

# Machine Learning algorithms

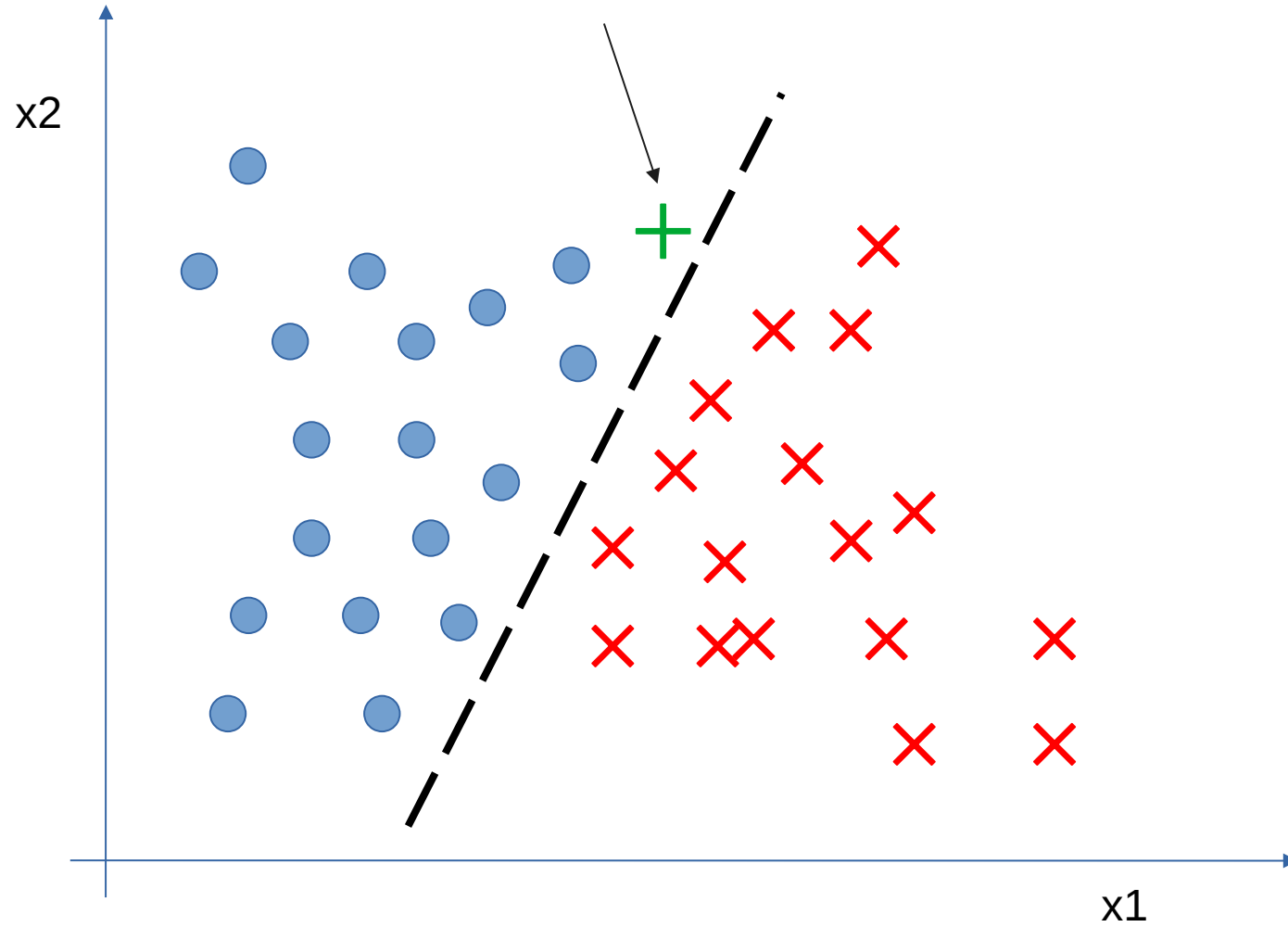
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MS DATA SCIENCE 2020-2021

