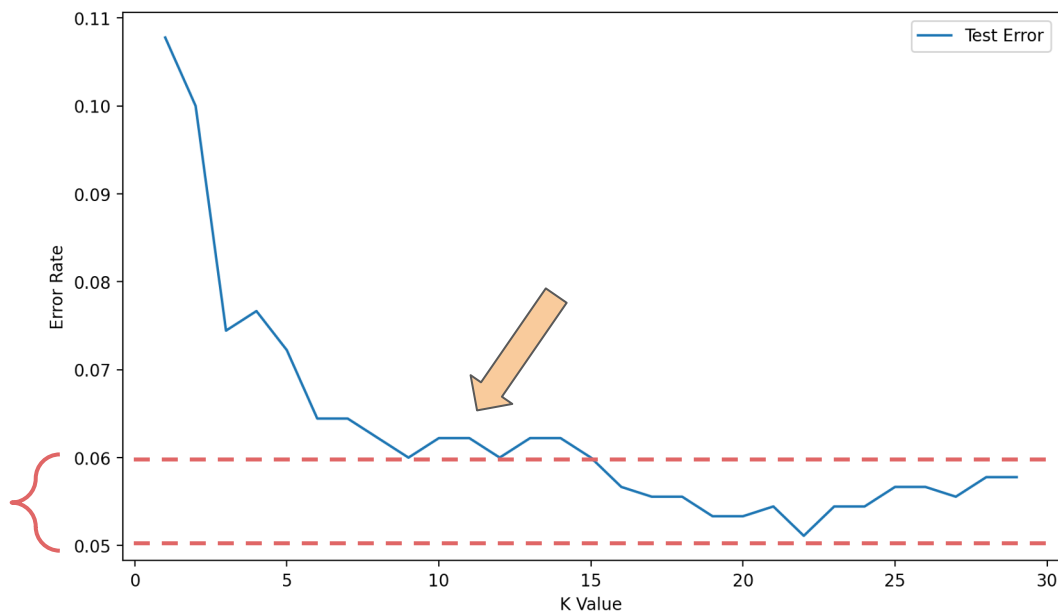
KNN

- Elbow method:



KNN

- Elbow method:



KNN

- Cross validation only takes into account the K value with the lowest error rate across multiple folds.
- This could result in a more complex model (higher value of K).
- Consider the context of the problem to decide if larger K values are an issue.

KNN

- KNN Algorithm
 - Choose K value.
 - Sort feature vectors (N dimensional space) by distance metric.
 - Choose class based on K nearest feature vectors.

KNN

- KNN Considerations:
 - Distance Metric
 - Many ways to measure distance:
 - Minkowski
 - Euclidean
 - Manhattan
 - Chebyshev

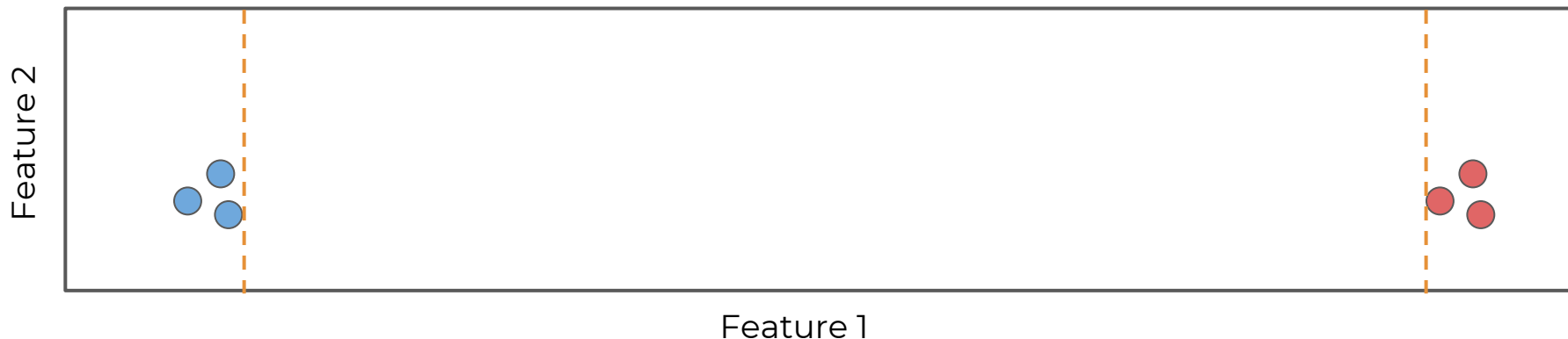
KNN

- KNN Considerations:
 - Scaling for Distance
 - Features could have vastly different value ranges!



