

A decorative graphic on the left side of the slide, consisting of a network of thin, dark blue lines and small circles, resembling a circuit board or a neural network diagram.

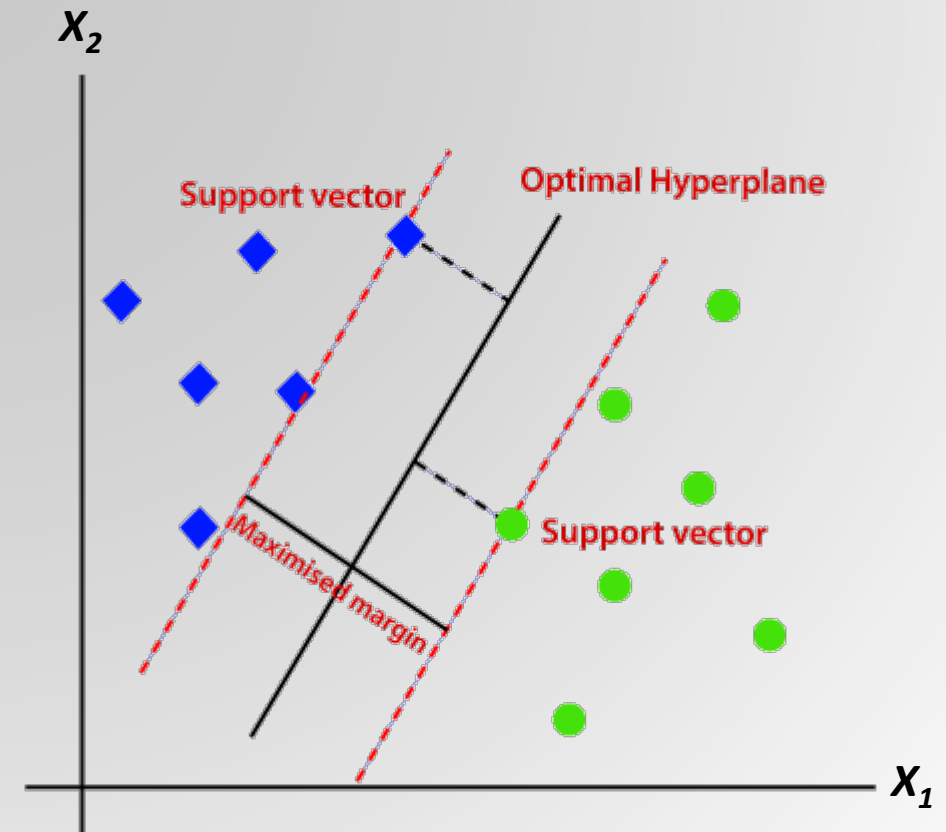
# CSC 462 – Machine Learning

## 3.4 Support Vector Machine

Dr. Sultan Alfarhood

# Support Vector Machine (SVM)

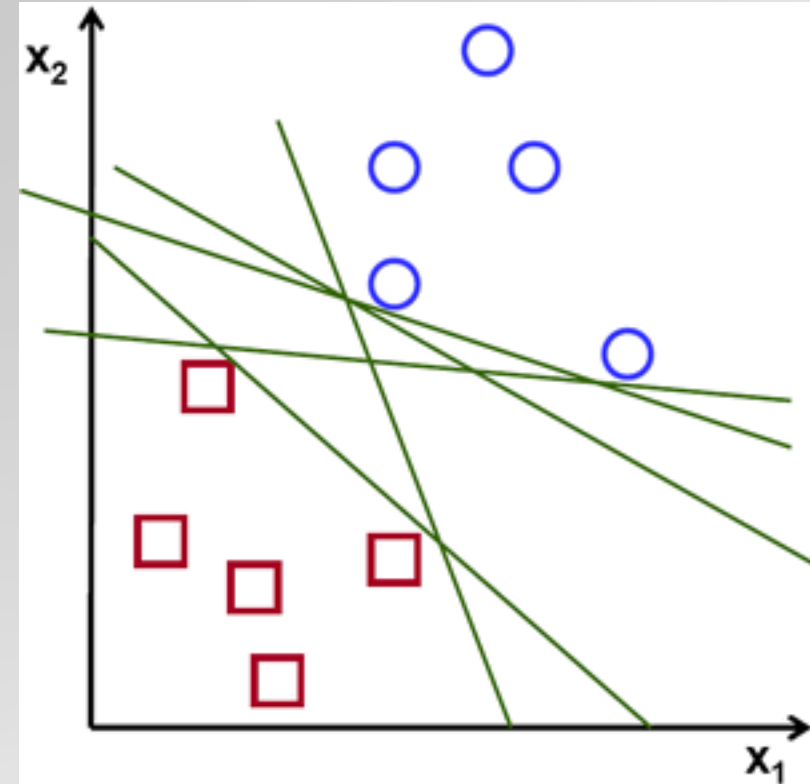
- Support Vector Machine (SVM) is one of the most popular Machine Learning Classifier.
