

Image =
$$\begin{bmatrix} 15 & 13 & 12 & 156 & 143 \\ 14 & 11 & 154 & 172 & 164 \\ 12 & 8 & 145 & 163 & 12 \\ 15 & 10 & 13 & 14 & 5 \end{bmatrix}$$

mask =
$$\begin{bmatrix} -1 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & -1 \end{bmatrix}$$

Result =
$$\begin{bmatrix} 15 & 13 & 12 & 156 & 143 \\ 14 & -138 & & & 164 \\ 12 & -11 & & & 12 \\ 15 & 10 & 13 & 14 & 5 \end{bmatrix}$$

$$-15 + 22 - 145$$

$$-14 + 16 - 13$$

Obj =
$$\begin{bmatrix} 0 & 0 & 2 & 2 & 3 \\ 0 & 0 & 2 & 2 & 3 \\ 1 & 1 & 2 & 2 & 3 \\ 1 & 1 & 3 & 3 & 3 \end{bmatrix}$$

RGB
 ↓
 grayscale
 ↓
 thresholding to get b/w
 ↓
 masking to remove noise
 ↓
 1 1 1 1 1

connected pixel
↓
obj

online

decision
action
decision
action
depth-first
informed

use when environment is dynamic (changing)

knows (or estimates)
distance to the
goal
use if information
is available
A*

offline

all decisions
all actions
optimal, but time and
computation consuming
breadth-first
uninformed

