# Machine Learning (CS60050)

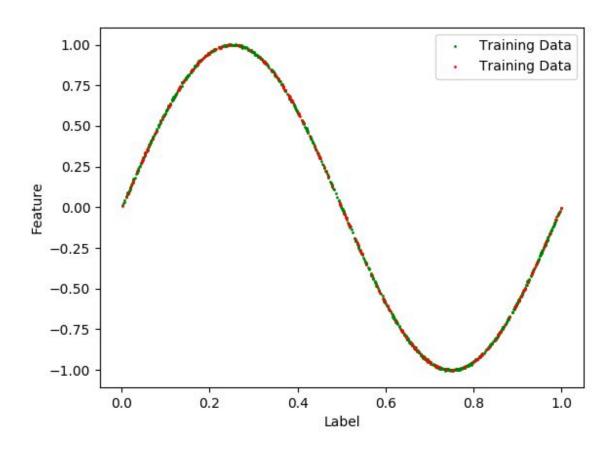
Assignment 1: Linear Regression

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## Part 1

## Feature VS. Lable Plot:



## Parameters for degree from 1 to 9:

**Degree = 1**: [0.916085945138095, -1.8551469696207938]

**Degree = 2**: [0.9737442801835667, -2.200960660808183, 0.3405230762705207]

**Degree = 3**: [-0.0767107622444632, 10.506474965090034, -31.22341989014763, 20.91009547657465] **Degree = 4**: [0.0830115193715623, 7.1789094988642725, -15.648234832635653, -3.949813465235262,

12.653836818948276]

**Degree = 5**: [0.19214933532351658, 5.409107429635385, -10.246347927458828, -4.97039620428527,

2.5245300301296223, 7.54786321066668]

**Degree = 6**: [0.07188411988445503, 7.232111300967344, -15.800752702779072, -2.2153445033059707, 7.143565739411272, 6.203138432158996, -2.3073032846351706]

**Degree = 7**: [0.034094258090568, 7.646632933024831, -16.094316704277727, -3.7246187987922417, 6.588480485119026, 8.273741039046474, 3.2233205566807532, -5.752931875305976]

**Degree = 8**: [0.036915992631697155, 7.446733468235627, -14.807237270859819, -4.961867156594991, 4.648066523434921, 7.862705552940834, 5.78870314623487, 0.48898897853438483, -6.414490117455645] **Degree = 9**: [0.056302115740820786, 7.054577620218062, -13.32569986899939, -5.573456046397797, 2.9354429400924467, 6.661375897484483, 6.242060454084398, 3.2162208799395717, -1.1709198384755746, -6.09583832577531]

#### **Test Error:**

#### Degree 1:

squared error on test data: 0.09553046711906686

Degree 2:

squared error on test data: 0.09579851086810237

Degree 3:

squared error on test data: 0.0032488537705092917

Degree 4:

squared error on test data: 0.00467533276939405

Degree 5:

squared error on test data: 0.008860184012000028

Degree 6:

squared error on test data: 0.004590779132340806

Degree 7:

squared error on test data : 0.0023335021565723587

Degree 8:

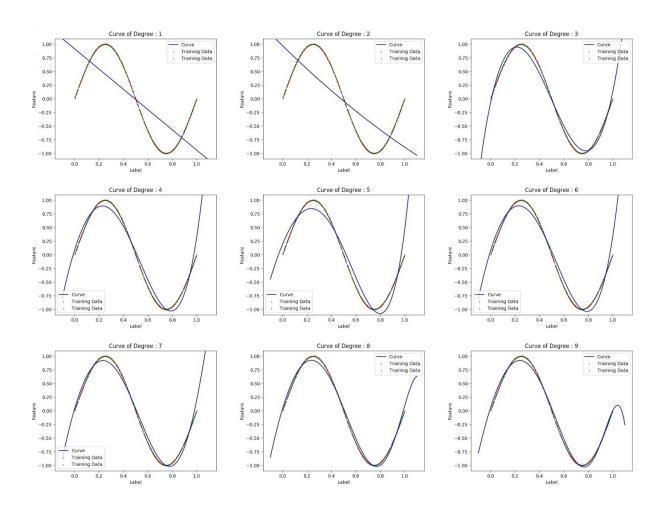
squared error on test data: 0.0014136056818152085