- It uses the concept of Margin to classify between classes.



Note: In this graph, both  $X_1$  and  $X_2$  axes represent **features**, and the **target** is represented by the sample color (blue or green).

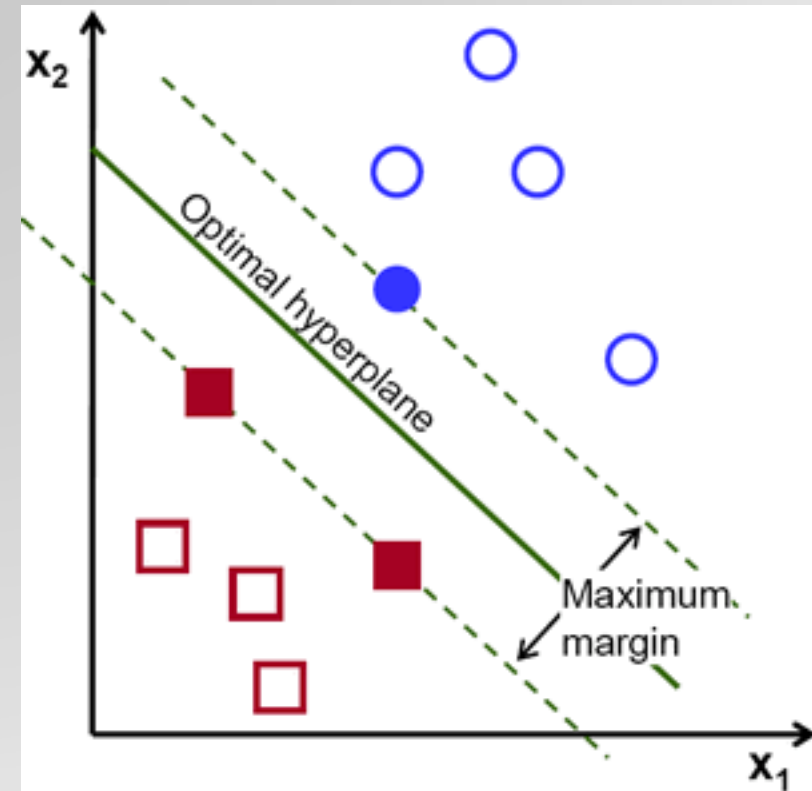
# Support Vector Machine (SVM)

It tries to find the decision boundary which separates the classes the best (i.e., with minimum error).



