

ECE 595: Homework 2

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(Spring 2019)

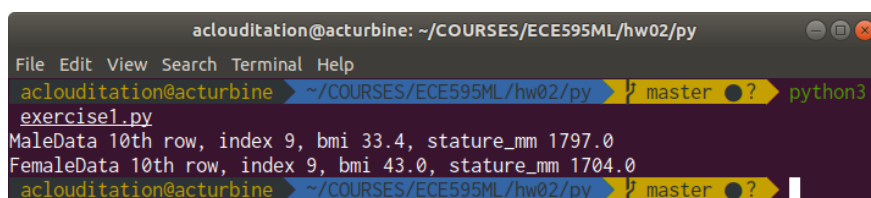
Exercise 1

(a)

Refer to code in the back.

(b)

The data was successfully read, shown by the screen below.



```
aclouditation@acturbine: ~/COURSES/ECE595ML/hw02/py
File Edit View Search Terminal Help
aclouditation@acturbine > ~/COURSES/ECE595ML/hw02/py > master > python3
exercise1.py
MaleData 10th row, index 9, bmi 33.4, stature_mm 1797.0
FemaleData 10th row, index 9, bmi 43.0, stature_mm 1704.0
aclouditation@acturbine > ~/COURSES/ECE595ML/hw02/py > master >
```

Figure 1: screenshot for reading data

Exercise 2

(a)

$$\begin{aligned} \begin{bmatrix} \boldsymbol{\omega}^* \\ \omega_0^* \end{bmatrix} &= \underset{\boldsymbol{\omega}, \omega_0}{\operatorname{argmin}} \sum_{j=1}^N (\boldsymbol{\omega}^T \mathbf{x}_j + \omega_0 - y_j)^2 \\ \text{set } \boldsymbol{\theta} &= \begin{bmatrix} \boldsymbol{\omega}^* \\ \omega_0^* \end{bmatrix} \\ \boldsymbol{\theta}^* &= \underset{\boldsymbol{\theta}}{\operatorname{argmin}} \sum_{j=1}^N ([\mathbf{x}_j^T \quad 1] \boldsymbol{\theta} - y_j)^2 \\ &\text{thus,} \end{aligned} \tag{1}$$

$$A = \begin{bmatrix} -\mathbf{x}_1^T & 1 \\ -\mathbf{x}_2^T & 1 \\ \vdots & \vdots \\ -\mathbf{x}_N^T & 1 \end{bmatrix} \quad \mathbf{b} = \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_N \end{bmatrix}$$

(b)

by least square

$$\begin{aligned} A^T A \boldsymbol{\theta}^* &= A^T \mathbf{b} \\ \boldsymbol{\theta}^* &= (A^T A)^{-1} A^T \mathbf{b} \end{aligned} \tag{2}$$

if $A^T A$ is invertible, A needs to be full column rank (or $\text{null}(A) = 0$). TODO: find out how to avoid this issue

(c)

Refer to the code in the back.

The optimal weight θ^* is:

$$\theta^* = \begin{bmatrix} -1.23396767e-2 \\ 6.67486843e-3 \\ -1.07017505e+1 \end{bmatrix}$$

(d)

Refer to the code in the back again.

The weight vector computed using CVXPY is the same as the one computed in the previous step.

$$\theta^* = \begin{bmatrix} -1.23396767e-2 \\ 6.67486843e-3 \\ -1.07017505e+1 \end{bmatrix}$$

Exercise 3

(a)

(i)

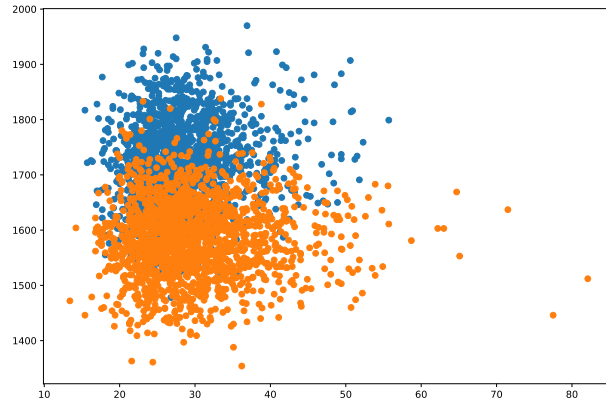


Figure 2: Training data

(ii)

$$\begin{aligned} \omega^{*T} \mathbf{x} + \omega_0^* &= 0 \\ \omega_1^* x_1 + \omega_2^* x_2 + \omega_0^* &= 0 \\ x_2 &= -\frac{\omega_1^* x_1 + \omega_0^*}{\omega_2^*} \end{aligned} \tag{3}$$

