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# NONLINEAR MARKOVIAN STOCHASTIC APPROXIMATION

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# 1 Preliminaries

**Notations** The Euclidean norm is denoted by  $\|\cdot\|$ . The lowercase letter  $c$  and its derivatives  $c', c_0$ , etc. denote universal numerical constants, whose value may change from line to line. As we are primarily interested in dependence of  $\alpha$  and  $k$ , we adopt the following big- $O$  notation:  $\|f\| = \mathcal{O}(h(\alpha, k))$  if it holds that  $\|f\| \leq s \cdot \|h(\alpha, k)\|$  for some constant  $s > 0$ .

We use of the following iteration scheme:

$$\theta_{t+1} = \theta_t + \alpha (g(\theta_t, X_{t+1}) + \xi_{t+1}(\theta_t))$$

## 1.1 Assumptions

**Assumption 1** For each  $X \in \mathcal{X}$ , the function  $g(\theta, X)$  is three times continuously differentiable in  $\theta$  with uniformly bounded first to third derivatives, i.e.,  $\sup_{\theta \in \mathbb{R}^d} \|g^{(i)}(\theta, X)\| < \infty$  for  $i = 1, 2, 3, X \in \mathcal{X}$ . Moreover, there exists a constant  $L_1 > 0$  such that (1)  $\|g^{(i)}(\theta, X) - g^{(i)}(\theta', X)\| \leq L_1$ , for all  $\theta, \theta' \in \mathbb{R}^d, i = 0, 1, 2$  and  $X \in \mathcal{X}$ , and (2)  $\|g(0, X)\| \leq L_1$  for all  $X \in \mathcal{X}$ .

Assumption 1 implies that  $g(\theta, X)$  is  $L_1$ -Lipschitz w.r.t  $\theta$  uniformly in  $X$ . The above assumption immediately implies that the growth of  $\|g\|$  and  $\|\bar{g}\|$  will be at most linear in  $\theta$ , i.e.,  $\|g(\theta, X)\| \leq L_1(\|\theta - \theta^*\| + 1)$  and  $\|\bar{g}(\theta)\| \leq L_1(\|\theta - \theta^*\| + 1)$ .

**Assumption 2** There exists  $\mu > 0$  such that  $\langle \theta - \theta', \bar{g}(\theta) - \bar{g}(\theta') \rangle \leq -\mu \|\theta - \theta'\|^2, \forall \theta, \theta' \in \mathbb{R}^d$ . Consequently, the target equation  $\bar{g}(\theta) = 0$  has a unique solution  $\theta^*$ .

Denote by  $\mathcal{F}_k$  the filtration generated by  $\{X_{t+1}, \theta_t, \xi_{t+1}\}_{t=0}^{k-1} \cup \{X_{k+1}, \theta_k\}$ .

**Assumption 3** Let  $p \in \mathbb{Z}_+$  be given. The noise sequence  $(\xi_k)_{k \geq 1}$  is a collection of i.i.d random fields satisfying the following conditions with  $L_{2,p} > 0$ :

$$\mathbb{E}[\xi_{k+1}(\theta) | \mathcal{F}_k] = 0 \quad \text{and} \quad \mathbb{E}^{1/(2p)}[\|\xi_1(\theta)\|^{2p}] \leq L_{2,p}(\|\theta - \theta^*\| + 1), \quad \forall \theta \in \mathbb{R}^d.$$

Define  $C(\theta) = \mathbb{E}[\xi_1(\theta)^{\otimes 2}]$  and assume that  $C(\theta)$  is at least twice differentiable. There also exists  $M_\epsilon, k_\epsilon \geq 0$  such that for  $\theta \in \mathbb{R}^d$ , we have  $\max_{i=1,2} \|C^{(i)}(\theta)\| \leq M_\epsilon \{1 + \|\theta - \theta^*\|^{k_\epsilon}\}$ . In the sequel, we set  $L := L_1 + L_2$ , and without loss of generality, we assume  $L \geq 2\mu$  for some technical reasons.

**Assumption 4** There exists a Borel measurable function  $\hat{g}: \mathbb{R}^d \times \mathcal{X} \rightarrow \mathbb{R}^d$  where for each  $\theta \in \mathbb{R}^d, X \in \mathcal{X}$ ,

$$\hat{g}(\theta, X) - P_\theta \hat{g}(\theta, X) = g(\theta, X) - \bar{g}(\theta).$$

**Assumption 5** There exists  $L_{PH}^{(0)} < \infty$  and  $L_{PH}^{(1)} < \infty$  such that, for all  $\theta \in \mathbb{R}^d$  and  $X \in \mathcal{X}$ , one has  $\|\hat{g}(\theta, X)\| \leq L_{PH}^{(0)}, \|P_\theta \hat{g}(\theta, X)\| \leq L_{PH}^{(0)}$ . Moreover, for  $(\theta, \theta') \in \mathcal{H}^2$ ,

$$\sup_{X \in \mathcal{X}} \|P_\theta \hat{g}(\theta, X) - P_{\theta'} \hat{g}(\theta', X)\| \leq L_{PH}^{(1)} \|\theta - \theta'\|.$$

**Assumption 6** For any  $\theta, \theta' \in \mathbb{R}^d$ , we have  $\sup_{X \in \mathcal{X}} \|P_\theta(X, \cdot) - P_{\theta'}(X, \cdot)\|_{TV} \leq L_P \|\theta - \theta'\|$ .

**Assumption 7** For any  $\theta, \theta' \in \mathbb{R}^d$ , we have  $\sup_{X \in \mathcal{X}} \|g(\theta, X) - g(\theta', X)\| \leq L_H \|\theta - \theta'\|$ .

**Assumption 8** *There exists  $\rho < 1$ ,  $K_P < \infty$  such that*

$$\sup_{\theta \in \mathbb{R}^d, X \in \mathcal{X}} \|P_\theta^n(X, \cdot) - \pi_\theta(\cdot)\|_{TV} \leq \rho^n K_P,$$

**Lemma 1** *Assume that assumptions 6-8 hold. Then, for any  $\theta \in \mathbb{R}^d$  and  $X \in \mathcal{X}$ ,*

$$\|\hat{g}(\theta, X)\| \leq \frac{\sigma K_P}{1 - \rho},$$

$$\|P_\theta \hat{g}(\theta, X)\| \leq \frac{\sigma \rho K_P}{1 - \rho}.$$

Moreover, for any  $\theta, \theta' \in \mathbb{R}^d$  and  $X \in \mathcal{X}$ ,

$$\|P_\theta \hat{g}(\theta, X) - P_{\theta'} \hat{g}(\theta', X)\| \leq \|\theta - \theta'\|,$$

where

$$L_{PH}^{(1)} = \frac{K_P^2 \sigma L_P}{(1 - \rho)^2} (2 + K_P) + \frac{K_P}{1 - \rho} L_H.$$

Proof of this lemma can be found in [2], Lemma 7.

## 2 Error Bound

### 2.1 Base Case

We first prove the following lemma because we are going to use that calculation in many different parts of the proof:

**Lemma2.** *Using Assumptions 1, 3, 5, and for sufficiently small  $\alpha$  and  $t \geq 1$ , we have*

$$\mathbb{E}[\|\theta_t - \theta_{t-1}\|] \leq \alpha L (\mathbb{E}[\|\theta_{t-1} - \theta^*\|] + 1).$$

**Proof.** We have

$$\begin{aligned} \mathbb{E}[\|\theta_t - \theta_{t-1}\|] &\leq \alpha \mathbb{E}[\|g(\theta_{t-1}, X_t) + \xi_t(\theta_{t-1})\|] \\ &\leq \alpha L_1 (\mathbb{E}[\|\theta_{t-1} - \theta^*\|] + 1) + \alpha L_2 (\mathbb{E}[\|\theta_{t-1} - \theta^*\|] + 1) \\ &\leq \alpha L (\mathbb{E}[\|\theta_{t-1} - \theta^*\|] + 1) \end{aligned}$$

where the first line follows from ??, second line from the Lipschitzness condition and the assumption of

$$\mathbb{E}^{1/2}[\|\xi_t(\theta_{t-1})\|^2 | \mathcal{F}_{t-1}] \leq L_2 (\mathbb{E}[\|\theta_{t-1} - \theta^*\|] + 1),$$

and the third line from ??.

