

Test1

Patel Femina

Machine Learning

N03094115

```
% converting wine.csv to wine.mat
```

```
Wine=csvread('wine.csv');  
save('wine.mat','Wine')
```

Question 1

```
% Taking the first 1280 data rows and calling them the training set  
trainingset=Wine(2:1281,:);
```

Question 2

```
% Taking the remaining rows and calling them the test set  
testset=Wine(1282:1600,:);
```

Question 3

```
% Model M1 using training set  
x=trainingset(:,[1 7]);  
% columns 1 and 7 are for acidity and sulphite level of wine respectively  
y=trainingset(:,9);  
% column 9 is for quality of wine  
[J_h,t_h]=testGradientDescend(x,y,1.0,0.5,0.5,0.01,2000)  
% Given values for  $\Theta_0$ ,  $\Theta_1$ ,  $\Theta_2$ ,  $\alpha$  and max Iterations are 1.0, 0.5, 0.5, 0.01, 2000 respectively
```

% value of J_h goes from 0.62720 to 0.35760 and the corresponding Θ values are 1, 0.5, 0.5 to %2.96316, 0.19935, 1.40583. Lower the value of cost function, better the model.

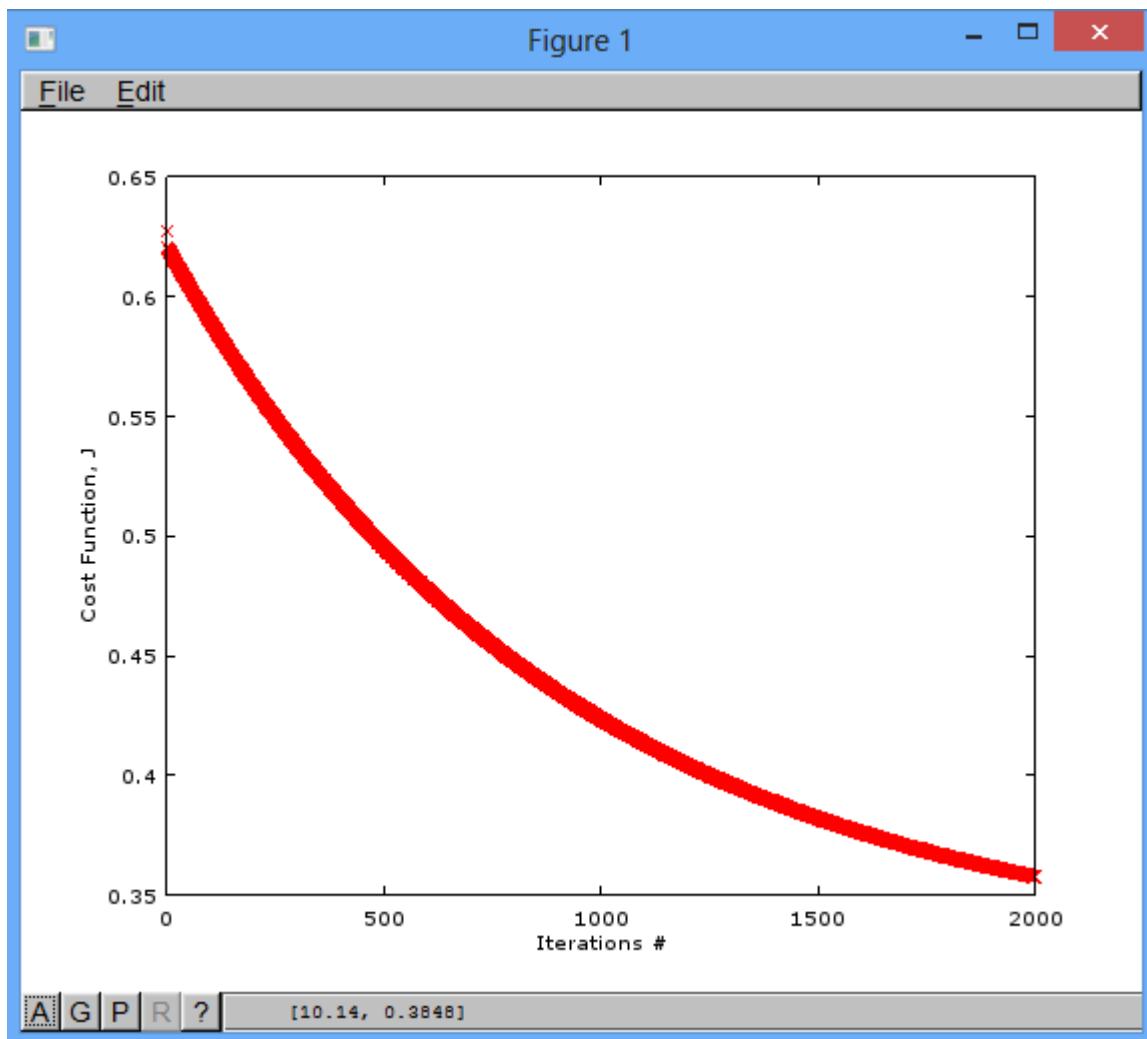
% So $\theta_0=2.96316$, $\theta_1=0.19935$ and $\theta_3=1.40583$

