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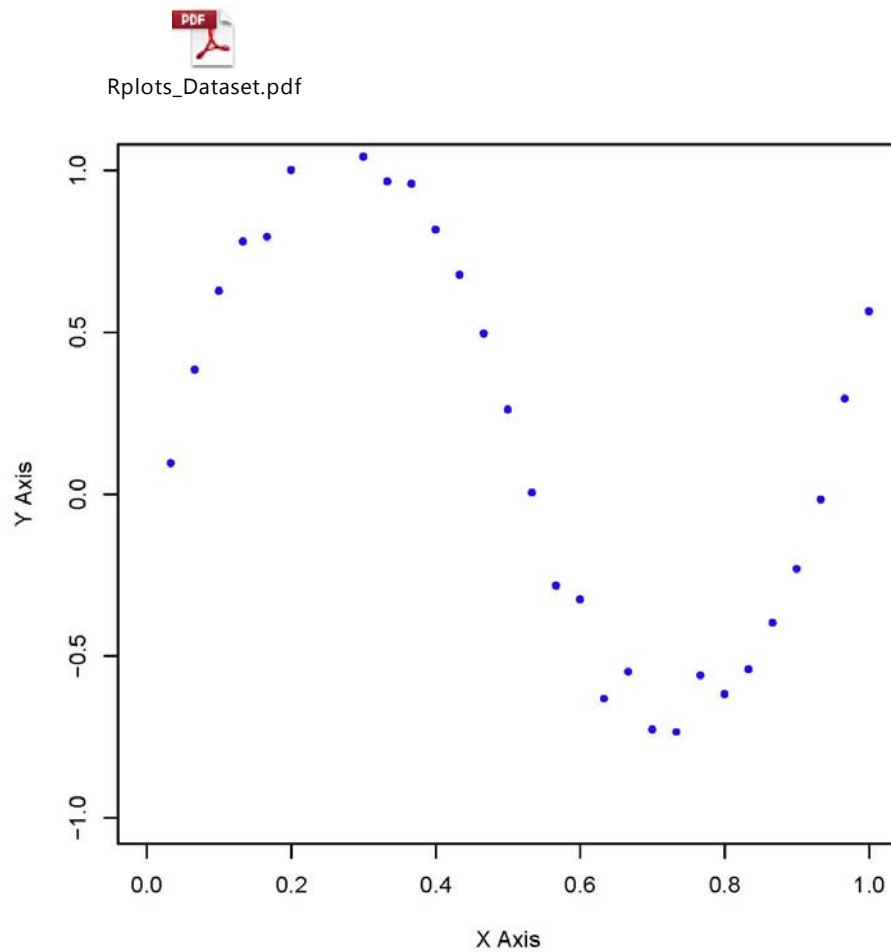
## Machine Learning -- Assignment 1

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5.

