

# MAD Assignment 1

dmx289, Jakob Legaard

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# 1 Exercise 1 (Weighted Average Loss)

## 1.1 a)

The weighted average loss is given by:

$$L = \frac{1}{N} \sum_{n=1}^N \alpha_n (w^T x_n - t_n)^2$$

- $\mathbf{X} \in \mathbb{R}^{N \cdot D}$  design matrix where each row is  $x_n^T$
- $\mathbf{t} \in \mathbb{R}^N$  atarget vector  $[t_1, t_2, \dots, t_N]^T$
- $\mathbf{w} \in \mathbb{R}^D$  a parameter vector
- $\mathbf{A} \in \mathbb{R}^{N \cdot N}$  diagonal weight matrix with  $A_{nn} = \alpha_n$

The loss function can be rewritten in matrix-vector form as:

$$L = \frac{1}{N} (\mathbf{X}\mathbf{w} - \mathbf{t})^T \mathbf{A} (\mathbf{X}\mathbf{w} - \mathbf{t})$$

expanding the matrix expression:

$$L = \frac{1}{N} [(\mathbf{X}\mathbf{w})^T \mathbf{A} (\mathbf{X}\mathbf{w}) - (\mathbf{X}\mathbf{w})^T \mathbf{A} \mathbf{t} - \mathbf{t}^T \mathbf{A} (\mathbf{X}\mathbf{w}) + \mathbf{t}^T \mathbf{A} \mathbf{t}]$$

$\mathbf{A}$  is symmetric and  $(\mathbf{X}\mathbf{w})^T \mathbf{A} \mathbf{t}$  is a scalar equal to  $\mathbf{t}^T \mathbf{A} (\mathbf{X}\mathbf{w})$ :

$$L = \frac{1}{N} [\mathbf{w}^T \mathbf{X}^T \mathbf{A} \mathbf{X} \mathbf{w} - 2\mathbf{w}^T \mathbf{X}^T \mathbf{A} \mathbf{t} + \mathbf{t}^T \mathbf{A} \mathbf{t}]$$

gradient with respect to  $\mathbf{w}$ :

$$\nabla_{\mathbf{w}} L = \frac{1}{N} [2\mathbf{X}^T \mathbf{A} \mathbf{X} \mathbf{w} - 2\mathbf{X}^T \mathbf{A} \mathbf{t}]$$

using the following matrix identities:

$$\frac{\partial}{\partial \mathbf{w}} (\mathbf{w}^T \mathbf{B} \mathbf{w}) = 2\mathbf{B} \mathbf{w} \text{ for symmetric } \mathbf{B}$$

$$\frac{\partial}{\partial \mathbf{w}} (\mathbf{w}^T \mathbf{c}) = \mathbf{c} \quad \text{for vector } \mathbf{c}$$

we now set the gradient to zero and solve it:

$$\frac{1}{N}[2\mathbf{X}^T \mathbf{A}\mathbf{X}\mathbf{w} - 2\mathbf{X}^T \mathbf{A}\mathbf{t}] = 0$$

Simplifying it:

$$\mathbf{X}^T \mathbf{A}\mathbf{X}\mathbf{w} = \mathbf{X}^T \mathbf{A}\mathbf{t}$$

isolate  $\mathbf{w}$ , by multiplying both sides by  $(\mathbf{X}^T \mathbf{A}\mathbf{X})^{-1}$ :

$$\hat{\mathbf{w}} = (\mathbf{X}^T \mathbf{A}\mathbf{X})^{-1} \mathbf{X}^T \mathbf{A}\mathbf{t}$$

This is the optimal solution for weighted least squares regression, where  $\mathbf{A} = \text{diag}(\alpha_1, \alpha_2, \dots, \alpha_N)$  is the diagonal matrix of weights.

## 1.2 b)

## 2 Exericise 2 Polynomial Fitting with Regularized Linear Regression and Cross-Validation

## 3 Exercise 3 Pdf and cdf

### 3.1 a)

we need to find the probability density function of

$$F(x) = \begin{cases} 0 & x \leq 0 \\ 1 - \exp(-\beta x^\alpha) & x > 0 \end{cases}$$