For the base case analysis, we can write:

$$\begin{aligned} & \mathbb{E}[\|\theta_{k+1} - \theta^*\|^2] - \mathbb{E}[\|\theta_k - \theta^*\|^2] = \\ & 2\alpha\mathbb{E}[\langle \theta_k - \theta^*, g(\theta_k, X_{k+1}) \rangle] + \alpha^2\mathbb{E}[\|g(\theta_k, X_{k+1})\|^2] + \alpha^2\mathbb{E}[\|\xi_{k+1}(\theta_k)\|^2] = \\ & 2\alpha\mathbb{E}[\langle \theta_k - \theta^*, g(\theta_k, X_{k+1}) - \bar{g}(\theta_k) \rangle] + 2\alpha\mathbb{E}[\langle \theta_k - \theta^*, \bar{g}(\theta_k) \rangle] + \alpha^2\mathbb{E}[\|g(\theta_k, X_{k+1})\|^2] + \alpha^2\mathbb{E}[\|\xi_{k+1}(\theta_k)\|^2]. \end{aligned}$$

It is easy to see that under Strong Monotonicity assumption, we have

$$\langle \theta_k - \theta^*, \bar{g}(\theta_k) \rangle = \langle \theta_k - \theta^*, \bar{g}(\theta_k) - \bar{g}(\theta^*) \rangle \leq -\mu\|\theta_k - \theta^*\|^2.$$

Additionally, under Assumption 1 and 3, we have the following upper bound

$$\begin{aligned} & \alpha^2(\mathbb{E}[\|g(\theta_k, X_{k+1})\|^2] + \mathbb{E}[\|\xi_{k+1}(\theta_k)\|^2]) \\ & \leq \alpha^2(L_1^2\mathbb{E}[(\|\theta_k - \theta^*\| + 1)^2] + L_2^2\mathbb{E}[(\|\theta_k - \theta^*\| + 1)^2]) \\ & \leq 2\alpha^2L^2(\mathbb{E}[\|\theta_k - \theta^*\|^2] + 1). \end{aligned}$$

Therefore, we have

$$\mathbb{E}[\|\theta_{k+1} - \theta^*\|^2] \leq (1 - 2\alpha(-\alpha L^2 + \mu))\mathbb{E}[\|\theta_k - \theta^*\|^2] + 2\alpha^2L^2 + 2\alpha\mathbb{E}[\langle \theta_k - \theta^*, g(\theta_k, X_{k+1}) - \bar{g}(\theta_k) \rangle]$$

Solving this recursion gives us the following inequality:

$$\begin{aligned} \mathbb{E}[\|\theta_{k+1} - \theta^*\|^2] & \leq (1 - 2\alpha(-\alpha L^2 + \mu))^{k+1}\mathbb{E}[\|\theta_0 - \theta^*\|^2] \\ & \quad + \sum_{t=0}^k (1 - 2\alpha(-\alpha L^2 + \mu))^t 2\alpha^2L^2 \\ & \quad + \sum_{t=0}^k 2\alpha(1 - 2\alpha(-\alpha L^2 + \mu))^{k-t}\mathbb{E}[\langle \theta_t - \theta^*, g(\theta_t, X_{t+1}) - \bar{g}(\theta_t) \rangle]. \end{aligned}$$

For notational simplicity we define  $\gamma_t := 2\alpha(1 - 2\alpha(-\alpha L^2 + \mu))^{k-t}$  for  $0 \leq t \leq k$ .

The second term above is just a geometric series. According to Lemma 12 of [1], this equals to  $\frac{\alpha L^2[1 - (1 - 2\alpha(-\alpha L^2 + \mu))^{k+1}]}{-\alpha L^2 + \mu}$ .

Now, we can upper bound the third summand using below decomposition:

$$\mathbb{E}\left[\sum_{t=0}^k \gamma_t \langle \theta_t - \theta^*, g(\theta_t, X_{t+1}) - \bar{g}(\theta_t) \rangle\right] = A_1 + A_2 + A_3 + A_4 + A_5$$

with

$$\begin{aligned}
A_1 &:= \mathbb{E} \left[ \sum_{t=1}^k \gamma_t \langle \theta_t - \theta^*, \hat{g}(\theta_t, X_{t+1}) - P_{\theta_t} \hat{g}(\theta_t, X_t) \rangle \right], \\
A_2 &:= \mathbb{E} \left[ \sum_{t=1}^k \gamma_t \langle \theta_t - \theta^*, P_{\theta_t} \hat{g}(\theta_t, X_t) - P_{\theta_{t-1}} \hat{g}(\theta_{t-1}, X_t) \rangle \right], \\
A_3 &:= \mathbb{E} \left[ \sum_{t=1}^k \gamma_t \langle \theta_t - \theta_{t-1}, P_{\theta_{t-1}} \hat{g}(\theta_{t-1}, X_t) \rangle \right], \\
A_4 &:= \mathbb{E} \left[ \sum_{t=1}^k (\gamma_t - \gamma_{t-1}) \langle \theta_{t-1} - \theta^*, P_{\theta_{t-1}} \hat{g}(\theta_{t-1} - \theta^*, X_t) \rangle \right], \\
A_5 &:= \mathbb{E} [\gamma_0 \langle \theta_0 - \theta^*, \hat{g}(\theta_0, X_0) \rangle] + \mathbb{E} [\gamma_k \langle \theta_k - \theta^*, P_{\theta_k} \hat{g}(\theta_k, X_{k+1}) \rangle]
\end{aligned}$$

For  $A_1$ , we note that  $\hat{g}(\theta_t, X_{t+1}) - P_{\theta_t} \hat{g}(\theta_t, X_t)$  is a martingale difference sequence [cf. ?] and therefore we have  $A_1 = 0$  by taking the total expectation.

For  $A_2$ , applying Cauchy-Schwarz inequality and ??, we have

$$\begin{aligned}
A_2 &\leq \sum_{t=1}^k L_{PH}^{(1)} \gamma_t \mathbb{E} [\|\theta_t - \theta^*\| \|\theta_t - \theta_{t-1}\|] \\
&= \sum_{t=1}^k \alpha L_{PH}^{(1)} \gamma_t \mathbb{E} [\|\theta_t - \theta^*\| \|g(\theta_{t-1}, X_t) + \xi_t(\theta_{t-1})\|] \\
&\leq \sum_{t=1}^k \alpha L_{PH}^{(1)} \gamma_t \mathbb{E} [(\|\theta_t - \theta_{t-1}\| + \|\theta_{t-1} - \theta^*\|) (\|g(\theta_{t-1}, X_t)\| + \|\xi_t(\theta_{t-1})\|)] \\
&\leq \sum_{t=1}^k \alpha L_{PH}^{(1)} \gamma_t (L_1 (\mathbb{E} [\|\theta_{t-1} - \theta^*\|^2] + \mathbb{E} [\|\theta_t - \theta_{t-1}\| \|\theta_{t-1} - \theta^*\|] + \mathbb{E} [\|\theta_{t-1} - \theta^*\|] + \mathbb{E} [\|\theta_t - \theta_{t-1}\|]) \\
&\quad + \mathbb{E} [\|\theta_t - \theta_{t-1}\| \|\xi_t(\theta_{t-1})\|] + \mathbb{E} [\|\theta_{t-1} - \theta^*\| \|\xi_t(\theta_{t-1})\|])
\end{aligned}$$

where the second line follows from ?? and the third line follows from the triangle inequality. Now we upper the compound terms in the last line's parentheses:

$$\begin{aligned}
\mathbb{E} [\|\theta_t - \theta_{t-1}\| \|\theta_{t-1} - \theta^*\|] &= \mathbb{E} [\mathbb{E} [\|\theta_t - \theta_{t-1}\| \|\theta_{t-1} - \theta^*\| | \mathcal{F}_{t-1}]] \\
&\leq \mathbb{E} [\alpha L (\|\theta_{t-1} - \theta^*\| + 1) \|\theta_{t-1} - \theta^*\|] \\
&\leq \frac{\alpha L (3 \mathbb{E} [\|\theta_{t-1} - \theta^*\|^2] + 1)}{2}
\end{aligned}$$

where in the second line we used Lemma 2 and in the last line we used  $u \leq \frac{u^2+1}{2}$ .

$$\begin{aligned}
\mathbb{E} [\|\theta_t - \theta_{t-1}\| \|\xi_t(\theta_{t-1})\|] &\leq \mathbb{E} [\alpha (\|g(\theta_{t-1}, X_t)\| + \|\xi_t(\theta_{t-1})\|) \|\xi_t(\theta_{t-1})\|] \\
&\leq \mathbb{E} [\alpha \|\xi_t(\theta_{t-1})\|^2 + \alpha L_1 (\|\theta_{t-1} - \theta^*\| + 1) \|\xi_t(\theta_{t-1})\|] \\
&\leq L (\mathbb{E} [\|\theta_{t-1} - \theta^*\|] + 1)
\end{aligned}$$

where the first inequality follows from ??, second line from Lemma 2 and in the last line we used

boundedness property of the noise and sufficiently small  $\alpha$ .

$$\begin{aligned}\mathbb{E}[\|\theta_{t-1} - \theta^*\| \|\xi_t(\theta_{t-1})\|] &= \mathbb{E}[\mathbb{E}[\|\theta_{t-1} - \theta^*\| \|\xi_t(\theta_{t-1})\| | \mathcal{F}_{t-1}]] \\ &\leq \mathbb{E}[L_2 \|\theta_{t-1} - \theta^*\| (\|\theta_{t-1} - \theta^*\| + 1)] \\ &\leq \frac{L_2(3\|\theta_{t-1} - \theta^*\|^2 + 1)}{2}\end{aligned}$$

where the second line follows from ?? and in the last line we used  $u \leq \frac{u^2+1}{2}$ .

