

# Decentralized Online Optimization

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## Abstract

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## 1 Notations and assumptions

Define the dynamic regret as

$$\mathcal{R}_T^{DOG} = \sum_{i=1}^n \sum_{t=1}^T \mathbb{E}_{\xi} f_{i,t}(\mathbf{x}_{i,t}; \xi) - \mathbb{E}_{\xi} f_{i,t}(\mathbf{x}_t^*; \xi),$$

where, for any  $\mathbf{x}$ ,

$$f_{i,t}(\mathbf{x}; \xi) := \beta g_{i,t}(\mathbf{x}) + (1 - \beta)f(\mathbf{x}; \xi)$$

with  $0 < \beta < 1$ , and  $\xi$  is a random variable drawn from an unknown distribution.  $g_{i,t}$  is an adversary loss function, and  $f$  is a given loss function.

The budget of the dynamics is defined as

$$\sum_{t=1}^T \|\mathbf{x}_{t+1}^* - \mathbf{x}_t^*\| \leq D.$$

**Assumption 1.** *We make the following assumptions.*

- For any  $i \in [n]$  and  $t \in [T]$ , we assume  $\max \left\{ \|\nabla g_{i,t}(\mathbf{x})\|^2, \|\nabla f(\mathbf{x}; \xi)\|^2, \|\nabla f_{i,t}(\mathbf{x}; \xi)\|^2 \right\} \leq G$ , and  $\|\nabla f(\mathbf{x}; \xi) - \nabla f(\mathbf{x})\|^2 \leq \sigma^2$ .
- For any  $\mathbf{x}$  and  $\mathbf{y}$ , we assume  $\|\mathbf{x} - \mathbf{y}\|^2 \leq R$ .
- For any  $i \in [n]$  and  $t \in [T]$ , we assume the function  $f_{i,t}(\mathbf{x}; \xi)$  is convex and differentiable.

## 2 Algorithm

**Theorem 1.** *Using Assumptions 1, and choosing  $\eta > 0$  in Algorithm 1, we have*

$$\begin{aligned} \mathcal{R}_T &\leq \sum_{t=1}^T \sum_{i=1}^n \mathbb{E}_{\xi} f_{i,t}(\mathbf{x}_{i,t}; \xi) - \mathbb{E}_{\xi} f_{i,t}(\mathbf{x}_t^*; \xi) \\ &\leq \left( 1 + 4\beta \left( \frac{1}{\sqrt{\beta^2 + \frac{1}{n}}} + 4 \right) \right) nT\beta G\eta + \left( L + \frac{2\eta L^2(1-\beta)^2}{\sqrt{\beta^2 + \frac{1}{n}}} + \frac{\sqrt{\beta^2 + \frac{1}{n}}}{2\eta} + 10\eta L^2 + \frac{\beta}{2\eta} \right) \frac{nT\eta^2 G}{(1-\rho)^2} \\ &\quad + n \left( \frac{1}{\sqrt{\beta^2 + \frac{1}{n}}} + 4 \right) (f(\bar{\mathbf{x}}_1) - f(\bar{\mathbf{x}}_{T+1})) + T\eta\sigma^2(1-\beta) \left( \frac{2}{\sqrt{\beta^2 + \frac{1}{n}}} + 9 \right) + \frac{n}{2\eta} (4\sqrt{RD} + R). \end{aligned}$$

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**Algorithm 1** DOG: Decentralized Online Gradient.

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**Require:** The learning rate  $\eta$ , number of iterations  $T$ , and the confusion matrix  $\mathbf{W}$ .

- 1: **for**  $t = 1, 2, \dots, T$  **do**  
    For the  $i$ -th node with  $i \in [n]$ :
    - 2:     Predict  $\mathbf{x}_{i,t}$ .
    - 3:     Observe the loss function  $f_{i,t}$ ,  
       and suffer loss  $f_{i,t}(\mathbf{x}_{i,t}; \xi)$ .
  - Update:
    - 4:     Query the gradient  $\nabla f_{i,t}(\mathbf{x}_{i,t}; \xi)$ .
    - 5:      $\mathbf{x}_{i,t+1} = \sum_{j=1}^n \mathbf{W}_{i,j} \mathbf{x}_{j,t} - \eta \nabla f_{i,t}(\mathbf{x}_{i,t}; \xi)$ .
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**Corollary 1.** Using Assumptions [1](#), and choosing

$$\eta = \sqrt{\frac{nD}{T \left( n\beta + n\sqrt{\frac{n\beta^2+1}{n}} + 2\sqrt{\frac{n}{n\beta^2+1}} + 9 \right)}}$$

in Algorithm [1](#), we have

$$\mathcal{R}_T^{DOG} \leq \mathcal{O} \left( \sqrt{n^2 \beta D T + n \sqrt{n} D T} \right).$$

Furthermore, when

$$\beta = n^{-a}, \text{ and } a \geq 0,$$

we have

$$\mathcal{R}_T^{DOG} \leq \mathcal{O} \left( n^{\max\{\frac{3}{4}, \frac{2-a}{2}\}} \sqrt{DT} \right).$$

## Appendix

### Proof details of Theorem [1](#):

*Proof.*

$$\begin{aligned} & \mathbb{E}_{\xi} \frac{1}{n} \sum_{i=1}^n f_{i,t}(\mathbf{x}_{i,t}; \xi) - f_t(\mathbf{x}_t^*; \xi) \\ &= \frac{1}{n} \sum_{i=1}^n \beta (g_{i,t}(\mathbf{x}_{i,t}) - g_{i,t}(\mathbf{x}_t^*)) + (1 - \beta) \mathbb{E}_{\xi} (f(\mathbf{x}_{i,t}; \xi) - f(\mathbf{x}_t^*; \xi)) \\ &\leq \frac{1}{n} \sum_{i=1}^n \beta \langle \nabla g_{i,t}(\mathbf{x}_{i,t}), \mathbf{x}_{i,t} - \mathbf{x}_t^* \rangle + (1 - \beta) \mathbb{E}_{\xi} \langle \nabla f(\mathbf{x}_{i,t}; \xi), \mathbf{x}_{i,t} - \mathbf{x}_t^* \rangle \\ &= \frac{1}{n} \sum_{i=1}^n \beta (\langle \nabla g_{i,t}(\mathbf{x}_{i,t}), \mathbf{x}_{i,t} - \bar{\mathbf{x}}_t \rangle + \langle \nabla g_{i,t}(\mathbf{x}_{i,t}), \bar{\mathbf{x}}_t - \bar{\mathbf{x}}_{t+1} \rangle + \langle \nabla g_{i,t}(\mathbf{x}_{i,t}), \bar{\mathbf{x}}_{t+1} - \mathbf{x}_t^* \rangle) \\ &\quad + \frac{1}{n} \sum_{i=1}^n (1 - \beta) \left( \mathbb{E}_{\xi} \langle \nabla f(\mathbf{x}_{i,t}; \xi), \mathbf{x}_{i,t} - \bar{\mathbf{x}}_t \rangle + \mathbb{E}_{\xi} \langle \nabla f(\mathbf{x}_{i,t}; \xi), \bar{\mathbf{x}}_t - \bar{\mathbf{x}}_{t+1} \rangle + \mathbb{E}_{\xi} \langle \nabla f(\mathbf{x}_{i,t}; \xi), \bar{\mathbf{x}}_{t+1} - \mathbf{x}_t^* \rangle \right) \end{aligned}$$

