

I. SUPPLEMENTARY MATERIAL

A. Proof of Lemma 1

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

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

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

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

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 $\text{soft}(\mathbf{W}\mathbf{x}_1, \varepsilon) = \mathbf{W}\mathbf{x}_1 + \varepsilon$, $\text{soft}(\mathbf{W}\mathbf{x}_2, \varepsilon) = \mathbf{W}\mathbf{x}_2 + \varepsilon$:

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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 $\mathbf{W}^\top \text{soft}(\mathbf{W}\mathbf{x}, \varepsilon)$ is $\lambda_{\max}(\mathbf{W}\mathbf{W})^\top$ -Lipschitz.