

# Assignment 3

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## 1 TASK 1

*Program is attached along with this file in the same folder.*

**for degree =1,  $\lambda$  =0:**

- $w_0 = -6.3872$
- $w_1 = 0.0276$
- $w_2 = 0.0432$
- $w_3 = 0.0126$
- $w_4 = 0.0176$
- $w_5 = 0.0080$
- $w_6 = -0.0058$
- $w_7 = -0.0081$
- $w_8 = 0.0714$
- $w_9 = -0.0153$
- $w_{10} = -0.0190$
- $w_{11} = 0.0117$
- $w_{12} = 0.0222$
- $w_{13} = -0.0018$
- $w_{14} = -0.0013$
- $w_{15} = 0.0091$
- $w_{16} = 0.0382$

Last line being:

ID= 3498, output= 3.8514, target value = 4.0000, squared error = 0.0221

**for degree = 1,  $\lambda$  = 1:**

- $w_0 = -6.2611$
- $w_1 = 0.0275$
- $w_2 = 0.0428$
- $w_3 = 0.0126$
- $w_4 = 0.0172$

- $w_5 = 0.0078$
- $w_6 = -0.0059$
- $w_7 = -0.0081$
- $w_8 = 0.0713$
- $w_9 = -0.0154$
- $w_{10} = -0.0191$
- $w_{11} = 0.0116$
- $w_{12} = 0.0221$
- $w_{13} = -0.0018$
- $w_{14} = -0.0017$
- $w_{15} = 0.0090$
- $w_{16} = 0.0383$

last line:

ID= 3498, output= 3.8528, target value = 4.0000, squared error = 0.0217

**for degree = 2  $\lambda = 0$ :**

- $w_0 = -7.5608$
- $w_1 = 0.0223$
- $w_2 = 0.0001$
- $w_3 = 0.0352$
- $w_4 = 0.0000$
- $w_5 = 0.0049$
- $w_6 = -0.0000$
- $w_7 = -0.0299$
- $w_8 = 0.0002$
- $w_9 = 0.0327$
- $w_{10} = -0.0001$
- $w_{11} = 0.0694$
- $w_{12} = -0.0004$
- $w_{13} = 0.0079$
- $w_{14} = -0.0002$
- $w_{15} = 0.0596$
- $w_{16} = -0.0003$
- $w_{17} = -0.0184$
- $w_{18} = -0.0000$
- $w_{19} = 0.0093$
- $w_{20} = 0.0002$
- $w_{21} = 0.0162$
- $w_{22} = -0.0000$
- $w_{23} = 0.0398$
- $w_{24} = -0.0002$
- $w_{25} = -0.0041$
- $w_{26} = 0.0001$
- $w_{27} = 0.0538$
- $w_{28} = -0.0007$

- $w_{29} = -0.0149$
- $w_{30} = 0.0002$
- $w_{31} = 0.1215$
- $w_{32} = -0.0007$

last line:

ID= 3498, output= 3.6074, target value = 4.0000, squared error = 0.1542

**for degree = 2  $\lambda = 1$ :**

- $w_0 = -7.0384$
- $w_1 = 0.0219$
- $w_2 = 0.0001$
- $w_3 = 0.0310$
- $w_4 = 0.0001$
- $w_5 = 0.0043$
- $w_6 = -0.0000$
- $w_7 = -0.0345$
- $w_8 = 0.0002$
- $w_9 = 0.0315$
- $w_{10} = -0.0001$
- $w_{11} = 0.0678$
- $w_{12} = -0.0004$
- $w_{13} = 0.0077$
- $w_{14} = -0.0002$
- $w_{15} = 0.0574$
- $w_{16} = -0.0003$
- $w_{17} = -0.0192$
- $w_{18} = -0.0000$
- $w_{19} = 0.0091$
- $w_{20} = 0.0002$
- $w_{21} = 0.0156$
- $w_{22} = -0.0000$
- $w_{23} = 0.0401$
- $w_{24} = -0.0002$
- $w_{25} = -0.0050$
- $w_{26} = 0.0001$
- $w_{27} = 0.0536$
- $w_{28} = -0.0007$
- $w_{29} = -0.0155$
- $w_{30} = 0.0002$
- $w_{31} = 0.1208$
- $w_{32} = -0.0007$

last line:

ID= 3498, output= 3.6001, target value = 4.0000, squared error = 0.1599

## 2 TASK 2

**Qsn:** We are given these training examples for a linear regression problem:

$$x_1 = 5.3, t_1 = 9.6 \quad x_2 = 7.1, t_2 = 4.2 \quad x_3 = 6.4, t_3 = 2.2$$

We just want to fit a line to this data, and we want to find the 2-dimensional vector  $w$  that minimizes  $E_D(w)$  as defined in slide 60 of the linear regression slides. What is the value of  $w$  in the limit where  $\lambda$  approaches positive infinity? Justify your answer. Correct answers with insufficient justification will not receive credit.