$$\begin{aligned}
&= \underbrace{\frac{1}{n} \sum_{i=1}^n \beta (\langle \nabla g_{i,t}(\mathbf{x}_{i,t}), \mathbf{x}_{i,t} - \bar{\mathbf{x}}_t \rangle + \langle \nabla g_{i,t}(\mathbf{x}_{i,t}), \bar{\mathbf{x}}_t - \bar{\mathbf{x}}_{t+1} \rangle)}_{I_1(t)} \\
&\quad + \underbrace{\frac{1}{n} \sum_{i=1}^n (1-\beta) \left( \langle \nabla f(\mathbf{x}_{i,t}), \mathbf{x}_{i,t} - \bar{\mathbf{x}}_t \rangle + \mathbb{E}_{\xi} \langle \nabla f(\mathbf{x}_{i,t}; \xi), \bar{\mathbf{x}}_t - \bar{\mathbf{x}}_{t+1} \rangle \right)}_{I_2(t)} \\
&\quad + \underbrace{\mathbb{E}_{\xi} \left\langle \frac{1}{n} \sum_{i=1}^n \nabla f_{i,t}(\mathbf{x}_{i,t}; \xi), \bar{\mathbf{x}}_{t+1} - \mathbf{x}_t^* \right\rangle}_{I_3(t)}
\end{aligned}$$

Now, we begin to bound  $I_1(t)$ .

$$\begin{aligned}
I_1(t) &\leq \frac{\beta}{n} \sum_{i=1}^n \left( \frac{\eta}{2} \|\nabla g_{i,t}(\mathbf{x}_{i,t})\|^2 + \frac{1}{2\eta} \|\mathbf{x}_{i,t} - \bar{\mathbf{x}}_t\|^2 + \frac{\eta}{2} \|\nabla f_{i,t}(\mathbf{x}_{i,t})\|^2 + \frac{1}{2\eta} \|\bar{\mathbf{x}}_t - \bar{\mathbf{x}}_{t+1}\|^2 \right) \\
&\leq \beta G\eta + \frac{\beta}{2n\eta} \sum_{i=1}^n \|\mathbf{x}_{i,t} - \bar{\mathbf{x}}_t\|^2 + \frac{\beta}{2\eta} \|\bar{\mathbf{x}}_t - \bar{\mathbf{x}}_{t+1}\|^2.
\end{aligned}$$

Now, we begin to bound  $I_2(t)$ .

$$I_2(t) = (1-\beta) \left( \underbrace{\frac{1}{n} \sum_{i=1}^n \langle \nabla f(\mathbf{x}_{i,t}), \mathbf{x}_{i,t} - \bar{\mathbf{x}}_t \rangle}_{J_1(t)} + \underbrace{\mathbb{E}_{\xi} \left\langle \frac{1}{n} \sum_{i=1}^n \nabla f(\mathbf{x}_{i,t}; \xi), \bar{\mathbf{x}}_t - \bar{\mathbf{x}}_{t+1} \right\rangle}_{J_2(t)} \right).$$

For  $J_1(t)$  and  $\nu > 0$ , we have

$$\begin{aligned}
J_1(t) &= \frac{1}{n} \sum_{i=1}^n \langle \nabla f(\mathbf{x}_{i,t}), \mathbf{x}_{i,t} - \bar{\mathbf{x}}_t \rangle \\
&= \frac{1}{n} \sum_{i=1}^n \langle \nabla f(\mathbf{x}_{i,t}) - \nabla f(\bar{\mathbf{x}}_t), \mathbf{x}_{i,t} - \bar{\mathbf{x}}_t \rangle + \frac{1}{n} \sum_{i=1}^n \langle \nabla f(\bar{\mathbf{x}}_t), \mathbf{x}_{i,t} - \bar{\mathbf{x}}_t \rangle \\
&\leq \frac{L}{n} \sum_{i=1}^n \|\mathbf{x}_{i,t} - \bar{\mathbf{x}}_t\|^2 + \frac{1}{n} \sum_{i=1}^n \langle \nabla f(\bar{\mathbf{x}}_t), \mathbf{x}_{i,t} - \bar{\mathbf{x}}_t \rangle \\
&\leq \frac{L}{n} \sum_{i=1}^n \|\mathbf{x}_{i,t} - \bar{\mathbf{x}}_t\|^2 + \frac{1}{n} \sum_{i=1}^n \left( \frac{\eta}{2\nu} \|\nabla f(\bar{\mathbf{x}}_t)\|^2 + \frac{\nu}{2\eta} \|\mathbf{x}_{i,t} - \bar{\mathbf{x}}_t\|^2 \right) \\
&\leq \frac{L}{n} \sum_{i=1}^n \|\mathbf{x}_{i,t} - \bar{\mathbf{x}}_t\|^2 + \frac{\eta}{2\nu} \|\nabla f(\bar{\mathbf{x}}_t)\|^2 + \frac{\nu}{2\eta n} \sum_{i=1}^n \|\mathbf{x}_{i,t} - \bar{\mathbf{x}}_t\|^2. \tag{1}
\end{aligned}$$

According to Lemma 1, we have

$$\begin{aligned}
&\frac{\eta}{2} \|\nabla f(\bar{\mathbf{x}}_t)\|^2 \\
&\leq \frac{\eta}{2} \|\nabla f(\bar{\mathbf{x}}_t)\|^2 + \left( \frac{\eta}{2} - \frac{L\eta^2}{2} \right) \left\| \frac{1}{n} \sum_{i=1}^n f_{i,t}(\mathbf{x}_{i,t}; \xi) \right\|^2
\end{aligned}$$

$$\leq f(\bar{\mathbf{x}}_t) - f(\bar{\mathbf{x}}_{t+1}) + 4G\eta\beta^2 + \frac{2\eta L^2(1-\beta)^2}{n} \sum_{i=1}^n \|\bar{\mathbf{x}}_t - \mathbf{x}_{i,t}\|^2 + \frac{2\eta\sigma^2(1-\beta)^2}{n}. \quad (2)$$