The PDF of this function is the derivative, so

$$pdf = f(x) = \frac{dF(x)}{dx}$$

For values of  $x > 0$  we have

$$F(x) = 1 - e^{-\beta x^\alpha}$$

Since  $F(x) = 0$  for  $x \leq 0$ , we have:

$$f(x) = \frac{d}{dx}[0] = 0$$

We need to differentiate  $F(x) = 1 - e^{(-\beta x^\alpha)}$ :

$$\begin{aligned}
 f(x) &= \frac{d}{dx} [1 - e^{(-\beta x^\alpha)}] \\
 &= 0 - \frac{d}{dx} [e^{(-\beta x^\alpha)}] \\
 &= -e^{(-\beta x^\alpha)} \cdot \frac{d}{dx} [-\beta x^\alpha] \quad (\text{chain rule}) \\
 &= -e^{(-\beta x^\alpha)} \cdot (-\beta \alpha x^{\alpha-1}) \quad (\text{power rule}) \\
 &= \alpha \beta x^{\alpha-1} e^{(-\beta x^\alpha)}
 \end{aligned}$$

which gives us an pdf of the function  $F(x)$ :

$$f(x) = \begin{cases} 0 & x \leq 0 \\ \alpha \beta x^{\alpha-1} e^{(-\beta x^\alpha)} & x > 0 \end{cases}$$

### 3.2 b)

We fix the parameters to  $\alpha = 2$  and  $\beta = \frac{1}{4}$ . our cdf becomes:

$$F(x) = 1 - e^{\left(-\frac{x^2}{4}\right)} \quad \text{for } x > 0$$

We need to find  $P(x > 4)$ .

$$P(x > 4) = 1 - F(4)$$

insert:

$$\begin{aligned}
 P(x > 4) &= 1 - (1 - e^{-4}) = e^{-4} \\
 &= 0.01831563888
 \end{aligned}$$

So the probability is 1.83% approximately.

We now need to find the probability that the chip stops working in the time terminal  $[5; 10]$ .  
So

$$P(5 \leq x \leq 10)$$

which means

$$P(5 \leq x \leq 10) = P(10) - P(5)$$

The parameters are still fixed so we insert the values of x

$$P(10) - P(5) = (1 - e^{-25}) - (1 - e^{-25/4})$$

$$\begin{aligned}
&= -e^{-25} + e^{-25/4} \\
&= 0.00193045415
\end{aligned}$$

SO the probability of a chips lifespan being in the interval of 5 to 10 years is approximately 0.193%

### 3.3 c)

We now need to find the median. The median  $m$  is the value where  $F(m) = 0.5$ .

Setting  $F(m) = 0.5$

$$1 - e^{(-\beta m^\alpha)} = 0.5$$

This means we solve for m:

$$1 - e^{(-\beta m^\alpha)} = 0.5$$

$$e^{(-\beta m^\alpha)} = 0.5$$

$$-\beta m^\alpha = \ln(0.5)$$

$$-\beta m^\alpha = -\ln(2)$$

$$\beta m^\alpha = \ln(2)$$

$$m^\alpha = \frac{\ln(2)}{\beta}$$

$$m = \left( \frac{\ln(2)}{\beta} \right)^{1/\alpha}$$

This means the median life span is  $median = \left( \frac{\ln(2)}{\beta} \right)^{1/\alpha}$

## 4 Exercise4 Conditional probability and expectations

### 4.1 a)

We need to calculate the outcomes of remaining silent or not when arrested by the police. The crime for which the police have arrested the suspect is punishable by 5 years in prison. in days, this is equal to

$$5 \cdot 365 = 1825 \text{ days}$$

We have two types of people: those who have never been convicted and those who have been convicted before. And they have two choices: remaining silent or talking to the police during the arrest. We are given some probabilities:

$$P(\text{Court}|\text{Silent}) = 0.001$$

$$P(\text{Court}|\text{NC talks}) = 0.0015$$

$$P(\text{Court}|\text{C talks}) = 0.005$$

then we have probabilities of the outcome if the case goes to court:

$$P(\text{Acquittal}|\text{NC talked}) = 0.8$$

$$P(\text{Acquittal}|\text{C talked}) = 0.2$$

$$P(\text{Acquittal}|\text{NC no talk}) = 0.2$$

$$P(\text{Acquittal}|\text{C no talk}) = 0.05$$

And the last thing is, if the person talked at the arrest stage and gets convicted, the sentence is reduced by 50%.

#### 4.1.1 Remain silent

The probabilities for remaining silent:

Case does not go to court:  $P(\text{No Court}) = 1 - 0.001 = 0.999 \Rightarrow 0 \text{ days}$

Case goes to court:  $P(\text{Court}) = 0.001$

Acquitted:  $P(\text{Acquittal} | \text{Court, Silent}) = 0.2 \Rightarrow 0 \text{ days}$

Convicted:  $P(\text{Conviction} | \text{Court, Silent}) = 1 - 0.2 = 0.8 \Rightarrow 1825 \text{ days}$