supervised

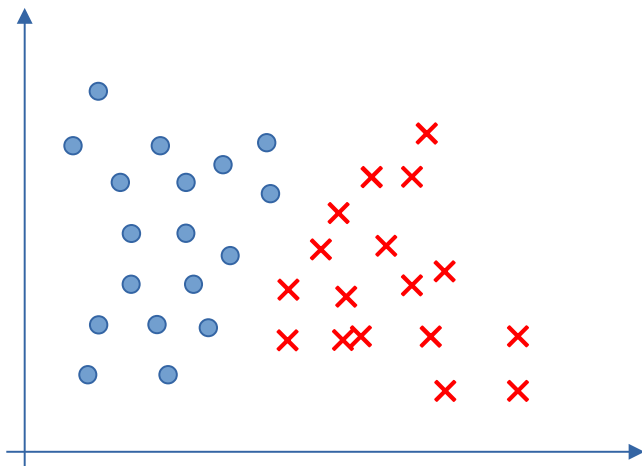
# classification

- Goal: predict the categorical class labels
  - discrete
  - unordered
  - group membership
- Binary classification
  - spam / no spam
  - cat / no cat
- Multi-class classification
  - handwritten digits

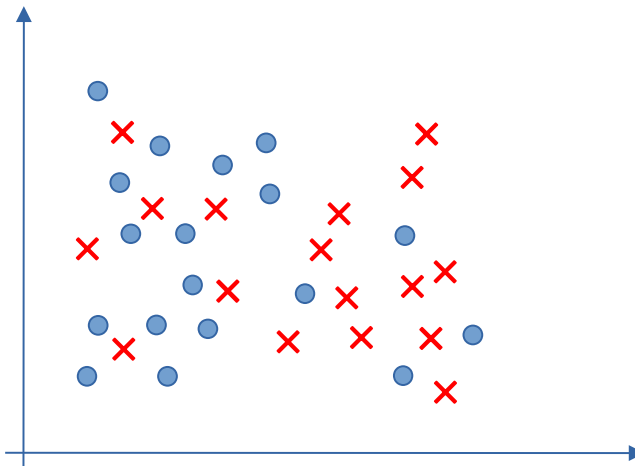
# classification



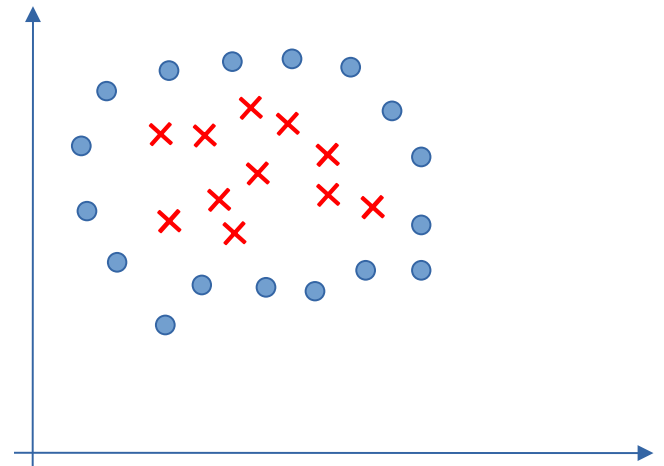
# classification



linearly separable



non linearly separable



non linearly separable

# classification

- logistic regression
- support vector machine
- decision tree
- random forest
- KNN

# logistic regression


- perfect for linearly separable
- can be extended to multiclass

$$\text{logit}(P) = \log \frac{P}{1 - P}$$

# logistic regression

- the logit function takes input in  $[0,1]$  and returns in  $(-\infty, +\infty)$
- express linear relationships between feature values and the log-odds

