COL774 Assignment-1

Q1 Part(a)

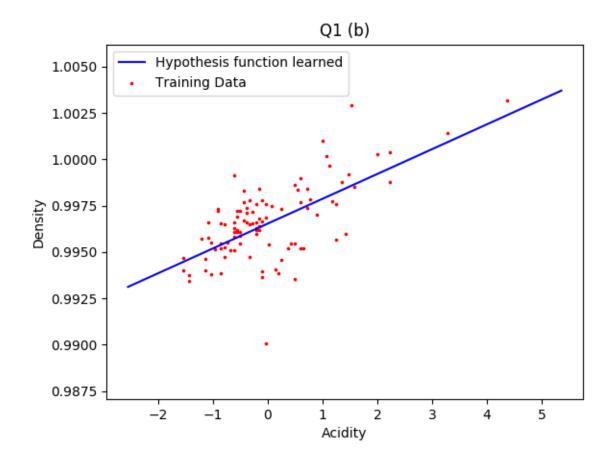
Learning rate: tried many values and chose η =**0.001**

Stopping criteria: If value of cost in 2 consecutive iterations changes by less than ϵ =**0.0000001** then stop.

I takes **89** iterations to converge at η =**0.001**

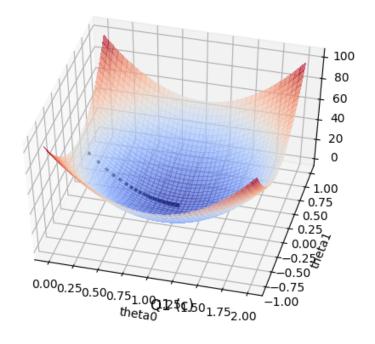
 θ comes out to be [0.99653574 0.00134008]

Part(b) Plotted training data and hypothesis function learned.

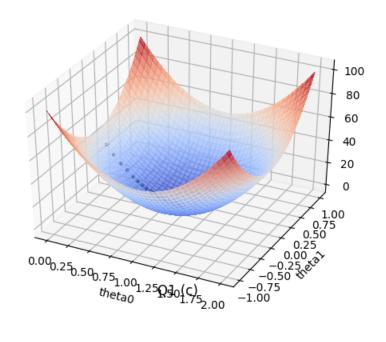


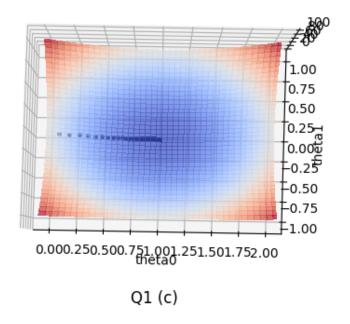
Part(c)

Plotted $J(\theta)$ and then plotted points at each iteration of gradient decent.