The expected number of days in prison is:

$$\begin{aligned}
 E[\text{Days} | \text{Silent, NC}] &= P(\text{No Court}) \cdot 0 \\
 &\quad + P(\text{Court}) \cdot P(\text{Acquittal} | \text{Court, Silent}) \cdot 0 \\
 &\quad + P(\text{Court}) \cdot P(\text{Conviction} | \text{Court, Silent}) \cdot 1825 \\
 &= 0.999 \cdot 0 + 0.001 \cdot 0.2 \cdot 0 + 0.001 \cdot 0.8 \cdot 1825 \\
 &= 0 + 0 + 1.46 \\
 &= 1.46 \text{ days}
 \end{aligned}$$

#### 4.1.2 talks to police

The probabilities for talking to the police:

Case does not go to court:  $P(\text{No Court}) = 1 - 0.0015 = 0.9985 \Rightarrow 0 \text{ days}$

Case goes to court:  $P(\text{Court}) = 0.0015$

Acquitted:  $P(\text{Acquittal} | \text{Court, Talk}) = 0.8 \Rightarrow 0 \text{ days}$

Convicted:  $P(\text{Conviction} | \text{Court, Talk}) = 1 - 0.8 = 0.2 \Rightarrow 1825 \cdot 0.5 = 912.5 \text{ days}$

$$\begin{aligned}
 E[\text{Days} | \text{Talk, NC}] &= P(\text{No Court}) \cdot 0 \\
 &\quad + P(\text{Court}) \cdot P(\text{Acquittal} | \text{Court, Talk}) \cdot 0 \\
 &\quad + P(\text{Court}) \cdot P(\text{Conviction} | \text{Court, Talk}) \cdot 912.5 \\
 &= 0.9985 \cdot 0 + 0.0015 \cdot 0.8 \cdot 0 + 0.0015 \cdot 0.2 \cdot 912.5 \\
 &= 0 + 0 + 0.27375 \\
 &= 0.27 \text{ days}
 \end{aligned}$$

This means that Talking to the police during the arrest would result in an expected lower amount of days in prison.

## 4.2 b)

Now we do the same, but with the history of convictions.

#### 4.2.1 Remain silent

The probabilities for remaining silent: Case does not go to court  $P(\text{No Court}) = 1 - 0.001 = 0.999 \Rightarrow 0 \text{ days}$

Case goes to court:  $P(\text{Court}) = 0.001$

Acquitted:  $P(\text{Acquittal}|\text{court}, \text{Silent}) = 0.05 \Rightarrow 0 \text{ days}$

Convicted:  $P(\text{Convicted}|\text{Court}, \text{silent}) = 1 - 0.05 = 0.95 \Rightarrow 1825 \text{ days}$

$$\begin{aligned}
E[\text{Days} | \text{Silent, C}] &= P(\text{No Court}) \cdot 0 \\
&\quad + P(\text{Court}) \cdot P(\text{Acquittal} | \text{Court, Silent}) \cdot 0 \\
&\quad + P(\text{Court}) \cdot P(\text{Conviction} | \text{Court, Silent}) \cdot 1825 \\
&= 0.999 \cdot 0 + 0.001 \cdot 0.05 \cdot 0 + 0.001 \cdot 0.95 \cdot 1825 \\
&= 0 + 0 + 1.73375 \\
&= 1.73 \text{ days}
\end{aligned}$$

#### 4.2.2 Talks to police

The probabilities for remaining silent:

Case does not go to court  $P(\text{No Court}) = 1 - 0.005 = 0.995 \Rightarrow 0 \text{ days}$

Case goes to court:  $P(\text{Court}) = 0.005$

Acquitted:  $P(\text{Acquittal}|\text{court}, \text{talks}) = 0.2 \Rightarrow 0 \text{ days}$

Convicted:  $P(\text{Convicted}|\text{Court}, \text{talks}) = 1 - 0.2 = 0.8 \Rightarrow 1825 \cdot 0.5 = 912.5 \text{ days}$

$$\begin{aligned}
E[\text{Days} | \text{Talk, C}] &= P(\text{No Court}) \cdot 0 \\
&\quad + P(\text{Court}) \cdot P(\text{Acquittal} | \text{Court, Talk}) \cdot 0 \\
&\quad + P(\text{Court}) \cdot P(\text{Conviction} | \text{Court, Talk}) \cdot 912.5 \\
&= 0.995 \cdot 0 + 0.005 \cdot 0.2 \cdot 0 + 0.005 \cdot 0.8 \cdot 912.5 \\
&= 0 + 0 + 3.65 \\
&= 3.65 \text{ days}
\end{aligned}$$

this means that a person with a conviction history has better odds by remaining silence when arrested by the police, as  $E[\text{Days} | \text{Talk, C}] > E[\text{Days} | \text{Silent, C}]$