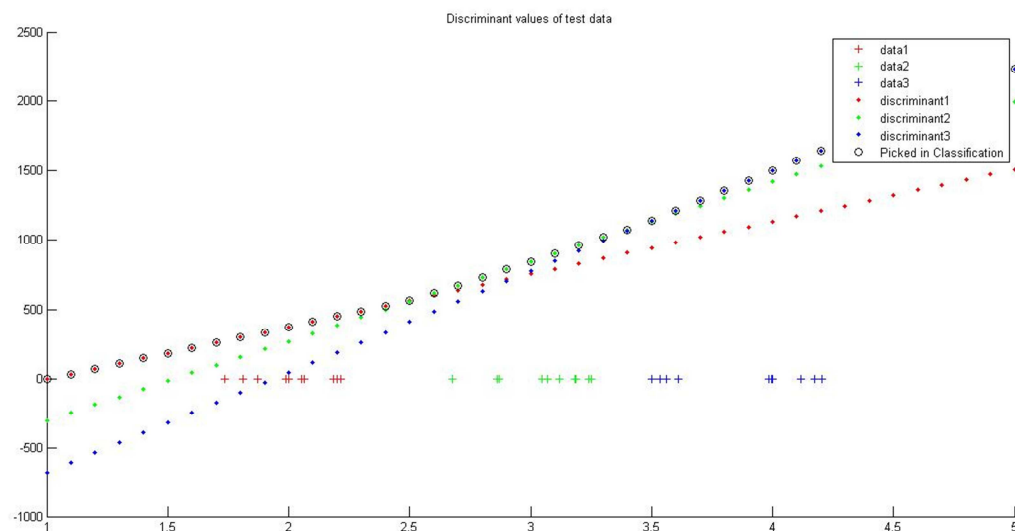
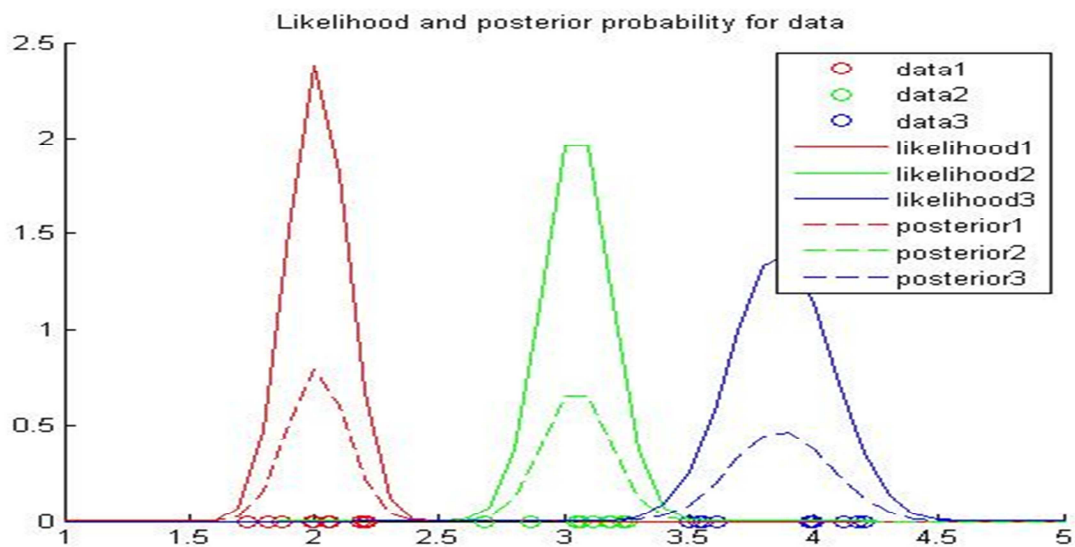


## CSL 407 Machine Learning Homework 2

## Question 1 :

The plots found in the question are :



## Question 2 :

2. The cost function without regularization is :

$$J(w) = \frac{1}{N} \sum_{n=1}^N (-y_n \log(g(x_n)) - (1-y_n) \log(1-g(x_n)))$$

where  $w$  is weight vector,  $N$  is number of  $x$ ,  $g(x_n) = \frac{1}{1+e^{-w^T x}}$

We know that:

$$\frac{\partial J(w)}{\partial w} = \frac{1}{N} \sum_{n=1}^N (g(x_n) - y_n) x_{d_n} \quad (i)$$