Substituting (2) into (1), we obtain

$$\begin{aligned} & J_1(t) \\ & \leq \frac{L}{n} \sum_{i=1}^n \|\mathbf{x}_{i,t} - \bar{\mathbf{x}}_t\|^2 + \left( \frac{1}{\nu} (f(\bar{\mathbf{x}}_t) - f(\bar{\mathbf{x}}_{t+1})) + \frac{4G\eta\beta^2}{\nu} + \frac{2\eta L^2(1-\beta)^2}{n\nu} \sum_{i=1}^n \|\bar{\mathbf{x}}_t - \mathbf{x}_{i,t}\|^2 + \frac{2\eta\sigma^2(1-\beta)^2}{n\nu} \right) \\ & \quad + \frac{\nu}{2n\eta} \sum_{i=1}^n \|\mathbf{x}_{i,t} - \bar{\mathbf{x}}_t\|^2 \\ & = \left( \frac{L}{n} + \frac{2\eta L^2(1-\beta)^2}{n\nu} + \frac{\nu}{2n\eta} \right) \sum_{i=1}^n \|\mathbf{x}_{i,t} - \bar{\mathbf{x}}_t\|^2 + \frac{1}{\nu} (f(\bar{\mathbf{x}}_t) - f(\bar{\mathbf{x}}_{t+1})) + \frac{4G\eta\beta^2}{\nu} + \frac{2\eta\sigma^2(1-\beta)^2}{n\nu}. \end{aligned}$$

For  $J_2(t)$ , we have

$$\begin{aligned} J_2(t) &= \mathbb{E}_{\xi} \left\langle \frac{1}{n} \sum_{i=1}^n \nabla f(\mathbf{x}_{i,t}; \xi), \bar{\mathbf{x}}_t - \bar{\mathbf{x}}_{t+1} \right\rangle \\ &\leq \left( \frac{\eta}{2} \mathbb{E}_{\xi} \left\| \frac{1}{n} \sum_{i=1}^n \nabla f(\mathbf{x}_{i,t}; \xi) \right\|^2 + \frac{1}{2\eta} \|\bar{\mathbf{x}}_t - \bar{\mathbf{x}}_{t+1}\|^2 \right) \\ &\leq \left( \frac{\eta}{2} \mathbb{E}_{\xi} \left\| \frac{1}{n} \sum_{i=1}^n \nabla f(\mathbf{x}_{i,t}; \xi) - \nabla f(\mathbf{x}_{i,t}) + \nabla f(\mathbf{x}_{i,t}) \right\|^2 + \frac{1}{2\eta} \|\bar{\mathbf{x}}_t - \bar{\mathbf{x}}_{t+1}\|^2 \right) \\ &\leq \left( \eta \mathbb{E}_{\xi} \left\| \frac{1}{n} \sum_{i=1}^n \nabla f(\mathbf{x}_{i,t}; \xi) - \nabla f(\mathbf{x}_{i,t}) \right\|^2 + \eta \left\| \frac{1}{n} \sum_{i=1}^n \nabla f(\mathbf{x}_{i,t}) \right\|^2 + \frac{1}{2\eta} \|\bar{\mathbf{x}}_t - \bar{\mathbf{x}}_{t+1}\|^2 \right) \\ &\leq \frac{\eta}{n} \sigma^2 + \eta \left\| \frac{1}{n} \sum_{i=1}^n \nabla f(\mathbf{x}_{i,t}) - \nabla f(\bar{\mathbf{x}}_t) + \nabla f(\bar{\mathbf{x}}_t) \right\|^2 + \frac{1}{2\eta} \|\bar{\mathbf{x}}_t - \bar{\mathbf{x}}_{t+1}\|^2 \\ &\leq \frac{\eta}{n} \sigma^2 + 2\eta \left\| \frac{1}{n} \sum_{i=1}^n \nabla f(\mathbf{x}_{i,t}) - \nabla f(\bar{\mathbf{x}}_t) \right\|^2 + 2\eta \|\nabla f(\bar{\mathbf{x}}_t)\|^2 + \frac{1}{2\eta} \|\bar{\mathbf{x}}_t - \bar{\mathbf{x}}_{t+1}\|^2 \\ &\leq \frac{\eta}{n} \sigma^2 + \frac{2\eta L^2}{n} \sum_{i=1}^n \|\mathbf{x}_{i,t} - \bar{\mathbf{x}}_t\|^2 + 2\eta \|\nabla f(\bar{\mathbf{x}}_t)\|^2 + \frac{1}{2\eta} \|\bar{\mathbf{x}}_t - \bar{\mathbf{x}}_{t+1}\|^2 \end{aligned}$$

Recall Lemma 1, and we have

$$\begin{aligned} & J_2(t) \\ & \leq \frac{\eta}{n} \sigma^2 + \frac{2\eta L^2}{n} \sum_{i=1}^n \|\mathbf{x}_{i,t} - \bar{\mathbf{x}}_t\|^2 + 4(f(\bar{\mathbf{x}}_t) - f(\bar{\mathbf{x}}_{t+1})) + 16G\eta\beta^2 + \frac{8\eta L^2(1-\beta)^2}{n} \sum_{i=1}^n \|\mathbf{x}_{i,t} - \bar{\mathbf{x}}_t\|^2 \\ & \quad + \frac{8\eta\sigma^2(1-\beta)^2}{n} + \frac{1}{2\eta} \|\bar{\mathbf{x}}_t - \bar{\mathbf{x}}_{t+1}\|^2 \\ & \leq \frac{\eta\sigma^2(1-8(1-\beta)^2)}{n} \sigma^2 + (2\eta L^2 + 8\eta L^2(1-\beta)^2) \frac{1}{n} \sum_{i=1}^n \|\mathbf{x}_{i,t} - \bar{\mathbf{x}}_t\|^2 + 4(f(\bar{\mathbf{x}}_t) - f(\bar{\mathbf{x}}_{t+1})) + 16G\eta\beta^2 + \frac{1}{2\eta} \|\bar{\mathbf{x}}_t - \bar{\mathbf{x}}_{t+1}\|^2 \end{aligned}$$

$$\leq \frac{9\eta\sigma^2}{n}\sigma^2 + (2\eta L^2 + 8\eta L^2(1-\beta)^2) \frac{1}{n} \sum_{i=1}^n \|\mathbf{x}_{i,t} - \bar{\mathbf{x}}_t\|^2 + 4(f(\bar{\mathbf{x}}_t) - f(\bar{\mathbf{x}}_{t+1})) + 16G\eta\beta^2 + \frac{1}{2\eta} \|\bar{\mathbf{x}}_t - \bar{\mathbf{x}}_{t+1}\|^2$$