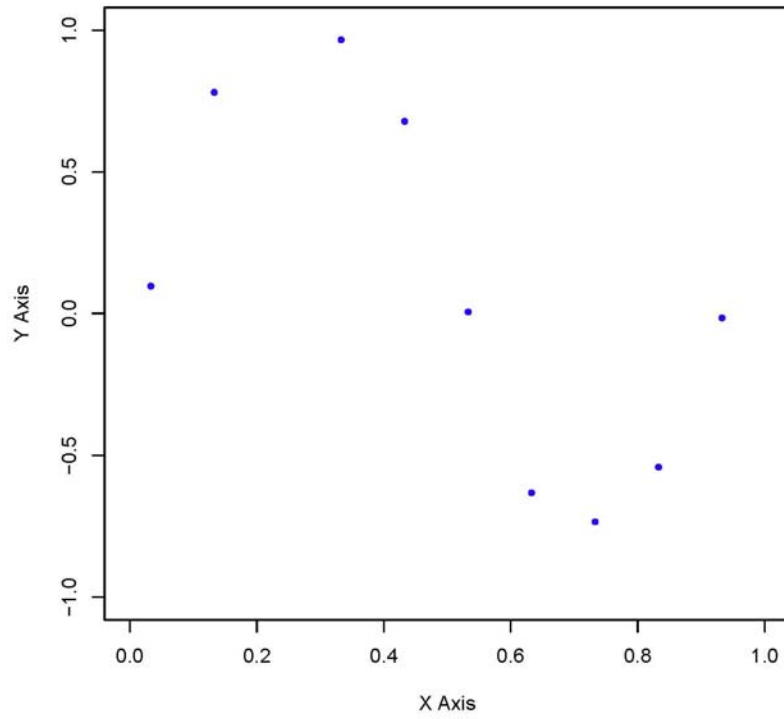
a. Plot for the complete Dataset



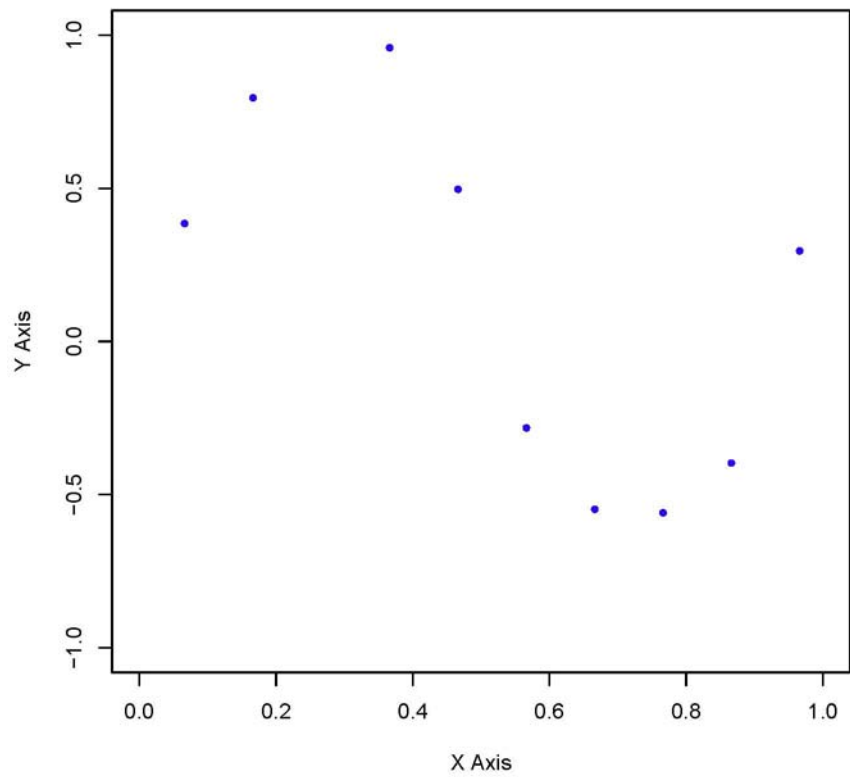
b. Plot for Train, Test and Validation data

