# Convergence Analysis

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#### Notation & Definitions

- t: time index and  $t \in \mathbb{Z}_+$ .
- $z_t$ : position of the target at time t and  $z_t \in \mathbb{R}^n$ .
- $x_t$ : position of the agent at time t and  $x_t \in \mathbb{R}^n$ .
- $\ell_t(x,z)$ : stochastic loss evaluated at time t, position of agent at x, and position of target at z for  $x,z \in \mathbb{R}^n$ .
- $\ell_{u,t}(x,z) := \mathbb{E}_u[\ell_t(x+\mu u,z)]$  for  $x,z,u\in\mathbb{R}^n$  and  $\mu\in\mathbb{R}$ .
- $\nabla \ell_{\mu,t}(x,z) := \mathbb{E}_u\left[g_{\mu,t}(x,z)\right]$  where  $g_{\mu,t}(x,z) = \frac{\ell_t(x+\mu u,z)-\ell_t(x,z)}{\mu}u$  for  $x,z,u\in\mathbb{R}^n$  and  $\mu\in\mathbb{R}$ .
- $\ell_t(x) := \mathbb{E}_z \left[ \ell_t(x, z) \right] x, z \in \mathbb{R}^n$ .

### Assumptions

**Assumption 1.** (Unbiased Stochastic Zeroth-Order Oracle) For any  $t \geq 1$  and  $x, z \in \mathbb{R}^d$ , we have

$$\mathbb{E}_z\left[\ell_t(x,z)\right] = \ell_t(x_t). \tag{1}$$

**Assumption 2.** (Unbiased Stochastic First-Order Oracle with Bounded Variance) For any  $t \ge 1$  and  $x, z \in \mathbb{R}^d$ , we have

$$\mathbb{E}_z \left[ \nabla \ell_t(x, z) \right] = \nabla \ell_t(x) \tag{2}$$

and

$$\mathbb{E}_z \left[ \|\nabla \ell_t(x, z) - \nabla \ell_t(x)\|^2 \right] \le \sigma^2. \tag{3}$$

**Assumption 3.** (L-smoothness) Each  $\ell_t(x, z)$  is continuously differentiable and L-smooth over x on  $\mathbb{R}^d$ , that is, there exists an  $L \geq 0$  such that for all  $x, y, z \in \mathbb{R}^d$  and  $t \geq 1$ , we have

$$\|\nabla \ell_t(x,z) - \nabla \ell_t(y,z)\| \le L\|x - y\|. \tag{4}$$

We denote this by  $\ell_t(x,z) \in C_L^{1,1}(\mathbb{R}^d)$ 

**Assumption 4.** (Contractive Compression) The compression function C is a contraction mapping, that is,

$$\mathbb{E}_{\mathcal{C}}\left[\|\mathcal{C}(x) - x\|^2 \mid x\right] \le (1 - \delta) \|x\|^2 \tag{5}$$

for all  $x \in \mathbb{R}^d$  where  $0 < \delta \le 1$ , and the expectation is over the randomness generated by compression C.

**Assumption 5.** (Bounded Stochastic Gradients) For any  $t \ge 1$  and  $x, z \in \mathbb{R}^d$ , there exist  $\sigma, M > 0$  such that

$$\mathbb{E}_z \left[ \|\nabla \ell_t(x, z)\|^2 \right] \le \sigma^2 + M \|\nabla \ell_{\mu, t}(x)\|^2. \tag{6}$$

**Assumption 6.** (Bounded Drift in Time) There exists a nonnegative sequence  $\{\omega_t\}_{t=1}^T$  such that for all  $t \geq 1$ ,  $|\ell_t(x,z) - \ell_{t+1}(x,z)| \leq \omega_t$  for any  $x,z \in \mathbb{R}^d$ . Note that in the case where  $\ell_{t+1} = \ell_t$ , this assumption holds with  $\omega_t = 0$ .

#### Lemmas

Suppose  $\ell_t(x,z) \in C_L^{1,1}(\mathbb{R}^d)$  over x. We have the following results:

Lemma 1.  $\ell_{\mu,t}(x,z) \in C_{L_{\mu}}^{1,1}(\mathbb{R}^d)$  over x, where  $L_{\mu} \leq L$ .

