

# Assignment 1

ELL784

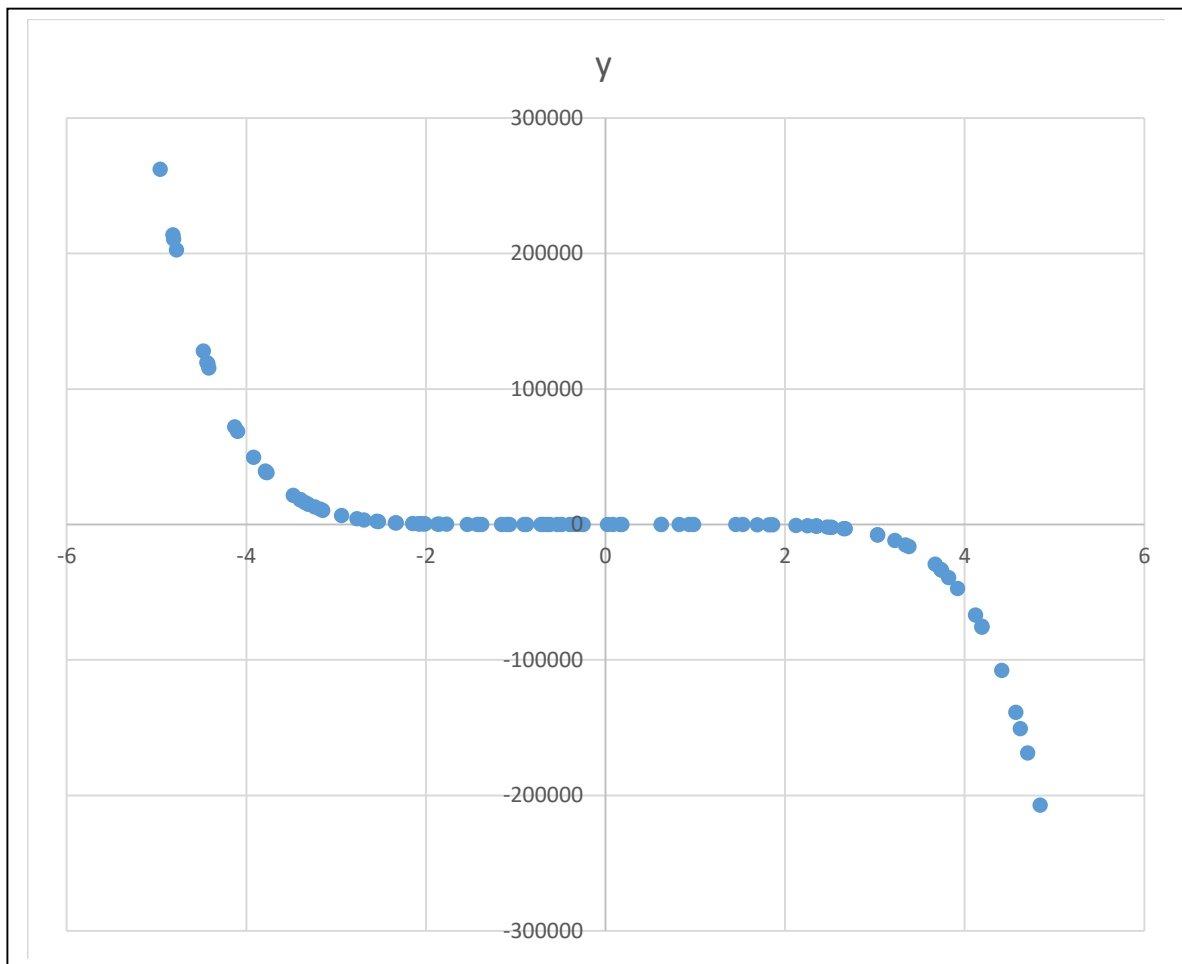
Submitted By: Harshdeep Gupta

Entry No- 2013MT60597

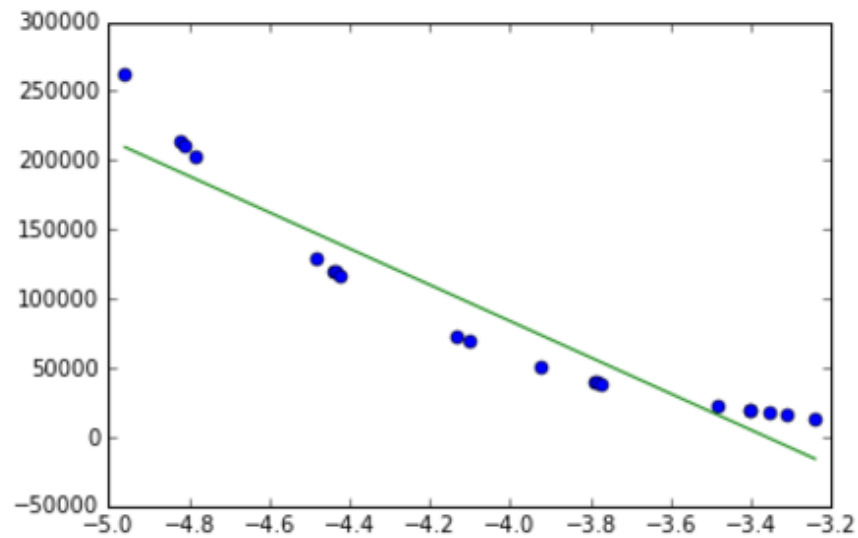
## Part 1

The Data:-

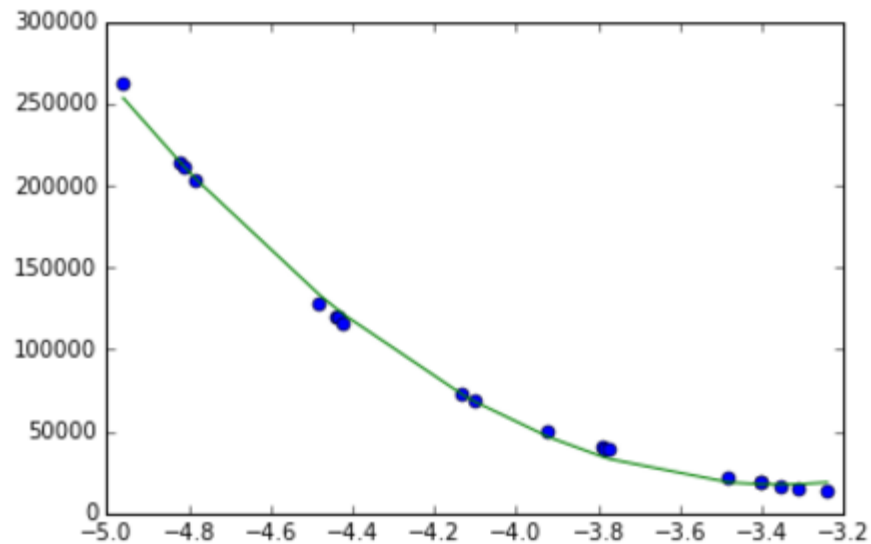
The plot of the data as given:



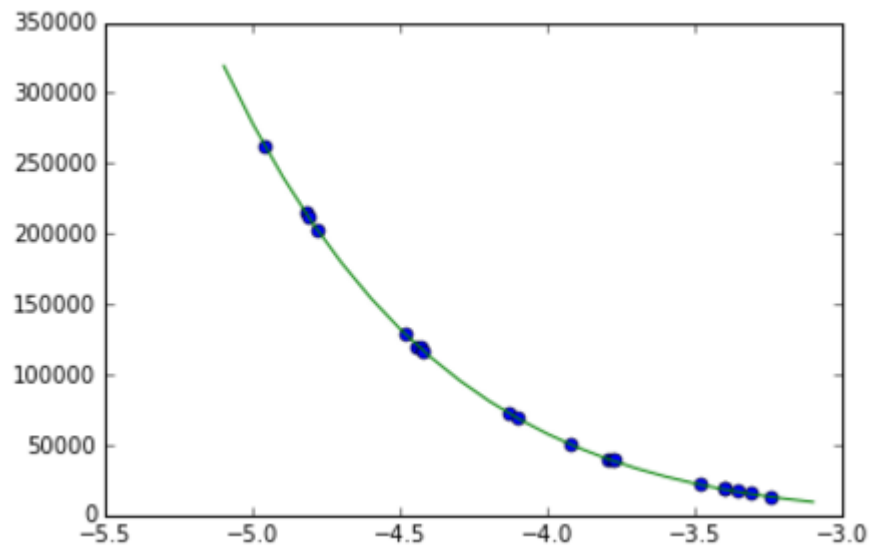
- Using the First twenty points only
  - **Regression with Linear Basis Functions, no regularization**
  - $\Theta = [-131101.94166835 \ -440682.09031097]$
  - RMS error = 103740.46909601474



- **Regression with Quadratic Basis Functions, no regularization**
- $\Theta = [92093.62943895 \ 618787.04644468 \ 1056948.65695317]$
- RMS error = 17797.548636733027



- **Regression with  $X^5$  Basis Functions, no regularization**
- $\Theta = [-1559.67288411 \ -22514.55413373 \ -142887.30678472 \ -475289.15934071$
- $-812414.81955637 \ -565601.85879763]$
- RMS error = 1560.3124785126483