Summing up all these bounds, we can write for  $A_2$ :

$$\begin{aligned}A_2 &\leq \sum_{t=1}^k \alpha L_{PH}^{(1)} \gamma_t (2L(\mathbb{E}[\|\theta_{t-1} - \theta^*\|] + 1) + 2L(3\mathbb{E}[\|\theta_{t-1} - \theta^*\|^2] + 1) + \mathbb{E}[\|\theta_{t-1} - \theta^*\|^2]) \\ &\leq \sum_{t=1}^k \alpha L L_{PH}^{(1)} \gamma_t (8\mathbb{E}[\|\theta_{t-1} - \theta^*\|^2] + 5)\end{aligned}$$

which in the last line we again used  $u \leq \frac{u^2+1}{2}$  property.

For  $A_3$ , we obtain

$$\begin{aligned}A_3 &\leq \sum_{t=1}^k \gamma_t \mathbb{E}[\|\theta_t - \theta_{t-1}\| \|P_{\theta_{t-1}} \hat{g}(\theta_{t-1}, X_t)\|] \\ &\leq \sum_{t=1}^k L_{PH}^{(0)} \gamma_t \mathbb{E}[\|g(\theta_{t-1}, X_t) + \xi_t(\theta_{t-1})\|] \\ &\leq \sum_{t=1}^k L_{PH}^{(0)} \gamma_t (L_1(\mathbb{E}[\|\theta_{t-1} - \theta^*\|] + 1) + L_2(\mathbb{E}[\|\theta_{t-1} - \theta^*\|] + 1)) \\ &\leq \sum_{t=1}^k \alpha L L_{PH}^{(0)} \gamma_t (\mathbb{E}[\|\theta_{t-1} - \theta^*\|] + 1)\end{aligned}$$

where second line follows from ?? and third line follows from ?? .

For  $A_4$ , we have

$$\begin{aligned}A_4 &\leq \sum_{t=1}^k |\gamma_t - \gamma_{t-1}| \mathbb{E}[\|\theta_{t-1} - \theta^*\| \|P_{\theta_{t-1}} \hat{g}(\theta_{t-1}, X_t)\|] \\ &\leq \sum_{t=1}^k L_{PH}^{(0)} |\gamma_t - \gamma_{t-1}| \mathbb{E}[\|\theta_{t-1} - \theta^*\|]\end{aligned}$$

Finally, for  $A_5$ , we obtain

$$A_5 \leq L_{PH}^{(0)} (\gamma_0 \mathbb{E}[\|\theta_0 - \theta^*\|] + \gamma_k \mathbb{E}[\|\theta_k - \theta^*\|])$$

which follows from Cacuhy-Swarz inequality and ??.

Combining the above terms gives us:

$$\begin{aligned}\mathbb{E}\left[\sum_{t=1}^k \gamma_t \langle \theta_t - \theta^*, g(\theta_t, X_{t+1} - \bar{g}(\theta_t)) \rangle\right] &\leq \sum_{t=0}^{k-1} \alpha L L_{PH}^{(1)} \gamma_{t+1} (5 + 8\mathbb{E}[\|\theta_t - \theta^*\|^2]) + \sum_{t=0}^{k-1} \alpha L L_{PH}^{(0)} \gamma_{t+1} (\mathbb{E}[\|\theta_{t-1} - \theta^*\|] + 1) + \\ &\quad \sum_{t=0}^{k-1} L_{PH}^{(0)} |\gamma_t - \gamma_{t+1}| \mathbb{E}[\|\theta_t - \theta^*\|] + L_{PH}^{(0)} (\gamma_0 \mathbb{E}[\|\theta_0 - \theta^*\|] + \gamma_k \mathbb{E}[\|\theta_k - \theta^*\|])\end{aligned}$$

now it should be noticed that as long as the  $\alpha$  satisfies  $\alpha \leq \frac{\mu}{L^2}$ , we have  $\gamma_t \leq \gamma_{t+1}$ . Thus, we can simplify

the above upper bound and write it this way:

$$\begin{aligned} \mathbb{E} \left[ \sum_{t=0}^k \gamma_t \langle \theta_t - \theta^*, g(\theta_t, X_{t+1} - \bar{g}(\theta_t)) \rangle \right] &\leq \sum_{t=0}^{k-1} \alpha L L_{PH}^{(1)} \gamma_{t+1} (5 + 8\mathbb{E}[\|\theta_t - \theta^*\|^2]) + \\ &\quad \sum_{t=0}^{k-1} L_{PH}^{(0)} ((\alpha L + 1) \gamma_{t+1} - \gamma_t) \mathbb{E}[\|\theta_t - \theta^*\|] + \\ &\quad \sum_{t=0}^{k-1} \alpha L L_{PH}^{(0)} \gamma_{t+1} + L_{PH}^{(0)} (\gamma_0 \mathbb{E}[\|\theta_0 - \theta^*\|] + \gamma_k \mathbb{E}[\|\theta_k - \theta^*\|]) \end{aligned}$$

Hence, using the derived upper bounds from the above terms, we have:

$$\begin{aligned} \mathbb{E}[\|\theta_{k+1} - \theta^*\|^2] &\leq \sum_{t=0}^{k-1} \alpha L L_{PH}^{(1)} \gamma_{t+1} (5 + 8\mathbb{E}[\|\theta_t - \theta^*\|^2]) + \sum_{t=0}^{k-1} L_{PH}^{(0)} ((\alpha L + 1) \gamma_{t+1} - \gamma_t) \mathbb{E}[\|\theta_t - \theta^*\|] + \\ &\quad (1 - 2\alpha(-\alpha L^2 + \mu)) \gamma_0 \mathbb{E}[\|\theta_0 - \theta^*\|^2] + L_{PH}^{(0)} \gamma_0 \mathbb{E}[\|\theta_0 - \theta^*\|] + L_{PH}^{(0)} \gamma_k \mathbb{E}[\|\theta_k - \theta^*\|] + \\ &\quad 2\alpha^2 L L_{PH}^{(0)} \left[ \frac{1 - (1 - 2\alpha(-\alpha L^2 + \mu))^k}{1 - (1 - 2\alpha(-\alpha L^2 + \mu))} \right] + \frac{\alpha L^2 [1 - (1 - 2\alpha(-\alpha L^2 + \mu))^{k+1}]}{-\alpha L^2 + \mu} \end{aligned}$$

for further notation simplicity we define  $c_{1,t} := 2\alpha^2 L L_{PH}^{(0)} \left[ \frac{1 - (1 - 2\alpha(-\alpha L^2 + \mu))^t}{1 - (1 - 2\alpha(-\alpha L^2 + \mu))} \right] + \frac{\alpha L^2 [1 - (1 - 2\alpha(-\alpha L^2 + \mu))^{t+1}]}{-\alpha L^2 + \mu}$  for  $0 \leq t \leq k$ . Now to write down this upper bound in a way in which it only depends on  $\|\theta_0 - \theta^*\|$  related terms and constants, we can write:

$$\begin{aligned} \mathbb{E}[\|\theta_{k+1} - \theta^*\|^2] &\leq \sum_{t=0}^{k-1} \left[ 8\alpha L L_{PH}^{(1)} \gamma_{t+1} \mathbb{E}[\|\theta_t - \theta^*\|^2] + 5\alpha L L_{PH}^{(1)} \gamma_{t+1} \right] + \sum_{t=0}^{k-1} L_{PH}^{(0)} ((\alpha L + 1) \gamma_{t+1} - \gamma_t) \mathbb{E}[\|\theta_t - \theta^*\|] + \\ &\quad (1 - 2\alpha(-\alpha L^2 + \mu)) \gamma_0 \mathbb{E}[\|\theta_0 - \theta^*\|^2] + L_{PH}^{(0)} \gamma_0 \mathbb{E}[\|\theta_0 - \theta^*\|] + L_{PH}^{(0)} \gamma_k \mathbb{E}[\|\theta_k - \theta^*\|] + c_{1,k} \\ &= \sum_{t=0}^{k-1} 8\alpha L L_{PH}^{(1)} \gamma_{t+1} \mathbb{E}[\|\theta_t - \theta^*\|^2] + 5\alpha L L_{PH}^{(1)} \sum_{t=0}^{k-1} \gamma_{t+1} + \sum_{t=0}^{k-1} L_{PH}^{(0)} ((\alpha L + 1) \gamma_{t+1} - \gamma_t) \mathbb{E}[\|\theta_t - \theta^*\|] + \\ &\quad (1 - 2\alpha(-\alpha L^2 + \mu)) \gamma_0 \mathbb{E}[\|\theta_0 - \theta^*\|^2] + L_{PH}^{(0)} \gamma_0 \mathbb{E}[\|\theta_0 - \theta^*\|] + L_{PH}^{(0)} \gamma_k \mathbb{E}[\|\theta_k - \theta^*\|] + c_{1,k} \\ &= \sum_{t=0}^{k-1} 8\alpha L L_{PH}^{(1)} \gamma_{t+1} \mathbb{E}[\|\theta_t - \theta^*\|^2] + \sum_{t=0}^{k-1} L_{PH}^{(0)} ((\alpha L + 1) \gamma_{t+1} - \gamma_t) \mathbb{E}[\|\theta_t - \theta^*\|] + \\ &\quad (1 - 2\alpha(-\alpha L^2 + \mu)) \gamma_0 \mathbb{E}[\|\theta_0 - \theta^*\|^2] + L_{PH}^{(0)} \gamma_0 \mathbb{E}[\|\theta_0 - \theta^*\|] + L_{PH}^{(0)} \gamma_k \mathbb{E}[\|\theta_k - \theta^*\|] + c_{1,k} + \\ &\quad \frac{10\alpha^2 L L_{PH}^{(1)} [1 - (1 - 2\alpha(-\alpha L^2 + \mu))^k]}{[1 - (1 - 2\alpha(-\alpha L^2 + \mu))]} \end{aligned}$$

where the last equality follows from the definition of  $\gamma_t$ s. Similarly we define  $c_{2,t} := \frac{10\alpha^2 L L_{PH}^{(1)} [1 - (1 - 2\alpha(-\alpha L^2 + \mu))^t]}{[1 - (1 - 2\alpha(-\alpha L^2 + \mu))]}$  for  $0 \leq t \leq k$ . So we can write it as

$$\begin{aligned} \mathbb{E}[\|\theta_{k+1} - \theta^*\|^2] &\leq \sum_{t=0}^{k-1} 8\alpha L L_{PH}^{(1)} \gamma_{t+1} \mathbb{E}[\|\theta_t - \theta^*\|^2] + \sum_{t=0}^{k-1} L_{PH}^{(0)} ((\alpha L + 1) \gamma_{t+1} - \gamma_t) \mathbb{E}[\|\theta_t - \theta^*\|] + \\ &\quad (1 - 2\alpha(-\alpha L^2 + \mu)) \gamma_0 \mathbb{E}[\|\theta_0 - \theta^*\|^2] + L_{PH}^{(0)} \gamma_0 \mathbb{E}[\|\theta_0 - \theta^*\|] + L_{PH}^{(0)} \gamma_k \mathbb{E}[\|\theta_k - \theta^*\|] + c_{1,k} + c_{2,k} \end{aligned}$$

Now for the second term on RHS, we note that since  $L \geq 2\mu$ ,

$$(\alpha L + 1)\gamma_{t+1} - \gamma_t \leq 2\alpha L\gamma_{t+1}, \quad \mathbb{E}[\|\theta_t - \theta^*\|] \leq \sqrt{\mathbb{E}[\|\theta_t - \theta^*\|^2]},$$

and consequently

$$\begin{aligned} & \frac{1}{(1 - 2\alpha(-\alpha L^2 + \mu))^k} \sum_{t=0}^{k-1} L_{PH}^{(0)}((\alpha L + 1)\gamma_{t+1} + \gamma_t) \mathbb{E}[\|\theta_t - \theta^*\|] \\ & \leq 4L_{PH}^{(0)} L\alpha^2 \sum_{t=0}^{k-1} \frac{1}{(1 - 2\alpha(-\alpha L^2 + \mu))^{t+1}} \sqrt{\mathbb{E}[\|\theta_t - \theta^*\|^2]} \\ & \leq 4L_{PH}^{(0)} L\alpha^2 \left( \sum_{t=0}^{k-1} \frac{1}{(1 - 2\alpha(-\alpha L^2 + \mu))^{t+1}} \right)^{1/2} \left( \sum_{t=1}^{k-1} \frac{1}{(1 - 2\alpha(-\alpha L^2 + \mu))^{t+1}} \mathbb{E}[\|\theta_t - \theta^*\|^2] \right)^{1/2} \\ & \leq 4L_{PH}^{(0)} L\alpha^2 \cdot \sum_{t=0}^{k-1} \frac{1}{(1 - 2\alpha(-\alpha L^2 + \mu))^{t+1}} \mathbb{E}[\|\theta_t - \theta^*\|^2] + \frac{1}{-\alpha L^2 + \mu} \cdot \frac{2\alpha L L_{PH}^{(0)}}{(1 - 2\alpha(-\alpha L^2 + \mu))^k}. \end{aligned}$$

We also note that

$$\frac{\gamma_k}{(1 - 2\alpha(-\alpha L^2 + \mu))^k} \mathbb{E}[\|\theta_k - \theta^*\|] \leq \alpha \frac{\mathbb{E}[\|\theta_k - \theta^*\|^2]}{(1 - 2\alpha(-\alpha L^2 + \mu))^k} + \frac{\alpha}{(1 - 2\alpha(-\alpha L^2 + \mu))^k}.$$

similarly

$$\frac{\gamma_0}{(1 - 2\alpha(-\alpha L^2 + \mu))^k} \mathbb{E}[\|\theta_0 - \theta^*\|] \leq \alpha \mathbb{E}[\|\theta_0 - \theta^*\|^2] + \alpha.$$

and we also define for  $0 \leq t \leq k$

$$c_{3,t} := \frac{1}{-\alpha L^2 + \mu} \frac{2\alpha L L_{PH}^{(0)}}{(1 - 2\alpha(-\alpha L^2 + \mu))^t} + \frac{\alpha L_{PH}^{(0)}}{(1 - 2\alpha(-\alpha L^2 + \mu))^t} + \alpha L_{PH}^{(0)}$$

to wrap up all the remainder terms.

Substituting back and rearranging with also defining  $c'_{2,t} := \frac{c_{2,t}}{(1 - 2\alpha(-\alpha L^2 + \mu))^t}$  and  $c'_{1,t} := \frac{c_{1,t}}{(1 - 2\alpha(-\alpha L^2 + \mu))^t}$ , yields

$$\begin{aligned} \frac{\mathbb{E}[\|\theta_{k+1} - \theta^*\|^2]}{(1 - 2\alpha(-\alpha L^2 + \mu))^k} & \leq \frac{\alpha}{(1 - 2\alpha(-\alpha L^2 + \mu))^k} \mathbb{E}[\|\theta_k - \theta^*\|^2] + \sum_{t=0}^{k-1} \frac{\alpha \left( 8L L_{PH}^{(1)} + 4\alpha(1 - 2\alpha(-\alpha L^2 + \mu))^{-1} L L_{PH}^{(0)} \right)}{(1 - 2\alpha(-\alpha L^2 + \mu))^t} \mathbb{E}[\|\theta_t - \theta^*\|^2] + \\ & \quad \alpha \mathbb{E}[\|\theta_0 - \theta^*\|^2] + c'_{1,k} + c'_{2,k} + c_{3,k}. \end{aligned}$$

for sufficiently small  $\alpha$ s, we have

$$\alpha \left( \frac{3}{2} L L_{PH}^{(1)} + 2\alpha(1 - 2\alpha(-\alpha L^2 + \mu))^{-1} L L_{PH}^{(0)} \right) \leq 4\alpha \left( (\alpha L + 1) L_{PH}^{(0)} + 2L L_{PH}^{(1)} \right)$$

using above simplification we can rewrite our upper bound as

$$\frac{\mathbb{E}[\|\theta_{k+1} - \theta^*\|^2]}{(1 - 2\alpha(-\alpha L^2 + \mu))^k} \leq 4\alpha \left( (\alpha L + 1) L_{PH}^{(0)} + 2L L_{PH}^{(1)} \right) \sum_{t=0}^k \frac{\mathbb{E}[\|\theta_t - \theta^*\|^2]}{(1 - 2\alpha(-\alpha L^2 + \mu))^t} + \alpha \mathbb{E}[\|\theta_0 - \theta^*\|^2] + c'_{1,k} + c'_{2,k} + c_{3,k}.$$