**Ans:** We know, as it is given in slide 61 that the value of  $w$  that minimizes  $E_D(w)$  is:

$$w = (\lambda I + \phi^T \phi)^{-1} \phi^T t \quad (2.1)$$

for  $\lambda = 0$ , it will just be like a linear regression without any regularization and it would go like:

$$w = (\phi^T \phi)^{-1} \phi^T t$$

for  $\lambda = 1, 2, 3$ , the same equation will be:

$$w = (I + \phi^T \phi)^{-1} \phi^T t$$

$$w = (2I + \phi^T \phi)^{-1} \phi^T t$$

$$w = (3I + \phi^T \phi)^{-1} \phi^T t$$

And, as  $\lim_{\lambda \rightarrow \infty}$ ,

$$w \rightarrow (\infty I + \phi^T \phi)^{-1} \phi^T t$$

$$w \rightarrow (\infty)^{-1} \phi^T t$$

A little sidenote here:

for  $A = \begin{pmatrix} j & 0 \\ 0 & j \end{pmatrix}$ ,  $A^{-1} = \begin{pmatrix} \frac{1}{j} & 0 \\ 0 & \frac{1}{j} \end{pmatrix}$  and as  $j$  approaches  $\infty$ ,  $A^{-1} = \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix}$  So,  $w$  will be a zero matrix of  $M \times 1$  dimensions where,  $M$  is the dimension of  $\phi$ . It will be a column matrix at the end, thanks to 't'. It can be seen as:

$$\lim_{\lambda \rightarrow \infty} w = \mathcal{O}_{M \times M} \phi^T t$$

$$\lim_{\lambda \rightarrow \infty} w = \mathcal{O}_{M \times M} t_{M \times 1}$$

$$\lim_{\lambda \rightarrow \infty} w = \mathcal{O}_{M \times 1}$$

So, to sum everything up, as  $\lim_{\lambda \rightarrow \infty}$ ,  $w \rightarrow \mathcal{O}_{M \times 1}$  where,  $\mathcal{O}_{M \times 1}$  is a zero matrix.

### 3 TASK 3

**Qsn:** We are given these training examples for a linear regression problem:

$$x_1 = 5.3, t_1 = 9.6$$

$$x_2 = 7.1, t_2 = 4.2$$

$$x_3 = 6.4, t_3 = 2.2$$

We are also given these two lines as possible solutions:

$$f(x) = 3.1x + 4.2 \quad (3.1)$$

$$f(x) = 2.4x - 1.5 \quad (3.2)$$

Which of these lines is a better solution according to the sum-of-squares criterion? This criterion is defined as function  $E_D(w)$  in slide 25 of the linear regression slides. Justify your answer. Correct answers with insufficient justification will not receive credit.

**Ans:** We will be using formula for standard sum of squares error for both equations 3.1 and 3.2.

We have the formula for  $E_D(w)$  as:

$$E_D(w) = \frac{1}{2} \sum_{n=1}^N \left[ (t_n - w^T \phi(x_n))^2 \right] \quad (3.3)$$

So, we start with equation 3.1 where  $w = \begin{pmatrix} 3.1 \\ 4.2 \end{pmatrix}$ . Now, we use each  $t_n$  and  $x_n$  and plug them into the formula.

Let's calculate  $t_n - w^T \phi(x_n)^2$  first for each data set given.

$$\begin{aligned} a) & \left( 9.6 - (3.1 \quad 4.2) \begin{pmatrix} 1 \\ 5.3 \end{pmatrix} \right)^2 \\ &= (9.6 - (3.1 + 4.2 \times 5.3))^2 \\ &= (-15.76)^2 \\ &= 248.3776 \end{aligned}$$

Similarly,

$$\begin{aligned} b) & \left( 4.2 - (3.1 \quad 4.2) \begin{pmatrix} 1 \\ 7.1 \end{pmatrix} \right)^2 \\ &= (4.2 - (3.1 + 4.2 \times 7.1))^2 \\ &= (-28.72)^2 \\ &= 824.8384 \end{aligned}$$

$$\begin{aligned}
& c) \left( 2.2 - (3.1 \quad 4.2) \begin{pmatrix} 1 \\ 6.4 \end{pmatrix} \right)^2 \\
& = (2.2 - (3.1 + 4.2 \times 6.4))^2 \\
& = (-27.78)^2 \\
& = 771.728
\end{aligned}$$

$$E_D(w) = \frac{1}{2} (248.3776 + 824.8384 + 771.728)$$

$$\boxed{E_D(w)_{3.1} = 922.4717}$$

The approach is the same for equation 3.2 as well so we can go ahead and calculate  $E_D(w)$  for this equation as well.

$$\begin{aligned}
& d) \left( 9.6 - (2.4 \quad -1.5) \begin{pmatrix} 1 \\ 5.3 \end{pmatrix} \right)^2 \\
& = (9.6 - 2.4 + 7.95)^2 \\
& = (15.15)^2 \\
& = 229.5225
\end{aligned}$$

$$\begin{aligned}
& e) \left( 4.2 - (2.4 \quad -1.5) \begin{pmatrix} 1 \\ 7.1 \end{pmatrix} \right)^2 \\
& = (4.2 - 2.4 + 10.65)^2 \\
& = (12.15)^2 \\
& = 155.0025
\end{aligned}$$

$$\begin{aligned}
& f) \left( 2.2 - (2.4 \quad -1.5) \begin{pmatrix} 1 \\ 6.4 \end{pmatrix} \right)^2 \\
& = (2.2 - 2.4 + 9.6)^2 \\
& = (9.4)^2 \\
& = 88.36
\end{aligned}$$

And, for the error calculation, we proceed as:

$$E_D(w) = \frac{1}{2} (229.5225 + 155.0025 + 88.36)$$

$$E_D(w)_{3.2} = 236.425$$

which is the sum of squares error for equation 3.2.

Upon comparison, we see equation 3.2 has much less standard sum of squares error which means **equation 3.2** is much better solution than equation 3.1.