

## 7 Appendix

### 7.1 Proof of Theorem 1

The Proof is based on [5, 16]. Before we prove the theorem, we need the following lemma.

**Lemma 1.** (*Bisla et al.*) Let  $L(w)$  be  $\alpha$ -Lipschitz continuous. Let  $\delta$  be distributed according to the distribution  $p$ . Then,

$$\begin{aligned} \|\nabla L_S(x) - \nabla L_S(y)\| &= \mathbb{E}_{\delta \sim p} [\nabla L_S(x + \delta) - \nabla L_S(y + \delta)] \\ &\leq \alpha \int |p(\delta - x) - p(\delta - y)| d\delta. \end{aligned} \quad (8)$$

Then we provide the proof of Theorem 1. We firstly illustrate that  $\nabla L_S(w)$  is  $\frac{\alpha}{\gamma}$ -Lipschitz continuous, and then demonstrate that  $\nabla L_S(w)$  is  $\beta$ -Lipschitz continuous. Finally, we can obtain that  $\nabla L_S(w)$  is  $\min\{\frac{\alpha}{\gamma}, \beta\}$ -Lipschitz continuous.

We can define  $\int |p(\delta - x) - p(\delta - y)| d\delta$  as:

$$\begin{aligned} &\int |p(\delta - x) - p(\delta - y)| d\delta \\ &= \int_{\delta: \|\delta - x\| \geq \|\delta - y\|} [p(\delta - x) - p(\delta - y)] d\delta + \int_{\delta: \|\delta - x\| \leq \|\delta - y\|} [p(\delta - y) - p(\delta - x)] d\delta \\ &= 2 \int_{\delta: \|\delta - x\| \leq \|\delta - y\|} [p(\delta - x) - p(\delta - y)] d\delta. \end{aligned} \quad (9)$$

We try to divide  $\int |p(\delta - x) - p(\delta - y)| d\delta$  into two different parts in the first equality. The second equality holds because they have the equal value when  $\|p(\delta - x)\| \geq \|p(\delta - y)\|$  and  $\|p(\delta - x)\| \leq \|p(\delta - y)\|$ .

Let  $\hat{\delta} = \delta - x$  for  $p(\delta - x)$  term and  $\hat{\delta} = \delta - y$  for  $p(\delta - y)$  term, we have:

$$\begin{aligned} &\int |p(\delta - x) - p(\delta - y)| d\delta \\ &= 2 \int_{\delta: \|\delta - x\| \leq \|\delta - y\|} p(\delta - x) d\delta - 2 \int_{\delta: \|\delta - x\| \leq \|\delta - y\|} p(\delta - y) d\delta \\ &= 2 \int_{\delta: \|\hat{\delta}\| \leq \|\hat{\delta} + (x - y)\|} p(\hat{\delta}) d\delta - 2 \int_{\delta: \|\hat{\delta}\| \geq \|\hat{\delta} - (x - y)\|} p(\hat{\delta}) d\delta \end{aligned} \quad (10)$$

We can rewrite Equation 10 as:

$$\begin{aligned} &\int |p(\delta - x) - p(\delta - y)| d\delta \\ &= 2\mathbb{P}_{\delta \sim p}(\|\delta\| \leq \|\delta + (x - y)\|) - 2\mathbb{P}_{\delta \sim p}(\|\delta\| \geq \|\delta - (x - y)\|). \end{aligned} \quad (11)$$

For the first part of the Equation 11, we can find:

$$\begin{aligned} &\mathbb{P}_{\delta \sim p}(\|\delta\| \leq \|\delta + (x - y)\|) \\ &= \mathbb{P}_{\delta \sim p}(\|\delta\|^2 \leq \|\delta + (x - y)\|^2) \\ &= \mathbb{P}_{\delta \sim p}(2\langle \delta, x - y \rangle \geq -\|x - y\|^2) \\ &= \mathbb{P}_{\delta \sim p}(2\langle \delta, \frac{x - y}{\|x - y\|} \rangle \geq -\|x - y\|). \end{aligned} \quad (12)$$

