
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 α 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 θ 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 θ uniformly in X . The above assumption immediately implies that the growth of $\|g\|$ and $\|\bar{g}\|$ will be at most linear in θ , 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 θ^* .

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 α 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 α .

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

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 Cacuchy-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 α 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 γ_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 α 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 α 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*

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

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

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

Proof. We have

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

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 α , 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 α . Therefore, when α 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

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

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

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

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

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

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

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

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:

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

Noticing that for $\alpha \leq \frac{\mu}{L^2}$, $\gamma_t \leq \gamma_{t+1}$. Consolidating the terms, for sufficiently small α , we have

$$\begin{aligned}\mathbb{E} [\|\theta_{k+1} - \theta^*\|^{2n}] &\leq \sum_{t=0}^{k-1} 2^{4n-4} \left[4\alpha L L_{PH}^{(1)} \gamma_{t+1} + L_{PH}^{(0)} ((n\alpha L + 1) \gamma_{t+1} - \gamma_t) \right] (\mathbb{E} [\|\theta_t - \theta^*\|] + 1) + \gamma_0 \mathbb{E} [\|\theta_0 - \theta^*\|^{2n}] + o(\alpha) \\ &\leq \sum_{t=0}^{k-1} 2^{4n-3} \alpha L \gamma_{t+1} \left[2L_{PH}^{(1)} + nL_{PH}^{(0)} \right] (\mathbb{E} [\|\theta_t - \theta^*\|] + 1) + L_{PH}^{(0)} (\gamma_0 \|\theta_0 - \theta^*\|^{2n-1} + \gamma_k \|\theta_k - \theta^*\|^{2n-1}) \\ &\quad + \gamma_0 \mathbb{E} [\|\theta_0 - \theta^*\|^{2n}] + o(\alpha)\end{aligned}$$

Now from the induction hypothesis we know that $\mathbb{E}[\|\theta_t - \theta^*\|^2] \leq \tilde{c}_1 \cdot (1 - 2\alpha\mu)^t + \tilde{c}_2 \cdot 2\alpha L \left(L_{PH}^{(0)} + L_{PH}^{(1)} \right)$. If we plug in this upper bound into the previous inequality (also using $u \leq \frac{u^2+1}{2}$), we will have

$$\begin{aligned} \mathbb{E}[\|\theta_{k+1} - \theta^*\|^{2n}] &\leq \sum_{t=0}^{k-1} \tilde{c}_1 \cdot 2^{4n-3} \alpha L \gamma_{t+1} \left[2L_{PH}^{(1)} + nL_{PH}^{(0)} \right] (1 - 2\alpha\mu)^t \mathbb{E}[\|\theta_0 - \theta^*\|^2] + 2^{2n} n\alpha \mathbb{E}[\|\theta_0 - \theta^*\|^{2n-1}] (1 - 2n\alpha\mu)^{k-1} \\ &\quad + \gamma_0 \mathbb{E}[\|\theta_0 - \theta^*\|^{2n}] + 2^{4n-3} \alpha L \left[2L_{PH}^{(1)} + nL_{PH}^{(0)} \right] \sum_{t=0}^{k-1} \left(3 + \tilde{c}_2 \cdot \alpha L \left(L_{PH}^{(0)} + \alpha L_{PH}^{(1)} \right) \right) \gamma_{t+1} + o(\alpha) \end{aligned}$$

To simplify the above, we first notice that

$$\begin{aligned} \gamma_{t+1} (1 - 2\alpha\mu)^t &= 2n\alpha (1 - 2n\alpha\mu)^{k-t-1} (1 - 2\alpha\mu)^t \\ &\leq 2^{t+1} n\alpha (1 - 2n\alpha\mu)^{k-1} \\ &\leq 2^k n\alpha (1 - 2n\alpha\mu)^{k-1} \end{aligned}$$

Also, we have that $\sum_{k=0}^{t-1} \gamma_{t+1} = 2n\alpha \left[\frac{1 - (1 - 2n\alpha\mu)^k}{1 - (1 - 2n\alpha\mu)} \right]$. Using these we can rewrite the final error bound as

$$\begin{aligned} \mathbb{E}[\|\theta_{k+1} - \theta^*\|^{2n}] &\leq \left(k \cdot \tilde{c}_1 \cdot 2^{4n+k-3} \alpha^2 L \left[2L_{PH}^{(1)} + 2^{2n} n\alpha \mathbb{E}[\|\theta_0 - \theta^*\|^{2n-1}] + nL_{PH}^{(0)} \right] \mathbb{E}[\|\theta_0 - \theta^*\|^2] + 2n\alpha [\|\theta_0 - \theta^*\|^{2n}] \right) (1 - 2n\alpha\mu)^{k-1} \\ &\quad + \tilde{c}_2 \cdot 4^n n\alpha L \left[2\alpha L_{PH}^{(1)} + nL_{PH}^{(0)} \right] \end{aligned}$$

In the end, for $\alpha < \frac{1}{4^n}$, we have

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

in which $\tilde{C}_{n,1}$ and $\tilde{C}_{n,2}$ are $\mathcal{O}(1)$ constants.

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