**Lemma 2.**  $\ell_{\mu,t}(x,z)$  has the following gradient with respect to x:

$$\nabla \ell_{\mu,t}(x,z) = \frac{1}{(2\pi)^{d/2}} \int \frac{\ell_t(x+\mu u,z) - \ell_t(x,z)}{\mu} u e^{(-\frac{1}{2}||u||^2)} du, \tag{7}$$

where  $u \sim \mathcal{N}(0, I_d)$ .

**Lemma 3.** For any  $x, z \in \mathbb{R}^d$ , we have

$$|\ell_{\mu,t}(x,z) - \ell_t(x,z)| \le \frac{\mu^2 L d}{2}.$$
 (8)

**Lemma 4.** For any  $x, z \in \mathbb{R}^d$ , we have

$$\|\nabla \ell_{\mu,t}(x,z) - \nabla \ell_t(x,z)\| \le \frac{\mu}{2} L(d+3)^{\frac{3}{2}},$$
 (9)

where the gradient is with respect to x.

**Lemma 5.** For any  $x, z \in \mathbb{R}^d$ , we have

$$\mathbb{E}_{u} \left[ \left\| \frac{\ell_{t}(x + \mu u, z) - \ell_{t}(x, z)}{\mu} u \right\|^{2} \right] \leq \frac{\mu^{2}}{2} L^{2} (d + 6)^{3} + 2(d + 4) \|\nabla \ell_{t}(x, z)\|^{2}, \tag{10}$$

where  $u \sim \mathcal{N}(0, I_d)$  and the gradient is with respect to x.

**Lemma 6.** (Young's inequality) For any  $x, y \in \mathbb{R}^d$  and  $\lambda > 0$ , we have

$$\langle x, y \rangle \le \frac{\|x\|^2}{2\lambda} + \frac{\|y\|^2 \lambda}{2}.\tag{11}$$

## Convergence Analysis

In the analysis, we assume that  $z_t \in \mathbb{R}^d$  are i.i.d. random variables for all  $t \geq 1$ .

Let  $\tilde{x}_t$  be defined as follows (following the analysis in [1]):

$$\tilde{x}_t = x_t - \eta e_t. \tag{12}$$

From the algorithm, we know that  $e_{t+1} = p_t - \mathcal{C}(p_t)$  and  $p_t = g_{\mu,t}(x_t, z_t) + e_t$ , so we can rewrite  $\tilde{x}_{t+1}$  as

$$\tilde{x}_{t+1} = x_{t+1} - \eta p_t + \eta \mathcal{C}(p_t) 
= x_t - \eta \mathcal{C}(p_t) - \eta g_{\mu,t}(x_t, z_t) - \eta e_t + \eta \mathcal{C}(p_t) 
= x_t - \eta e_t - \eta g_{\mu,t}(x_t, z_t) 
= \tilde{x}_t - \eta g_{\mu,t}(x_t, z_t),$$
(13)

where  $g_{\mu,t}(x_t, z_t) := \frac{\ell_t(x_t + \mu u_t, z_t) - \ell_t(x_t, z_t)}{\mu} u_t$  and  $u_t \sim \mathcal{N}(0, I_d)$ .

By definition,  $\ell_{\mu,t}(x_t, z_t) := \mathbb{E}_{u_t} [\ell_t(x_t + \mu u_t, z_t)]$ , so by assumption 3, we can write the following:

$$\ell_{\mu,t}(\tilde{x}_{t+1}, z_{t+1}) \le \ell_{\mu,t}(\tilde{x}_t, z_t) + \langle \nabla \ell_{\mu,t}(\tilde{x}_t, z_t), \tilde{x}_{t+1} - \tilde{x}_t \rangle + \frac{L}{2} \|\tilde{x}_{t+1} - \tilde{x}_t\|^2.$$
(14)

Now by assumption 6, we get:

$$\ell_{\mu,t+1}(\tilde{x}_{t+1}, z_{t+1}) \le \ell_{\mu,t}(\tilde{x}_t, z_t) - \eta \langle g_{\mu,t}(x_t, z_t), \nabla \ell_{\mu,t}(\tilde{x}_t, z_t) \rangle + \frac{L\eta^2}{2} \|g_{\mu,t}(x_t, z_t)\|^2 + \omega_t$$
 (15)