In addition,  $\frac{x-y}{\|x-y\|}$  has norm 1 and  $\delta \sim \mathbb{N}(0, \gamma^2 I)$  implies  $\langle \delta, \frac{x-y}{\|x-y\|} \rangle \sim \mathbb{N}(0, \gamma^2)$ . Therefore,

$$\begin{aligned} & \mathbb{P}_{\delta \sim p}(\|\delta\| \leq \|\delta + (x - y)\|) \\ &= \mathbb{P}_{\delta \sim p}(\langle \delta, \frac{x-y}{\|x-y\|} \rangle \geq -\frac{\|x-y\|}{2}) \\ &= \int_{-\frac{\|x-y\|}{2}}^{+\infty} \frac{1}{\sqrt{2\pi\gamma^2}} \exp(-\frac{\hat{\delta}^2}{2\gamma^2}) d\hat{\delta}. \end{aligned} \quad (13)$$

For the second part in Equation 11, we can obtain that:

$$\begin{aligned} & \mathbb{P}_{\delta \sim p}(\|\delta\| \geq \|\delta - (x - y)\|) \\ &= \mathbb{P}_{\delta \sim p}(\|\delta\|^2 \geq \|\delta - (x - y)\|^2) \\ &= \mathbb{P}_{\delta \sim p}(2\langle \delta, x - y \rangle \geq \|x - y\|^2) \\ &= \mathbb{P}_{\delta \sim p}(2\langle \delta, \frac{x-y}{\|x-y\|} \rangle \geq \|x - y\|). \end{aligned} \quad (14)$$

We can also get the similar distribution  $\langle \delta, \frac{x-y}{\|x-y\|} \rangle \sim \mathbb{N}(0, \gamma^2)$ . Therefore, we can find:

$$\begin{aligned} & \mathbb{P}_{\delta \sim p}(\|\delta\| \geq \|\delta - (x - y)\|) \\ &= \mathbb{P}_{\delta \sim p}(\langle \delta, \frac{x-y}{\|x-y\|} \rangle \geq \frac{\|x-y\|}{2}) \\ &= \int_{\frac{\|x-y\|}{2}}^{+\infty} \frac{1}{\sqrt{2\pi\gamma^2}} \exp(-\frac{\hat{\delta}^2}{2\gamma^2}) d\hat{\delta}. \end{aligned} \quad (15)$$

Then, we can combine Equation 13 and 15 to Equation 11:

$$\begin{aligned} & \int |p(\delta - x) - p(\delta - y)| d\delta \\ &= 2 \int_{-\frac{\|x-y\|}{2}}^{+\infty} \frac{1}{\sqrt{2\pi\gamma^2}} \exp(-\frac{\hat{\delta}^2}{2\gamma^2}) d\hat{\delta} - 2 \int_{\frac{\|x-y\|}{2}}^{+\infty} \frac{1}{\sqrt{2\pi\gamma^2}} \exp(-\frac{\hat{\delta}^2}{2\gamma^2}) d\hat{\delta} \\ &= 2 \int_{-\frac{\|x-y\|}{2}}^{\frac{\|x-y\|}{2}} \frac{1}{\sqrt{2\pi\gamma^2}} \exp(-\frac{\hat{\delta}^2}{2\gamma^2}) d\hat{\delta} \\ &\leq \frac{\sqrt{2}\|x-y\|}{\gamma\sqrt{\pi}} \end{aligned} \quad (16)$$

Therefore, we can combine Equation 8 and Equation 16:

$$\begin{aligned} \|\nabla L_S(x) - \nabla L_S(y)\| &\leq \alpha \int |p(\delta - x) - p(\delta - y)| d\delta \\ &\leq \alpha \frac{\sqrt{2}\|x-y\|}{\gamma\sqrt{\pi}} \\ &\leq \frac{\alpha}{\gamma} \|x-y\| \end{aligned} \quad (17)$$