Therefore, we obtain

$$\begin{aligned} & I_2(t) \\ &= (1-\beta)(J_1(t) + J_2(t)) \\ &\leq (1-\beta) \left( \left( \frac{L}{n} + \frac{2\eta L^2(1-\beta)^2}{n\nu} + \frac{\nu}{2n\eta} + \frac{\eta L^2(2+8(1-\beta)^2)}{n} \right) \sum_{i=1}^n \|\mathbf{x}_{i,t} - \bar{\mathbf{x}}_t\|^2 + \left( \frac{1}{\nu} + 4 \right) (f(\bar{\mathbf{x}}_t) - f(\bar{\mathbf{x}}_{t+1})) \right) \\ &\quad + (1-\beta) \left( 4G\eta\beta^2 \left( \frac{1}{\nu} + 4 \right) + \frac{1}{2\eta} \|\bar{\mathbf{x}}_t - \bar{\mathbf{x}}_{t+1}\|^2 \right) + \frac{2\eta\sigma^2(1-\beta)^3}{n\nu} + \frac{9\eta\sigma^2(1-\beta)}{n} \\ &\leq (1-\beta) \left( \left( \frac{L}{n} + \frac{2\eta L^2(1-\beta)^2}{n\nu} + \frac{\nu}{2n\eta} + \frac{10\eta L^2}{n} \right) \sum_{i=1}^n \|\mathbf{x}_{i,t} - \bar{\mathbf{x}}_t\|^2 \right) + \left( \frac{1}{\nu} + 4 \right) (f(\bar{\mathbf{x}}_t) - f(\bar{\mathbf{x}}_{t+1})) \\ &\quad + (1-\beta) \left( 4G\eta\beta^2 \left( \frac{1}{\nu} + 4 \right) + \frac{1}{2\eta} \|\bar{\mathbf{x}}_t - \bar{\mathbf{x}}_{t+1}\|^2 \right) + \frac{2\eta\sigma^2(1-\beta)^3}{n\nu} + \frac{9\eta\sigma^2(1-\beta)}{n} \end{aligned}$$

Combine those bounds of  $I_1(t)$  and  $I_2(t)$ . We thus have

$$\begin{aligned} & I_1(t) + I_2(t) \\ &\leq \beta G\eta + \frac{\beta}{2n\eta} \sum_{i=1}^n \|\mathbf{x}_{i,t} - \bar{\mathbf{x}}_t\|^2 + \frac{\beta}{2\eta} \|\bar{\mathbf{x}}_t - \bar{\mathbf{x}}_{t+1}\|^2 \\ &\quad + (1-\beta) \left( \left( \frac{L}{n} + \frac{2\eta L^2(1-\beta)^2}{n\nu} + \frac{\nu}{2n\eta} + \frac{10\eta L^2}{n} \right) \sum_{i=1}^n \|\mathbf{x}_{i,t} - \bar{\mathbf{x}}_t\|^2 \right) + \left( \frac{1}{\nu} + 4 \right) (f(\bar{\mathbf{x}}_t) - f(\bar{\mathbf{x}}_{t+1})) \\ &\quad + (1-\beta) \left( 4G\eta\beta^2 \left( \frac{1}{\nu} + 4 \right) + \frac{1}{2\eta} \|\bar{\mathbf{x}}_t - \bar{\mathbf{x}}_{t+1}\|^2 \right) + \frac{2\eta\sigma^2(1-\beta)^3}{n\nu} + \frac{\eta\sigma^2(1-\beta)}{n} \\ &= \left( 1 + 4\beta(1-\beta) \left( \frac{1}{\nu} + 4 \right) \right) \beta G\eta + \left( (1-\beta) \left( \frac{L}{n} + \frac{2\eta L^2(1-\beta)^2}{n\nu} + \frac{\nu}{2n\eta} + \frac{10\eta L^2}{n} \right) + \frac{\beta}{2n\eta} \right) \sum_{i=1}^n \|\mathbf{x}_{i,t} - \bar{\mathbf{x}}_t\|^2 \\ &\quad + \left( \frac{1}{\nu} + 4 \right) (f(\bar{\mathbf{x}}_t) - f(\bar{\mathbf{x}}_{t+1})) + \frac{1}{2\eta} \|\bar{\mathbf{x}}_t - \bar{\mathbf{x}}_{t+1}\|^2 + \frac{\eta\sigma^2(1-\beta)}{n} \left( \frac{2(1-\beta)^2}{\nu} + 9 \right) \\ &\leq \left( 1 + 4\beta \left( \frac{1}{\nu} + 4 \right) \right) \beta G\eta + \left( \frac{L}{n} + \frac{2\eta L^2(1-\beta)^2}{n\nu} + \frac{\nu}{2n\eta} + \frac{10\eta L^2}{n} + \frac{\beta}{2n\eta} \right) \sum_{i=1}^n \|\mathbf{x}_{i,t} - \bar{\mathbf{x}}_t\|^2 \\ &\quad + \left( \frac{1}{\nu} + 4 \right) (f(\bar{\mathbf{x}}_t) - f(\bar{\mathbf{x}}_{t+1})) + \frac{1}{2\eta} \|\bar{\mathbf{x}}_t - \bar{\mathbf{x}}_{t+1}\|^2 + \frac{\eta\sigma^2(1-\beta)}{n} \left( \frac{2}{\nu} + 9 \right). \end{aligned}$$

According to Lemma 2, we have

$$\frac{1}{n} \sum_{i=1}^n \sum_{t=1}^T \|\mathbf{x}_{i,t} - \bar{\mathbf{x}}_t\|^2 \leq \frac{T\eta^2 G}{(1-\rho)^2}.$$

Therefore,

$$\begin{aligned} & \sum_{t=1}^T (I_1(t) + I_2(t)) \\ &\leq \left( 1 + 4\beta \left( \frac{1}{\nu} + 4 \right) \right) T\beta G\eta + \left( \frac{L}{n} + \frac{2\eta L^2(1-\beta)^2}{n\nu} + \frac{\nu}{2n\eta} + \frac{10\eta L^2}{n} + \frac{\beta}{2n\eta} \right) \sum_{t=1}^T \sum_{i=1}^n \|\mathbf{x}_{i,t} - \bar{\mathbf{x}}_t\|^2 \end{aligned}$$

$$\begin{aligned}
& + \left( \frac{1}{\nu} + 4 \right) (f(\bar{\mathbf{x}}_1) - f(\bar{\mathbf{x}}_{T+1})) + \frac{1}{2\eta} \sum_{t=1}^T \|\bar{\mathbf{x}}_t - \bar{\mathbf{x}}_{t+1}\|^2 + \frac{T\eta\sigma^2(1-\beta)}{n} \left( \frac{2}{\nu} + 9 \right) \\
& \leq \left( 1 + 4\beta \left( \frac{1}{\nu} + 4 \right) \right) T\beta G\eta + \left( L + \frac{2\eta L^2(1-\beta)^2}{\nu} + \frac{\nu}{2\eta} + 10\eta L^2 + \frac{\beta}{2\eta} \right) \frac{T\eta^2 G}{(1-\rho)^2} \\
& + \left( \frac{1}{\nu} + 4 \right) (f(\bar{\mathbf{x}}_1) - f(\bar{\mathbf{x}}_{T+1})) + \frac{1}{2\eta} \sum_{t=1}^T \|\bar{\mathbf{x}}_t - \bar{\mathbf{x}}_{t+1}\|^2 + \frac{T\eta\sigma^2(1-\beta)}{n} \left( \frac{2}{\nu} + 9 \right).
\end{aligned}$$