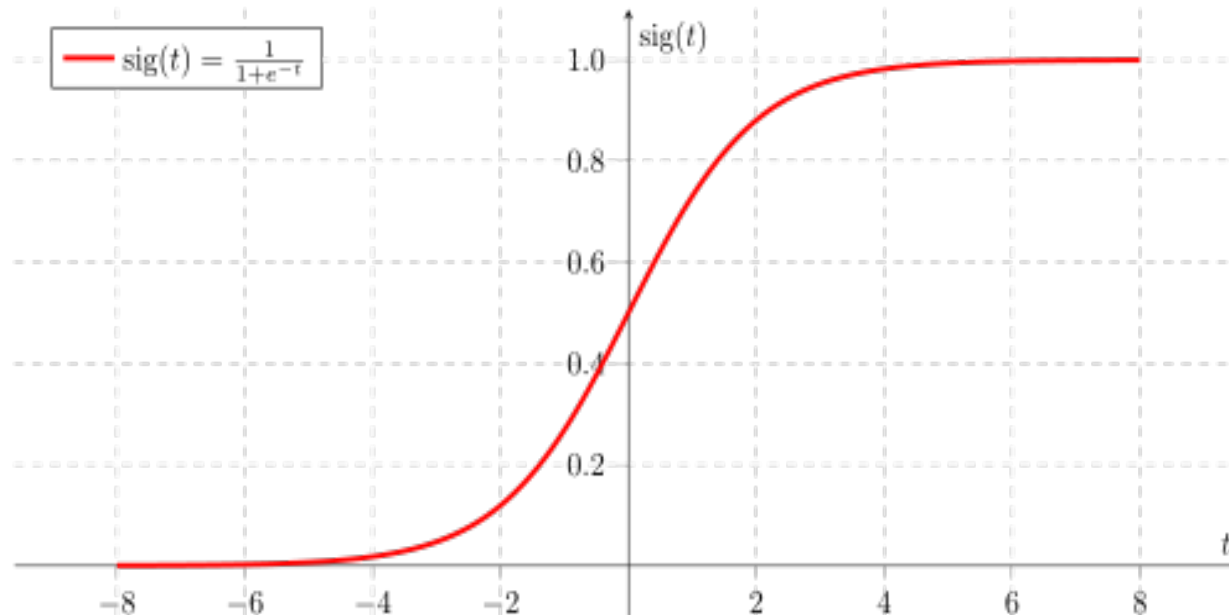
$$\text{logit}(P(y=1|x)) = \sum(W_i X_i) = W^T X$$

- where  is the conditional probability that a particular sample belongs to class 