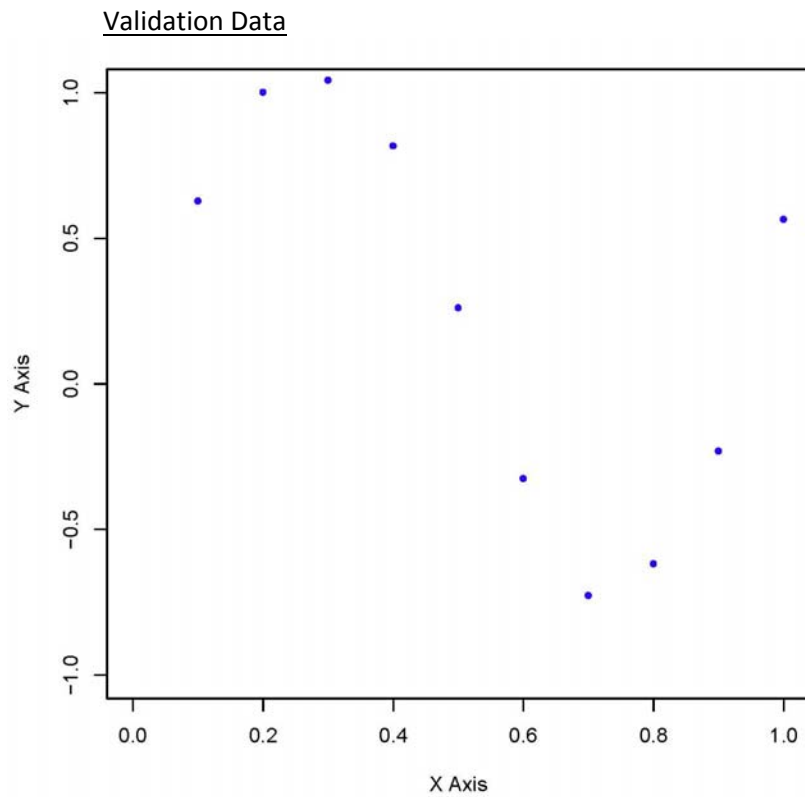


Training Data

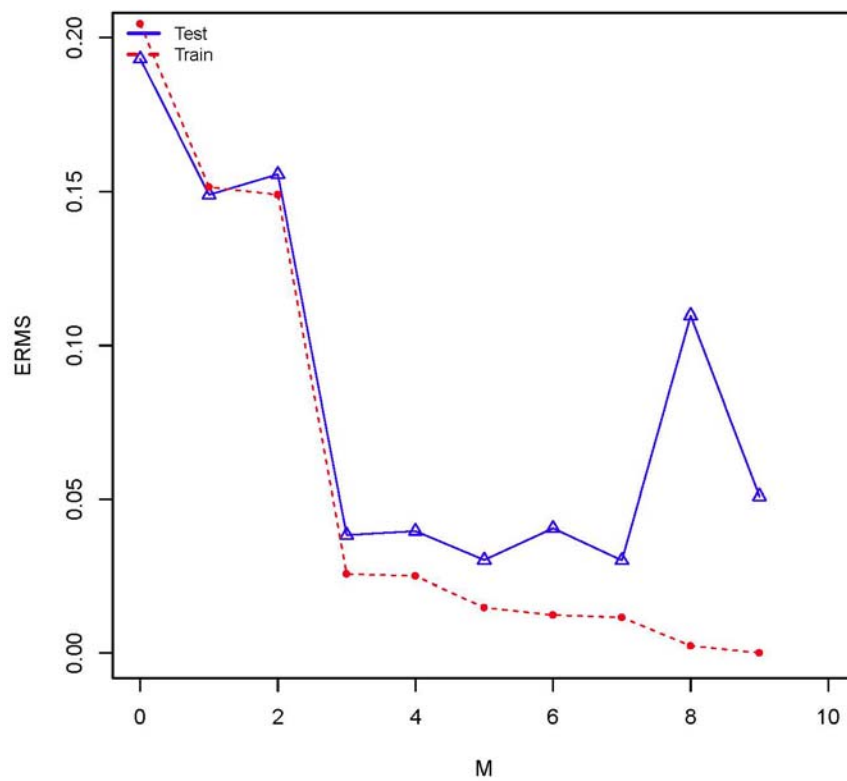


Test Data

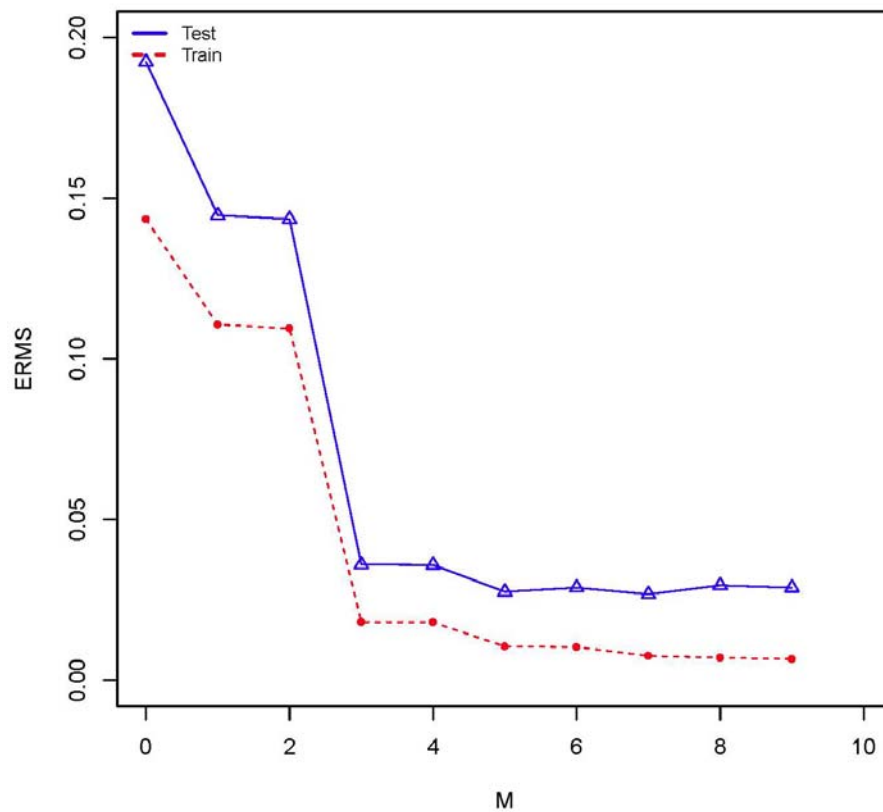




c. 1. Without Regularization – M vs ERMS values for Training and Test Data



c. 2. Without Regularization – M vs ERMS values for Training + Validation Data and Test Data



 Rplots\_MvsERMS\_WithoutReg\_Train&