Now, we begin to bound  $I_3(t)$ . Recall that the update rule is

$$\mathbf{x}_{i,t+1} = \sum_{j=1}^n \mathbf{W}_{ij} \mathbf{x}_{j,t} - \eta \nabla f_{i,t}(\mathbf{x}_{i,t}; \xi).$$

By taking average over  $i \in [n]$  on both sides, we have

$$\bar{\mathbf{x}}_{t+1} = \bar{\mathbf{x}}_t - \eta \left( \frac{1}{n} \sum_{i=1}^n \nabla f_{i,t}(\mathbf{x}_{i,t}; \xi) \right). \quad (3)$$

Denote a new auxiliary function  $h(\mathbf{z})$  as

$$h(\mathbf{z}) = \left\langle \frac{1}{n} \sum_{i=1}^n \nabla f_{i,t}(\mathbf{x}_{i,t}; \xi), \mathbf{z} \right\rangle + \frac{1}{2\eta} \|\mathbf{z} - \bar{\mathbf{x}}_t\|^2.$$

Note that (3) is equivalent to

$$\begin{aligned}
\bar{\mathbf{x}}_{t+1} &= \underset{\mathbf{z} \in \mathbb{R}^d}{\operatorname{argmin}} h(\mathbf{z}) \\
&= \underset{\mathbf{z} \in \mathbb{R}^d}{\operatorname{argmin}} \left\langle \frac{1}{n} \sum_{i=1}^n \nabla f_{i,t}(\mathbf{x}_{i,t}; \xi), \mathbf{z} \right\rangle + \frac{1}{2\eta} \|\mathbf{z} - \bar{\mathbf{x}}_t\|^2.
\end{aligned}$$

Furthermore, denote a new auxiliary variable  $\bar{\mathbf{x}}_\tau$  as

$$\bar{\mathbf{x}}_\tau = \bar{\mathbf{x}}_{t+1} + \tau (\mathbf{x}_t^* - \bar{\mathbf{x}}_{t+1}),$$

where  $0 \leq \tau \leq 1$ . According to the optimality of  $\bar{\mathbf{x}}_{t+1}$ , we have

$$\begin{aligned}
0 &\leq h(\bar{\mathbf{x}}_\tau) - h(\bar{\mathbf{x}}_{t+1}) \\
&= \left\langle \frac{1}{n} \sum_{i=1}^n \nabla f_{i,t}(\mathbf{x}_{i,t}; \xi), \bar{\mathbf{x}}_\tau - \bar{\mathbf{x}}_{t+1} \right\rangle + \frac{1}{2\eta} \left( \|\bar{\mathbf{x}}_\tau - \bar{\mathbf{x}}_t\|^2 - \|\bar{\mathbf{x}}_{t+1} - \bar{\mathbf{x}}_t\|^2 \right) \\
&= \left\langle \frac{1}{n} \sum_{i=1}^n \nabla f_{i,t}(\mathbf{x}_{i,t}; \xi), \tau (\mathbf{x}_t^* - \bar{\mathbf{x}}_{t+1}) \right\rangle + \frac{1}{2\eta} \left( \|\bar{\mathbf{x}}_{t+1} + \tau (\mathbf{x}_t^* - \bar{\mathbf{x}}_{t+1}) - \bar{\mathbf{x}}_t\|^2 - \|\bar{\mathbf{x}}_{t+1} - \bar{\mathbf{x}}_t\|^2 \right) \\
&= \left\langle \frac{1}{n} \sum_{i=1}^n \nabla f_{i,t}(\mathbf{x}_{i,t}; \xi), \tau (\mathbf{x}_t^* - \bar{\mathbf{x}}_{t+1}) \right\rangle + \frac{1}{2\eta} \left( \|\tau (\mathbf{x}_t^* - \bar{\mathbf{x}}_{t+1})\|^2 + 2 \langle \tau (\mathbf{x}_t^* - \bar{\mathbf{x}}_{t+1}), \bar{\mathbf{x}}_{t+1} - \bar{\mathbf{x}}_t \rangle \right).
\end{aligned}$$

Dividing  $\tau$  on both sides, and letting  $\tau$  be close to 0, we have

$$I_3(t) = \left\langle \frac{1}{n} \sum_{i=1}^n \nabla f_{i,t}(\mathbf{x}_{i,t}; \xi), \bar{\mathbf{x}}_{t+1} - \mathbf{x}_t^* \right\rangle$$

$$\begin{aligned}
&\leq \frac{1}{2\eta} \left( \lim_{\tau \rightarrow 0} \tau \|(\mathbf{x}_t^* - \bar{\mathbf{x}}_{t+1})\|^2 + 2 \langle \mathbf{x}_t^* - \bar{\mathbf{x}}_{t+1}, \bar{\mathbf{x}}_{t+1} - \bar{\mathbf{x}}_t \rangle \right) \\
&= \frac{1}{\eta} \langle \mathbf{x}_t^* - \bar{\mathbf{x}}_{t+1}, \bar{\mathbf{x}}_{t+1} - \bar{\mathbf{x}}_t \rangle \\
&= \frac{1}{2\eta} \left( \|\mathbf{x}_t^* - \bar{\mathbf{x}}_t\|^2 - \|\mathbf{x}_t^* - \bar{\mathbf{x}}_{t+1}\|^2 - \|\bar{\mathbf{x}}_t - \bar{\mathbf{x}}_{t+1}\|^2 \right). \tag{4}
\end{aligned}$$