Degree 9 :

squared error on test data: 0.0012161062834012406

## Part 2

## Plots for Degree from 1 to 9:



#### Squared error on both train and test data for Degree from 1 to 9:

#### Degree 1:

squared error on train data: 0.09968054192477488 squared error on test data: 0.09553046711906686

Degree 2:

squared error on train data : 0.09914021585733448 squared error on test data : 0.09579851086810237

Degree 3:

squared error on train data : 0.0032392440501344487 squared error on test data : 0.0032488537705092917

Degree 4:

squared error on train data : 0.004617754619299829 squared error on test data : 0.00467533276939405

Degree 5:

squared error on train data: 0.008652737686857016 squared error on test data: 0.008860184012000028

Degree 6:

squared error on train data: 0.004544174186980676 squared error on test data: 0.004590779132340806

Degree 7:

squared error on train data : 0.0023375452518882484 squared error on test data : 0.0023335021565723587

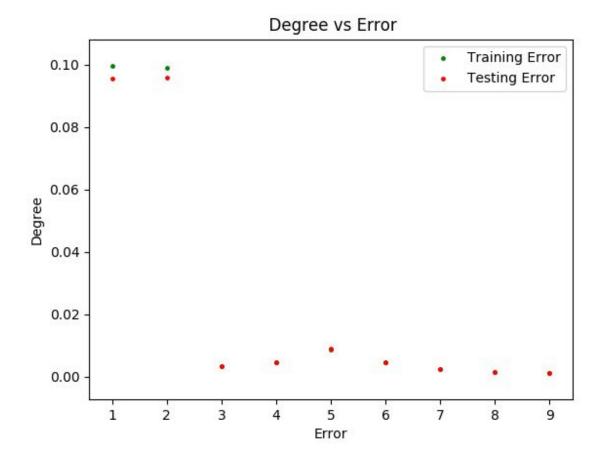
Degree 8:

squared error on train data: 0.0014320874611304841 squared error on test data: 0.0014136056818152085

Degree 9:

squared error on train data: 0.0012228187405432117 squared error on test data: 0.0012161062834012406

### PLOT for Squared error VS Degree :



N=9 is Most suitable for this data.

Because it has least Squared Error on Test set as well as Train set.

And From all plots also we can see that it's (N=9) Best fitted plot.

## Part 3

Degree 1 and 9 have maximum and minimum training error Respectively. We will apply Regulerisation on Degree 1 and 9.

#### Error for Lasso Regression :

#### lambda = 0.25

```
squared error on train data when deg = 1 is: 0.09969039597658721 squared error on test data when deg = 1 is: 0.09547685812533711 squared error on train data when deg = 9 is: 0.00833958390752358 squared error on test data when deg = 9 is: 0.008674067531898472 lambda = 0.5
```

squared error on train data when deg = 1 is: 0.09968771024528741 squared error on test data when deg = 1 is: 0.09548584359464295 squared error on train data when deg = 9 is: 0.008328697086353483 squared error on test data when deg = 9 is: 0.00867062036937101 lambda = 0.75

squared error on train data when deg = 1 is: 0.09968536369618507 squared error on test data when deg = 1 is: 0.0954953911756938 squared error on train data when deg = 9 is: 0.008320493681063953 squared error on test data when deg = 9 is: 0.008669776731771826 lambda = 1

squared error on train data when deg = 1 is: 0.09968349788979701 squared error on test data when deg = 1 is: 0.09550480050902876 squared error on train data when deg = 9 is: 0.008314873979916327 squared error on test data when deg = 9 is: 0.008671431348204252

