

Analysis Report

1 Normal Equation and Gradient Descent Algorithm without regularizer :-

1.1 Hypothesis parameters using Normal Equation Algorithm are :-

We got idea that if our matrix is singular then we could make it invertible by eliminating redundant features or we could make a new feature using the existing feature to eliminate the problem of singularity of matrix. Following are the results obtained after using mean normalization.

$$\theta = [[1.37658083e + 03][2.15081925e - 01][1.11110312e + 02][8.69446160e + 02][3.97676261e + 02][4.05611236e + 02][2.73607644e + 02][3.30685106e + 02][7.78219890e + 02][7.66179987e + 02][2.57447262e + 02][5.68257407e + 02]]$$

1.2 Hypothesis parameters using Gradient Descent Algorithm are:-

We used different $\alpha = 0.01, 0.001, 0.0001$ and epoch = 100, 1000, 10000 respectively. Finally our algorithm converged best at $\alpha = 0.0001$ and epoch = 10000 with minimal possible error.

$$\theta = [[-0.14265545][0.13801749][0.0412469][0.04138947][0.04141092][0.04133055][0.04139053][0.0413724][0.0413582][0.04144458][0.04146955][0.04140794]]$$

1.3 Performance of predictor/hypothesis in both the cases :-

Ans. Performance of the predictor in normal equation was found to be better than gradient descent since the calculated squared error in gradient was 0.42740064225945834 whereas in normal equation it was 0.42128128593783093. In general we got the idea that if the number of features are more and data points are less then gradient descent will perform much better as the time complexity of normal equation is $O(n^3)$ but if the number of features are less and data points are more then normal equation will perform better.