Since  $\nabla \ell_{\mu,t}(x_t, z_t) = \mathbb{E}_{u_t}[g_{\mu,t}(x_t, z_t)]$ , we have the following:

$$\mathbb{E}_{u_{t}}\left[\langle g_{\mu,t}(x_{t}, z_{t}), \nabla \ell_{\mu,t}(\tilde{x}_{t}, z_{t})\rangle\right] = \langle \nabla \ell_{\mu,t}(x_{t}, z_{t}), \nabla \ell_{\mu,t}(\tilde{x}_{t}, z_{t})\rangle$$

$$= \frac{1}{2} \|\nabla \ell_{\mu,t}(x_{t}, z_{t})\|^{2} + \frac{1}{2} \|\nabla \ell_{\mu,t}(\tilde{x}_{t}, z_{t})\|^{2} - \frac{1}{2} \|\nabla \ell_{\mu,t}(x_{t}, z_{t}) - \nabla \ell_{\mu,t}(\tilde{x}_{t}, z_{t})\|^{2}.$$
(16)

In the last step, we use the fact that  $2\langle a,b\rangle=\|a\|^2+\|b\|^2-\|a-b\|^2$ . Plugging this into (15), we get:

$$\ell_{\mu,t+1}(\tilde{x}_{t+1}, z_{t+1}) \leq \ell_{\mu,t}(\tilde{x}_t, z_t) - \frac{\eta}{2} \|\nabla \ell_{\mu,t}(x_t, z_t)\|^2 - \frac{\eta}{2} \|\nabla \ell_{\mu,t}(\tilde{x}_t, z_t)\|^2 + \frac{L^2 \eta}{2} \|x_t - \tilde{x}_t\|^2 + \frac{L \eta^2}{2} \|g_{\mu,t}(x_t, z_t)\|^2 + \omega_t.$$
(17)

Note that  $\|\nabla \ell_{\mu,t}(x_t, z_t) - \nabla \ell_{\mu,t}(\tilde{x}_t, z_t)\|^2 \le L^2 \|x_t - \tilde{x}_t\|^2$  because of lemma 1. Also, we can drop  $-\frac{\eta}{2} \|\nabla \ell_{\mu,t}(\tilde{x}_t, z_t)\|^2$  because it is always nonpositive. Using the fact that  $\tilde{x}_t - x_t = \eta e_t$ , we get:

$$\underbrace{\frac{\eta}{2} \|\nabla \ell_{\mu,t}(x_t, z_t)\|^2}_{\text{Term III}} \leq \underbrace{\left[\ell_{\mu,t}(\tilde{x}_t, z_t) - \ell_{\mu,t+1}(\tilde{x}_{t+1}, z_{t+1})\right]}_{\text{Term II}} + \underbrace{\frac{L\eta^2}{2} \|g_{\mu,t}(x_t, z_t)\|^2}_{\text{Term IV}} + \underbrace{\frac{\eta^3 L^2}{2} \|e_t\|^2}_{\text{Term IV}} + \omega_t. \quad (18)$$

We will put an upper bound to the terms I, II, IV and a lower bound to term III. Starting with term I, by lemma 5, we know that

$$\mathbb{E}_{u_t, z_t} \left[ \|g_{\mu, t}(x_t, z_t)\|^2 \right] \le 2(d+4) \mathbb{E}_{z_t} \left[ \|\nabla \ell_t(x_t, z_t)\|^2 \right] + \frac{\mu^2 L^2}{2} (d+6)^3, \tag{19}$$

where  $\mathbb{E}_{z_t}[\|\nabla \ell_t(x_t, z_t)\|^2] \leq \|\nabla \ell_t(x_t)\|^2 + \sigma^2$  by assumption 2.