After adding regularization, ~~term~~ on the cost function becomes :

$$J(w) = \frac{1}{N} \sum_{n=1}^N (-y_n \log(g(x_n)) - (1-y_n) \log(1-g(x_n))) + \frac{\lambda}{2N} \|w\|^2 \quad \left\{ \|w\|^2 = \sum_{j=1}^m w_j^2 \right\}$$

so when we differentiate:-

for  $w_j = 0$   $\frac{\partial J(w)}{\partial w_j}$  is same as (i)

But for  $j \neq 0$ , we add differentiation of regularization term:

$$\begin{aligned} \frac{\partial J(w)}{\partial w_j} &= \frac{1}{N} \sum_{n=1}^N (g(x_n) - y_n) x_{d_n} + \frac{\partial}{\partial w_j} \left( \frac{\lambda}{2N} \sum_{j=1}^m w_j^2 \right) \\ &= \frac{1}{N} \sum_{n=1}^N (g(x_n) - y_n) x_{d_n} + \left( \frac{\lambda}{2N} \cdot 2w_j = \frac{\lambda w_j}{N} \right) \end{aligned}$$

Therefore the weight update equation is:

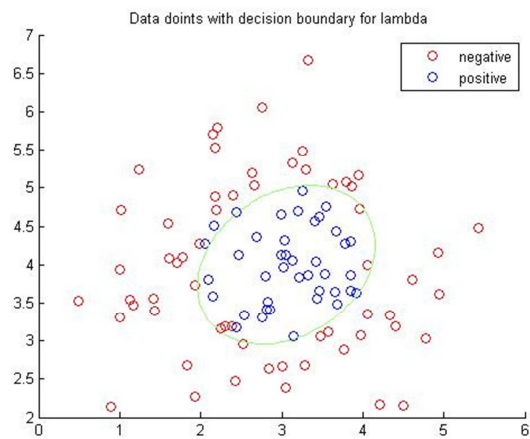
$$w_j = \begin{cases} w_j - \alpha \frac{1}{N} \left( \sum_{n=1}^N (g(x_n) - y_n) x_{d_n} \right) & \text{for } j=0 \\ w_j - \alpha \frac{1}{N} \left( \sum_{n=1}^N (g(x_n) - y_n) x_{d_n} + \lambda w_j \right) & \text{for } 1 \leq j \leq m \end{cases}$$

### Question 3 :

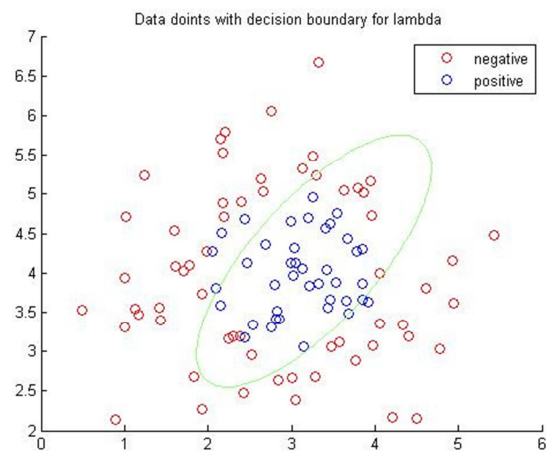
Observation : Graph for varying degree and lambda .

