

# **Non - Linear SVM**



### Introduction to SVM: Why SVM?

 Working with neural networks for supervised and unsupervised learning showed good results while used for such learning applications.

MLP's uses feed forward and recurrent networks.

 Multilayer perceptron (MLP) properties include universal approximation of continuous nonlinear functions and include learning with input-output patterns and also involve advanced network architectures with multiple inputs and outputs.



### Introduction to SVM: Why SVM?

 There can be some issues noticed. Some of them are having many local minima and also finding how many neurons might be needed for a task is another issue which determines whether optimality of that NN is reached.

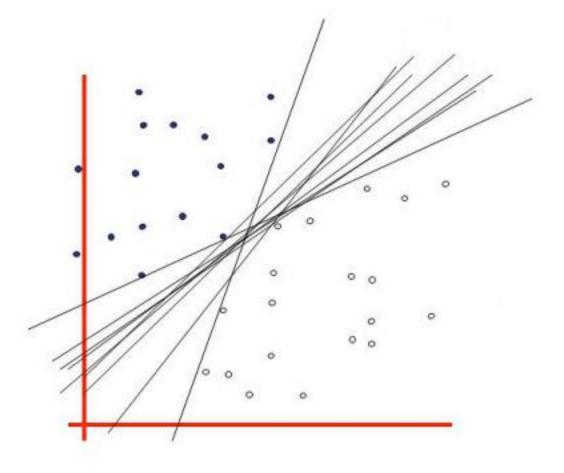
 Another thing to note is that even if the neural network solutions used tends to converge, this may not result in a unique solution.



### Introduction to SVM: Why SVM?

 Now let us look at another example where we plot the data and try to classify it and we see that there are many hyper planes which can classify it.

But which one is better?





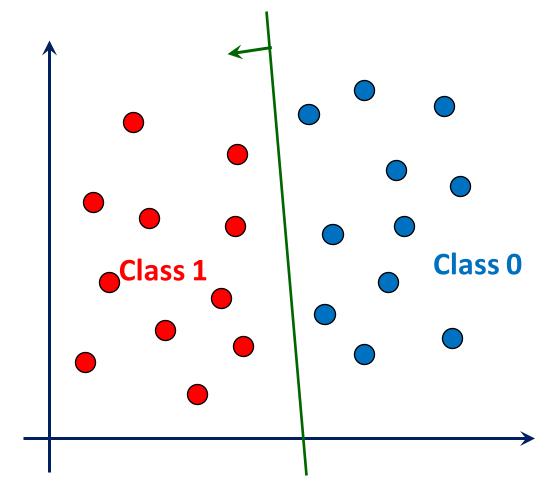
#### **Linear Classifier**

Decision boundary: Hyperplane

$$w^T x = 0$$

• Class 1 lies on the positive side  $w^T x > 0$ 

• Class 0 lies on the negative side  $w^T x < 0$ 

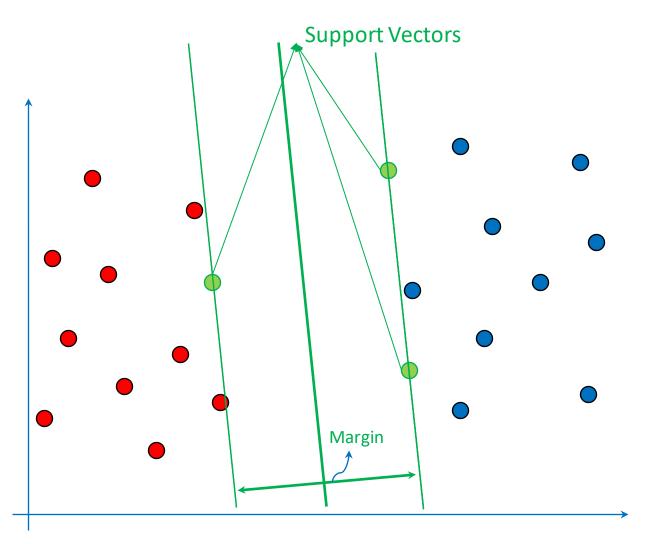




### **Summary: Max-Margin Classification**

- A Large Margin will reduce the chance of misclassifying future test samples
- In other words, large-margin classifiers will generalize better.

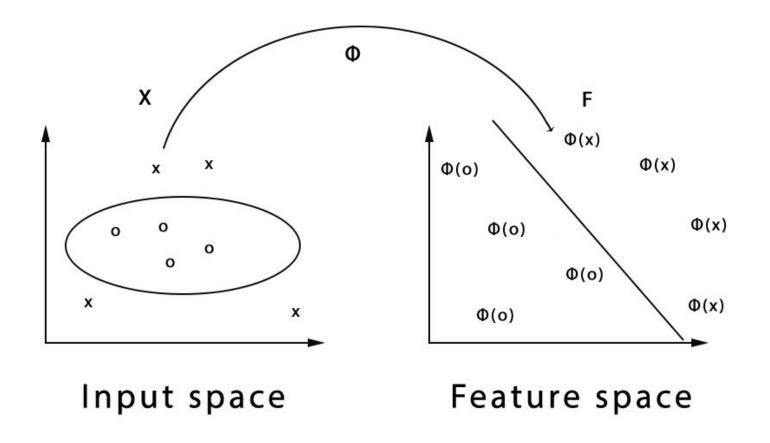
 Samples at the boundary support the margin: called Support Vectors





#### **Non Linear SVM**

- Some data points are not linear separable
- Intuition: to transform the data to a high dimension space

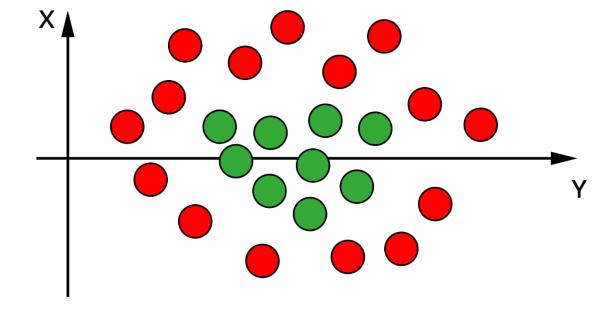




### **Support Vector Machines**

• SVM uses **kernel** to transform the data and then based on these transformations it finds an optimal hyperplane that distinctly classifies the data points.

- Example:
  - No Straight line that can separate the two groups





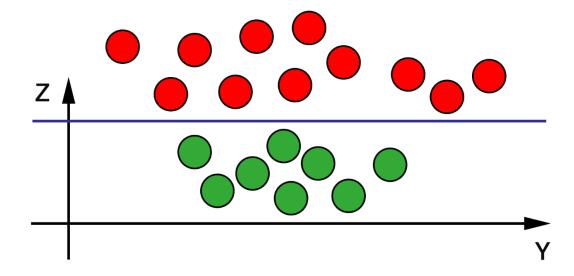
#### **Transform**

 Transform the data by adding one more dimension as z,

- Where 
$$z = x^2 + y^2$$

• If we plot in z-y axis, a clear separation is visible and a line can be drawn.

 The line is the hyperplane we want and the transformation is a kind of kernel.





### **Demo Experiments**

#### Demo\_Non\_Linear\_SVM\_multi-class

The aim of this experiment is to perform binary and multi-class classification on Iris dataset and get the support vectors that are the basis for the max margin.

#### Demo\_Non-Linear\_SVM\_RBF

The aim of this experiment is to understand how non-linear separable data can be visualized linearly in a higher dimensional space



## Thanks!!

**Questions?**