○

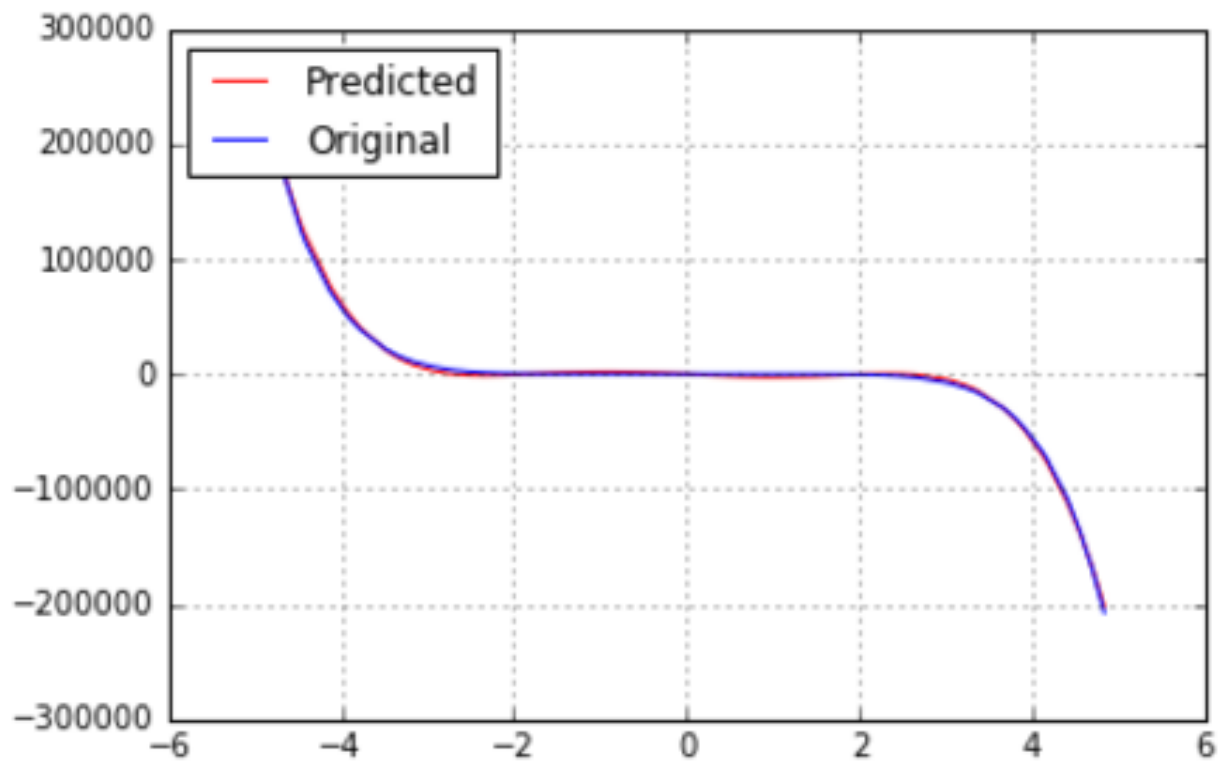
- Estimation of Beta:
- **Using all the 100 points in the dataset**
  - For obtaining optimal results, this is what I did:
    - Ran regression for all polynomials with degree 1 to 10
    - For each polynomial, put alpha from 0 to 30, adding 3 to each previous value to generate 10 values of alpha

#### ○ *Result*

For the given data, I found that a fifth order polynomial, with regularization coefficient set to 10 works pretty well (gives the least cross validation error). The parameters of my model are  $[a_0, a_1, \dots, a_5]$ :

$[-117.57479698 \ -3575.76286303 \ -199.62516982 \ 1448.89434622 \ 22.37773588 \ -134.19724626]$ .

The final fit looks like:



Estimation of Beta

$$\frac{1}{\beta_{\text{ML}}} = \frac{1}{N} \sum_{n=1}^N \{y(x_n, \mathbf{w}_{\text{ML}}) - t_n\}^2.$$

Therefore, using the above model, and plugging the values we get our beta as :-

0.08457689868069149

## Part 2:

Here, we use a multivariate regression model. The parameters which are important for obtaining low errors are:

- Suitable choice of basis functions
- Choosing a suitable value of Regularization parameter ( $\alpha$ )

- To choose a suitable  $\alpha$ , we need to apply proper cross validation techniques and use that alpha that gives the least error according to some error function.
- **Results**
- In my final model I use upto quadratic basis functions of all the variables to capture nonlinear dependencies on the data. However I don't multiply two different variables with one another.
- For finding a suitable value of  $\alpha$ :
  - I used 10-fold cross validation
  - Scanned for values of from  $10^{-15}$  to 1, and using that one which gives the lowest cross validation error.
- When I tried to include cubic basis functions into my model, I saw that my error on test data increased, which shows that my model was over fitting my training data. So I switched back to using quadratic basis functions.
- During cross validation, I found that setting alpha too low ( $<10^{-14}$ ) or too high ( $>10^{-12}$ ) increased the amount of error. The lowest amount of cross validation error that I got was when alpha was set to  $10^{-13}$ .

So the best results are obtained using a quadratic model with ridge shrinkage. During the time of writing this report, my rank of leaderboard is seven, my best rank being five.0