We can put the following upper bound to **term II** by means of a telescoping sum and subsequently applying lemma 3:

$$\sum_{t=1}^{T} \left[ \ell_{\mu,t}(\tilde{x}_{t}, z_{t}) - \ell_{\mu,t+1}(\tilde{x}_{t+1}, z_{t+1}) \right] = \ell_{\mu,1}(\tilde{x}_{1}, z_{1}) - \ell_{\mu,T+1}(\tilde{x}_{T+1}, z_{T+1}) 
\leq \mu L^{2} d + \ell_{1}(\tilde{x}_{1}, z_{1}) - \ell_{T+1}(\tilde{x}_{T+1}, z_{T+1}) 
= \mu L^{2} d + \ell_{1}(x_{1}, z_{1}) - \ell_{T+1}(\tilde{x}_{T+1}, z_{T+1}),$$
(20)

where we use the fact that  $\ell(x_1, z_1) = \ell_1(\tilde{x}_1, z_1)$  because  $\tilde{x}_1 = x_1$  by definition. If we take the expectation of both sides with respect to  $z_{1:T+1} = \{z_1, z_2, ..., z_{T+1}\}$ , owing to the fact that  $z_t$ 's are i.i.d., we get

$$\ell_{\mu,1}(\tilde{x}_1) - \ell_{\mu,T+1}(\tilde{x}^*) \le \mu L^2 d + \ell_1(x_1) - \ell_{T+1}(\tilde{x}_{T+1}) \le \mu L^2 d + \ell_1(x_1) - \ell_{T+1}(x_{T+1}^*),$$
(21)

where  $x_{T+1}^* = \arg\min_x \ell_{T+1}(x)$ .

We can put the following lower bound to **term III** by using lemma 4 and lemma 6:

$$\frac{1}{2} \|\nabla \ell_t(x_t, z_t)\|^2 - \frac{\mu^2 L^2}{4} (d+3)^3 \le \|\nabla \ell_{\mu, t}(x_t, z_t)\|^2.$$
 (22)

Lastly, we can put the following upper bound to **term IV** by assumption 4 and lemma 6:

$$\mathbb{E}_{u_{1:T},z_{1:T},\mathcal{C}_{1:T}}[\|e_{t+1}\|^{2}] = \|p_{t} - \mathcal{C}_{t}(p_{t})\|^{2} \leq (1-\delta)\|p_{t}\|^{2} = (1-\delta)\|e_{t} + g_{\mu,t}(x_{t},z_{t})\|^{2} \\
\leq (1-\delta)(1+\varphi)\mathbb{E}_{u_{1:T},z_{1:T},\mathcal{C}_{1:T}}\left[\|e_{t}\|^{2}\right] + (1-\delta)(1+\frac{1}{\varphi})\mathbb{E}_{u_{1:T},z_{1:T}}\left[\|g_{\mu,t}(x_{t},z_{t})\|^{2}\right] \\
= \sum_{i=1}^{t} \left[ (1-\delta)(1+\varphi)\right]^{t-i} (1-\delta)(1+\frac{1}{\varphi})\mathbb{E}_{u_{i},z_{i}}\left[\|g_{\mu,i}(x_{i},z_{i})\|^{2}\right], \tag{23}$$

for some  $\varphi > 0$ ,  $z_t, x_t, \mathcal{C}_t$  are *i.i.d.*, and  $\mathbb{E}_{\mathcal{C}_t}[\cdot]$  denotes the expectation over the randomness at time t due to the compression used. Note that by assumption 5 and using lemma 5,

$$\mathbb{E}_{u_t, z_t}[\|g_{\mu, t}(x_t, z_t)\|^2] \le A\|\nabla \ell_t(x_t)\|^2 + B, \tag{24}$$

where

$$B = 2\sigma^2(d+4) + \frac{\mu^2 L^2}{2}(d+6)^3$$
 and  $A = 2M(d+4)$ . (25)

So we can rewrite (23) as follows:

$$\mathbb{E}_{u_{1:T}, z_{1:T}, \mathcal{C}_{1:T}} \left[ \|e_{t+1}\|^2 \right] \le \sum_{i=1}^t \left[ (1-\delta)(1+\varphi) \right]^{t-i} (1-\delta)(1+\frac{1}{\varphi}) \left[ A \|\nabla \ell_i(x_i)\|^2 + B \right]. \tag{26}$$

If we set  $\varphi := \frac{\delta}{2(1-\delta)}$ , then  $1 + \frac{1}{\varphi} \leq \frac{2}{\delta}$  and  $(1-\delta)(1+\varphi) = (1-\frac{\delta}{2})$ , so we get:

$$\mathbb{E}_{u_{1:T}, z_{1:T}, \mathcal{C}_{1:T}} \left[ \|e_{t+1}\|^2 \right] \le \sum_{i=1}^t \left( 1 - \frac{\delta}{2} \right)^{t-i} \left[ A \|\nabla \ell_i(x_i)\|^2 + B \right] \frac{2(1-\delta)}{\delta}. \tag{27}$$