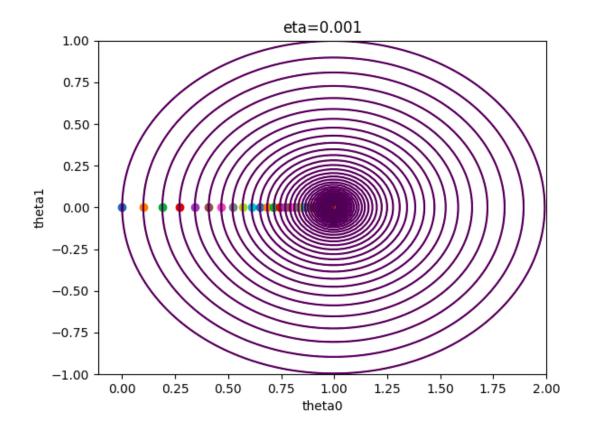


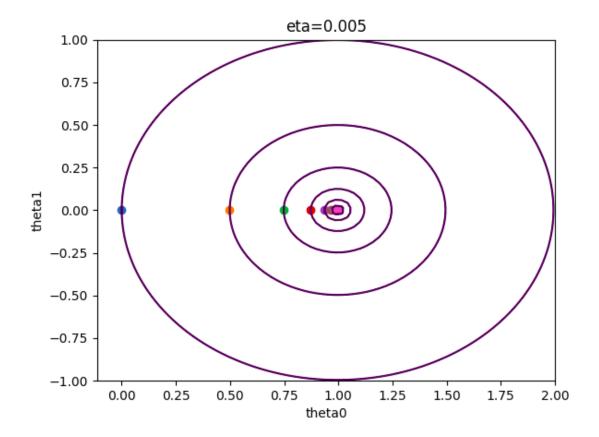
Plots of $J(\theta)$ vs θ

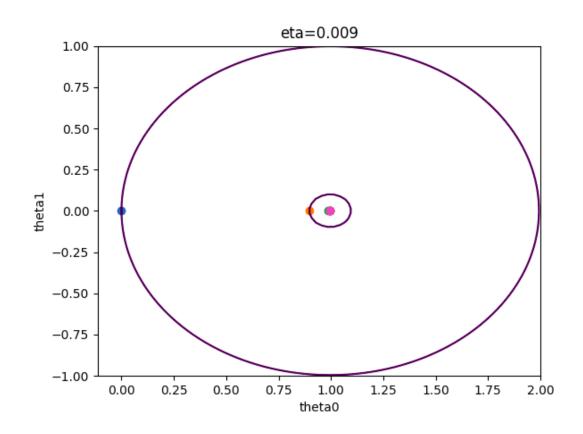


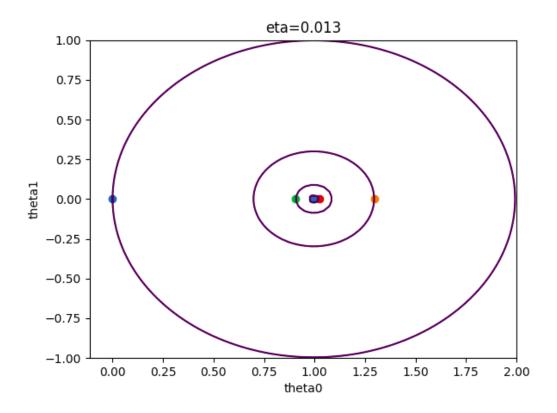


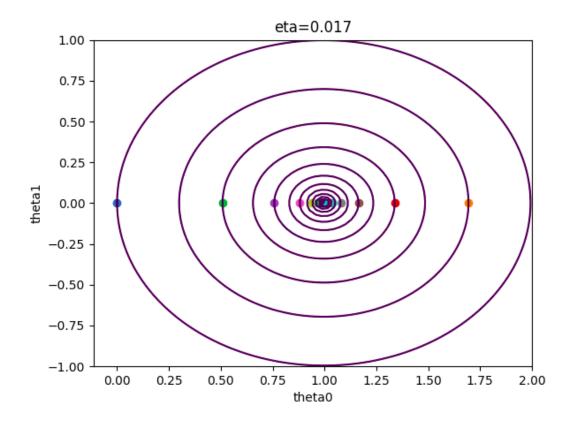
Part(d)Plotted contour for cost function at each iteration.











Part(e)

The step size increases from η =0.001,0.005,0.009 and it take less iterations to converge.

Algorithm starts oscillating initially when η =0.013

For η =0.013 and 0.017 it oscillates initially but converges eventually.

At η =0.021 and 0.025 it keeps diverging and does not converge.

$\mathbf{Q}\mathbf{2}$

Part(a)

Calculated θ using :

$$\theta = (X^TX)^{-1}X^TY$$

$$0$$

$$0$$

$$-1$$

$$-2.0 \quad -1.5 \quad -1.0 \quad -0.5 \quad 0.0 \quad 0.5 \quad 1.0 \quad 1.5 \quad 2.0$$

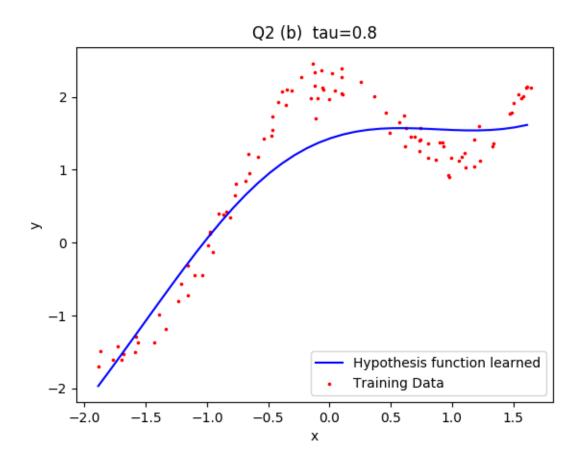
Part(b)

Calculated θ for each query point x where $\theta^{\scriptscriptstyle T} x$ has to be calculated using:

