

## I. SUPPLEMENTARY MATERIAL

### A. Proof of Lemma 1

According to the paradigm compatibility, we can construct the following inequality:

$$\begin{aligned}
& \| \mathbf{W}^T \text{soft}(\mathbf{W} \mathbf{x}_1, \varepsilon) - \mathbf{W}^T \text{soft}(\mathbf{W} \mathbf{x}_2, \varepsilon) \|_2^2 \\
&= \| \mathbf{W}^T [\text{soft}(\mathbf{W} \mathbf{x}_1, \varepsilon) - \text{soft}(\mathbf{W} \mathbf{x}_2, \varepsilon)] \|_2^2 \\
&\leq \| \mathbf{W}^T \|_2^2 \| \text{soft}(\mathbf{W} \mathbf{x}_1, \varepsilon) - \text{soft}(\mathbf{W} \mathbf{x}_2, \varepsilon) \|_2^2 \quad (1) \\
&= \lambda_{\max}(\mathbf{W} \mathbf{W}^T) \| \text{soft}(\mathbf{W} \mathbf{x}_1, \varepsilon) - \text{soft}(\mathbf{W} \mathbf{x}_2, \varepsilon) \|_2^2 \\
&= \lambda_{\max}(\mathbf{W} \mathbf{W}^T) \| \text{soft}(\mathbf{W} \mathbf{x}_2, \varepsilon) - \text{soft}(\mathbf{W} \mathbf{x}_1, \varepsilon) \|_2^2
\end{aligned}$$

Let  $(\mathbf{W}\mathbf{x})_i$  represent the  $i$ -th element of  $\mathbf{W}\mathbf{x}$ . The soft thresholding operator  $soft(\mathbf{W}\mathbf{x}, \epsilon)$  is defined as:

$$soft(\mathbf{W}\mathbf{x}, \varepsilon) = \begin{cases} (\mathbf{W}\mathbf{x})_i + \varepsilon_i, & (\mathbf{W}\mathbf{x})_i < -\varepsilon_i \\ 0, & |(\mathbf{W}\mathbf{x})_i| \leq \varepsilon_i \\ (\mathbf{W}\mathbf{x})_i - \varepsilon_i, & (\mathbf{W}\mathbf{x})_i > \varepsilon_i \end{cases} \quad (2)$$

We divide our discussion of Eqn. (1) into the following scenarios:

$$(i)(\mathbf{W}\mathbf{x}_1)_i < (\mathbf{W}\mathbf{x}_2)_i$$

a. if  $(\mathbf{W}\mathbf{x}_1)_i < (\mathbf{W}\mathbf{x}_2)_i < -\varepsilon_i$ , then  $\text{soft}(\mathbf{W}\mathbf{x}_1, \varepsilon) = \mathbf{W}\mathbf{x}_1 + \varepsilon$ ,  $\text{soft}(\mathbf{W}\mathbf{x}_2, \varepsilon) = \mathbf{W}\mathbf{x}_2 + \varepsilon$ ;

$$\begin{aligned}
& ||soft(\mathbf{W}\mathbf{x}_2, \varepsilon) - soft(\mathbf{W}\mathbf{x}_1, \varepsilon)||_2^2 \\
&= ||(\mathbf{W}\mathbf{x}_2 + \varepsilon) - (\mathbf{W}\mathbf{x}_1 + \varepsilon)||_2^2 \\
&= ||\mathbf{W}\mathbf{x}_2 - \mathbf{W}\mathbf{x}_1||_2^2 \\
&\leq ||\mathbf{W}||_2^2 ||\mathbf{x}_2 - \mathbf{x}_1||_2^2 \\
&= ||\mathbf{x}_2 - \mathbf{x}_1||_2^2
\end{aligned} \tag{3}$$

b. if  $(\mathbf{W}\mathbf{x}_1)_i < -\varepsilon_i \leq (\mathbf{W}\mathbf{x}_2)_i \leq \varepsilon_i$ , then  $\text{soft}(\mathbf{W}\mathbf{x}_1, \varepsilon) = \mathbf{W}\mathbf{x}_1 + \varepsilon, \text{soft}(\mathbf{W}\mathbf{x}_2, \varepsilon) = 0$ ;