(iii)

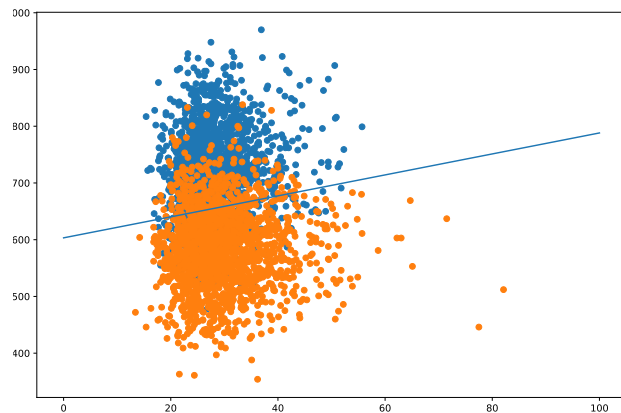


Figure 3: Training data with decision boundary

(b)

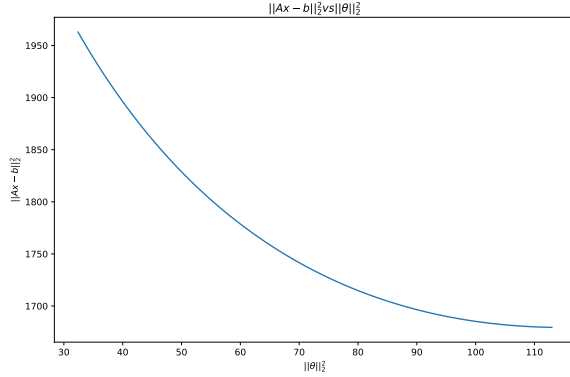
Refer to the code in the back, the success rate is:

$$\text{success rate} = 83.93\%$$

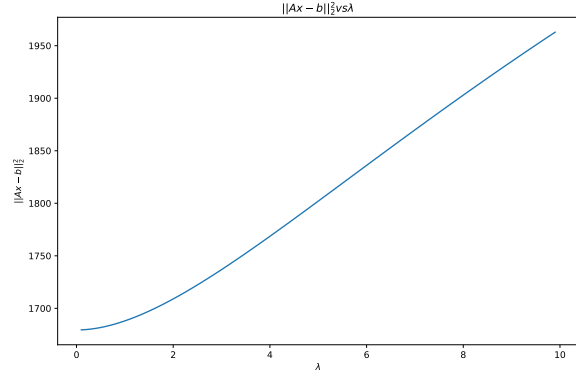
Exercise 4

(a)

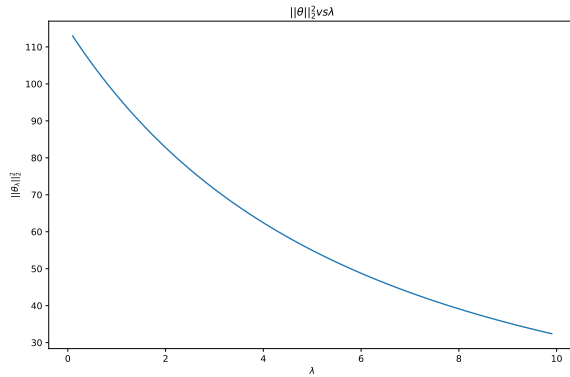
(i)



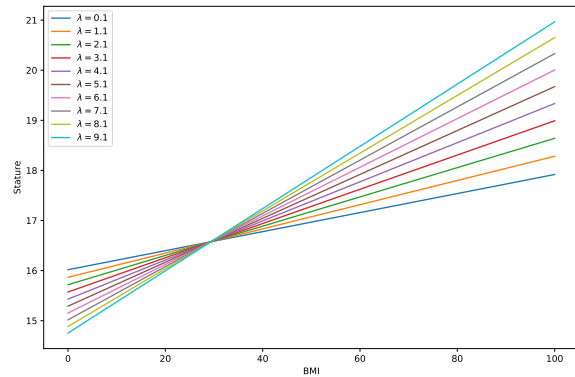
(a) 1



(b) 2



(c) 3



(d) 4

(ii)