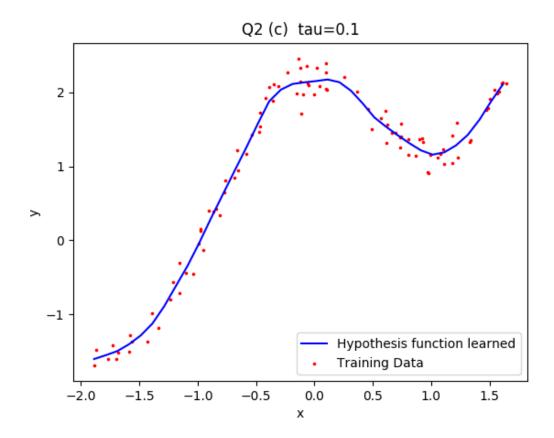
$$\theta = ((X^T W X)^{-1}) X^T W Y$$

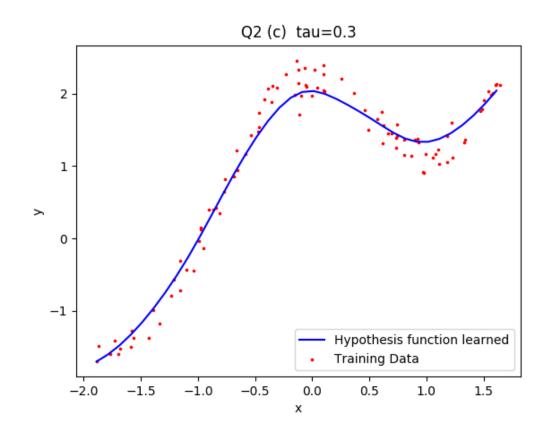
Where W is a m*m diagonal matrix, W is calculated for each query point x

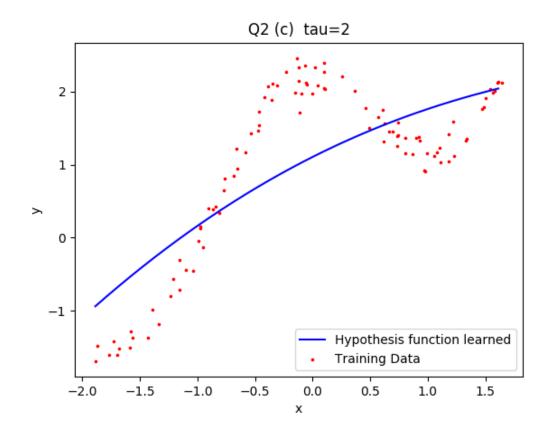
$$W_{ii}(x) = exp(-\frac{(x - x^{(i)})^2}{2\tau^2})$$
 $0 \le i < m$

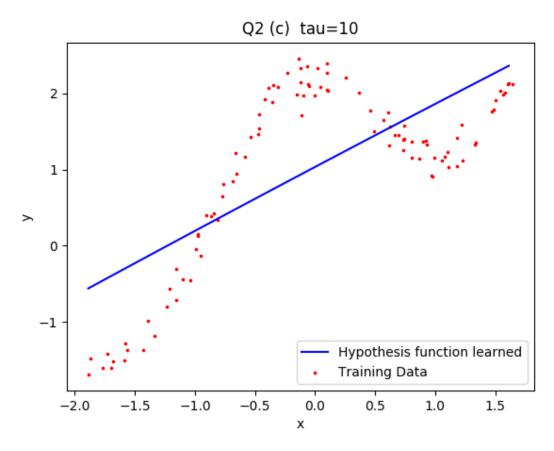


Part(c) Tried tau={0.1,0.3,0.8,2,10} and plotted the hypothesis function learned.









Out of tau={0.1,0.3,0.8,2,10}, I think tau=0.3 is best, because there is no overfitting and no underfitting.

When tau is too small, tau=0.1 then overfitting happens.

When tau is too large, tau=10 then underfitting happens and the function learned is almost a straight line.

For tau=2,10 the curve does not represent the data at all.

$\mathbf{Q}\mathbf{3}$

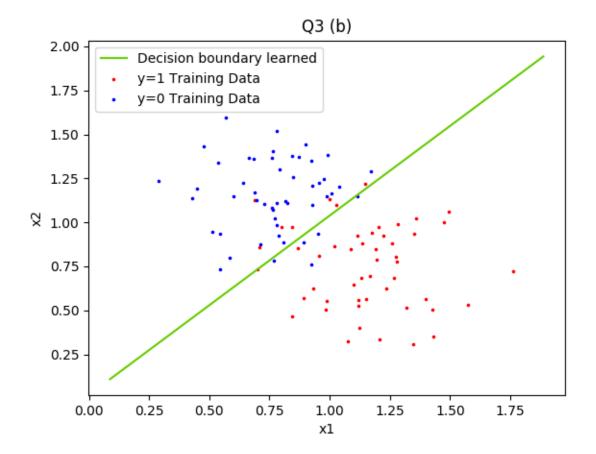
Part(a)

Implement Newton's method for optimizing $L(\theta)$. Used Newton's method to find roots of $L'(\theta)$

Convergence condition used: If L2 norm of difference between theta of 2 consecutive iterations is less than 0.000001 then stop

Part(b)

Plotted decision boundary learned with the training data.



Q4

Alaska is 0 and Canada is 1

Part (a)

Calculated μ 0, μ 1, Σ using the formulas told in class.