If we sum through all  $\mathbb{E}_{u_{1:T},z_{1:T},\mathcal{C}_{1:T}}[\|e_t\|^2]$ , we get:

$$\sum_{t=1}^{T} \mathbb{E}_{u_{1:T}, z_{1:T}, \mathcal{C}_{1:T}} \left[ \|e_{t}\|^{2} \right] \leq \sum_{t=1}^{T} \sum_{i=1}^{t-1} \left( 1 - \frac{\delta}{2} \right)^{t-i} \left[ A \|\nabla \ell_{i}(x_{i})\|^{2} + B \right] \frac{2(1-\delta)}{\delta} \\
\leq \sum_{t=1}^{T} \left[ A \|\nabla \ell_{t}(x_{t})\|^{2} + B \right] \sum_{i=0}^{\infty} \left( 1 - \frac{\delta}{2} \right)^{i} \frac{2(1-\delta)}{\delta} \\
\leq \sum_{t=1}^{T} \left[ A \|\nabla \ell_{t}(x_{t})\|^{2} + B \right] C, \tag{28}$$

where  $C = \frac{2(1-\delta)}{\delta} \frac{2}{\delta} \leq \frac{4}{\delta^2}$ . If we combine the upper bounds derived in (19), (20), (23), and the lower bound derived in (22) and plug them in (18), we get the following:

$$\sum_{t=1}^{T} \frac{\eta}{4} \mathbb{E}_{z_{t}} \left[ \|\nabla \ell_{t}(x_{t}, z_{t})\|^{2} \right] - \frac{\eta \mu^{2} L^{2}}{8} (d+3)^{3} T$$

$$\leq \mu L^{2} d + \Delta + \frac{T \mu^{2} L^{3} \eta^{2}}{4} (d+6)^{3} + \frac{L \eta^{2}}{2} \sigma^{2} T \times 2(d+4)$$

$$+ \frac{L \eta^{2}}{2} \times 2M(d+4) \sum_{t=1}^{T} \mathbb{E}_{z_{t}} \left[ \|\nabla \ell_{t}(x_{t}, z_{t})\|^{2} \right] + \frac{\eta^{3} L^{2}}{2} \times \frac{4}{\delta^{2}} T \left[ 2\sigma^{2} (d+4) + \frac{\mu^{2} L^{2}}{2} (d+6)^{3} \right]$$

$$+ \frac{\eta^{3} L^{2}}{2} \times \frac{4}{\delta^{2}} \sum_{t=1}^{T} 2M(d+4) \mathbb{E}_{z_{t}} \left[ \|\nabla \ell_{t}(x_{t}, z_{t})\|^{2} \right] + \sum_{t=1}^{T} \omega_{t},$$
(29)

where  $\Delta = \ell_1(x_1) - \ell_{T+1}(x_{T+1}^*)$ .

Now, since  $z_t$ 's are *i.i.d.* for all t, we have:

$$\frac{E}{T} \sum_{t=1}^{T} \mathbb{E}_{z_{t}} \left[ \|\nabla \ell_{t}(x_{t}, z_{t})\|^{2} \right] \leq \frac{\mu L^{2} d + \left[\ell_{1}(x_{1}) - \ell_{T+1}(x_{T+1}^{*})\right]}{T} + \frac{\eta^{2} L^{3} \mu^{2} (d+6)^{3}}{4} + L \eta^{2} \sigma^{2} (d+4) + \frac{\eta \mu^{2} L^{2} (d+3)^{3}}{8} + \frac{\eta^{3} L^{2}}{\delta^{2}} 4 \sigma^{2} (d+4) + \frac{\eta^{3} L^{2}}{\delta^{2}} \mu^{2} L^{2} (d+6)^{3} + \frac{1}{T} \sum_{t=1}^{T} \omega_{t}, \tag{30}$$

where

$$E = \frac{\eta}{4} - LM\eta^{2}(d+4) - \frac{L^{2}\eta^{3}}{\delta^{2}} 4M(d+4)$$

$$= \eta \left[ \frac{1}{4} - LM\eta(d+4) \left( 1 + \frac{4L\eta}{\delta^{2}} \right) \right].$$
(31)