TODO: fix this

(b)

(i)

for equation (8), $\theta_\lambda = \underset{\theta}{\operatorname{argmin}} ||A\theta - b||_2^2 + \lambda ||\theta||_2^2$:

$$\mathcal{L}(\theta) = ||A\theta - b||_2^2 + \lambda ||\theta||_2^2 \quad (4)$$

Since no constraint, the vertex can be found at:

$$\begin{aligned} \nabla_{\theta} ||A\theta - b||_2^2 + \lambda \nabla_{\theta} ||\theta||_2^2 &= 0 \\ 2A^T(A\theta - b) + 2\lambda\theta &= 0 \\ A^T A\theta + \lambda\theta &= A^T b \\ \theta_\lambda &= (A^T A + \lambda I)^{-1} A^T b \end{aligned} \quad (5)$$

for equation (9), $\boldsymbol{\theta}_\alpha = \underset{\boldsymbol{\theta}}{\operatorname{argmin}} \|\mathbf{A}\boldsymbol{\theta} - \mathbf{b}\|_2^2$ subject to $\|\boldsymbol{\theta}\|_2^2 \leq \alpha$, $\alpha - \|\boldsymbol{\theta}\|_2^2 \geq 0$:

$$\mathcal{L}(\boldsymbol{\theta}, \gamma) = \|\mathbf{A}\boldsymbol{\theta} - \mathbf{b}\|_2^2 - \gamma(\alpha - \|\boldsymbol{\theta}\|_2^2) \quad (6)$$

KKT conditions:

$$\begin{aligned} \nabla_{\boldsymbol{\theta}} \|\mathbf{A}\boldsymbol{\theta} - \mathbf{b}\|_2^2 + \gamma \nabla_{\boldsymbol{\theta}} (\|\boldsymbol{\theta}\|_2^2 - \alpha) &= 0 & \text{stationarity} \\ \alpha - \|\boldsymbol{\theta}\|_2^2 &\geq 0 & \text{primal feasibility} \\ \gamma &\geq 0 & \text{dual feasibility} \\ \gamma(\alpha - \|\boldsymbol{\theta}\|_2^2) &= 0 & \text{complementary slackness} \end{aligned} \quad (7)$$

for equation (10), $\boldsymbol{\theta}_\epsilon = \underset{\boldsymbol{\theta}}{\operatorname{argmin}} \|\boldsymbol{\theta}\|_2^2$ subject to $\|\mathbf{A}\boldsymbol{\theta} - \mathbf{b}\|_2^2 \leq \epsilon$, $\epsilon - \|\mathbf{A}\boldsymbol{\theta} - \mathbf{b}\|_2^2 \geq 0$:

$$\mathcal{L}(\boldsymbol{\theta}, \gamma) = \|\boldsymbol{\theta}\|_2^2 - \gamma(\epsilon - \|\mathbf{A}\boldsymbol{\theta} - \mathbf{b}\|_2^2) \quad (8)$$

KKT conditions:

$$\begin{aligned} \nabla_{\boldsymbol{\theta}} \|\boldsymbol{\theta}\|_2^2 + \gamma \nabla_{\boldsymbol{\theta}} (\|\mathbf{A}\boldsymbol{\theta} - \mathbf{b}\|_2^2 - \epsilon) &= 0 & \text{stationarity} \\ \epsilon - \|\mathbf{A}\boldsymbol{\theta} - \mathbf{b}\|_2^2 &\geq 0 & \text{primal feasibility} \\ \gamma &\geq 0 & \text{dual feasibility} \\ \gamma(\epsilon - \|\mathbf{A}\boldsymbol{\theta} - \mathbf{b}\|_2^2) &= 0 & \text{complementary slackness} \end{aligned} \quad (9)$$

