

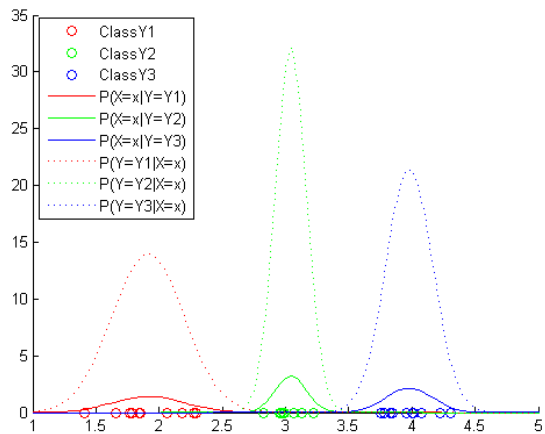
CSL407 Machine Learning

HW2

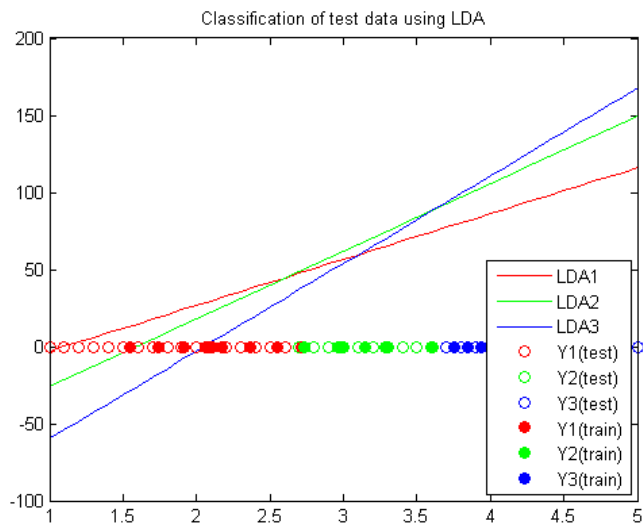
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Ans1

Plot1



Plot2



Ans2

Regularized Logistic Regression Cost Function

$$J(\theta) = -\frac{1}{m} \sum_{i=1}^m y^{(i)} \log h_{\theta}(x^{(i)}) + (1 - y^{(i)}) \log (1 - h_{\theta}(x^{(i)})) + \frac{\lambda}{2m} \sum_{j=1}^n \theta_j^2$$

Derivation of weight update equation for the Logistic regression function without regularization

$$\frac{\delta J}{\delta \theta_j} = -\frac{1}{m} \sum_{i=1}^m \left(y^i \left(\frac{1}{h_\theta(x^i)} \right) + (1 - y^i) \left(\frac{-1}{1 - h_\theta(x^i)} \right) \right) \frac{\delta h_\theta(x^i)}{\delta \theta_j}$$

Solving which gives

$$1) \frac{\delta J}{\delta \theta_j} = -\frac{1}{m} \sum_{i=1}^m \frac{y^i - h_\theta(x^i)}{h_\theta(x^i)(1 - h_\theta(x^i))} \frac{\delta h_\theta(x^i)}{\delta \theta_j}$$

Since,

$$h_\theta(x^i) = \frac{1}{1 + e^{-\theta^T x^i}}$$

Solving for $\frac{\delta h_\theta(x^i)}{\delta \theta_j}$

$$\begin{aligned} 2) \frac{\delta h_\theta(x^i)}{\delta \theta_j} &= \frac{-1}{(1 + e^{-\theta^T x^i})^2} (-x_j^i) (e^{-\theta^T x^i}) = (h_\theta(x^i))^2 (x_j^i) \left(\frac{1}{h_\theta(x^i)} - 1 \right) \\ &= h_\theta(x^i)(1 - h_\theta(x^i))x_j^i \end{aligned}$$

Substituting 2) in 1)

We get

$$3) \frac{\delta J}{\delta \theta_j} = -\frac{1}{m} \sum_{i=1}^m \frac{y^i - h_\theta(x^i)}{h_\theta(x^i)(1 - h_\theta(x^i))} h_\theta(x^i)(1 - h_\theta(x^i))x_j^i = \frac{1}{m} \sum_{i=1}^m (h_\theta(x^i) - y^i)x_j^i$$

For regularized logistic regression,

To 3) we add the derivative of the regularized term

So,

For j=0

$$\frac{\delta J}{\delta \theta_0} = \frac{1}{m} \sum_{i=1}^m (h_\theta(x^i) - y^i)x_0^i$$

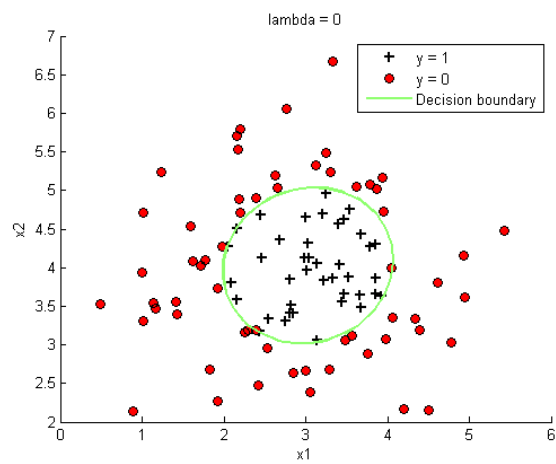
And for j=1,2,3,...,n

$$\frac{\delta J}{\delta \theta_j} = \frac{1}{m} \sum_{i=1}^m (h_\theta(x^i) - y^i)x_j^i + \frac{\lambda}{m} \theta_j$$