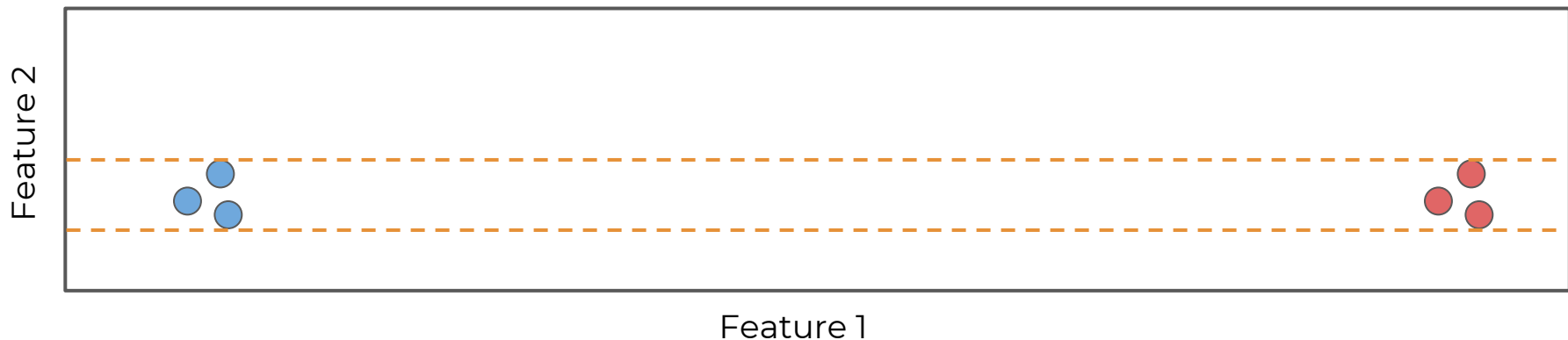
KNN

- KNN Considerations:
 - Scaling for Distance
 - Features could have vastly different value ranges!



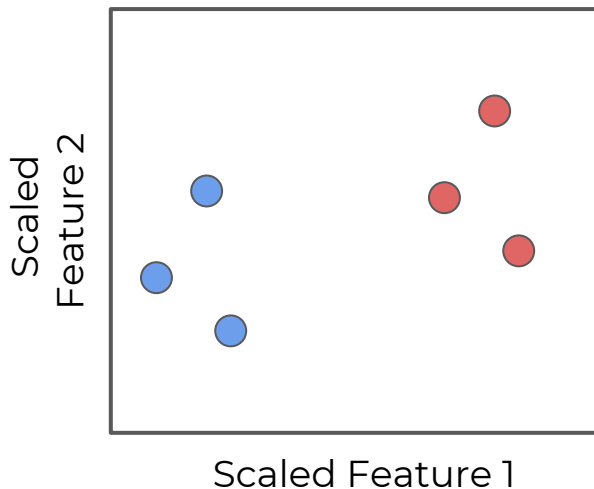
KNN

- KNN Considerations:
 - Scaling for Distance
 - Features could have vastly different value ranges!



KNN

- KNN Considerations:
 - Scaling is necessary for KNN.



KNN

- While the KNN Algorithm is relatively simple, keep in mind the following considerations:
 - Choosing the optimal K value.
 - Scaling features.
- Let's continue to explore how to perform KNN for classification!

KNN Classification

Coding Part One: Data and Model

KNN

- Let's test your new skills on a real data set.
- We'll be analyzing sonar frequencies to help distinguish between rocks or sea mines!