 μ 0=[0.80651536 1.07283018]

 $\mu 1 = [1.14318082 \ 0.91362299]$

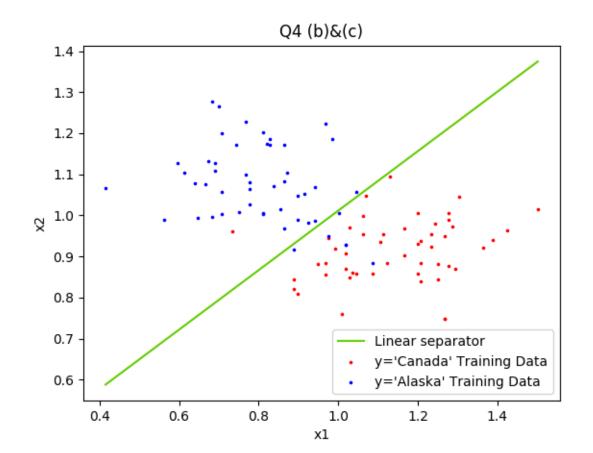
 Σ =[[0.0213353 -0.00058194] [-0.00058194 0.00716423]]

Part (b)&(c)

When $\Sigma 0 = \Sigma 1$ the equation of separator is:

$$X^{T} \Sigma^{-1} (\mu_{0} - \mu_{1}) + (\mu_{0} - \mu_{1})^{T} \Sigma^{-1} X + (\mu_{1}^{T} \Sigma^{-1} \mu_{1} - \mu_{0}^{T} \Sigma^{-1} \mu_{0}) = 2log_{e}(\frac{\phi}{1 - \phi})$$

This is the equation of a straight line, plotted the above line.



Part(d)

 $\mu 0$ and $\mu 1$ are same as before.

Calculated $\Sigma 0$ and $\Sigma 1$ using the formula given in assignment pdf.

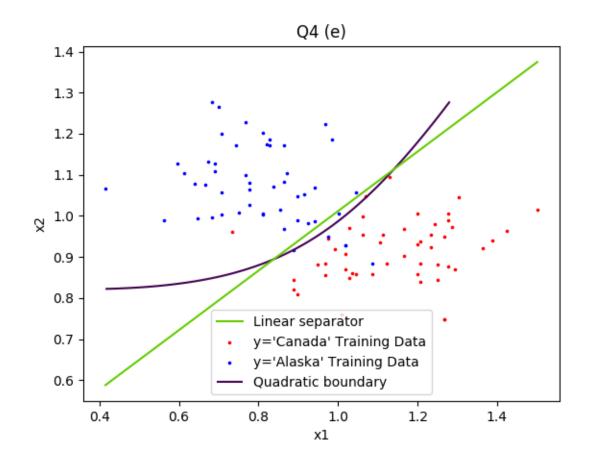
```
\Sigma 0=
[[ 0.01895402 -0.00401041]
[ -0.00401041 0.00874507]]
\Sigma 1=
[[ 0.02371657 0.00284652]
[ 0.00284652 0.00558338]]
```

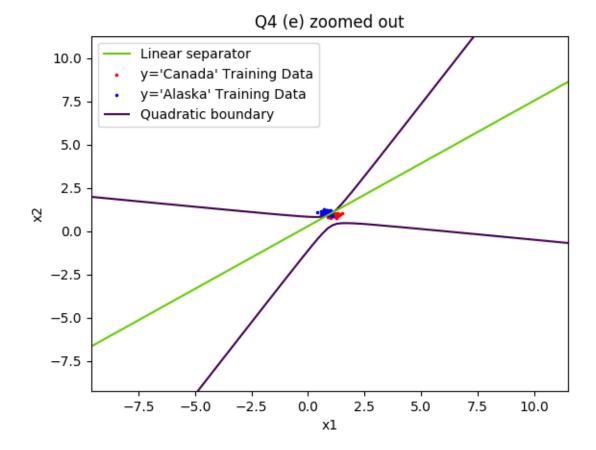
Part(e)

When $\Sigma 0 \neq \Sigma 1$ the equation of separator is:

$$(X - \mu_1)^T \Sigma_1^{-1} (X - \mu_1) - (X - \mu_0)^T \Sigma_0^{-1} (X - \mu_0) - 2log_e(\frac{\phi}{1 - \phi}) + log_e(\frac{|\Sigma_1|}{|\Sigma_0|}) = 0$$

Plotted quadratic separator as a contour of above equation at Z=0





Part(f)

When $\Sigma 0 = \Sigma 1$ the equation of separator is a straight line because the quadratic terms cancel. When $\Sigma 0 \neq \Sigma 1$ the equation of separator is quadratic in x1 and x2.

The quadratic separator is actually a hyperbola, and it intersects linear separator at two points.

The linear separator classifies few 'Alaska' points as 'Canada', but quadratic separator bends and correctly classifies those points as 'Alaska'.