doesn't

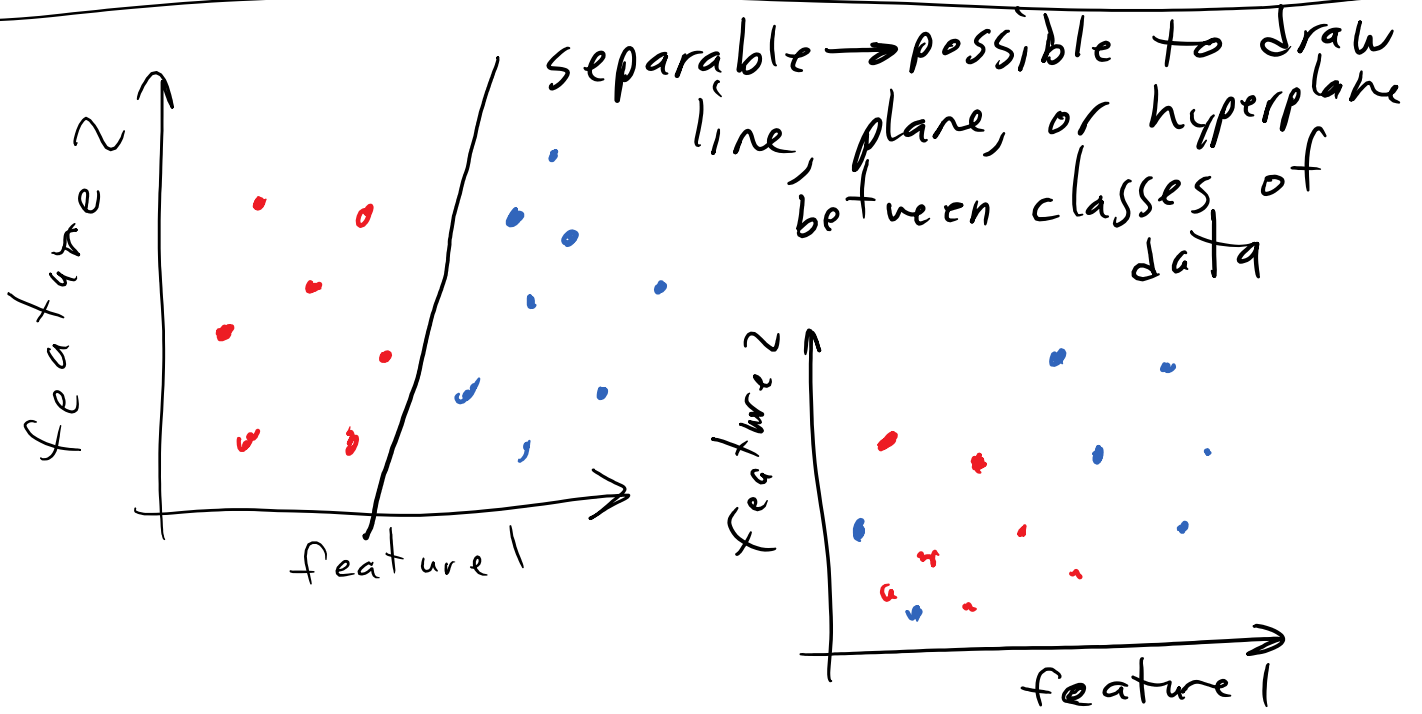
supervised

labeled training data
use if labels are
available
+ vector

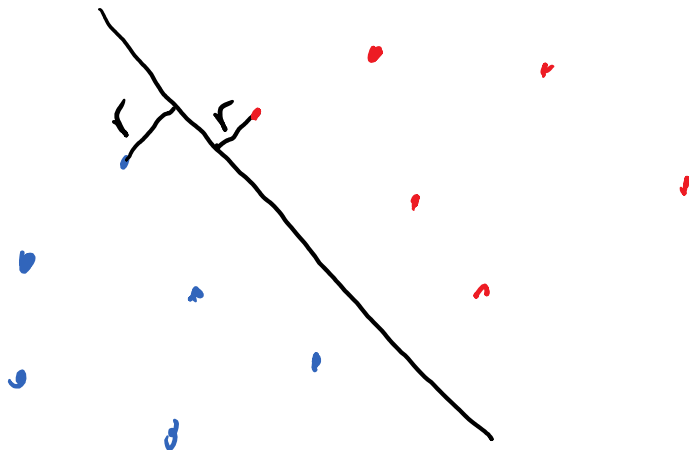
unsupervised

unlabeled training data
use if labels
are unknown
K-means clustering

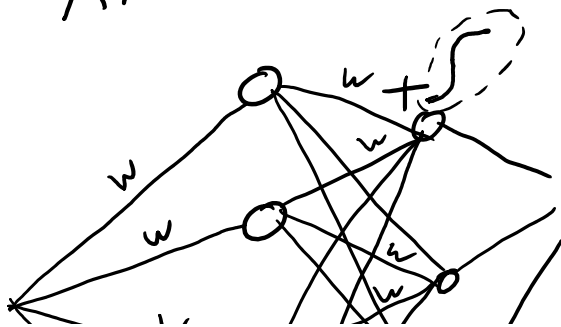
SVM \rightarrow support vector
ANN \rightarrow artificial neural network



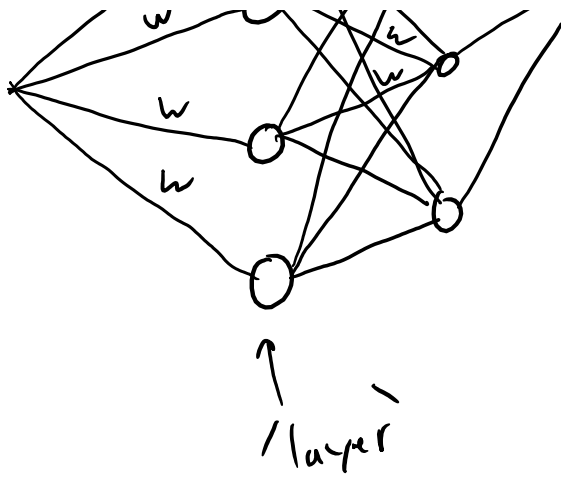
SVM



ANN



pro \rightarrow non-separable data
con \rightarrow computationally intensive



optical flow

based on masking

pro → can tell direction and speed

con → computationally-intensive

difference

based on subtraction

rule 1: has to be possible to get the feature from the raw data

rule 2: useful for increasing separability in the data

rule 3: use the fewest number of features that allows the algorithm to learn the data