Besides, we have

$$\begin{aligned}
&\|\mathbf{x}_{t+1}^* - \bar{\mathbf{x}}_{t+1}\|^2 - \|\mathbf{x}_t^* - \bar{\mathbf{x}}_{t+1}\|^2 \\
&= \|\mathbf{x}_{t+1}^*\|^2 - \|\mathbf{x}_t^*\|^2 - 2 \langle \bar{\mathbf{x}}_{t+1}, -\mathbf{x}_t^* + \mathbf{x}_{t+1}^* \rangle \\
&= (\|\mathbf{x}_{t+1}^*\| - \|\mathbf{x}_t^*\|) (\|\mathbf{x}_{t+1}^*\| + \|\mathbf{x}_t^*\|) - 2 \langle \bar{\mathbf{x}}_{t+1}, -\mathbf{x}_t^* + \mathbf{x}_{t+1}^* \rangle \\
&\leq \|\mathbf{x}_{t+1}^* - \mathbf{x}_t^*\| (\|\mathbf{x}_{t+1}^*\| + \|\mathbf{x}_t^*\|) - 2 \|\bar{\mathbf{x}}_{t+1}\| \|\mathbf{x}_{t+1}^* - \mathbf{x}_t^*\| \\
&\leq 4\sqrt{R} \|\mathbf{x}_{t+1}^* - \mathbf{x}_t^*\|. \quad (\text{due to } \|\mathbf{x} - \mathbf{y}\|^2 \leq R, \forall \mathbf{x}, \mathbf{y} \in \mathcal{X})
\end{aligned}$$

Thus, telescoping  $I_3(t)$  over  $t \in [T]$ , we have

$$\begin{aligned}
\sum_{t=1}^T I_3(t) &\leq \frac{1}{2\eta} \left( 4\sqrt{R} \sum_{t=1}^T \|\mathbf{x}_{t+1}^* - \mathbf{x}_t^*\| + \|\bar{\mathbf{x}}_1^* - \bar{\mathbf{x}}_1\|^2 - \|\bar{\mathbf{x}}_T^* - \bar{\mathbf{x}}_{T+1}\|^2 \right) - \frac{1}{2\eta} \sum_{t=1}^T \|\bar{\mathbf{x}}_t - \bar{\mathbf{x}}_{t+1}\|^2 \\
&\leq \frac{1}{2\eta} \left( 4\sqrt{R} \sum_{t=1}^T D + R \right) - \frac{1}{2\eta} \sum_{t=1}^T \|\bar{\mathbf{x}}_t - \bar{\mathbf{x}}_{t+1}\|^2.
\end{aligned}$$

Combining those bounds of  $I_1(t)$ ,  $I_2(t)$  and  $I_3(t)$  together, we finally obtain

$$\begin{aligned}
&\mathbb{E}_\xi \sum_{t=1}^T \sum_{i=1}^n f_{i,t}(\mathbf{x}_{i,t}; \xi) - f_t(\mathbf{x}_t^*; \xi) \\
&\leq n \sum_{t=1}^T (I_1(t) + I_2(t) + I_3(t)) \\
&\leq \left( 1 + 4\beta \left( \frac{1}{\nu} + 4 \right) \right) nT\beta G\eta + \left( L + \frac{2\eta L^2(1-\beta)^2}{\nu} + \frac{\nu}{2\eta} + 10\eta L^2 + \frac{\beta}{2\eta} \right) \frac{nT\eta^2 G}{(1-\rho)^2} \\
&\quad + n \left( \frac{1}{\nu} + 4 \right) (f(\bar{\mathbf{x}}_1) - f(\bar{\mathbf{x}}_{T+1})) + T\eta\sigma^2(1-\beta) \left( \frac{2}{\nu} + 9 \right) + \frac{n}{2\eta} (4\sqrt{R}D + R).
\end{aligned}$$

Let  $\nu = \sqrt{\beta^2 + \frac{1}{n}}$ , and thus we have

$$\begin{aligned}
&\sum_{t=1}^T \sum_{i=1}^n \mathbb{E}_\xi f_{i,t}(\mathbf{x}_{i,t}; \xi) - \mathbb{E}_\xi f_t(\mathbf{x}_t^*; \xi) \\
&\leq \left( 1 + 4\beta \left( \frac{1}{\sqrt{\beta^2 + \frac{1}{n}}} + 4 \right) \right) nT\beta G\eta + \left( L + \frac{2\eta L^2(1-\beta)^2}{\sqrt{\beta^2 + \frac{1}{n}}} + \frac{\sqrt{\beta^2 + \frac{1}{n}}}{2\eta} + 10\eta L^2 + \frac{\beta}{2\eta} \right) \frac{nT\eta^2 G}{(1-\rho)^2} \\
&\quad + n \left( \frac{1}{\sqrt{\beta^2 + \frac{1}{n}}} + 4 \right) (f(\bar{\mathbf{x}}_1) - f(\bar{\mathbf{x}}_{T+1})) + T\eta\sigma^2(1-\beta) \left( \frac{2}{\sqrt{\beta^2 + \frac{1}{n}}} + 9 \right) + \frac{n}{2\eta} (4\sqrt{R}D + R).
\end{aligned}$$

It completes the proof.  $\square$

**Lemma 1.**

$$\begin{aligned} & \mathbb{E}_\xi \frac{\eta}{2} \|\nabla f(\bar{\mathbf{x}}_t)\|^2 + \left( \frac{\eta}{2} - \frac{L\eta^2}{2} \right) \left\| \frac{1}{n} \sum_{i=1}^n \nabla f_{i,t}(\mathbf{x}_{i,t}; \xi) \right\|^2 \\ & \leq f(\bar{\mathbf{x}}_t) - f(\bar{\mathbf{x}}_{t+1}) + 4G\eta\beta^2 + \frac{2\eta L^2(1-\beta)^2}{n} \sum_{i=1}^n \|\bar{\mathbf{x}}_t - \mathbf{x}_{i,t}\|^2 + \frac{2\eta\sigma^2(1-\beta)^2}{n}. \end{aligned}$$