For solving the above recursion, we first define  $S_t := 4\alpha \left( (\alpha L + 1) L_{PH}^{(0)} + 2L L_{PH}^{(1)} \right) \sum_{l=0}^t \frac{\mathbb{E}[\|\theta_l - \theta^*\|^2]}{(1 - 2\alpha(-\alpha L^2 + \mu))^l}$  for



$0 \leq t \leq k$ . Also we use  $C_t := c'_{1,t} + c'_{2,t} + c_{3,t}$  and for  $0 \leq t \leq k$ , defining constant terms. Now we can write

$$\frac{\mathbb{E}[\|\theta_{t+1} - \theta^*\|^2]}{(1 - 2\alpha(-\alpha L^2 + \mu))^t} \leq S_t + \alpha \mathbb{E}[\|\theta_0 - \theta^*\|^2] + C_t.$$

using this expansion, we should first notice that

$$\begin{aligned} \frac{S_t}{S_{t-1}} &= \frac{S_{t-1} + 4\alpha \left( (\alpha L + 1) L_{PH}^{(0)} + 2LL_{PH}^{(1)} \right) \frac{\mathbb{E}[\|\theta_t - \theta^*\|^2]}{(1 - 2\alpha(-\alpha L^2 + \mu))^t}}{S_{t-1}} \\ &= 1 + 4\alpha \left( (\alpha L + 1) L_{PH}^{(0)} + 2LL_{PH}^{(1)} \right) \frac{S_{t-1} + \alpha \mathbb{E}[\|\theta_0 - \theta^*\|^2] + C_{t-1}}{S_{t-1}} \\ &\leq 1 + 8\alpha \left( (\alpha L + 1) L_{PH}^{(0)} + 2LL_{PH}^{(1)} \right). \end{aligned}$$

Now, since we have  $S_0 = 4\alpha \left( (\alpha L + 1) L_{PH}^{(0)} + 2LL_{PH}^{(1)} \right) \mathbb{E}[\|\theta_0 - \theta^*\|^2]$ , thus

$$S_t \leq 4\alpha \left( (\alpha L + 1) L_{PH}^{(0)} + 2LL_{PH}^{(1)} \right) \left[ 1 + 8\alpha \left( (\alpha L + 1) L_{PH}^{(0)} + 2LL_{PH}^{(1)} \right) \right]^t \mathbb{E}[\|\theta_0 - \theta^*\|^2].$$

Substituting this upper bound into previous equations we get

$$\begin{aligned} \mathbb{E}[\|\theta_{k+1} - \theta^*\|^2] &\leq \\ (1 - 2\alpha(-\alpha L^2 + \mu))^k &\left[ 4\alpha \left( (\alpha L + 1) L_{PH}^{(0)} + 2LL_{PH}^{(1)} \right) \left( 1 + 8\alpha \left( (\alpha L + 1) L_{PH}^{(0)} + 2LL_{PH}^{(1)} \right) \right)^k \mathbb{E}[\|\theta_0 - \theta^*\|^2] + \alpha \mathbb{E}[\|\theta_0 - \theta^*\|^2] + C_k \right] \end{aligned}$$

Choosing  $\alpha$  sufficiently small, we can simplify the above inequality and write it as follows

$$\mathbb{E}[\|\theta_{k+1} - \theta^*\|^2] \leq \tilde{c}_1 \cdot (1 - 2\alpha\mu)^{k+1} \mathbb{E}[\|\theta_0 - \theta^*\|^2] + \tilde{c}_2 \cdot 2\alpha L \left( L_{PH}^{(0)} + \alpha L_{PH}^{(1)} \right)$$

where  $\tilde{c}_1$  and  $\tilde{c}_2$  are  $\mathcal{O}(1)$  constants.

## 2.2 General Case

Similar to the previous case, we first prove a useful lemma:

**Lemma3.** *Using Assumptions 1, 3, 5, and  $m, n \geq 1$ , we have*

$$\mathbb{E}[\|\theta_t - \theta_{t-1}\|^m \|\theta_{t-1} - \theta^*\|^n] \leq 2^{2m+n-2} \alpha^m L^m (\mathbb{E}[\|\theta_{t-1} - \theta^*\|^{m+n}] + 1).$$

Also when  $m = 0$ , this upper bound can be written as

$$\mathbb{E}[\|\theta_{t-1} - \theta^*\|^n] \leq 2^{n-1} (\mathbb{E}[\|\theta_{t-1} - \theta^*\|^n] + 1). \quad (1)$$

**Proof.** We have

$$\begin{aligned}
\mathbb{E} [\|\theta_t - \theta_{t-1}\|^m \|\theta_{t-1} - \theta^*\|^n] &\leq \alpha^m \mathbb{E} [\|g(\theta_{t-1}, X_t) + \xi_t(\theta_{t-1})\|^m \|\theta_{t-1} - \theta^*\|^n] \\
&\leq 2^{m-1} \alpha^m \mathbb{E} [(\|g(\theta_{t-1}, X_t)\|^m + \|\xi_t(\theta_{t-1})\|^m) \|\theta_{t-1} - \theta^*\|^n] \\
&= 2^{m-1} \alpha^m \mathbb{E} [\mathbb{E} [(\|g(\theta_{t-1}, X_t)\|^m + \|\xi_t(\theta_{t-1})\|^m) | \mathcal{F}_{t-1}] \|\theta_{t-1} - \theta^*\|^n] \\
&\leq 2^{m-1} \alpha^m \mathbb{E} [L_1^m \mathbb{E} [(\|\theta_{t-1} - \theta^*\| + 1)^m] + L_2^m \mathbb{E} [(\|\theta_{t-1} - \theta^*\| + 1)^m] \|\theta_{t-1} - \theta^*\|^n] \\
&\leq 2^{m-1} \alpha^m L^m \mathbb{E} [(\|\theta_{t-1} - \theta^*\| + 1)^{m+n}] \\
&\leq 2^{2m+n-2} \alpha^m L^m (\mathbb{E} [\|\theta_{t-1} - \theta^*\|^{m+n}] + 1)
\end{aligned}$$

where the second and the last line follows from the mclaurin's inequality. Also the fourth line follows from ???. Proof for the other case is a trivial consequence of these calculations.

Now to do the proof in this case, we assume that the moment bound in [??] has been proven for  $k \leq n-1$ , we now proceed to show that the desired moment convergence holds for  $n$  with  $2 \leq n \leq p$ .

We start with the following decomposition of  $\|\theta_{k+1} - \theta^*\|^{2n}$

$$\begin{aligned}
\|\theta_{k+1} - \theta^*\|^{2n} &= (\|\theta_k - \theta^*\|^2 + 2\alpha \langle \theta_k - \theta^*, g(\theta_k, X_{k+1}) + \xi_{k+1}(\theta_k) \rangle + \alpha^2 \|g(\theta_k, X_{k+1}) + \xi_{k+1}(\theta_k)\|^2)^n \\
&= \sum_{\substack{i,j,l \\ i+j+l=n}} \binom{n}{i,j,l} \|\theta_k - \theta^*\|^{2i} (2\alpha \langle \theta_k - \theta^*, g(\theta_k, X_{k+1}) + \xi_{k+1}(\theta_k) \rangle)^j (\alpha \|g(\theta_k, X_{k+1}) + \xi_{k+1}(\theta_k)\|)^{2l}
\end{aligned}$$

We note the following cases.

1.  $i = n, j = l = 0$ . In this case, the summand is simply  $\|\theta_k - \theta^*\|^{2i}$ .
2. When  $i = n-1, j = 1$  and  $l = 0$ . In this case, the summand is of order  $\alpha$ , i.e.,

$$\alpha \cdot 2n \langle \theta_k - \theta^*, g(\theta_k, X_{k+1}) + \xi_{k+1}(\theta_k) \rangle^j \|\theta_k - \theta^*\|^{2(n-1)}.$$

We can further decompose it as

$$\begin{aligned}
&2n\alpha \langle \theta_k - \theta^*, g(\theta_k, X_{k+1}) + \xi_{k+1}(\theta_k) \rangle \|\theta_k - \theta^*\|^{2(n-1)} \\
&= \underbrace{2n\alpha \langle \theta_k - \theta^*, g(\theta_k, X_{k+1}) - \bar{g}(\theta_k) + \xi_{k+1}(\theta_k) \rangle \|\theta_k - \theta^*\|^{2(n-1)}}_{T_1} + \underbrace{2n\alpha \langle \theta_k - \theta^*, \bar{g}(\theta_k) \rangle \|\theta_k - \theta^*\|^{2(n-1)}}_{T_2}.
\end{aligned}$$

Note that, when  $(X_k)$  is i.i.d or from a martingale noise, we have

$$\mathbb{E}[T_1 | \theta_k] = 0$$

However, when  $(X_k)$  is Markovian, the above inequality does not hold and  $T_1$  requires careful analysis.