Finally, we finish the proof of  $\frac{\alpha}{\gamma}$ -Lipschitz continuous for  $\nabla L_S(w)$ . Then, we try to show the proof of  $\beta$ -Lipschitz continuous for  $\nabla L_S(w)$ :

$$\begin{aligned}
\|\nabla L_S(x) - \nabla L_S(y)\| &= \|\nabla \mathbb{E}_{\delta \sim p}[L(x + \delta)] - \nabla \mathbb{E}_{\delta \sim p}[L(x + \delta)]\| \\
&= \|\mathbb{E}_{\delta \sim p}[\nabla L(x + \delta)] - \nabla L(y + \delta)\| \\
&= \left\| \int [\nabla L(x + \delta)] - \nabla L(y + \delta)] p(\delta) d\delta \right\| \\
&\leq \int \|\nabla L(x + \delta) - \nabla L(y + \delta)\| p(\delta) d\delta \\
&\leq \int \beta \|(x + \delta)\| \int p(\delta) d\delta \\
&= \beta \|x - y\| \int p(\delta) d\delta \\
&= \beta \|x - y\|
\end{aligned} \tag{18}$$

Therefore, we can obtain  $\nabla L_S(w)$  is  $\beta$ -Lipschitz continuous.

Finally, we finish the proof that  $\nabla L_S(w)$  is  $\min\{\frac{\alpha}{\gamma}, \beta\}$ -Lipschitz continuous.

## 7.2 Compared with LPF-SGD

As shown in Table 1 and Table 2, we try to compare R-SAM with LPF-SGD on CIFAR-10 and CIFAR-100. In this section, we also try to compare R-SAM with LPF-SGD on ImageNet.

Table 5: Accuracy of ViT on ImageNet-1k for 300 epoch. The base optimizer for SAM and R-SAM is AdamW. Batch Size is 4096. We use Inception-style preprocessing method for input image.

Model	Resolution	AdamW	LPF-SGD	SAM	R-SAM
<b>ViT-B-16</b>	224	74.7	75.9	79.8	<b>80.7</b>
<b>ViT-S-16</b>	224	74.9	75.8	77.9	<b>78.7</b>

## 7.3 The sensitivity Analysis of $\gamma$

In this section, we try to analyze the sensitivity of  $\gamma$  for R-SAM. The result is shown in Table 6.

Table 6: The sensitivity Analysis about  $\gamma$

Model	2e-3	1.5e-3	1e-3	5e-4	1e-4	5e-5
<b>WRN-28-10</b>	84.7	85.1	85.3	85.4	85.1	84.5
<b>ViT-B-16</b>	79.7	80.2	80.7	80.7	80.4	79.7
<b>ViT-S-16</b>	77.3	77.9	78.5	78.2	77.9	77.8

## 7.4 Hyperparameters

Table 7: Architectures of Vision Transformer

Model	Params	Patch Resolution	Sequence Length	Hidden Size	Heads	Layers
ViT-B-16	87M	$16 \times 16$	196	768	12	12
ViT-S-16	22M	$16 \times 16$	196	384	6	12

Table 8: Parameter Settings of ViT from Scratch on ImageNet

Model	Input Resolution	Batch Size	Epoch	Warmup Steps	Peak LR	LR Decay	Optimizer	$\rho$	$\lambda$	Weight Decay	Gradient Clipping
ViT-B-16	224	4096	300	10000	3e-3	cosine	AdamW	/	/	0.3	1.0
ViT-S-16	224	4096	300	10000	3e-3	cosine	AdamW	/	/	0.3	1.0
ViT-B-16 + SAM	224	4096	300	10000	3e-3	cosine	AdamW	0.18	/	0.3	1.0
ViT-S-16 + SAM	224	4096	300	10000	3e-3	cosine	AdamW	0.1	/	0.2	1.0
ViT-B-16 + R-SAM	224	4096	300	10000	3e-3	linear	AdamW	1.4	2	0.3	1.0
ViT-S-16 + R-SAM	224	4096	300	10000	3e-3	linear	AdamW	0.4	1	0.2	1.0