*Proof.*

$$\begin{aligned} & \mathbb{E}_\xi f(\bar{\mathbf{x}}_{t+1}) \\ & \leq f(\bar{\mathbf{x}}_t) + \mathbb{E}_\xi \langle \nabla f(\bar{\mathbf{x}}_t), \bar{\mathbf{x}}_{t+1} - \bar{\mathbf{x}}_t \rangle + \frac{L}{2} \mathbb{E}_\xi \|\bar{\mathbf{x}}_{t+1} - \bar{\mathbf{x}}_t\|^2 \\ & = f(\bar{\mathbf{x}}_t) + \mathbb{E}_\xi \left\langle \nabla f(\bar{\mathbf{x}}_t), -\frac{\eta}{n} \sum_{i=1}^n \nabla f_{i,t}(\mathbf{x}_{i,t}; \xi) \right\rangle + \frac{L}{2} \mathbb{E}_\xi \left\| \frac{\eta}{n} \sum_{i=1}^n \nabla f_{i,t}(\mathbf{x}_{i,t}; \xi) \right\|^2 \\ & = f(\bar{\mathbf{x}}_t) + \frac{\eta}{2} \left( \mathbb{E}_\xi \left\| \nabla f(\bar{\mathbf{x}}_t) - \frac{1}{n} \sum_{i=1}^n \nabla f_{i,t}(\mathbf{x}_{i,t}; \xi) \right\|^2 - \|\nabla f(\bar{\mathbf{x}}_t)\|^2 - \mathbb{E}_\xi \left\| \frac{1}{n} \sum_{i=1}^n \nabla f_{i,t}(\mathbf{x}_{i,t}; \xi) \right\|^2 \right) \\ & \quad + \frac{L}{2} \mathbb{E}_\xi \left\| \frac{\eta}{n} \sum_{i=1}^n \nabla f_{i,t}(\mathbf{x}_{i,t}; \xi) \right\|^2 \\ & = f(\bar{\mathbf{x}}_t) + \frac{\eta}{2} \mathbb{E}_\xi \left\| \nabla f(\bar{\mathbf{x}}_t) - \frac{1}{n} \sum_{i=1}^n \nabla f_{i,t}(\mathbf{x}_{i,t}; \xi) \right\|^2 - \frac{\eta}{2} \|\nabla f(\bar{\mathbf{x}}_t)\|^2 - \left( \frac{\eta}{2} - \frac{L\eta^2}{2} \right) \mathbb{E}_\xi \left\| \frac{1}{n} \sum_{i=1}^n \nabla f_{i,t}(\mathbf{x}_{i,t}; \xi) \right\|^2 \\ & \leq f(\bar{\mathbf{x}}_t) + \frac{\eta}{2} \mathbb{E}_\xi \left\| \nabla f(\bar{\mathbf{x}}_t) - \frac{1}{n} \sum_{i=1}^n \nabla f_{i,t}(\mathbf{x}_{i,t}; \xi) \right\|^2 - \frac{\eta}{2} \|\nabla f(\bar{\mathbf{x}}_t)\|^2 - \left( \frac{\eta}{2} - \frac{L\eta^2}{2} \right) \left\| \frac{1}{n} \sum_{i=1}^n \nabla f_{i,t}(\mathbf{x}_{i,t}; \xi) \right\|^2. \quad (3) \end{aligned}$$

Additionally, we have

$$\begin{aligned} & \mathbb{E}_\xi \left\| \nabla f(\bar{\mathbf{x}}_t) - \frac{1}{n} \sum_{i=1}^n \nabla f_{i,t}(\mathbf{x}_{i,t}; \xi) \right\|^2 \\ & = \mathbb{E}_\xi \left\| \nabla f(\bar{\mathbf{x}}_t) - \frac{1}{n} \sum_{i=1}^n (\beta \nabla g_{i,t}(\mathbf{x}_{i,t}) + (1-\beta) \nabla f(\mathbf{x}_{i,t}; \xi)) \right\|^2 \\ & \leq 2\beta^2 \left\| \nabla f(\bar{\mathbf{x}}_t) - \frac{1}{n} \sum_{i=1}^n \nabla g_{i,t}(\mathbf{x}_{i,t}) \right\|^2 + 2(1-\beta)^2 \mathbb{E}_\xi \left\| \nabla f(\bar{\mathbf{x}}_t) - \frac{1}{n} \sum_{i=1}^n \nabla f(\mathbf{x}_{i,t}; \xi) \right\|^2 \\ & \leq 2\beta^2 \left( 2 \|\nabla f(\bar{\mathbf{x}}_t)\|^2 + 2 \left\| \frac{1}{n} \sum_{i=1}^n \nabla g_{i,t}(\mathbf{x}_{i,t}) \right\|^2 \right) + 2(1-\beta)^2 \mathbb{E}_\xi \left\| \nabla f(\bar{\mathbf{x}}_t) - \frac{1}{n} \sum_{i=1}^n \nabla f(\mathbf{x}_{i,t}; \xi) \right\|^2 \\ & \leq 8G\beta^2 + 2(1-\beta)^2 \mathbb{E}_\xi \left\| \nabla f(\bar{\mathbf{x}}_t) - \frac{1}{n} \sum_{i=1}^n \nabla f(\mathbf{x}_{i,t}) + \frac{1}{n} \sum_{i=1}^n \nabla f(\mathbf{x}_{i,t}) - \frac{1}{n} \sum_{i=1}^n \nabla f(\mathbf{x}_{i,t}; \xi) \right\|^2 \\ & \leq 8G\beta^2 + 4(1-\beta)^2 \left\| \nabla f(\bar{\mathbf{x}}_t) - \frac{1}{n} \sum_{i=1}^n \nabla f(\mathbf{x}_{i,t}) \right\|^2 + 4(1-\beta)^2 \mathbb{E}_\xi \left\| \frac{1}{n} \sum_{i=1}^n \nabla f(\mathbf{x}_{i,t}) - \frac{1}{n} \sum_{i=1}^n \nabla f(\mathbf{x}_{i,t}; \xi) \right\|^2 \\ & \leq 8G\beta^2 + \frac{4(1-\beta)^2}{n} \sum_{i=1}^n \|\nabla f(\bar{\mathbf{x}}_t) - \nabla f(\mathbf{x}_{i,t})\|^2 + \frac{4(1-\beta)^2}{n^2} \sum_{i=1}^n \mathbb{E}_\xi \|\nabla f(\mathbf{x}_{i,t}) - \nabla f(\mathbf{x}_{i,t}; \xi)\|^2 \end{aligned}$$



$$\leq 8G\beta^2 + \frac{4L^2(1-\beta)^2}{n} \sum_{i=1}^n \|\bar{\mathbf{x}}_t - \mathbf{x}_{i,t}\|^2 + \frac{4\sigma^2(1-\beta)^2}{n}. \quad (4)$$

Substituting (4) into (3), we obtain

$$\begin{aligned} & f(\bar{\mathbf{x}}_{t+1}) \\ & \leq f(\bar{\mathbf{x}}_t) + \frac{\eta}{2} \left( 8G\beta^2 + \frac{4L^2(1-\beta)^2}{n} \sum_{i=1}^n \|\bar{\mathbf{x}}_t - \mathbf{x}_{i,t}\|^2 + \frac{4\sigma^2(1-\beta)^2}{n} \right) \\ & \quad - \frac{\eta}{2} \|\nabla f(\bar{\mathbf{x}}_t)\|^2 - \left( \frac{\eta}{2} - \frac{L\eta^2}{2} \right) \left\| \frac{1}{n} \sum_{i=1}^n \nabla f_{i,t}(\mathbf{x}_{i,t}) \right\|^2. \end{aligned}$$