### Error for ridge Regression

#### lambda = 0.25

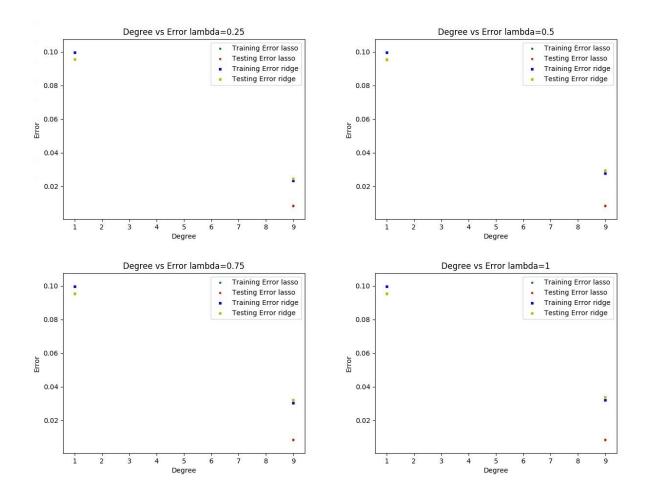
squared error on train data when deg = 1 is: 0.09970294075305704 squared error on test data when deg = 1 is: 0.09544740250140471 squared error on train data when deg = 9 is: 0.02342223373954379 squared error on test data when deg = 9 is: 0.02467554945441318 lambda = 0.5

squared error on train data when deg = 1 is: 0.09971469545303899 squared error on test data when deg = 1 is: 0.09542908883407403 squared error on train data when deg = 9 is: 0.02783351642416298 squared error on test data when deg = 9 is: 0.029390701313720194 lambda = 0.75

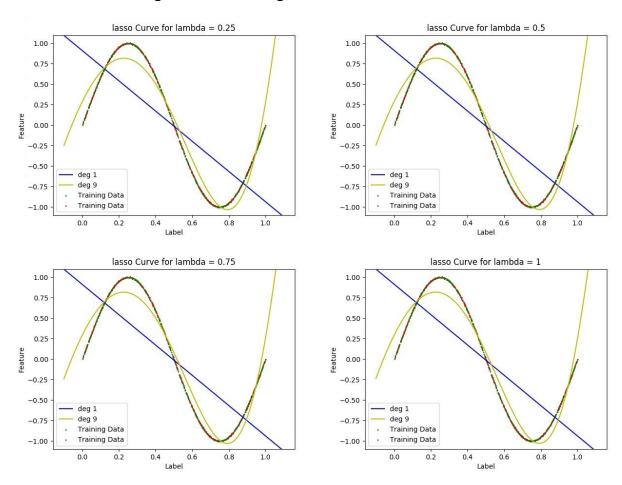
squared error on train data when deg = 1 is: 0.09972855171380436 squared error on test data when deg = 1 is: 0.09541332468978343 squared error on train data when deg = 9 is: 0.03024756106826543 squared error on test data when deg = 9 is: 0.031967315840444585 lambda = 1

squared error on train data when deg = 1 is: 0.0997448124619883 squared error on test data when deg = 1 is: 0.09539964783083132 squared error on train data when deg = 9 is: 0.031996045171430836 squared error on test data when deg = 9 is: 0.03381972342843529

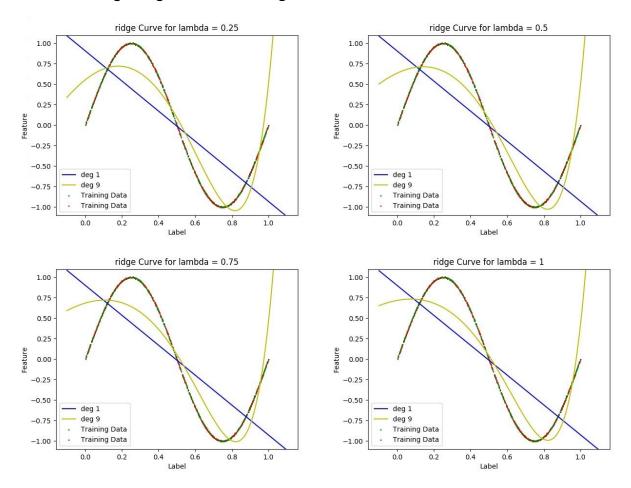
## PLOT for Squared error VS Degree :



## Plot for Lasso regression on degree 1 and 9:



#### Plot for Ridge regression on degree 1 and 9:



Lasso regression tries to reduce value of parameter i.e. w can be large -ve provided it's local minima.

But Ridge regression tries to reduce magnitude of parameter.

I will prefer Lasso regression Because it has less Error than ridge and also from plots we can see Lasso regression Model fits data well.