$$\begin{aligned}
& \|soft(\mathbf{W}\mathbf{x}_2, \varepsilon) - soft(\mathbf{W}\mathbf{x}_1, \varepsilon)\|_2^2 \\
&= \|0 - (\mathbf{W}\mathbf{x}_1 + \varepsilon)\|_2^2 \\
&= \|-\mathbf{W}\mathbf{x}_1 - \varepsilon\|_2^2 \\
&= \|-\varepsilon - \mathbf{W}\mathbf{x}_1\|_2^2 \\
&\leq \|\mathbf{W}\mathbf{x}_2 - \mathbf{W}\mathbf{x}_1\|_2^2 \\
&\leq \|\mathbf{W}\|_2^2 \|\mathbf{x}_2 - \mathbf{x}_1\|_2^2 \\
&= \|\mathbf{x}_2 - \mathbf{x}_1\|_2^2
\end{aligned} \tag{4}$$

c. if  $(\mathbf{W}\mathbf{x}_1)_i < -\varepsilon_i < \varepsilon_i < (\mathbf{W}\mathbf{x}_2)_i$ , then  $\text{soft}(\mathbf{W}\mathbf{x}_1, \varepsilon) = \mathbf{W}\mathbf{x}_1 + \varepsilon, \text{soft}(\mathbf{W}\mathbf{x}_2, \varepsilon) = \mathbf{W}\mathbf{x}_2 - \varepsilon$ ;

$$\begin{aligned}
& \|soft(\mathbf{W}\mathbf{x}_2, \varepsilon) - soft(\mathbf{W}\mathbf{x}_1, \varepsilon)\|_2^2 \\
&= \|(\mathbf{W}\mathbf{x}_2 - \varepsilon) - (\mathbf{W}\mathbf{x}_1 + \varepsilon)\|_2^2 \\
&= \|\mathbf{W}\mathbf{x}_2 - \mathbf{W}\mathbf{x}_1 - 2\varepsilon\|_2^2 \\
&\leq \|\mathbf{W}\mathbf{x}_2 - \mathbf{W}\mathbf{x}_1\|_2^2 \\
&\leq \|\mathbf{W}\|_2^2 \|\mathbf{x}_2 - \mathbf{x}_1\|_2^2 \\
&= \|\mathbf{x}_2 - \mathbf{x}_1\|_2^2
\end{aligned} \tag{5}$$

d. if  $-\varepsilon_i \leq (\mathbf{W}\mathbf{x}_1)_i < (\mathbf{W}\mathbf{x}_2)_i \leq \varepsilon_i$ , then  $\text{soft}(\mathbf{W}\mathbf{x}_1, \varepsilon) = 0, \text{soft}(\mathbf{W}\mathbf{x}_2, \varepsilon) = 0, :$

$$\begin{aligned} & ||soft(\mathbf{W}\mathbf{x}_2, \varepsilon) - soft(\mathbf{W}\mathbf{x}_1, \varepsilon)||_2^2 \\ &= ||0 - 0||_2^2 \\ &= 0 \end{aligned} \tag{6}$$

e. if  $-\varepsilon_i \leq (\mathbf{W}\mathbf{x}_1)_i \leq \varepsilon_i < (\mathbf{W}\mathbf{x}_2)_i$ , then  $soft(\mathbf{W}\mathbf{x}_1, \varepsilon) = 0, soft(\mathbf{W}\mathbf{x}_2, \varepsilon) = \mathbf{W}\mathbf{x}_2 - \varepsilon$ ;

$$\begin{aligned}
& ||\text{soft}(\mathbf{W}\mathbf{x}_2, \varepsilon) - \text{soft}(\mathbf{W}\mathbf{x}_1, \varepsilon)||_2^2 \\
&= ||(\mathbf{W}\mathbf{x}_2 - \varepsilon) - 0||_2^2 \\
&\leq ||\mathbf{W}\mathbf{x}_2 - \mathbf{W}\mathbf{x}_1||_2^2 \\
&\leq ||\mathbf{W}||_2^2 ||\mathbf{x}_2 - \mathbf{x}_1||_2^2 \\
&= ||\mathbf{x}_2 - \mathbf{x}_1||_2^2
\end{aligned} \tag{7}$$

f. if  $\varepsilon_i < (\mathbf{W}\mathbf{x}_1)_i < (\mathbf{W}\mathbf{x}_2)_i$ , then  $\text{soft}(\mathbf{W}\mathbf{x}_1, \varepsilon) = \mathbf{W}\mathbf{x}_1 - \varepsilon, \text{soft}(\mathbf{W}\mathbf{x}_2, \varepsilon) = \mathbf{W}\mathbf{x}_2 - \varepsilon, :$