1 given its features  $x$ .



# sigmoid function

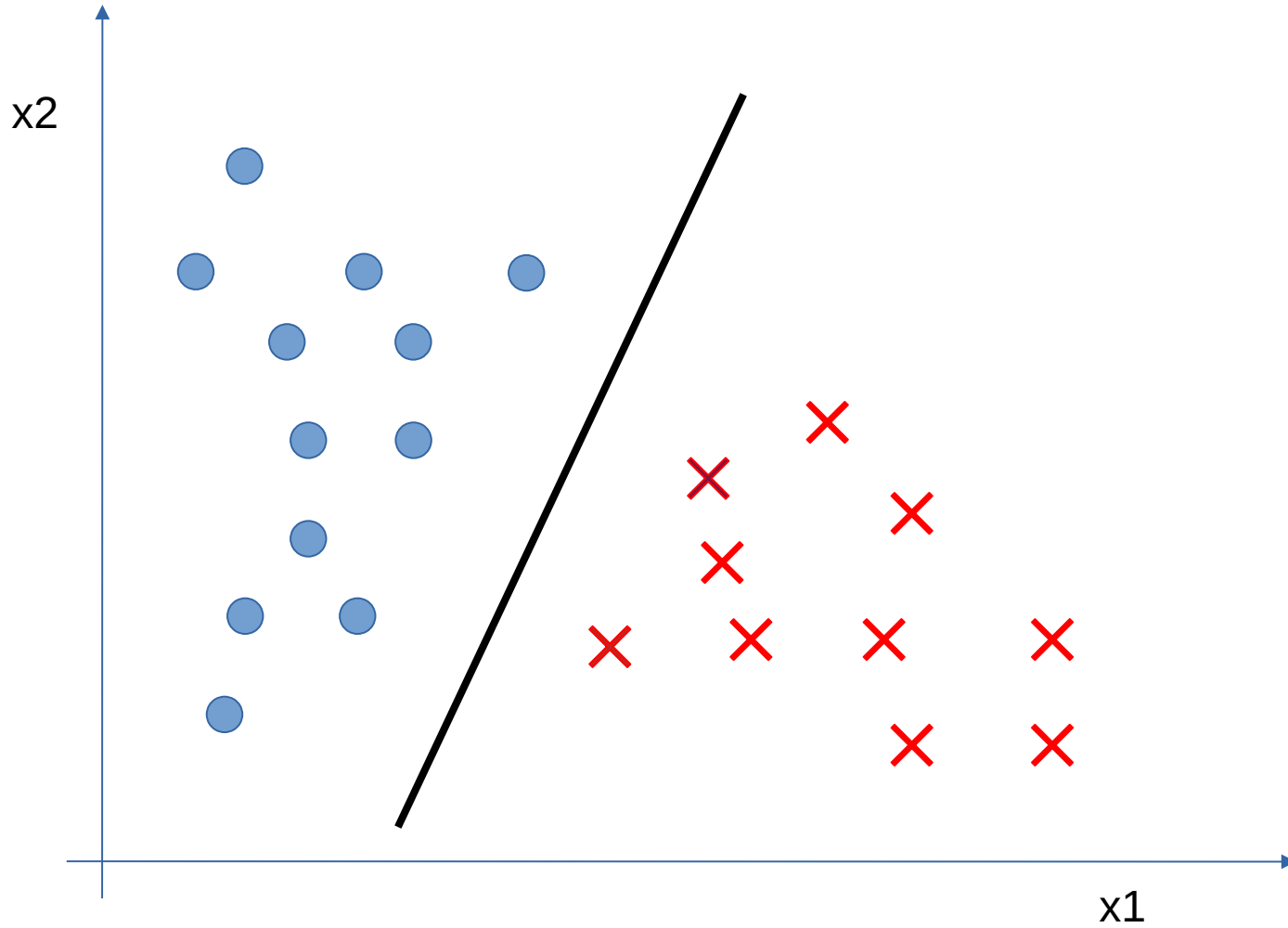
- the inverse of the logit function
- $\text{sigmoid}(\text{logit}(p)) = p$