(ii)

for equation (9), stationarity suggests that:

$$\nabla_{\boldsymbol{\theta}} \|\mathbf{A}\boldsymbol{\theta} - \mathbf{b}\|_2^2 + \gamma \nabla_{\boldsymbol{\theta}} (\|\boldsymbol{\theta}\|_2^2 - \alpha) = 0 \quad (10)$$

by doing the algebra, we get:

$$\begin{aligned} 2\mathbf{A}^T(\mathbf{A}\boldsymbol{\theta} - \mathbf{b}) + 2\gamma\boldsymbol{\theta} &= 0 \\ \boldsymbol{\theta}_\alpha &= (\mathbf{A}^T\mathbf{A} + \gamma\mathbf{I})^{-1}\mathbf{A}^T\mathbf{b} \end{aligned} \quad (11)$$

if we set $\gamma_\alpha = \lambda$, referring to equation (5) above:

$$\begin{aligned} \boldsymbol{\theta}_\alpha &= (\mathbf{A}^T\mathbf{A} + \lambda\mathbf{I})^{-1}\mathbf{A}^T\mathbf{b} \\ &= \boldsymbol{\theta}_\lambda \end{aligned} \quad (12)$$

where primal feasibility:

$$\alpha - \|\boldsymbol{\theta}_\alpha\|_2^2 \geq 0$$

is satisfied if we choose $\alpha = \|\boldsymbol{\theta}_\lambda\|_2^2$ since $\boldsymbol{\theta}_\alpha = \boldsymbol{\theta}_\lambda$,

$$\alpha - \|\boldsymbol{\theta}_\alpha\|_2^2 = \|\boldsymbol{\theta}_\alpha\|_2^2 - \|\boldsymbol{\theta}_\alpha\|_2^2 = 0 \geq 0$$

Dual feasibility is automatically satisfied since we fix $\gamma_\alpha = \lambda > 0$

Complementary slackness is satisfied since $\alpha - \|\boldsymbol{\theta}_\alpha\|_2^2 = 0$

Thus, the two problems (8) and (9) are equivalent.

(iii)

for equation (9), stationarity suggests that:

$$\nabla_{\theta} \|\theta\|_2^2 + \gamma \nabla_{\theta} (\|\mathbf{A}\theta - \mathbf{b}\|_2^2 - \epsilon) = 0 \quad (13)$$

by doing the algebra, we get:

$$\begin{aligned} 2\theta + 2\gamma \mathbf{A}^T (\mathbf{A}\theta - \mathbf{b}) &= 0 \\ \theta_{\epsilon} &= (\gamma \mathbf{A}^T \mathbf{A} + \mathbf{I})^{-1} \mathbf{A}^T \gamma \mathbf{b} \\ \theta_{\epsilon} &= (\gamma^{-1} \gamma \mathbf{A}^T \mathbf{A} + \gamma^{-1} \mathbf{I})^{-1} \mathbf{A}^T \mathbf{b} \\ \theta_{\epsilon} &= (\mathbf{A}^T \mathbf{A} + \gamma^{-1} \mathbf{I})^{-1} \mathbf{A}^T \mathbf{b} \end{aligned} \quad (14)$$

if we set $\gamma_{\epsilon} = \lambda^{-1}$, referring to equation (5) above:

$$\begin{aligned} \theta_{\epsilon} &= (\mathbf{A}^T \mathbf{A} + \lambda \mathbf{I})^{-1} \mathbf{A}^T \mathbf{b} \\ &= \theta_{\lambda} \end{aligned} \quad (15)$$

where primal feasibility:

$$\epsilon - \|\mathbf{A}\theta_{\epsilon} - \mathbf{b}\|_2^2 \geq 0$$

is satisfied if we choose $\epsilon = \|\mathbf{A}\theta_{\lambda} - \mathbf{b}\|_2^2$ since $\theta_{\epsilon} = \theta_{\lambda}$,

$$\epsilon - \|\mathbf{A}\theta_{\epsilon} - \mathbf{b}\|_2^2 = 0 \geq 0$$

Dual feasibility is automatically satisfied since we fix $\gamma_{\epsilon}^{-1} = \lambda > 0$, thus, $\gamma_{\epsilon} > 0$

Complementary slackness is satisfied since $\epsilon - \|\mathbf{A}\theta_{\epsilon} - \mathbf{b}\|_2^2 = 0$

Thus, the two problems (8) and (10) are also equivalent.

(c)

(i)

(ii)

(d)