If  $\eta \leq \frac{1}{4L}$ , instead first upper bound will be:

$$1 + \frac{4L\eta}{\delta^2} \le 1 + \frac{1}{\delta^2} = \frac{\delta^2 + 1}{\delta^2} \le \frac{2}{\delta^2}.$$
 (32)

Find  $\eta$  such that

$$LM\eta(d+4) \times \frac{2}{\delta^2} \le \frac{1}{8} \tag{33}$$

Then, we get

$$\eta \le \frac{\delta^2}{16LM(d+4)} \tag{34}$$

which implies  $E \geq \frac{\eta}{8}$ . Multiply all terms in the bound by

$$\frac{1}{T} \sum_{t=1}^{T} \mathbb{E}_{z_{t}} \left[ \|\nabla \ell_{t}(x_{t}, z_{t})\|^{2} \right] \leq \frac{8(\ell_{1} - \ell^{*})}{(\eta T)} + \frac{8\mu L^{2}d}{\eta T} + 2\eta L^{3}\mu^{2}(d+6)^{3} 
+ 8L\eta\sigma^{2}(d+4) + \mu^{2}L^{2}(d+3)^{3} 
+ \frac{32\eta^{2}L^{2}}{\delta^{2}}\sigma^{2}(d+4) + \frac{8\eta^{2}L^{4}\mu^{2}(d+6)^{3}}{\delta^{2}} + \frac{8}{\eta T} \sum_{t=1}^{T} \omega_{t}$$
(35)

Let

$$\eta = \frac{1}{\sigma\sqrt{(d+4)MTL}} \quad \text{and} \quad \mu = \frac{1}{(d+4)\sqrt{T}}$$
(36)

Then, we have

$$\frac{1}{CT} \sum_{t=1}^{T} \mathbb{E}_{z_t} \left[ \|\nabla \ell_t(x_t, z_t)\|^2 \right] \le \frac{1}{\delta^2} \frac{dL\Delta}{T} + \sigma \sqrt{\frac{d}{T} L \Delta M} + \frac{1}{T} \sum_{t=1}^{T} \omega_t$$
 (37)

for a numerical constant C > 0, where  $\Delta = \ell_1(x_1) - \ell_{T+1}(x_{T+1}^*)$  for  $x_{T+1}^* = \arg\min_{x \in \mathbb{R}^d} \ell_{T+1}(x)$ . If  $\omega_t \neq 0$  for  $t \in \mathbb{Z}_+$ , then using (29), the number of times steps T to obtain a  $\xi$ -first order solution is

$$T = \mathcal{O}\left(\frac{d\sigma^2 L \Delta M}{\xi^2} + \frac{dL\Delta}{\delta^2 \xi} + \frac{\bar{\omega}\sigma^2 (d+4)ML}{\xi^2}\right)$$
(38)

where  $\bar{\omega} = \sum_{t=1}^{T} \omega_t$  and it is constant.

**Remark:** Choosing  $\eta = \frac{1}{\sigma\sqrt{(d+4)MTL}}$ , we assumed that it satisfies (34). For this to hold, T can be increased arbitrarily but it should not exceed the bound we found in (38). (36) and (34) imply that T should be

$$T = \mathcal{O}\left(\frac{(d+4)LM}{\delta^4 \sigma^2}\right) \tag{39}$$

Since  $\xi \ll \delta$ , this term is smaller than (37). For instance, consider a very bad compression scheme with  $\delta = 0.1$  and we want to find the solution with  $10^{-3}$  precision. Even in that scenario, (38) is greater than (39). In fact, this is demonstrated by our experiments.

Lastly, if  $\omega_t = 0$ , i.e., in the case where  $\ell_{t+1} = \ell_t$ , the number of time steps T to should be

$$T = \mathcal{O}\left(\frac{d\sigma^2 L \Delta M}{\xi^2} + \frac{dL\Delta}{\delta^2 \xi}\right). \tag{40}$$

#### References

[1] S. P. Karimireddy, Q. Rebjock, S. U. Stich, and M. Jaggi, "Error feedback fixes signsgd and other gradient compression schemes," 2019.