$$\begin{aligned}
& \| \text{soft}(\mathbf{W}\mathbf{x}_2, \varepsilon) - \text{soft}(\mathbf{W}\mathbf{x}_1, \varepsilon) \|_2^2 \\
&= \| (\mathbf{W}\mathbf{x}_2 - \varepsilon) - (\mathbf{W}\mathbf{x}_1 - \varepsilon) \|_2^2 \\
&= \| \mathbf{W}\mathbf{x}_2 - \mathbf{W}\mathbf{x}_1 \|_2^2 \\
&\leq \| \mathbf{W} \|_2^2 \| \mathbf{x}_2 - \mathbf{x}_1 \|_2^2 \\
&= \| \mathbf{x}_2 - \mathbf{x}_1 \|_2^2
\end{aligned} \tag{8}$$

$$(ii) (\mathbf{W} \mathbf{x}_2)_i < (\mathbf{W} \mathbf{x}_1)_i$$

a. if  $(Wx_2)_i < (Wx_1)_i < -\varepsilon_i$ , then  $\text{soft}(Wx_2, \varepsilon) = Wx_2 + \varepsilon, \text{soft}(Wx_1, \varepsilon) = Wx_1 + \varepsilon$ ;

$$\begin{aligned}
& ||\text{soft}(\mathbf{W}\mathbf{x}_2, \varepsilon) - \text{soft}(\mathbf{W}\mathbf{x}_1, \varepsilon)||_2^2 \\
&= ||(\mathbf{W}\mathbf{x}_2 + \varepsilon) - (\mathbf{W}\mathbf{x}_1 + \varepsilon)||_2^2 \\
&= ||\mathbf{W}\mathbf{x}_2 - \mathbf{W}\mathbf{x}_1||_2^2 \\
&\leq ||\mathbf{W}||_2^2 ||\mathbf{x}_2 - \mathbf{x}_1||_2^2 \\
&= ||\mathbf{x}_2 - \mathbf{x}_1||_2^2
\end{aligned} \tag{9}$$

b. if  $(\mathbf{W}\mathbf{x}_2)_i < -\varepsilon_i \leq (\mathbf{W}\mathbf{x}_1)_i \leq \varepsilon_i$ , then  
 $\text{soft}(\mathbf{W}\mathbf{x}_2, \varepsilon) = \mathbf{W}\mathbf{x}_2 + \varepsilon, \text{soft}(\mathbf{W}\mathbf{x}_1, \varepsilon) = 0$ ;

$$\begin{aligned}
& \|soft(\mathbf{W}\mathbf{x}_2, \boldsymbol{\varepsilon}) - soft(\mathbf{W}\mathbf{x}_1, \boldsymbol{\varepsilon})\|_2^2 \\
&= \|(\mathbf{W}\mathbf{x}_2 + \boldsymbol{\varepsilon}) - 0\|_2^2 \\
&= \|\mathbf{W}\mathbf{x}_2 + \boldsymbol{\varepsilon}\|_2^2 \\
&= \|\mathbf{W}\mathbf{x}_2 - (-\boldsymbol{\varepsilon})\|_2^2 \\
&\leq \|\mathbf{W}\mathbf{x}_2 - \mathbf{W}\mathbf{x}_1\|_2^2 \\
&\leq \|\mathbf{W}\|_2^2 \|\mathbf{x}_2 - \mathbf{x}_1\|_2^2 \\
&= \|\mathbf{x}_2 - \mathbf{x}_1\|_2^2
\end{aligned} \tag{10}$$

c. if  $(\mathbf{W}\mathbf{x}_2)_i < -\varepsilon_i < \varepsilon_i < (\mathbf{W}\mathbf{x}_1)_i$ , then  $\text{soft}(\mathbf{W}\mathbf{x}_2, \varepsilon) = \mathbf{W}\mathbf{x}_2 + \varepsilon, \text{soft}(\mathbf{W}\mathbf{x}_1, \varepsilon) = \mathbf{W}\mathbf{x}_1 - \varepsilon, :$

