Supplementary Materials

1 Proof of Lemma 4.1

Lemma 4.1. The scaled soft-plus function $\gamma_s(x) = s \log (1 + \exp(x/s))$ (s > 0) is convex and $\log (\gamma_s(x))$ is concave.

Proof. Since s is a positive constant, we only need to show that the soft-plus function $\gamma(x) = \gamma_1(x)$ is convex and log concave. Then it is straightforward to show that the scaled version is also convex and log concave. To this end, we first observe that

$$\gamma(x) = \log(1 + \exp(x)) = -\log(\sigma(-x))$$

where $\sigma(x) = 1/(1 + \exp(-x))$ is the sigmoid activation function. We then take the gradient of $\gamma(x)$,

$$\frac{\mathrm{d}\gamma(x)}{\mathrm{d}x} = -\frac{1}{\sigma(-x)}\sigma(-x)(1-\sigma(-x))(-1) = \sigma(x). \tag{1}$$

Note that we have used a known fact that $\frac{d\sigma(x)}{dx} = \sigma(x)(1 - \sigma(x))$. Next, we take the second derivative,

$$\frac{\mathrm{d}^2 \gamma(x)}{\mathrm{d}x^2} = \sigma(x)(1 - \sigma(x)).$$

Since $\forall x \in \mathbb{R}$, we have $0 \le \sigma(x) \le 1$, we must have $\frac{d^2 \gamma(x)}{dx^2} \ge 0$. Therefore, $\gamma(x)$ is convex.

Now, let us look at $h(x) = \log (\gamma(x))$. First, we can derive the first derivative based on (1),

$$\frac{\mathrm{d}h(x)}{\mathrm{d}x} = \frac{1}{\gamma(x)} \frac{\mathrm{d}\gamma(x)}{\mathrm{d}x} = \frac{\sigma(x)}{\gamma(x)}.$$

Then, the second derivative is

$$\frac{\mathrm{d}^2 h(x)}{\mathrm{d}x^2} = \frac{\frac{\mathrm{d}\sigma(x)}{\mathrm{d}x}\gamma(x) - \sigma(x)\frac{\mathrm{d}\gamma(x)}{\mathrm{d}x}}{\left(\gamma(x)\right)^2} = \frac{\sigma(x)\cdot g(x)}{\left(\gamma(x)\right)^2} \tag{2}$$

where

$$g(x) = (1 - \sigma(x))\gamma(x) - \sigma(x).$$

From (2), we can see that $\sigma(x) \geq 0$ and $(\gamma(x))^2 \geq 0$. Therefore, we only need to check if $g(x) \leq 0$ to show the concavity of $h(\cdot)$. Since $\gamma(x) = -\log(\sigma(-x)) = -\log(1-\sigma(x))$, we can view g(x) as a function of $t=1-\sigma(x)$, namely,

$$g(x) = g(t) = -t\log(t) - (1-t) = t(1-\log(t)) - 1,$$

and $0 \le t \le 1$. Note that g(t) = 0 when t = 1. We take the derivative of $g(\cdot)$ w.r.t t,

$$\frac{dg(t)}{dt} = 1 - \log(t) + t(-\frac{1}{t}) = -\log(t) \ge 0.$$

Therefore, g(t) is monotonically increasing with t. Since $0 \le t \le 1$, we always have $g(t) \le g(t=1) = 0$. Hence, $\forall x, g(x) \le 0$. From (2), we have $\frac{\mathrm{d}^2 h(x)}{\mathrm{d} x^2} \le 0$, and hence the log soft-plus function is concave.

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