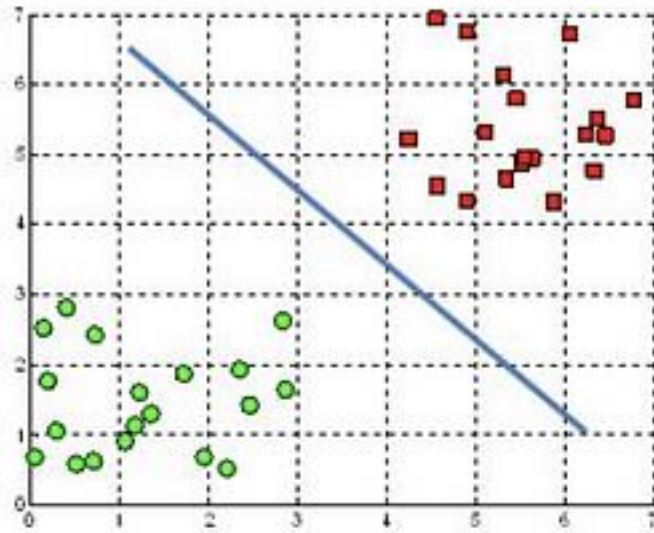
# Support Vector Machine (SVM)

- A **hyperplane** is a decision plane which separates between a set of objects having different class memberships.
- The nearest points from the line are called **support vectors**.
- The most optimal line is the one with **maximum margin**.

