

Towards better understanding of decentralized optimization using row stochastic matrix

Liyuan Liang

1 Related Works

[?, ?] suggests using only row-stochastic matrix for decentralized optimization. There are some works following this idea:...

2 Assumptions and Notations

Data heterogeneity: There exists a constant b such that $\|\nabla f_i(x) - \nabla f(x)\| \leq b^2, \forall i = 1, 2, \dots, n$.

Notation. Suppose that we have primitive row-stochastic matrix A satisfying $\pi^\top A = \pi^\top, A\mathbf{1}_n = \mathbf{1}_n$. Here π is the Perron vector whose sum is 1. We define $V_0 = I_n, D_k = \text{diag}(V_k) + \underbrace{\epsilon_k I_n}_{\text{New!}}, A_\infty = \lim_{k \rightarrow \infty} V_k = \mathbf{1}_n \pi^\top$. We use bold symbol to express parameter matrix. $\mathbf{x}^{(k)} := [x_1^{(k)T}, \dots, x_n^{(k)T}]^T \in \mathbb{R}^{n \times d}, \nabla f(\mathbf{x}^{(k)}) := \nabla f(\mathbf{x}^{(k)}) = [\nabla f_1(x_1^{(k)})^T; \dots; \nabla f_n(x_n^{(k)})^T]^T \in \mathbb{R}^{n \times d}, \mathbf{w}^{(k)} = A_\infty \mathbf{x}^{(k)}, w^{(k)} = \pi^T \mathbf{x}^{(k)},$

3 Generalized Pull-Sum Gradient Tracking

To generalize the pull-diging method, we introduce a decay weight for the nonlinear divisor.

Algorithm.

$$\mathbf{x}^{(k+1)} = A\mathbf{x}^{(k)} - \alpha \mathbf{y}^{(k)} \quad (1)$$

$$V_{k+1} = AV_k \quad (2)$$

$$\mathbf{y}^{(k+1)} = A\mathbf{y}^{(k)} + D_{k+1}^{-1} \nabla f(\mathbf{x}^{(k+1)}) - D_k^{-1} \nabla f(\mathbf{x}^{(k)}) \quad (3)$$

hereon we set $\mathbf{y}^{(0)} = \nabla f(\mathbf{x}^{(0)})$.

3.1 Estimate of Gradient Consensus Error(Lemma 7 in [?])

[LLY: Xinyi, please complete it following the proof in Lemma B.5 in my paper]

Lemma 1. When $\alpha \leq \frac{1-\beta}{12L\kappa_\pi(1+2\beta^2)(\frac{T}{1-\beta} + D_2 + 1)}$, we have:

$$\begin{aligned} \sum_{k=0}^T \|\mathbf{y}^{(k+1)} - A_\infty \mathbf{y}^{(k+1)}\|_\pi^2 &\leq \frac{2\kappa_\pi^2(1+\beta)}{(1-\beta)^4} \|\mathbf{y}^{(0)}\|_\pi^2 \\ &+ \frac{6L^2\kappa_\pi^2(1+2\beta^2)}{(1-\beta)} \left(\frac{T}{1-\beta} + D_2 + 1\right) \sum_{k=0}^T \|\mathbf{x}^{(k)} - \mathbf{w}^{(k)}\|_\pi^2 + \frac{6\alpha^2 L^2 \kappa_\pi^2}{1-\beta} \left(\frac{T}{1-\beta} + D_2 + 1\right) \sum_{k=0}^T \|A_\infty \mathbf{y}^{(k)}\|_\pi^2 \end{aligned} \quad (4)$$

Proof. To bound $\mathbf{y}^{(k-i)} - A_\infty \mathbf{y}^{(k-i)}$, we left multiply A_∞ on each side of (??), we have

$$\hat{\mathbf{y}}^{(k+1)} = \hat{\mathbf{y}}^{(k)} + A_\infty D_{k+1}^{-1} \nabla f(\mathbf{x}^{(k+1)}) - A_\infty D_k^{-1} \nabla f(\mathbf{x}^{(k)}) \quad (5)$$

Remarks. Unfold (??), we have $\hat{\mathbf{y}}^{(k)} = \hat{\mathbf{y}}_0 + A_\infty D_k^{-1} \nabla f(\mathbf{x}^{(k)}) - A_\infty D_0^{-1} \nabla f(\mathbf{x}^{(0)}) = A_\infty D_k^{-1} \nabla f_k$. When $k \rightarrow \infty$, $\epsilon_k \rightarrow 0$, $\hat{\mathbf{y}}^{(k)} \rightarrow \mathbf{1}_n \pi^\top \text{diag}(\pi)^{-1} \nabla f(\mathbf{x}^{(k)}) = \mathbf{1}_n \mathbf{1}_n^\top \nabla f(\mathbf{x}^{(k)})$. This is why we use D_k as the divisor.