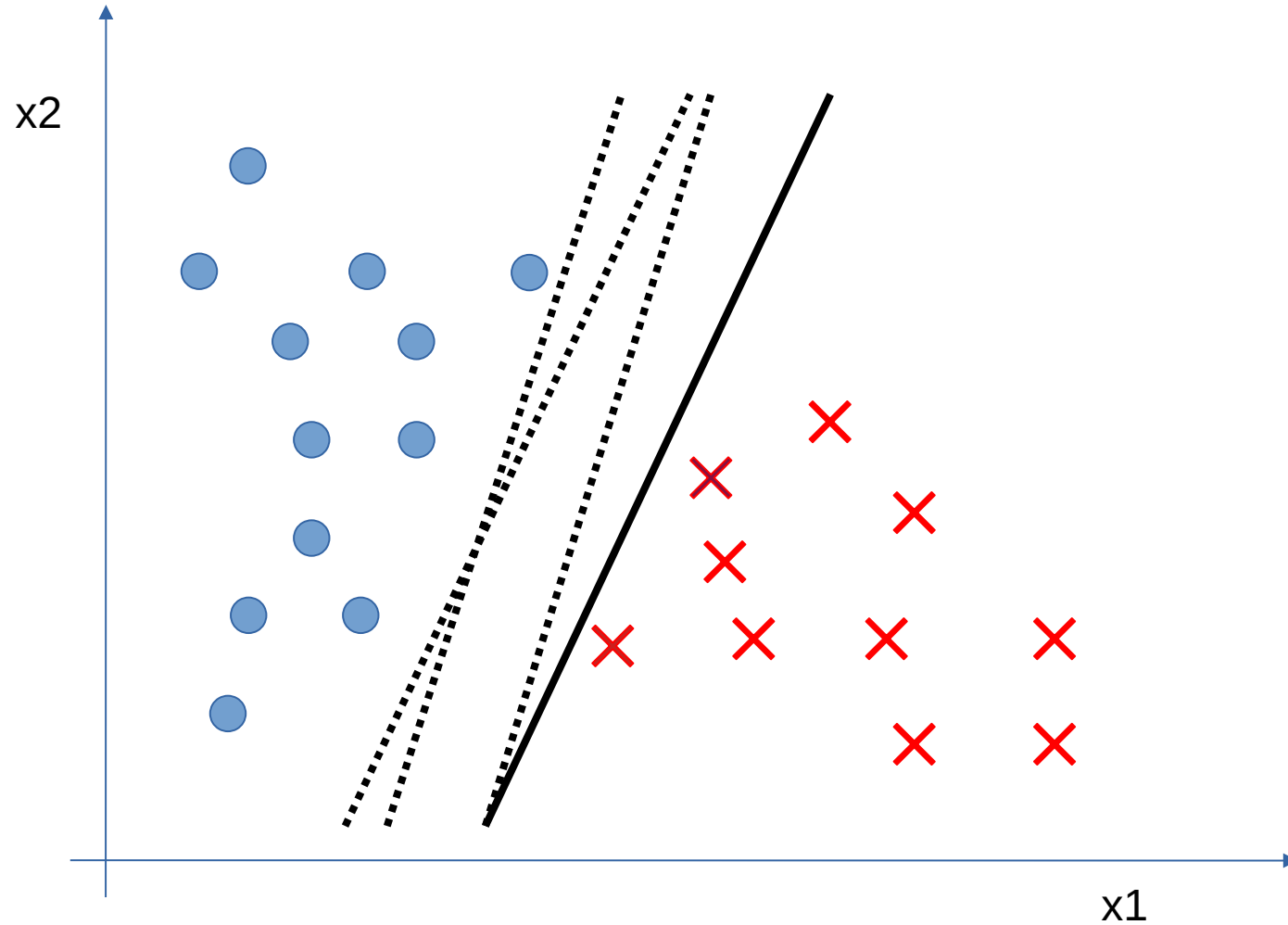
# sigmoid

- from  $(-\infty, +\infty)$  to  $[0,1]$
- takes real values and transform them in the  $[0,1]$  range with an intercept at 0.5
- THIS IS WHAT THE logit function does while trained.
- the output of the sigmoid is the probability of a certain sample to be of class 1, given its feature  $x$  parametrized by the weights  $w$