Nonetheless, under the strong monotonicity assumption, we have

$$T_2 \leq -2n\alpha\mu \|\theta_k - \theta^*\|^{2n}.$$

3. For the remaining terms, we see that they are of higher orders of  $\alpha$ . Therefore, when  $\alpha$  is selected sufficiently small, these terms do not raise concern.

Therefore, to prove the desired moment bound, we spend the remaining section analyzing  $T_1$ . Immediately, we note that

$$\begin{aligned}\mathbb{E}[T_1] &= \mathbb{E}\left[2n\alpha\langle\theta_k - \theta^*, g(\theta_k, X_{k+1}) - \bar{g}(\theta_k) + \mathbb{E}[\xi_{k+1}(\theta_k)|\theta_k]\rangle\|\theta_k - \theta^*\|^{2(n-1)}\right] \\ &= 2n\alpha\mathbb{E}\left[\underbrace{\langle\theta_k - \theta^*, g(\theta_k, X_{k+1}) - \bar{g}(\theta_k)\rangle\|\theta_k - \theta^*\|^{2(n-1)}}_{T'_1}\right].\end{aligned}$$

Subsequently, we focus on analyzing  $T'_1$ ; but before that, we write the general recursion of the error bound. First, we define  $T'_{1,t} := \langle\theta_t - \theta^*, g(\theta_t, X_{t+1}) - \bar{g}(\theta_t)\rangle\|\theta_t - \theta^*\|^{2(n-1)}$  to make  $T'_1$  dependent on the iteration index. Now, following the above decomposition and taking the expectations, we have:

$$\mathbb{E}[\|\theta_{k+1} - \theta^*\|^{2n}] \leq \mathbb{E}[\|\theta_k - \theta^*\|^{2n}] + 2n\alpha\mathbb{E}[T'_{1,k}] - 2n\alpha\mu\mathbb{E}[\|\theta_k - \theta^*\|^{2n}] + o(\alpha) = (1 - 2n\alpha\mu)\mathbb{E}[\|\theta_k - \theta^*\|^{2n}] + 2n\alpha\mathbb{E}[T'_{1,k}] + o(\alpha)$$

similarly to the previous case we define  $\gamma_t := 2n\alpha(1 - 2n\alpha\mu)^{k-t}$  for  $0 \leq t \leq k$ . Solving the above recursion will give us

$$\mathbb{E}[\|\theta_{k+1} - \theta^*\|^{2n}] \leq \sum_{t=0}^k \gamma_t \mathbb{E}[T'_{1,t}] + \gamma_0 \mathbb{E}[\|\theta_0 - \theta^*\|^{2n}] + o(\alpha)$$

We have to upper bound the first term in the RHS above. For this purpose, we use a similar decomposition to our base case analysis:

$$\sum_{t=0}^k \gamma_t \mathbb{E}[T'_{1,t}] = \mathbb{E}\left[\sum_{t=0}^k \gamma_t \langle\theta_t - \theta^*, g(\theta_t, X_{t+1}) - \bar{g}(\theta_t)\rangle\|\theta_t - \theta^*\|^{2(n-1)}\right] = A_1 + A_2 + A_3 + A_4 + A_5$$

with

$$\begin{aligned}A_1 &:= \mathbb{E}\left[\sum_{t=1}^k \gamma_t \langle\theta_t - \theta^*, \hat{g}(\theta_t, X_{t+1}) - P_{\theta_t} \hat{g}(\theta_t, X_t)\rangle\|\theta_t - \theta^*\|^{2(n-1)}\right], \\ A_2 &:= \mathbb{E}\left[\sum_{t=1}^k \gamma_t \langle\theta_t - \theta^*, P_{\theta_t} \hat{g}(\theta_t, X_t) - P_{\theta_{t-1}} \hat{g}(\theta_{t-1}, X_t)\rangle\|\theta_t - \theta^*\|^{2(n-1)}\right], \\ A_3 &:= \mathbb{E}\left[\sum_{t=1}^k \gamma_t \langle\theta_t - \theta_{t-1}, P_{\theta_{t-1}} \hat{g}(\theta_{t-1}, X_t)\rangle\|\theta_t - \theta^*\|^{2(n-1)}\right], \\ A_4 &:= \mathbb{E}\left[\sum_{t=1}^k (\gamma_t - \gamma_{t-1}) \langle\theta_{t-1} - \theta^*, P_{\theta_{t-1}} \hat{g}(\theta_{t-1} - \theta^*, X_t)\rangle\|\theta_t - \theta^*\|^{2(n-1)}\right], \\ A_5 &:= \mathbb{E}\left[\sum_{t=1}^k \gamma_{t-1} \langle\theta_{t-1} - \theta^*, P_{\theta_{t-1}} \hat{g}(\theta_{t-1} - \theta^*, X_t)\rangle(\|\theta_t - \theta^*\|^{2(n-1)} - \|\theta_{t-1} - \theta^*\|^{2(n-1)})\right], \\ A_6 &:= \mathbb{E}[\gamma_0 \langle\theta_0 - \theta^*, \hat{g}(\theta_0, X_0)\rangle\|\theta_0 - \theta^*\|^{2(n-1)}] + \mathbb{E}[\gamma_k \langle\theta_k - \theta^*, P_{\theta_k} \hat{g}(\theta_k, X_{k+1})\rangle\|\theta_k - \theta^*\|^{2(n-1)}].\end{aligned}$$

For  $A_1$ , we note that  $\hat{g}(\theta_t, X_{t+1}) - P_{\theta_t} \hat{g}(\theta_t, X_t)$  is a martingale difference sequence [cf. ?] and therefore we have  $A_1 = 0$  by taking the total expectation.

For  $A_2$ , applying Cauchy-Schwarz inequality and ??, we have

$$\begin{aligned}
A_2 &\leq \sum_{t=1}^k L_{PH}^{(1)} \gamma_t \mathbb{E} [\|\theta_t - \theta_{t-1}\| \|\theta_t - \theta^*\|^{2n-1}] \\
&\leq \sum_{t=1}^k L_{PH}^{(1)} \gamma_t \mathbb{E} [\|\theta_t - \theta_{t-1}\| (\|\theta_t - \theta_{t-1}\| + \|\theta_{t-1} - \theta^*\|)^{2n-1}] \\
&\leq \sum_{t=1}^k 2^{2n-2} L_{PH}^{(1)} \gamma_t \mathbb{E} [\|\theta_t - \theta_{t-1}\|^{2n} + \|\theta_t - \theta_{t-1}\| \|\theta_{t-1} - \theta^*\|^{2n-1}] \\
&\leq \sum_{t=1}^k 2^{2n-2} L_{PH}^{(1)} \gamma_t (2^{4n} \alpha^{2n} L^{2n} (\mathbb{E} [\|\theta_{t-1} - \theta^*\|^{2n}] + 1) + 2^{2n-1} \alpha L (\mathbb{E} [\|\theta_{t-1} - \theta^*\|^{2n}] + 1)) \\
&\leq \sum_{t=1}^k 2^{4n-2} \alpha L L_{PH}^{(1)} \gamma_t (\mathbb{E} [\|\theta_{t-1} - \theta^*\|^{2n}] + 1)
\end{aligned}$$

where the second line follows from triangle inequality, third line from mclaurin's inequality, fourth line from Lemma 3, and the last line for  $\alpha < \frac{1}{4L}$ .

For  $A_3$ , we obtain

$$\begin{aligned}
A_3 &\leq \sum_{t=1}^k \gamma_t \mathbb{E} [\|\theta_t - \theta_{t-1}\| \|P_{\theta_{t-1}} \hat{g}(\theta_{t-1}, X_t)\| \|\theta_t - \theta^*\|^{2(n-1)}] \\
&\leq \sum_{t=1}^k L_{PH}^{(0)} \gamma_t \mathbb{E} [\|\theta_t - \theta_{t-1}\| (\|\theta_t - \theta_{t-1}\| + \|\theta_{t-1} - \theta^*\|)^{2(n-1)}] \\
&\leq \sum_{t=1}^k 2^{2n-3} L_{PH}^{(0)} \gamma_t \mathbb{E} [\|\theta_t - \theta_{t-1}\|^{2n-1} + \|\theta_t - \theta_{t-1}\| \|\theta_{t-1} - \theta^*\|^{2(n-1)}] \\
&\leq \sum_{t=1}^k 2^{2n-3} L_{PH}^{(0)} \gamma_t (2^{4n-1} \alpha^{2n-1} L^{2n-1} (\mathbb{E} [\|\theta_{t-1} - \theta^*\|^{2n-1}] + 1) + 2^{2n-2} \alpha L (\mathbb{E} [\|\theta_{t-1} - \theta^*\|^{2n-1}] + 1)) \\
&\leq \sum_{t=1}^k 2^{4n-4} \alpha L L_{PH}^{(0)} \gamma_t (\mathbb{E} [\|\theta_{t-1} - \theta^*\|^{2n-1}] + 1)
\end{aligned}$$

where second line follows from triangle inequality, third line from mclaurin's inequality, fourth line from Lemma 3, and the last line for  $\alpha < \frac{1}{4L}$ .

