

SUPPORT VECTOR MACHINE



- What is SVM?
- Ideology behind SVM
- Intuition Development
- Terminologies used in SVM
- How does it work?
- What is Kernel trick?

- Types of kernel
 - Polynomial Kernel
 - Gaussian RBF Kernel
- Support Vector Regression
- Pros and Cons of SVM
- Data preparation for SVM
- Use Case House Prices



What is SVM?

Supervised learning algorithm which can be used for,

Classification - Support Vector Classification (SVC)

Regression - Support Vector Regression (SVR)



What is SVM?

Used for smaller dataset as it takes too long to process.



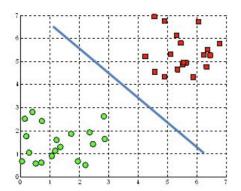
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The ideology behind SVM

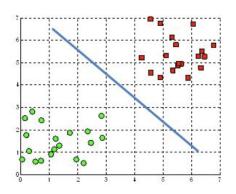
SVM is based on the idea of finding a **hyperplane** that best separates the features into different domains.

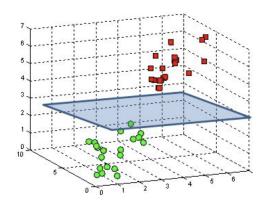




What is Hyperplane?

We can draw a line in 2-D, a plane in 3-D, anything beyond is a hyperplane.







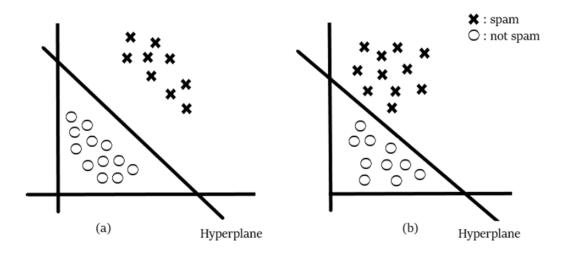
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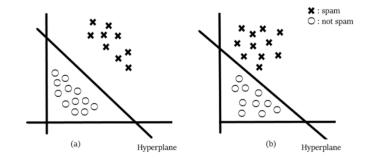
Classify emails as spam or ham using a function(hyperplane)

Two cases in which the hyperplane are drawn, which one will you pick and why?





I guess you would have picked the fig(a).
 Did you think why have you picked the fig(a)?

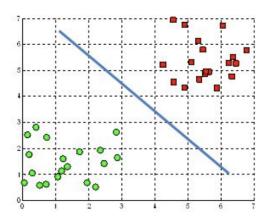


Because the emails in fig(a) are clearly classified and
 you are more confident about that as compared to fig(b).



Hyperplane(Decision Surface)

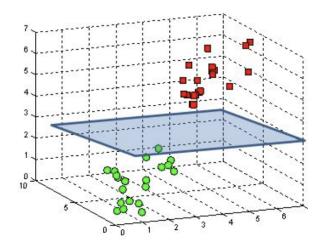
- The hyperplane is a function which is used to differentiate between features.
- In a 2-D data, the function used to classify is a LINE.





Hyperplane(Decision Surface)

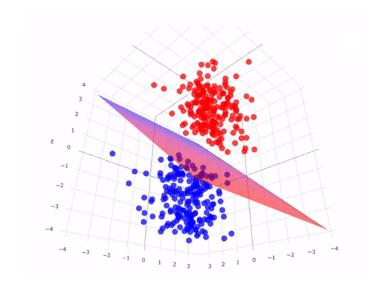
- The function used to classify the features in a 3-D data is a PLANE.
- Beyond 3-D its a HYPERPLANE.





Basically, SVM is composed of the idea of coming up with an Optimal hyperplane which will clearly classify the different classes.

(In this case they are binary classes)





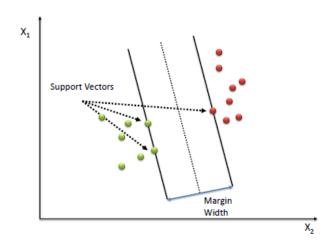
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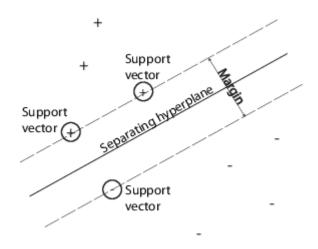
Terminologies used in SVM

- The points closest to the hyperplane are called SUPPORT VECTORS.
- The distance of the vectors from the hyperplane is called MARGIN.





Terminologies used in SVM





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How does it work?

We got accustomed to the process of segregating the two classes with a hyper-plane.

Now the burning question is "How can we identify the right hyper-plane?".

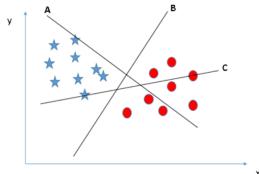


How does it work?

Scenario-1 (Linearly Separable Data)

Identify the right hyper-plane:

- Here, we have three hyper-planes (A, B and C).
- Now, identify the right hyper-plane to classify star and circle.



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How does it work?

Scenario-1 (Linearly Separable Data)

You need to remember a thumb rule to identify the right hyper-plane

- Select the hyper-plane which segregates the two classes best.
- In this scenario, hyper-plane "B"
 has excellently performed this job.