Varying lambda for degree = 2

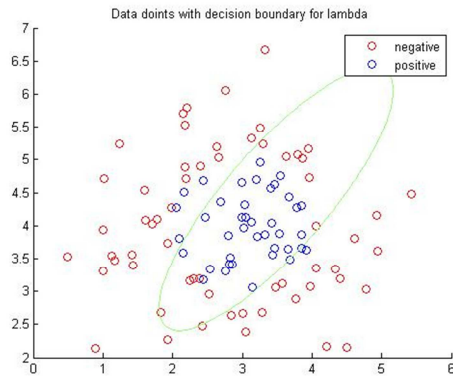
Lambda = 0.0



Lambda = 0.5

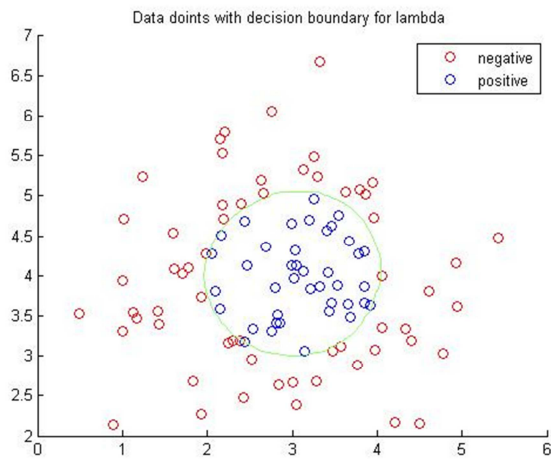


Lambda = 1.0

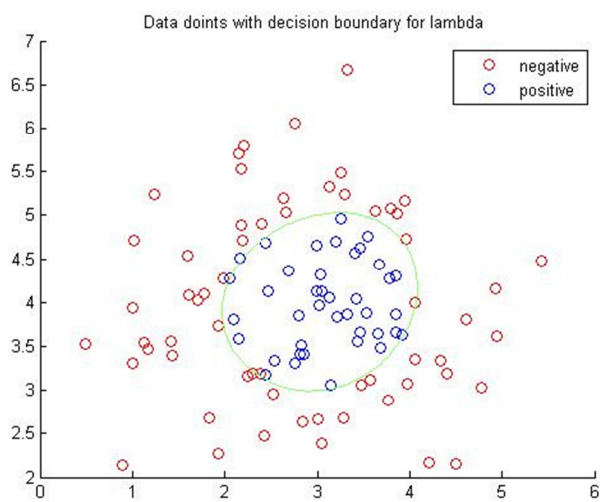


For Degree = 3

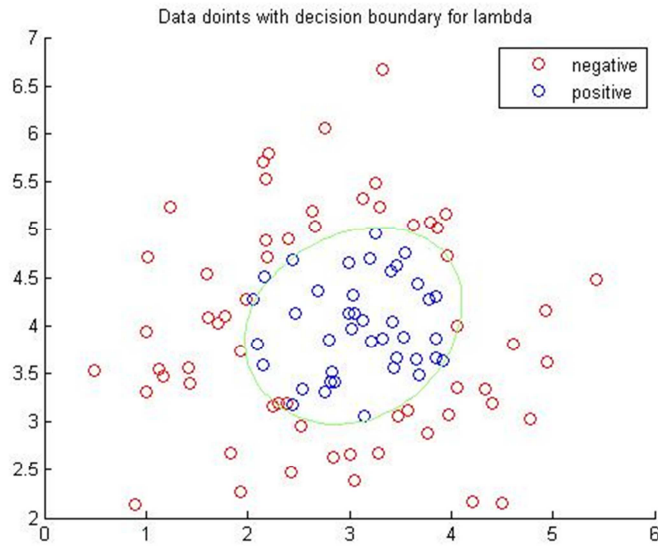
Lambda = 0.0



Lambda = 0.5

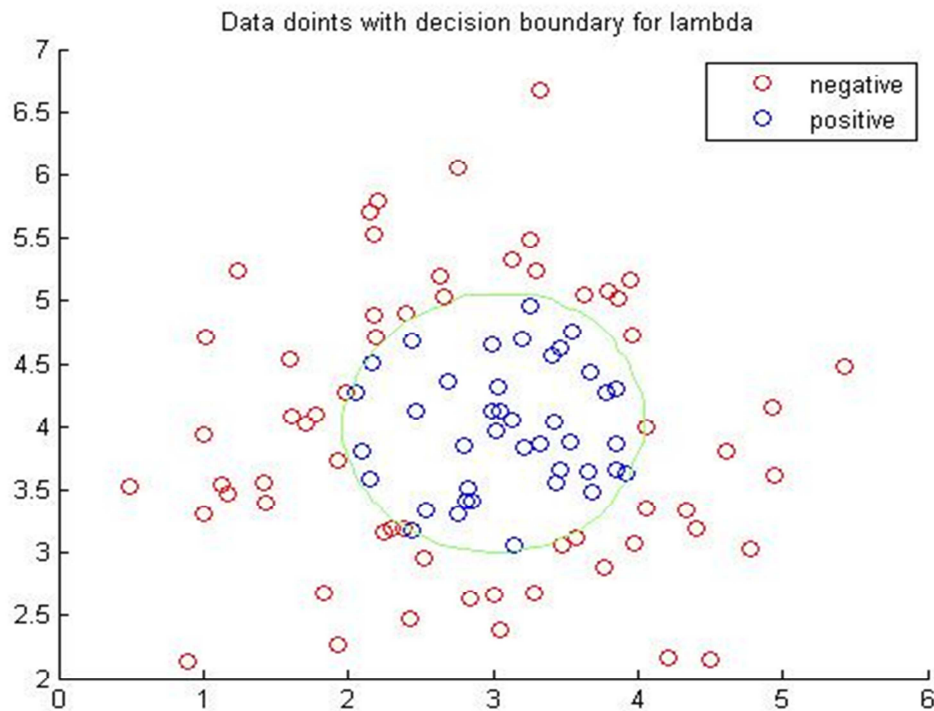


Lambda = 1.0



**(ii)** Overfitting happens in the case where value of lambda is low  
.Example given Lambda = 0.0 Degree 4

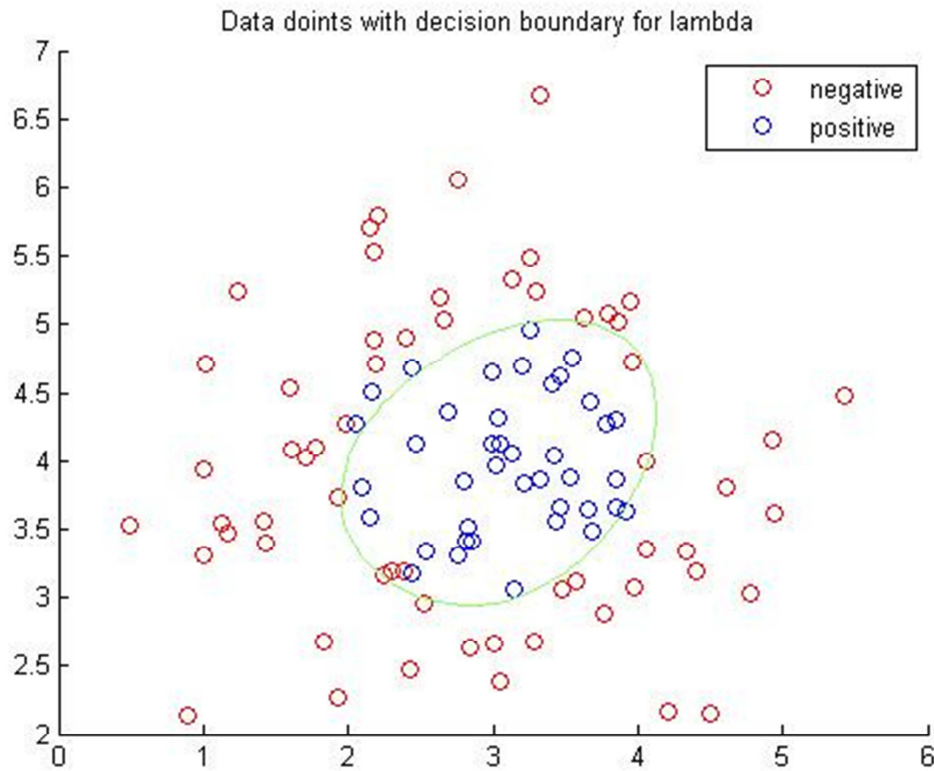
We can notice that all the points are separated by decision boundary.





Under fitting happens when there is extra regularization .  
Example given Lambda = 100.0 Degree = 4.0

We can notice that some significant amount of points are missed by decision boundary.



## Question 4 :

**Solution :**  $x_1$  = hours studied,  $x_2$  = undergrad GPA, and  $Y$  = receive an A

Estimated coefficient ,  $w_0 = -8$  ,  $w_1 = 0.05$ ,  $w_2 = 1$

(a)  $x_1 = 5$   $x_2 = 7.5$

$$w^T x = -8 + (0.05 * 5) + (1 * 7.5) = -0.25$$

$$g(w^T x) = g(-0.25) = 0.4378.$$

(b) Probability of student getting A in class = 0.6

$$\text{Therefore } g(w^T x) = 0.6$$

$$\text{So, } w^T x = 0.4054$$

$$-8 + (0.05 * x_1) + (1 * 7.5) = 0.4054$$

$$x_1 = 18.108$$

Student need to study for 18.108 hours .