# Support Vector Machine



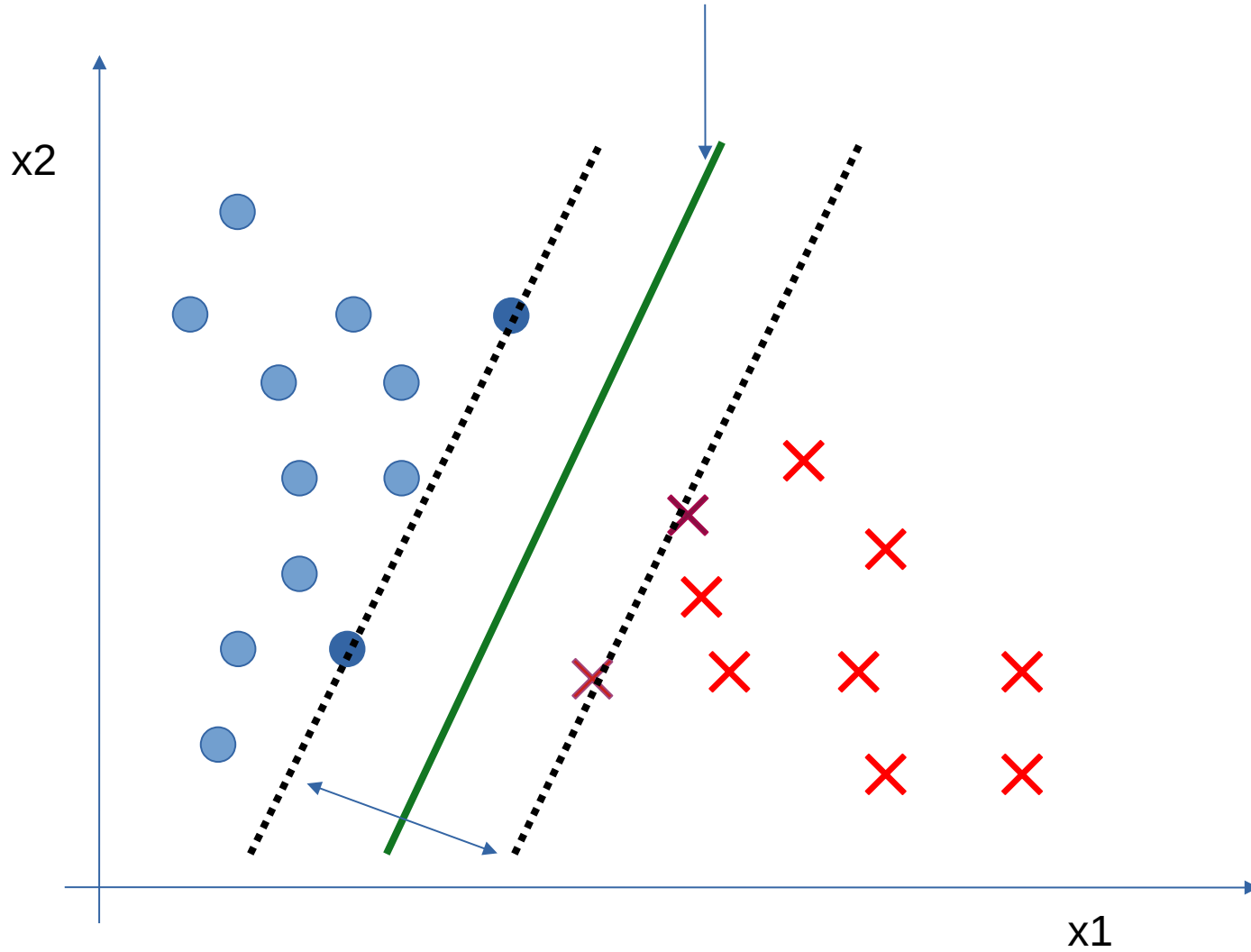
# Support Vector Machine



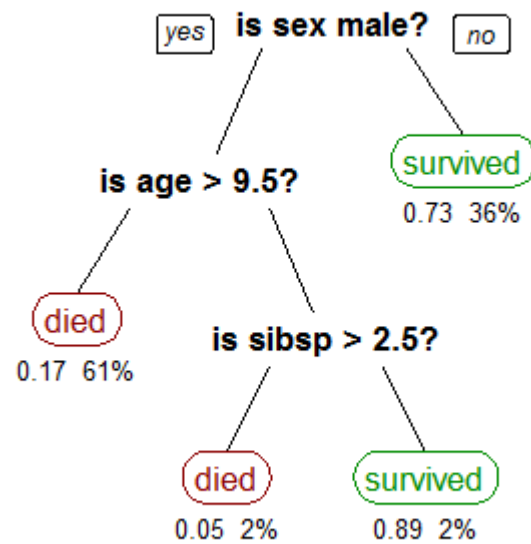
# Support Vector Machine

- find a hyperplane in an N-dimensional space that distinctly classifies the data points.
- many possible hyperplanes that could be chosen.
- find a plane that has the maximum margin, i.e., the maximum distance between data points of both classes.