% Graph of the cost function as the iterations proceed for Model M1

```
plot(J_h,'rx');
```

```
xlabel('Iterations #');
```

```
ylabel('Cost Function, J');
```



Hypothesis function for model M1, $h_{\theta}(x) = \Theta_0 * x_0 + \Theta_1 * x_1 + \Theta_2 * x_2$

By default we take $x_0=1$

So equation is $\Theta_0 + \Theta_1 * x_1 + \Theta_2 * x_2$

Placing the values of Θ_0 , Θ_1 and Θ_2 from above we get

$$h_{\theta}(x) = 2.96316 + 0.19935 * x_1 + 1.40583 * x_2$$

Question 4

```
% Model M2 using training set
```

```
x1=trainingset(:,[3 6]);
```

```
% columns 3 and 6 are for sugar and pH level of wine respectively
```

```
y1=trainingset(:,9);
```

```
% column 9 is for quality of wine
```

```
[J_h,t_h]=testGradientDescend(x1,y1,1.0,0.5,0.5,0.01,2000)
```

```
% Given values for  $\Theta_0$ ,  $\Theta_1$ ,  $\Theta_2$ ,  $\alpha$  and max Iterations are 1.0, 0.5, 0.5, 0.01, 2000 respectively
```

```
% value of J_h goes from 2.04119 to 0.35864 and the corresponding  $\Theta$  values are 1, 0.5, 0.5 to  
% 1.450857, 0.038766, 1.243855. Lower the value of cost function, better the model.
```