$$\begin{aligned}
& \|soft(\mathbf{W}\mathbf{x}_2, \varepsilon) - soft(\mathbf{W}\mathbf{x}_1, \varepsilon)\|_2^2 \\
&= \|(\mathbf{W}\mathbf{x}_2 + \varepsilon) - (\mathbf{W}\mathbf{x}_1 + \varepsilon)\|_2^2 \\
&= \|\mathbf{W}\mathbf{x}_2 - \mathbf{W}\mathbf{x}_1 + 2\varepsilon\|_2^2 \\
&\leq \|\mathbf{W}\mathbf{x}_2 - \mathbf{W}\mathbf{x}_1\|_2^2 \\
&\leq \|\mathbf{W}\|_2^2 \|\mathbf{x}_2 - \mathbf{x}_1\|_2^2 \\
&= \|\mathbf{x}_2 - \mathbf{x}_1\|_2^2
\end{aligned} \tag{11}$$

d. if  $-\varepsilon_i \leq (\mathbf{W}\mathbf{x}_2)_i < (\mathbf{W}\mathbf{x}_1)_i \leq \varepsilon_i$ , then  $\text{soft}(\mathbf{W}\mathbf{x}_2, \varepsilon) = 0, \text{soft}(\mathbf{W}\mathbf{x}_1, \varepsilon) = 0$ ;

$$\begin{aligned} & ||soft(\mathbf{W}\mathbf{x}_2, \varepsilon) - soft(\mathbf{W}\mathbf{x}_1, \varepsilon)||_2^2 \\ &= ||0 - 0||_2^2 \\ &= 0 \end{aligned} \tag{12}$$

e. if  $-\varepsilon_i \leq (\mathbf{W}\mathbf{x}_2)_i \leq \varepsilon_i < (\mathbf{W}\mathbf{x}_1)_i$ , then  $\text{soft}(\mathbf{W}\mathbf{x}_2, \varepsilon) = 0, \text{soft}(\mathbf{W}\mathbf{x}_1, \varepsilon) = \mathbf{W}\mathbf{x}_1 - \varepsilon, :$

$$\begin{aligned}
& \|\text{soft}(\mathbf{W}\mathbf{x}_2, \varepsilon) - \text{soft}(\mathbf{W}\mathbf{x}_1, \varepsilon)\|_2^2 \\
&= \|0 - (\mathbf{W}\mathbf{x}_1 - \varepsilon)\|_2^2 \\
&= \|\varepsilon - \mathbf{W}\mathbf{x}_1\|_2^2 \\
&\leq \|\mathbf{W}\mathbf{x}_2 - \mathbf{W}\mathbf{x}_1\|_2^2 \\
&\leq \|\mathbf{W}\|_2^2 \|\mathbf{x}_2 - \mathbf{x}_1\|_2^2 \\
&= \|\mathbf{x}_2 - \mathbf{x}_1\|_2^2
\end{aligned} \tag{13}$$

f. if  $\varepsilon_i < (\mathbf{W}\mathbf{x}_2)_i < (\mathbf{W}\mathbf{x}_1)_i$ , then  $\text{soft}(\mathbf{W}\mathbf{x}_2, \varepsilon) = \mathbf{W}\mathbf{x}_2 - \varepsilon, \text{soft}(\mathbf{W}\mathbf{x}_1, \varepsilon) = \mathbf{W}\mathbf{x}_1 - \varepsilon, :$

$$\begin{aligned}
& \|\text{soft}(\mathbf{W}\mathbf{x}_2, \varepsilon) - \text{soft}(\mathbf{W}\mathbf{x}_1, \varepsilon)\|_2^2 \\
&= \|(\mathbf{W}\mathbf{x}_2 - \varepsilon) - (\mathbf{W}\mathbf{x}_1 - \varepsilon)\|_2^2 \\
&= \|\mathbf{W}\mathbf{x}_2 - \mathbf{W}\mathbf{x}_1\|_2^2 \\
&\leq \|\mathbf{W}\|_2^2 \|\mathbf{x}_2 - \mathbf{x}_1\|_2^2 \\
&= \|\mathbf{x}_2 - \mathbf{x}_1\|_2^2
\end{aligned} \tag{14}$$

From the above inequalities, it can be seen that the upper bound of  $\|\text{soft}(\mathbf{W}\mathbf{x}_2, \varepsilon) - \text{soft}(\mathbf{W}\mathbf{x}_1, \varepsilon)\|_2^2$  is either 0 or  $\|\mathbf{x}_2 - \mathbf{x}_1\|_2^2$ . The  $\mathbf{W}^T \text{soft}(\mathbf{W}\mathbf{x}, \varepsilon)$  is  $\lambda_{max}(\mathbf{W}\mathbf{W})^T$ -Lipschitz.