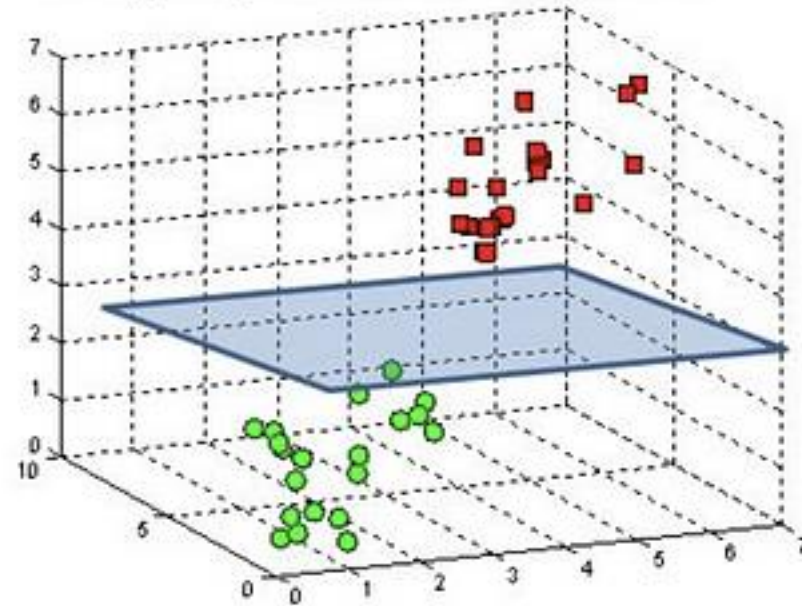


# Hyperplanes in 2D and 3D feature space

A hyperplane in  $\mathbb{R}^2$  is a line

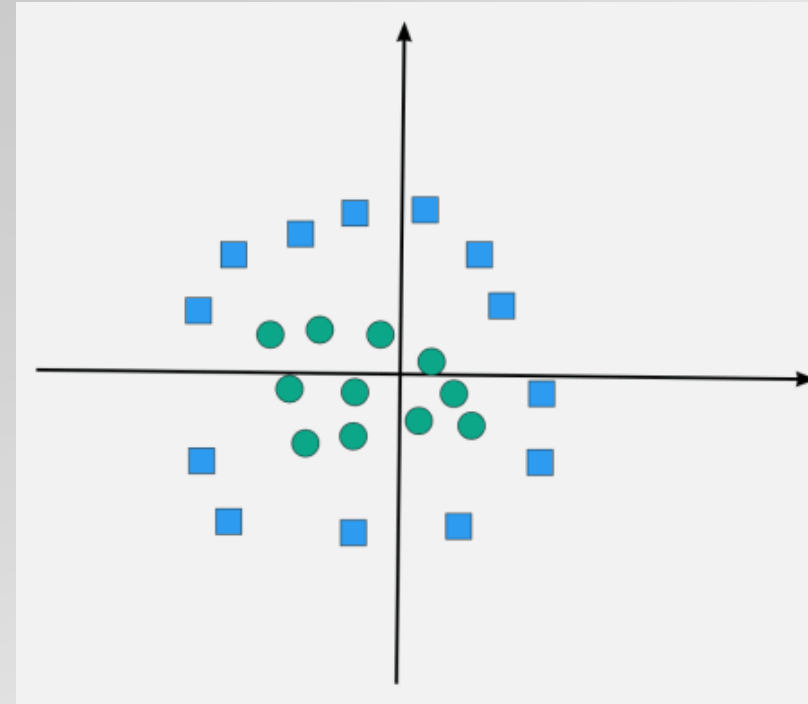


A hyperplane in  $\mathbb{R}^3$  is a plane



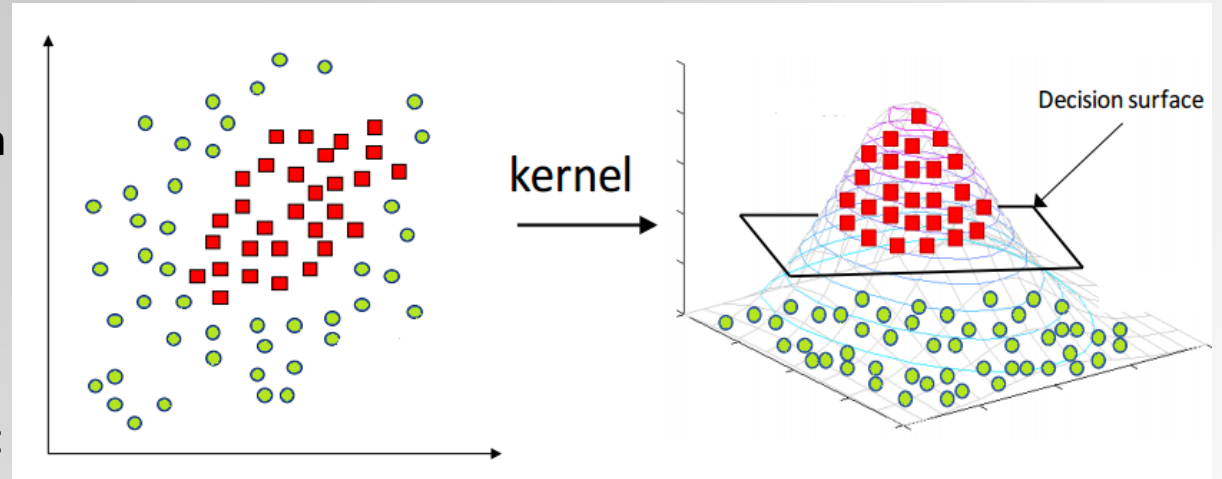
# Support Vector Machine (SVM)

- What if the data is inseparable?
  - Things can become difficult when the data is non-linear.
- SVM can use the kernel function to make non-linear data linear.