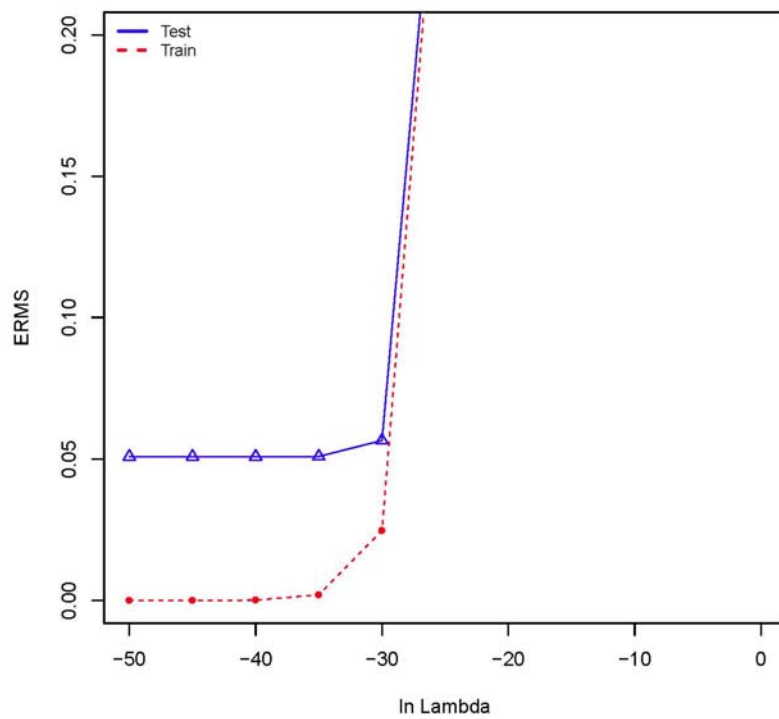
In Without Regularization, with just use *Train.dat* (i.e. training data). We can see that M-ERMS Graph for Test data is consistent in  $M = [3 \text{ to } 7]$  with some distortion. But when the training data is increased to Train + Validation data these distortions are gone & is smooth.