# Support Vector Machine



# Decision Tree

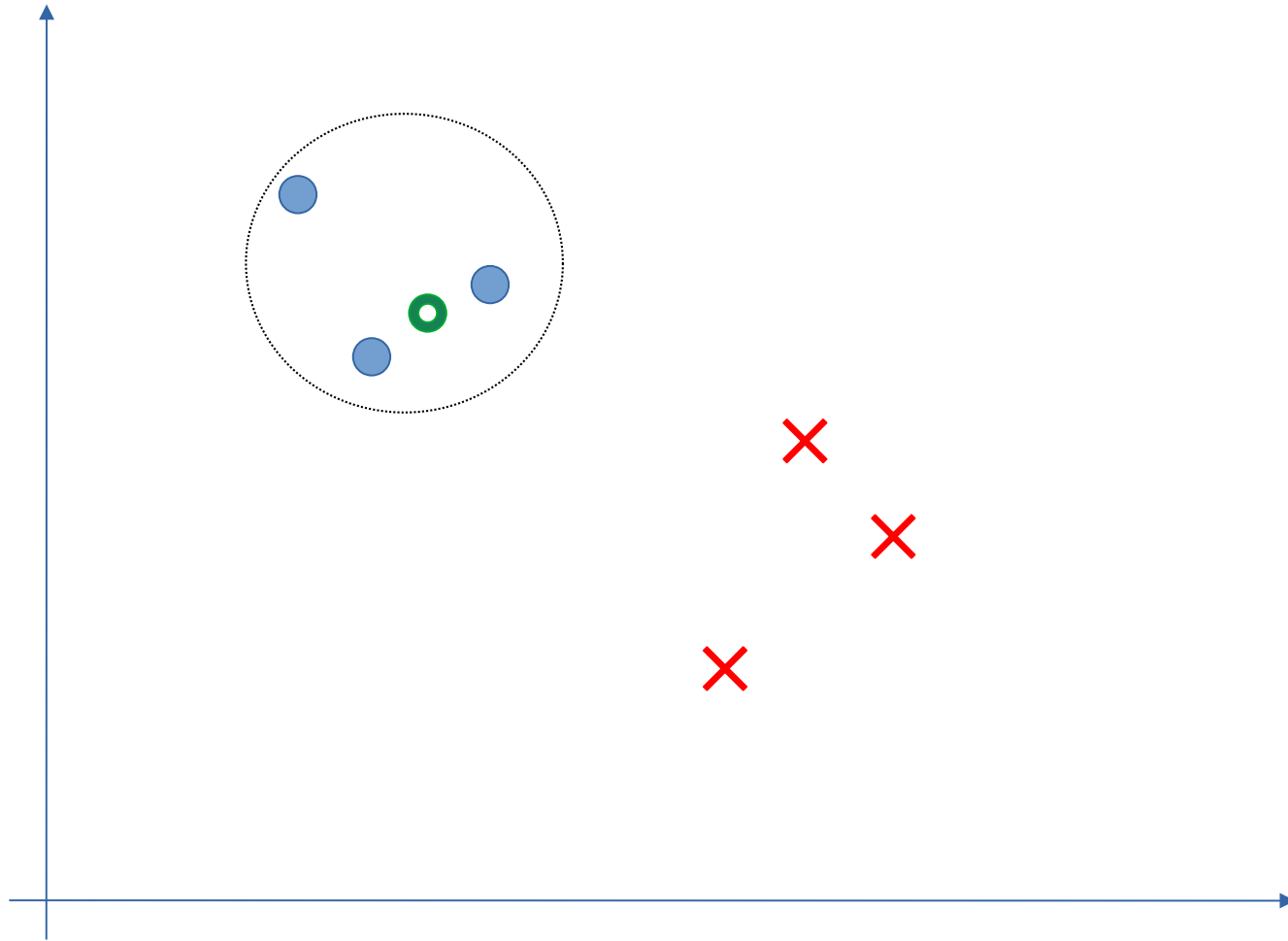


# Decision Tree

- feature importance is KEY
- $n$  features  $\rightarrow$   $n$  candidates splits
- calculate how much accuracy is lost for each split
- the split that costs least is chosen
- WHEN DO WE STOP???
- max depth
- min number of training inputs for each leaf
- ...



# KNN



# KNN

- Load the data
- Choose **K**
- For each point **p** in test data:
  - Compute distance between **p** and each training data
  - Sort in ascending order
  - Choose the top **K** rows
  - Assign the most frequent class
- Done.

unsupervised

# unsupervised

- No labels given
- GOAL: find structure
  - discovering hidden patterns in data

# unsupervised

- trickier
  - no answer labels (no ground truth)
  - external evaluation vs internal evaluation
    - experts vs objective function
- but:
  - annotating large datasets is very costly (Speech Recognition)
  - we don't know how many classes can be (Data Mining)
  - gain some insight into the structure of the data before designing a classifier

# clustering

- more problems:
  - define distance
  - define similarity
  - define clusters
- Examples:
  - Kmeans
  - Fuzzy Kmeans
  - GMM
  - Hierarchical
  - ...

# K-means

- Group input data into K groups
- Define K centers
- While “not converged”:
  - Take each point and assign it to the “closest” center
  - Recompute centers
    - minimize inter-cluster distances