For  $A_4$ , we have

$$\begin{aligned}
A_4 &\leq \sum_{t=1}^k |\gamma_t - \gamma_{t-1}| \mathbb{E} [\|\theta_{t-1} - \theta^*\| \|P_{\theta_{t-1}} \hat{g}(\theta_{t-1}, X_t)\| \|\theta_t - \theta^*\|^{2(n-1)}] \\
&\leq \sum_{t=1}^k L_{PH}^{(0)} |\gamma_t - \gamma_{t-1}| \mathbb{E} [\|\theta_{t-1} - \theta^*\| \|\theta_t - \theta^*\|^{2(n-1)}] \\
&\leq \sum_{t=1}^k L_{PH}^{(0)} |\gamma_t - \gamma_{t-1}| \mathbb{E} [\|\theta_{t-1} - \theta^*\| (\|\theta_t - \theta_{t-1}\| + \|\theta_{t-1} - \theta^*\|)^{2(n-1)}] \\
&\leq \sum_{t=1}^k 2^{2n-3} L_{PH}^{(0)} |\gamma_t - \gamma_{t-1}| \mathbb{E} [\|\theta_t - \theta_{t-1}\|^{2(n-1)} \|\theta_{t-1} - \theta^*\| + \|\theta_{t-1} - \theta^*\|^{2n-1}] \\
&\leq \sum_{t=1}^k 2^{2n-3} L_{PH}^{(0)} |\gamma_t - \gamma_{t-1}| (2^{4n-5} \alpha^{2n-2} L^{2n-2} (\mathbb{E} [\|\theta_{t-1} - \theta^*\|^{2n-1}] + 1) + 2^{2n-2} (\mathbb{E} [\|\theta_{t-1} - \theta^*\|^{2n-1}] + 1)) \\
&\leq \sum_{t=1}^k 2^{4n-4} L_{PH}^{(0)} |\gamma_t - \gamma_{t-1}| (\mathbb{E} [\|\theta_{t-1} - \theta^*\|^{2n-1}] + 1)
\end{aligned}$$

where the third line follows from triangle inequality, fourth line from mclaurin's inequality, fifth line from Lemma 3, and the last line for  $\alpha < \frac{1}{4L}$ .

Now for  $A_5$ , we have to first note that, using mean-value theorem and with  $a \in [0, 1]$ , we'll get

$$\begin{aligned}\|\theta_t - \theta^*\|^{2(n-1)} - \|\theta_{t-1} - \theta^*\|^{2(n-1)} &= \|\theta_t - \theta_{t-1}\| \cdot 2(n-1) \|a(\theta_t - \theta^*) + (1-a)(\theta_{t-1} - \theta^*)\|^{2n-3} \\ &\leq \|\theta_t - \theta_{t-1}\| \cdot 2(n-1) \|a(\theta_t - \theta_{t-1}) + \theta_{t-1} - \theta^*\|^{2n-3} \\ &\leq 2^{2n-3}(n-1) \|\theta_t - \theta_{t-1}\| (\|\theta_t - \theta_{t-1}\|^{2n-3} + \|\theta_{t-1} - \theta^*\|^{2n-3})\end{aligned}$$

where the last line follows using the mclaurin's inequality. Plugging in the above upper bound to  $A_5$  gives us

$$\begin{aligned}A_5 &\leq \sum_{t=1}^k 2^{2n-3}(n-1) \gamma_{t-1} \mathbb{E} [\langle \theta_{t-1} - \theta^*, P_{\theta_{t-1}} \hat{g}(\theta_{t-1} - \theta^*, X_t) \rangle \|\theta_t - \theta_{t-1}\| (\|\theta_t - \theta_{t-1}\|^{2n-3} + \|\theta_{t-1} - \theta^*\|^{2n-3})] \\ &\leq \sum_{t=1}^k 2^{2n-3}(n-1) \gamma_{t-1} \mathbb{E} [\|\theta_{t-1} - \theta^*\| \|P_{\theta_{t-1}} \hat{g}(\theta_{t-1}, X_t)\| \|\theta_t - \theta_{t-1}\| (\|\theta_t - \theta_{t-1}\|^{2n-3} + \|\theta_{t-1} - \theta^*\|^{2n-3})] \\ &\leq \sum_{t=1}^k 2^{2n-3}(n-1) L_{PH}^{(0)} \gamma_{t-1} \mathbb{E} [\|\theta_{t-1} - \theta^*\| \|\theta_t - \theta_{t-1}\| (\|\theta_t - \theta_{t-1}\|^{2n-3} + \|\theta_{t-1} - \theta^*\|^{2n-3})] \\ &\leq \sum_{t=1}^k 2^{2n-3}(n-1) L_{PH}^{(0)} \gamma_{t-1} \mathbb{E} [\|\theta_t - \theta_{t-1}\|^{2n-2} \|\theta_{t-1} - \theta^*\| + \|\theta_t - \theta_{t-1}\| \|\theta_{t-1} - \theta^*\|^{2n-2}] \\ &\leq \sum_{t=1}^k 2^{2n-3}(n-1) L_{PH}^{(0)} \gamma_{t-1} (2^{4n-5} \alpha^{2n-2} L^{2n-2} (\mathbb{E} [\|\theta_{t-1} - \theta^*\|^{2n-1}] + 1) + 2^{2n-2} \alpha L (\mathbb{E} [\|\theta_{t-1} - \theta^*\|^{2n-1}] + 1)) \\ &\leq \sum_{t=1}^k 2^{4n-4}(n-1) \alpha L L_{PH}^{(0)} \gamma_{t-1} (\mathbb{E} [\|\theta_{t-1} - \theta^*\|^{2n-1}] + 1)\end{aligned}$$

where the fifth line follows from Lemma 3 and the last line follows for  $\alpha < \frac{1}{4L}$ .

Finally, for  $A_6$ , we obtain

$$A_6 \leq L_{PH}^{(0)} (\gamma_0 \|\theta_0 - \theta^*\|^{2n-1} + \gamma_k \|\theta_k - \theta^*\|^{2n-1})$$

which follows from Cauchy-Schwarz inequality and ??.

Combining the above terms gives us:

$$\begin{aligned}\sum_{t=0}^k \gamma_t \mathbb{E} [T'_{1,t}] &= \mathbb{E} \left[ \sum_{t=0}^k \gamma_t \langle \theta_t - \theta^*, g(\theta_t, X_{t+1}) - \bar{g}(\theta_t) \rangle \|\theta_t - \theta^*\|^{2(n-1)} \right] \\ &\leq \sum_{t=0}^{k-1} 2^{4n-2} \alpha L L_{PH}^{(1)} \gamma_{t+1} (\mathbb{E} [\|\theta_t - \theta^*\|^{2n}] + 1) + \sum_{t=0}^{k-1} 2^{4n-4} \alpha L L_{PH}^{(0)} \gamma_{t+1} (\mathbb{E} [\|\theta_t - \theta^*\|^{2n-1}] + 1) \\ &\quad + \sum_{t=0}^{k-1} 2^{4n-4} L_{PH}^{(0)} |\gamma_{t+1} - \gamma_t| (\mathbb{E} [\|\theta_t - \theta^*\|^{2n-1}] + 1) + \sum_{t=0}^{k-1} 2^{4n-4} (n-1) \alpha L L_{PH}^{(0)} \gamma_{t+1} (\mathbb{E} [\|\theta_t - \theta^*\|^{2n-1}] + 1) \\ &\quad + L_{PH}^{(0)} (\gamma_0 \|\theta_0 - \theta^*\|^{2n-1} + \gamma_k \|\theta_k - \theta^*\|^{2n-1}) \\ &= \sum_{t=0}^{k-1} 2^{4n-2} \alpha L L_{PH}^{(1)} \gamma_{t+1} (\mathbb{E} [\|\theta_t - \theta^*\|^{2n}] + 1) + \sum_{t=0}^{k-1} 2^{4n-4} L_{PH}^{(0)} ((n\alpha L + 1) \gamma_{t+1} - \gamma_t) (\mathbb{E} [\|\theta_t - \theta^*\|^{2n-1}] + 1) \\ &\quad + L_{PH}^{(0)} (\gamma_0 \|\theta_0 - \theta^*\|^{2n-1} + \gamma_k \|\theta_k - \theta^*\|^{2n-1})\end{aligned}$$

where in the last equality we used the fact that for  $\alpha \leq \frac{\mu}{L^2}$ ,  $\gamma_t \leq \gamma_{t+1}$ .