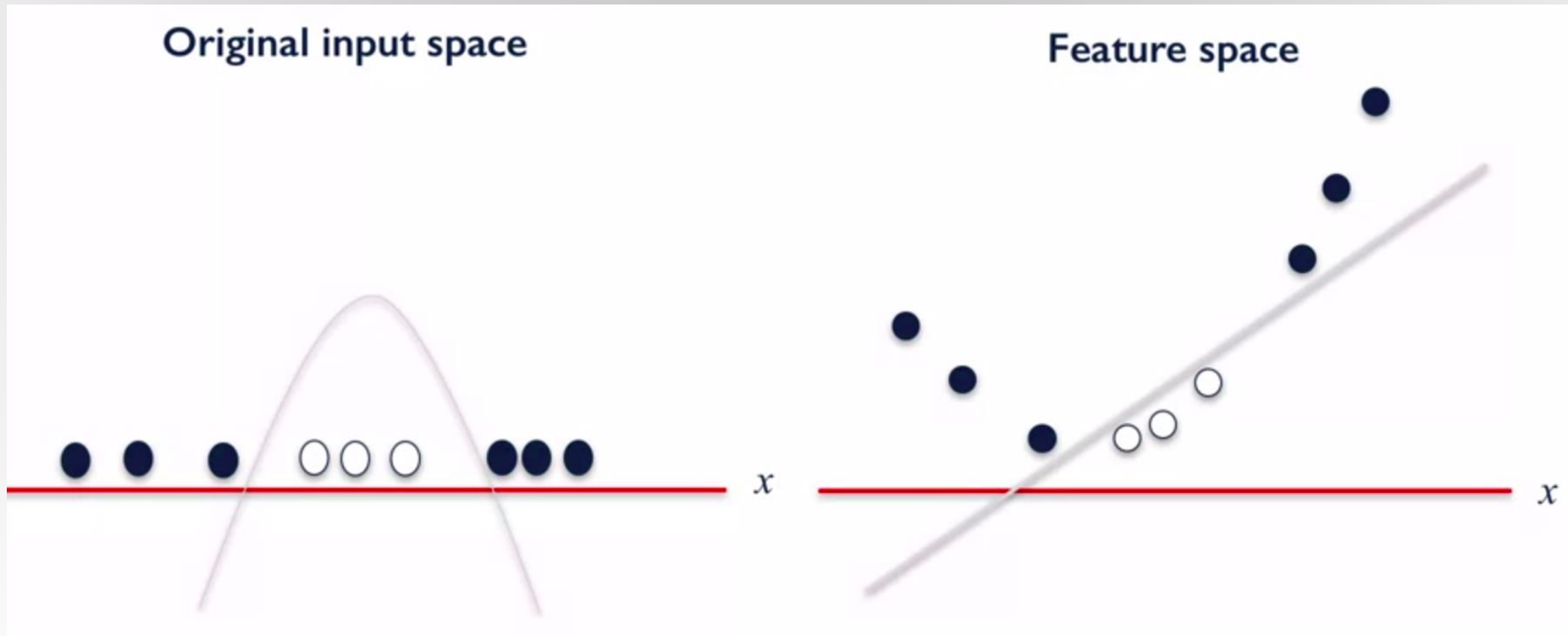


# Kernel Function

- The **kernel function** separates the data by **adding dimensions to the problem**.
- Kernel Function generally transforms the training set of data so that a non-linear decision surface can transform to a linear equation in a higher number of dimension spaces.
- The idea of the **kernel trick** and the reason why we use kernel functions in SVM is the following:
  - Although we transformed the data to a higher dimension, the kernel only calculates the relationship between the data as if its in a higher dimension but **doesn't change it**.

