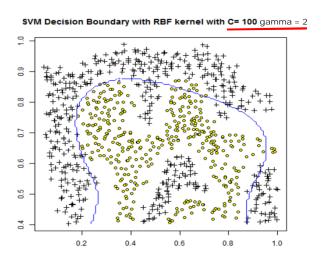
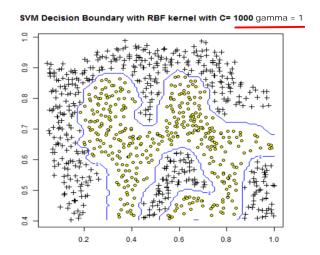


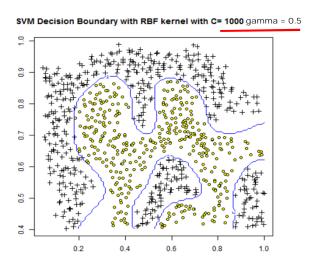
Observe the change in Decision Boundary for different values of C and gamma

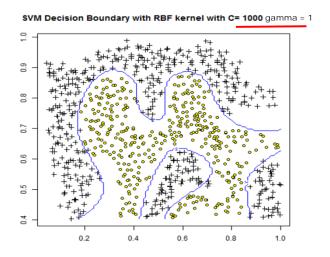






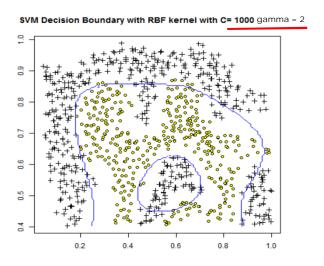
Observe the change in Decision Boundary for different values of C and gamma







Observe the change in Decision Boundary for different values of C and gamma





Appropriate values for C and gamma are arrived after examining Confusion Matrix



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

- Types of kernels
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Support Vector Regression (SVR)

As of now everyone is familiar with SVM or Support Vector Machine.

But SVR is different from SVM.

As the name suggests **SVR** is a **regression algorithm**, so we can use **SVR** for **Continuous values** instead of Classification which is SVC.



Terminologies related to SVR

Kernel

• The function used to map a lower dimensional data to higher dimensional data.

Hyper Plane

- In SVM this is basically the separation line between the data classes.
- In SVR we are going to define it as the line that will help us predict the continuous value or target value.

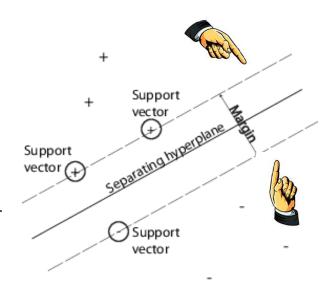


Terminologies related to SVR

Boundary line

 In SVM, there are two lines other than Hyper Plane which creates a margin.

 The support vectors can be on the Boundary lines or outside it. This boundary line separates the two classes. In SVR the concept is same.

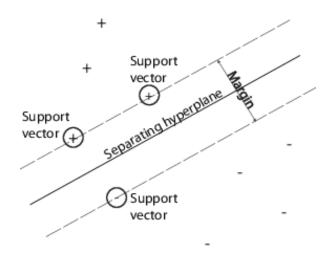




Terminologies related to SVR

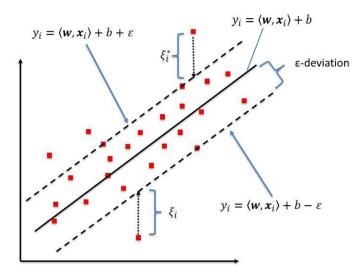
Support vectors

 This are the data points which are closest to the boundary. The distance of the points is minimum or least.





SVR



'ξ' (epsilon) is a hyper-parameter



SVR

Basically we are trying to decide a decision boundary at ' ξ ' distance from the original hyper plane such that data points closest to the hyper plane or the support vectors are within that boundary line.



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Pros and cons of SVM

Pros

- It is really effective in the higher dimension.
- Effective when the number of features are more than training examples.
- Best algorithm when classes are separable.
- The hyperplane is affected by only the support vectors thus outliers have less impact.



Pros and cons of SVM

Cons

- For larger dataset, it requires a large amount of time to process.
- Does not perform well in case of overlapped classes.
- Selecting, appropriately hyper parameters of the SVM that will allow for sufficient generalization performance.
- Selecting the appropriate kernel function can be tricky.



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Preparing data for SVM

Numerical Conversion

- SVM assumes that your inputs are numerical instead of categorical.
- So you can convert them using one of the most commonly used one hot encoding or label-encoding etc.



Preparing data for SVM

Binary Conversion

Since SVM is able to classify only binary data so you would need to convert
the multi-dimensional dataset into binary form using (one vs the rest method /
one vs one method) conversion method.



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Use Case: House Prices

Ask a home buyer to describe their dream house, and they probably won't begin
with the height of the basement ceiling or the proximity to an east-west railroad.

But this dataset proves that much more influences price negotiations than
the number of bedrooms or a white-picket fence.



Use Case: House Prices

 With 81 explanatory variables describing (almost) every aspect of residential homes in Ames, lowa, this dataset challenges you to predict the final price of each home.