Now, if we make use of the inequality  $2|x|^3 \leq x^2 + x^4$  and consolidating the terms, for sufficiently small

$\alpha$ , we have

$$\begin{aligned}
\mathbb{E}[\|\theta_{k+1} - \theta^*\|^{2n}] &\leq \sum_{t=0}^{k-1} 2^{4n-5} \left[ 8\alpha L L_{PH}^{(1)} \gamma_{t+1} + L_{PH}^{(0)} ((n\alpha L + 1) \gamma_{t+1} - \gamma_t) \right] (\mathbb{E}[\|\theta_t - \theta^*\|^{2n}] + 1) \\
&\quad + \sum_{t=0}^{k-1} 2^{4n-5} L_{PH}^{(0)} ((n\alpha L + 1) \gamma_{t+1} - \gamma_t) (\mathbb{E}[\|\theta_t - \theta^*\|^{2(n-1)}] + 1) + \frac{L_{PH}^{(0)}}{2} (\gamma_0 \|\theta_0 - \theta^*\|^{2(n-1)} + \gamma_k \|\theta_k - \theta^*\|^{2(n-1)}) \\
&\quad + \left( \frac{L_{PH}^{(0)}}{2} + 1 \right) \gamma_0 \mathbb{E}[\|\theta_0 - \theta^*\|^{2n}] + \frac{L_{PH}^{(0)}}{2} \gamma_k \mathbb{E}[\|\theta_k - \theta^*\|^{2n}] + o(\alpha) \\
&\leq \sum_{t=0}^{k-1} 2^{4n-4} \alpha L \gamma_{t+1} \left[ 4L_{PH}^{(1)} + nL_{PH}^{(0)} \right] \mathbb{E}[\|\theta_t - \theta^*\|^{2n}] + \sum_{t=0}^{k-1} 2^{4n-4} n\alpha L L_{PH}^{(0)} \mathbb{E}[\|\theta_t - \theta^*\|^{2(n-1)}] \\
&\quad + \frac{L_{PH}^{(0)}}{2} (\gamma_0 \|\theta_0 - \theta^*\|^{2(n-1)} + \gamma_k \|\theta_k - \theta^*\|^{2(n-1)}) + \left( \frac{L_{PH}^{(0)}}{2} + 1 \right) \gamma_0 \mathbb{E}[\|\theta_0 - \theta^*\|^{2n}] + \frac{L_{PH}^{(0)}}{2} \gamma_k \mathbb{E}[\|\theta_k - \theta^*\|^{2n}] + \tilde{c}_{1,k} + o(\alpha)
\end{aligned}$$

where the last inequality follows for  $(n\alpha L + 1) \gamma_{t+1} - \gamma_t \leq 2n\alpha L \gamma_{t+1}$  and also with

$$\tilde{c}_{1,t} := \sum_{l=0}^{t-1} 2^{4n-4} n\alpha L \gamma_{l+1} \left[ 4L_{PH}^{(1)} + nL_{PH}^{(0)} \right] + \sum_{l=0}^{t-1} 2^{4n-4} n\alpha L L_{PH}^{(0)}$$

being defined for  $1 \leq t \leq k-1$ . It can be seen that for sufficiently small  $\alpha$ ,  $\tilde{c}_{1,t}$  would be of order  $4^n n\alpha L (nL_{PH}^{(0)} + 4\alpha L_{PH}^{(1)})$ . Now from induction hypothesis we know that

$$\mathbb{E}[\|\theta_t - \theta^*\|^{2(n-1)}] \leq \tilde{C}_{n-1,1} \cdot (1 - 2n\alpha\mu)^t \mathbb{E}[\|\theta_0 - \theta^*\|^{2n}] + \tilde{C}_{n-1,2} \cdot 4^{n-1} (n-1) \alpha L \left( (n-1) L_{PH}^{(0)} + 2\alpha L_{PH}^{(1)} \right).$$

Thus, if we plug in this upper bound into the previous inequality, for  $\alpha < \frac{1}{4^n}$  we will have

$$\begin{aligned}
\mathbb{E}[\|\theta_{k+1} - \theta^*\|^{2n}] &\leq \sum_{t=0}^{k-1} 2^{4n-4} \alpha L \gamma_{t+1} \left[ 4L_{PH}^{(1)} + nL_{PH}^{(0)} \right] \mathbb{E}[\|\theta_t - \theta^*\|^{2n}] \\
&\quad + \frac{L_{PH}^{(0)}}{2} \gamma_k \mathbb{E}[\|\theta_k - \theta^*\|^{2n}] + \tilde{c}_{2,k} \mathbb{E}[\|\theta_0 - \theta^*\|^{2n}] + \tilde{c}_{1,k} + \tilde{c}_{3,k} + o(\alpha)
\end{aligned}$$

in which  $\tilde{c}_{3,k}$  would be of order  $4^n n\alpha L (nL_{PH}^{(0)} + 4\alpha L_{PH}^{(1)})$  and  $\tilde{c}_{2,k}$  is an  $\mathcal{O}(1)$  constant dependent to  $k$ .

To solve the above recursion, we define  $\tilde{S}_t := \frac{L_{PH}^{(0)}}{2} \gamma_t \mathbb{E}[\|\theta_t - \theta^*\|^{2n}] + \sum_{l=0}^{t-1} 2^{4n-4} \alpha L \gamma_{l+1} \left[ 4L_{PH}^{(1)} + nL_{PH}^{(0)} \right] \mathbb{E}[\|\theta_l - \theta^*\|^{2n}]$  for  $1 \leq t \leq k$ . Using this we can write

$$\mathbb{E}[\|\theta_{t+1} - \theta^*\|^{2n}] \leq \tilde{S}_t + \tilde{c}_{2,t} \mathbb{E}[\|\theta_0 - \theta^*\|^{2n}] + \tilde{c}_{1,t} + \tilde{c}_{3,t} + o(\alpha).$$

Now notice that we have (recall that  $\gamma_t = 2n\alpha(1 - 2n\alpha\mu)^{k-t}$ .)

$$\begin{aligned}
\frac{\tilde{S}_t}{\tilde{S}_{t-1}} &\leq \frac{\tilde{S}_{t-1} + \left( 2^{4n-4} \alpha L \gamma_t \left[ 4L_{PH}^{(1)} + nL_{PH}^{(0)} \right] - \frac{L_{PH}^{(0)}}{2} \gamma_{t-1} \right) \mathbb{E}[\|\theta_{t-1} - \theta^*\|^{2n}] + \frac{L_{PH}^{(0)}}{2} \gamma_t \mathbb{E}[\|\theta_t - \theta^*\|^{2n}]}{\tilde{S}_{t-1}} \\
&= \frac{\tilde{S}_{t-1} + \left( 2^{4n-4} \alpha L \gamma_t \left[ 4L_{PH}^{(1)} + nL_{PH}^{(0)} \right] - \frac{L_{PH}^{(0)}}{2} \gamma_{t-1} \right) \mathbb{E}[\|\theta_{t-1} - \theta^*\|^{2n}] + \frac{L_{PH}^{(0)}}{2} \gamma_t [\tilde{S}_{t-1} + \tilde{c}_{2,t-1} \mathbb{E}[\|\theta_0 - \theta^*\|^{2n}] + \tilde{c}_{1,t} + \tilde{c}_{3,t} + o(\alpha)]}{\tilde{S}_{t-1}}
\end{aligned}$$

it is not very hard to see that the if we take  $\alpha$  small enough above fraction could get as close as possible

to 1. This means that we can write our final error bound as the following

$$\mathbb{E} [\|\theta_{k+1} - \theta^*\|^{2n}] \leq \tilde{C}_{n,1} \cdot (1 - 2n\alpha\mu)^{k+1} \mathbb{E} [\|\theta_0 - \theta^*\|^{2n}] + \tilde{C}_{n,2} \cdot 4^n n\alpha L \left( nL_{PH}^{(0)} + 4\alpha L_{PH}^{(1)} \right)$$

in which  $\tilde{C}_{n,1}$  and  $\tilde{C}_{n,2}$  are  $\mathcal{O}(1)$  constants and also  $\alpha$  has been taken small enough.

## References

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