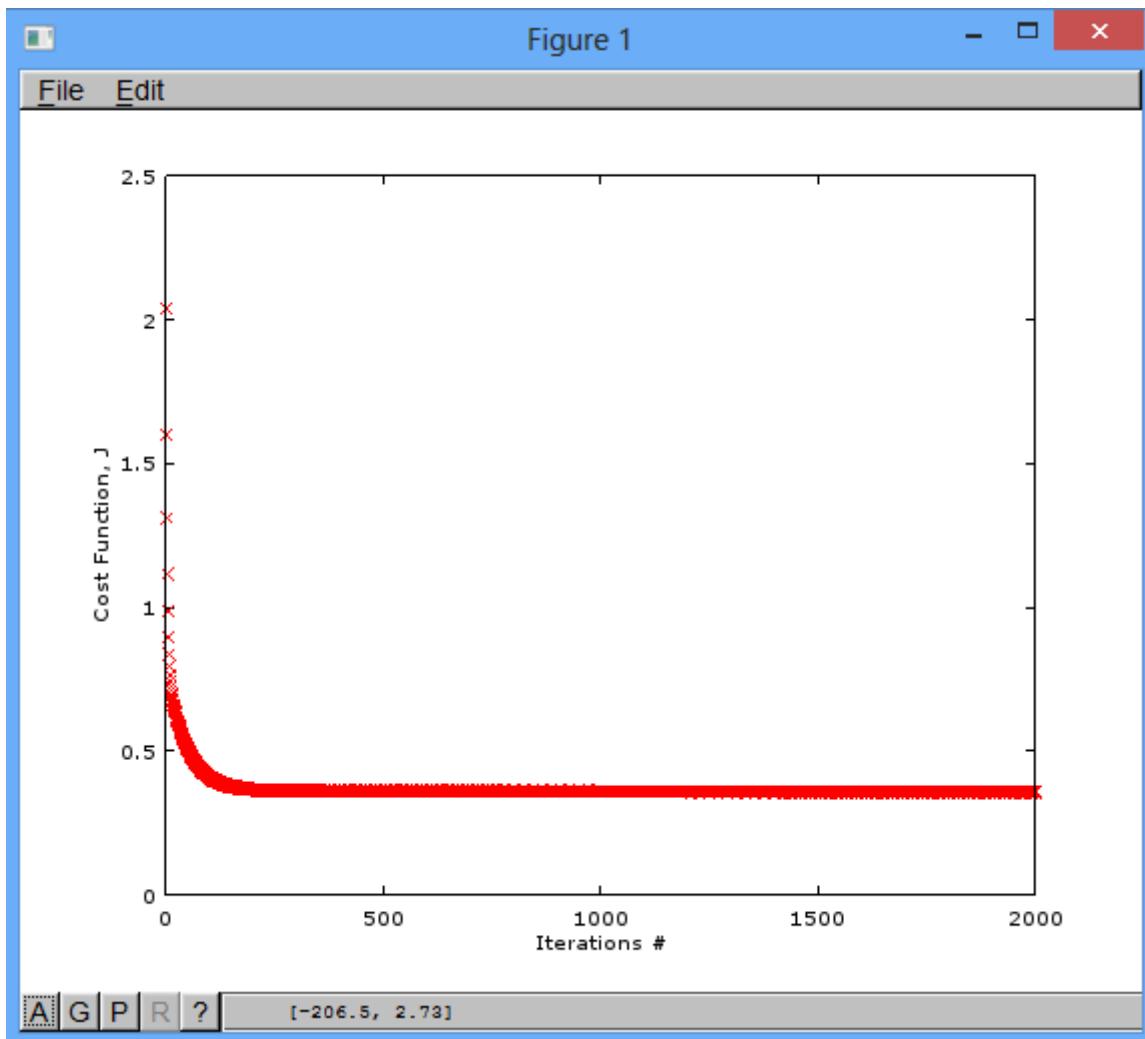
```
% So  $\theta_0=1.450857$ ,  $\theta_1=0.038766$  and  $\theta_2=1.243855$ 
```

```
% Graph of the cost function as the iterations proceed for Model M2
```

```
plot(J_h,'rx');
```

```
xlabel('Iterations #');
```

```
ylabel('Cost Function, J');
```



Hypothesis function for model M2, $h_{\theta}(x) = \Theta_0 * x_0 + \Theta_1 * x_1 + \Theta_2 * x_2$

By default we take $x_0=1$

So equation is $\Theta_0 + \Theta_1 * x_1 + \Theta_2 * x_2$

Placing the values of Θ_0 , Θ_1 and Θ_2 from above we get

$$h_{\theta}(x) = 1.450857 + 0.038766 * x_1 + 1.243855 * x_2$$

Question 5

```
% Evaluating cost function using Θ values of M1 to predict the quality of wine in the test data set
```

```
x3=testset(:,[1 7]);
```

```
y3=testset(:,9);
```

```
costFunForMultivariable(x3, y3, 2.96316,0.19935,1.40583)
```

```
ans = 0.32174
```

```
% Evaluating cost function using Θ values of M2 to predict the quality of wine in the test data set
```

```
x4=testset(:,[3 6]);
```

```
y4=testset(:,9);
```

```
costFunForMultivariable(x4, y4, 1.450857,0.038766,1.243855)
```

```
ans = 0.33412
```

As the value of cost function of Model M1 is lower than that of Model M2, **M1 is a better model.**