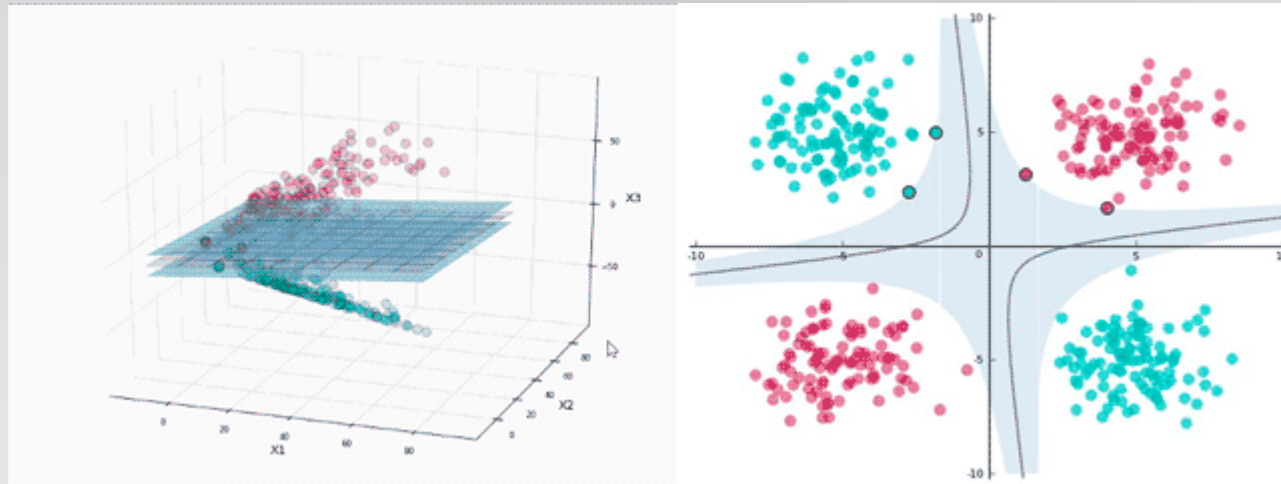


# Kernel Transformation Example



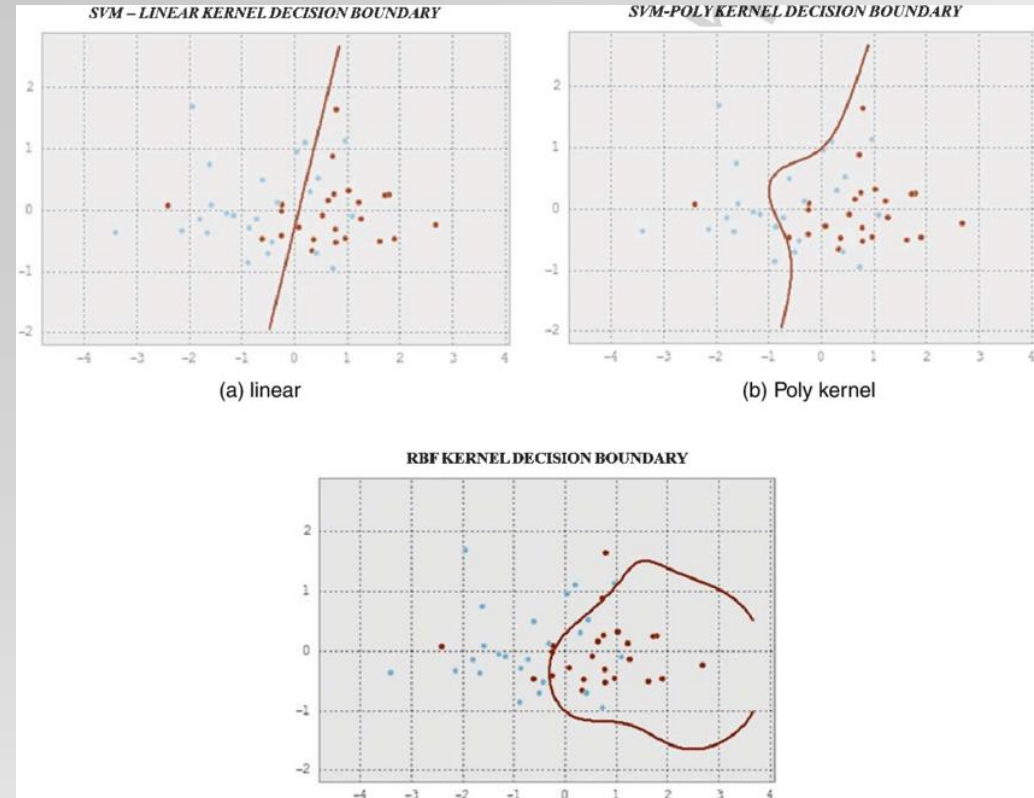


# Kernel Transformation Example 2 (Animated)

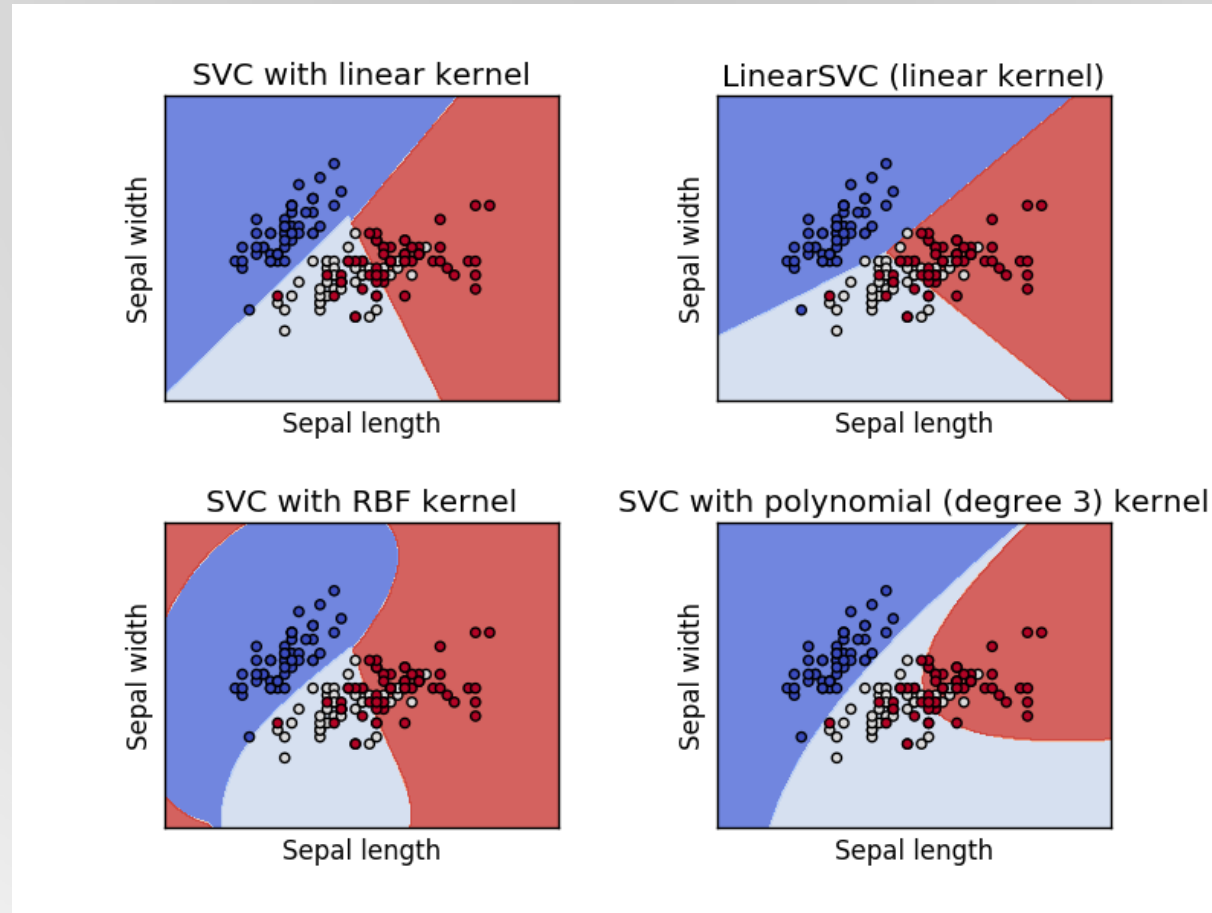


# Kernels

- There are many kernel functions; most common:
  - Linear Kernel
  - Polynomial Kernel
  - Radial Basis Function (RBF) Kernel



# Kernels



**Plot different SVM classifiers in the iris dataset**

[https://scikit-learn.org/0.18/auto\\_examples/svm/plot\\_iris.html](https://scikit-learn.org/0.18/auto_examples/svm/plot_iris.html)

# SVM Advantages & Disadvantages

- Advantages

- High accuracy
- Works well with non-linear data.

- Disadvantages

- Creating the model takes a long time (especially in high degree polynomial)
- The kernel used can greatly change the accuracy of the model, so the right kernel must be chosen.



# Thank You