Equivalently, we obtain

$$\begin{aligned} & \frac{\eta}{2} \|\nabla f(\bar{\mathbf{x}}_t)\|^2 + \left( \frac{\eta}{2} - \frac{L\eta^2}{2} \right) \left\| \frac{1}{n} \sum_{i=1}^n \nabla f_{i,t}(\mathbf{x}_{i,t}) \right\|^2 \\ & \leq f(\bar{\mathbf{x}}_t) - f(\bar{\mathbf{x}}_{t+1}) + 4G\eta\beta^2 + \frac{2\eta L^2(1-\beta)^2}{n} \sum_{i=1}^n \|\bar{\mathbf{x}}_t - \mathbf{x}_{i,t}\|^2 + \frac{2\eta\sigma^2(1-\beta)^2}{n}. \end{aligned}$$

It completes the proof.  $\square$

**Lemma 2.**

$$\frac{1}{n} \sum_{i=1}^n \sum_{t=1}^T \|\mathbf{x}_{i,t} - \bar{\mathbf{x}}_t\|^2 \leq \frac{1}{(1-\rho)^2} \sum_{t=1}^T \eta^2 G.$$

*Proof.* Recall that

$$\mathbf{x}_{i,t+1} = \sum_{j=1}^n \mathbf{W}_{ij} \mathbf{x}_{j,t} - \eta \nabla f_{i,t}(\mathbf{x}_{i,t}; \xi),$$

and

$$\bar{\mathbf{x}}_{t+1} = \bar{\mathbf{x}}_t - \eta \left( \frac{1}{n} \sum_{i=1}^n \nabla f_{i,t}(\mathbf{x}_{i,t}; \xi) \right).$$

Denote

$$\begin{aligned} \mathbf{X}_t &= [\mathbf{x}_{1,t}, \mathbf{x}_{2,t}, \dots, \mathbf{x}_{n,t}] \in \mathbb{R}^{d \times n}, \\ \mathbf{G}_t &= [\nabla f_{1,t}(\mathbf{x}_{1,t}; \xi), \nabla f_{2,t}(\mathbf{x}_{2,t}; \xi), \dots, \nabla f_{n,t}(\mathbf{x}_{n,t}; \xi)] \in \mathbb{R}^{d \times n}. \end{aligned}$$

By letting  $\mathbf{x}_{i,1} = \mathbf{0}$  for any  $i \in [n]$ , the update rule is re-formulated as

$$\mathbf{X}_{t+1} = \mathbf{X}_t \mathbf{W} - \eta \mathbf{G}_t = - \sum_{s=1}^t \eta \mathbf{G}_s \mathbf{W}^{t-s}.$$

Similarly, denote  $\bar{\mathbf{G}}_t = \frac{1}{n} \sum_{i=1}^n \nabla f_{i,t}(\mathbf{x}_{i,t}; \xi)$ , and we have

$$\bar{\mathbf{x}}_{t+1} = \bar{\mathbf{x}}_t - \eta \left( \frac{1}{n} \sum_{i=1}^n \nabla f_{i,t}(\mathbf{x}_{i,t}; \xi) \right) = - \sum_{s=1}^t \eta \bar{\mathbf{G}}_s. \quad (5)$$

Therefore,

$$\begin{aligned}
& \sum_{i=1}^n \|\mathbf{x}_{i,t} - \bar{\mathbf{x}}_t\|^2 \\
& \stackrel{\textcircled{1}}{=} \sum_{i=1}^n \left\| \sum_{s=1}^{t-1} \eta \bar{\mathbf{G}}_s - \eta \mathbf{G}_s \mathbf{W}^{t-s-1} \mathbf{e}_i \right\|^2 \\
& \stackrel{\textcircled{2}}{=} \left\| \sum_{s=1}^{t-1} \eta \mathbf{G}_s \mathbf{v}_1 \mathbf{v}_1^\top - \eta \mathbf{G}_s \mathbf{W}^{t-s-1} \right\|_F^2 \\
& \stackrel{\textcircled{3}}{\leq} \left( \eta \rho^{t-s-1} \left\| \sum_{s=1}^{t-1} \mathbf{G}_s \right\|_F \right)^2 \\
& \leq \left( \sum_{s=1}^{t-1} \eta \rho^{t-s-1} \|\mathbf{G}_s\|_F \right)^2.
\end{aligned}$$

① holds due to  $\mathbf{e}_i$  is a unit basis vector, whose  $i$ -th element is 1 and other elements are 0s. ② holds due to  $\mathbf{v}_1 = \frac{\mathbf{1}_n}{\sqrt{n}}$ . ③ holds due to Lemma 3.

According to Lemma 4, we have

$$\begin{aligned}
\frac{1}{n} \sum_{i=1}^n \sum_{t=1}^T \|\mathbf{x}_{i,t} - \bar{\mathbf{x}}_t\|^2 & \leq \frac{1}{n(1-\rho)^2} \sum_{t=1}^T \eta^2 \|\mathbf{G}_s\|_F^2 \\
& \leq \frac{1}{(1-\rho)^2} \sum_{t=1}^T \eta^2 G.
\end{aligned}$$

It completes the proof.  $\square$

**Lemma 3** (Appeared in Lemma 5 in [Tang et al., 2018]). *For any matrix  $\mathbf{X}_t \in \mathbb{R}^{d \times n}$ , decompose the confusion matrix  $\mathbf{W}$  as  $\mathbf{W} = \sum_{i=1}^n \lambda_i \mathbf{v}_i \mathbf{v}_i^\top = \mathbf{P} \mathbf{\Lambda} \mathbf{P}^\top$ , where  $\mathbf{P} = [\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n] \in \mathbb{R}^{n \times n}$ ,  $\mathbf{v}_i$  is the normalized eigenvector of  $\lambda_i$ .  $\mathbf{\Lambda}$  is a diagonal matrix, and  $\lambda_i$  be its  $i$ -th element. We have*

$$\|\mathbf{X}_t \mathbf{W}^t - \mathbf{X}_t \mathbf{v}_1 \mathbf{v}_1^\top\|_F^2 \leq \|\rho^t \mathbf{X}_t\|_F^2,$$

where  $\rho = \max\{|\lambda_2(\mathbf{W})|, |\lambda_n(\mathbf{W})|\}$ .

**Lemma 4** (Appeared in Lemma 6 in [Tang et al., 2018]). *Given two non-negative sequences  $\{a_t\}_{t=1}^\infty$  and  $\{b_t\}_{t=1}^\infty$  that satisfying*

$$a_t = \sum_{s=1}^t \rho^{t-s} b_s,$$

with  $\rho \in [0, 1)$ , we have

$$\sum_{t=1}^k a_t^2 \leq \frac{1}{(1-\rho)^2} \sum_{t=1}^k b_t^2.$$

## References

H. Tang, S. Gan, C. Zhang, T. Zhang, and J. Liu. Communication Compression for Decentralized Training. *arXiv.org*, Mar. 2018.