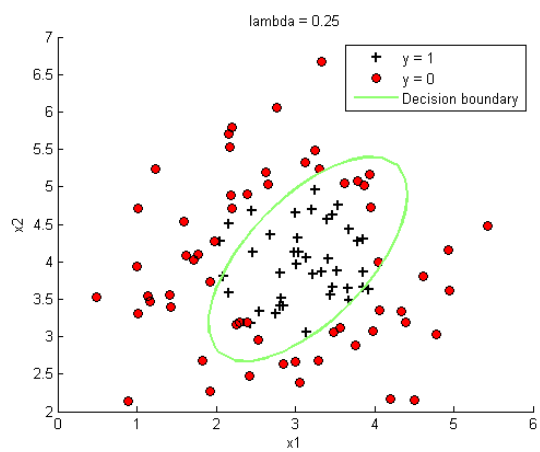
Ans3

Varying the lamda values in the set {0,0.25,0.5,0.75,1}

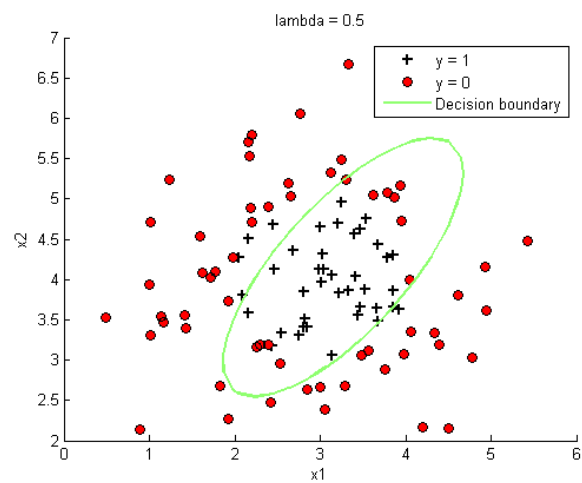
Lambda=0, Training Accuracy=98%(Overfitting)



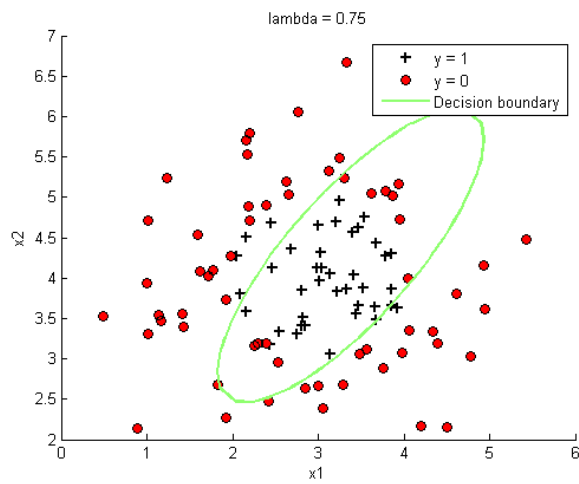
Lambda=0.25, Training Accuracy-86%



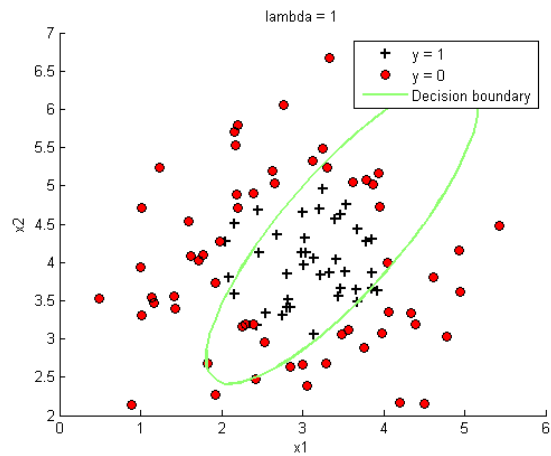
Lambda=0.50, Training Accuracy=83%



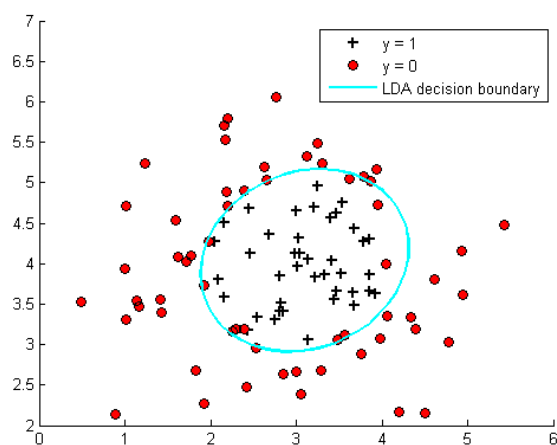
Lambda=0.75, Training Accuracy=82%



Lambda=1, Training Accuracy=81%(Underfitting)



The linear decision boundary(very similar to plotdecisionboundary where lambda=0)



Ans4

a)

$x_1=5\text{hrs}, x_2=7.5\text{gpa}$

So,

$$\theta^T X = -8 + .05(5) + 7.5 = -0.25$$

Thus,

$$h_{\theta}(x) = \frac{1}{1 + e^{0.25}} = 0.4378$$

b)

Since,

$$h_{\theta}(x) = \frac{1}{1 + e^{-\theta^T X}} = 0.6$$

$$\theta^T X = -\ln\left(\frac{1}{.6} - 1\right) = \ln\left(\frac{3}{2}\right) = -8 + 0.05(x_1) + 7.5$$

Solving for x_1

We get $x_1=(0.5+\ln(3/2))/0.05=18.11\text{hrs}$