When  $M = 8/9$  with just *Train.dat* data, ERMS value is high but after introducing validation data into training we can observe that it is smoothed out and is consistent with the rest of graph.

One more observation would be the ERMS values of the training data prediction is considerably reduced after introducing validation data into training data set.

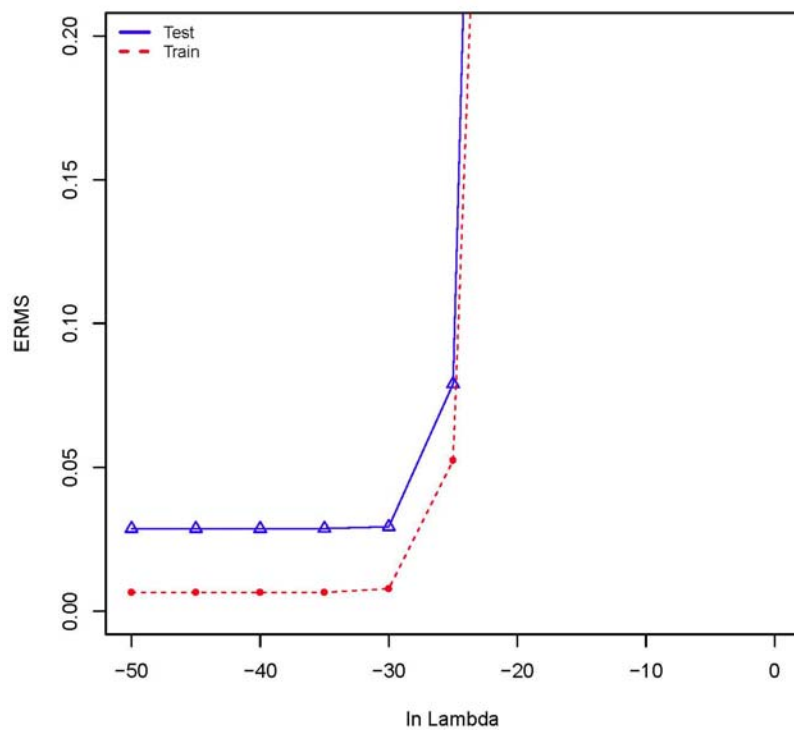
So, we can see that introducing more data reduced the ERMS values i.e. deviation of the predicted value from the original is getting reduced. It even stabilized the error rate as well from  $M = 3$  to  $M = 9$  for this data set.

c. 3. With Regularization –  $\ln \lambda$  vs ERMS value for Training Data & Test Data (M=9)



 Rplots\_MvsERMS\_  
WithReg\_TrainingD:

With Regularization –  $\ln \lambda$  vs ERMS value for Training + Validation Data & Test Data (M=9)



 Rplots\_MvsERMS\_  
WithReg\_Train&Vali

We introduce regularization if we need a good prediction even from few training examples. Here we can see for  $\ln \lambda = -50$  to  $\ln \lambda = -30$ , ERMS values are consistent and then it shoots to really high values. After introducing the validation data into training data we observe that the ERMS value almost reduced by 50%. i.e. our purpose of introducing  $\ln \lambda$  is served.

An interesting observation along with this is that the ERMS values of the training + validation data increased than that of the training data, this might be caused due to the extra data which might be introducing noise.

One more observation would be the ERMS value for  $\ln \lambda = -25$  is visible in the graph (i.e within range of 0.2) for the training + validation data.

If we have to choose an  $\ln \lambda$  value then I would suggest to do it in between -50 to -30 as this range has the least & consistent ERMS values.