Define $\Delta^{(k)} = \nabla f(\mathbf{x}^{(k+1)}) - \nabla f(\mathbf{x}^{(k)})$, $\delta_k = D_{k+1}^{-1} - D_k^{-1}$, Subtract (??) from (??), we have the following equation:

$$\begin{aligned} &\mathbf{y}^{(k+1)} - A_\infty \mathbf{y}^{(k+1)} \\ &= (A - A_\infty)(\mathbf{y}^{(k)} - A_\infty \mathbf{y}^{(k)}) + (D_{k+1}^{-1} - A_\infty D_{k+1}^{-1}) \nabla f(\mathbf{x}^{(k+1)}) - (D_k^{-1} - A_\infty D_k^{-1}) \nabla f(\mathbf{x}^{(k)}) \\ &= (A - A_\infty)(\mathbf{y}^{(k)} - A_\infty \mathbf{y}^{(k)}) + (D_{k+1}^{-1} - A_\infty D_{k+1}^{-1}) \Delta^{(k)} + (I_n - A_\infty) \delta_k \nabla f(\mathbf{x}^{(k)}) \\ &= (A - A_\infty)^2 (\mathbf{y}^{(k-1)} - A_\infty \mathbf{y}^{(k-1)}) + (I_n - A_\infty) D_{k+1}^{-1} \Delta^{(k)} + (I_n - A_\infty) [\delta_k + (A - A_\infty)(D_k^{-1} - A_\infty D_k^{-1})] \Delta^{(k-1)} \\ &\quad + [(I_n - A_\infty) \delta_k + (A - A_\infty)(I_n - A_\infty) \delta_{k-1}] \nabla f(\mathbf{x}^{(k-1)}) \\ &= \dots \\ &= (A - A_\infty)^{k+1} (\mathbf{y}^{(0)} - A_\infty \mathbf{y}^{(0)}) + \sum_{i=0}^k C_i \Delta^{(k-i)} + \sum_{i=0}^k (A - A_\infty)^i (I_n - A_\infty) \delta_{k-i} \mathbf{y}^{(0)} \\ &= [(A - A_\infty)^{k+1} + \sum_{i=0}^k (A - A_\infty)^i (I_n - A_\infty) \delta_{k-i}] \mathbf{y}^{(0)} + \sum_{i=0}^k C_i \Delta^{(k-i)} \end{aligned} \quad (6)$$

where

$$C_i = (A - A_\infty)^i (I_n - A_\infty) D_{k+1-i}^{-1} + \sum_{j=0}^i (A - A_\infty)^j (I_n - A_\infty) \delta_{k-j}$$

Suppose that we have a constant κ_π satisfying

$$\|D_k^{-1}\| \leq \kappa_\pi, \|\delta_k\| \leq \kappa_\pi \beta^k$$

Let $\beta = \|A - A_\infty\|_\pi$, then

$$\begin{aligned} \|A^k - A_\infty\|_F^2 &= \|(A - A_\infty)^k\|_F^2 = \sum_{i=1}^n \|(A - A_\infty)^k \mathbf{e}_i\|_F^2 \\ &= \sum_{i=1}^n \|\Pi^{-\frac{1}{2}} \Pi^{\frac{1}{2}} (A - A_\infty)^k \mathbf{e}_i\|_F^2 \leq \sum_{i=1}^n \|\Pi^{-\frac{1}{2}}\|_2^2 \|(A - A_\infty)^k \mathbf{e}_i\|_\pi^2 \leq \pi^{-1} \sum_{i=1}^n \|(A - A_\infty)^k\|_\pi^2 \|\mathbf{e}_i\|_\pi^2 \\ &\leq \sum_{i=1}^n \frac{1}{\pi} \beta^{2k} \pi_i = \frac{1}{\pi} \beta^{2k} \end{aligned} \quad (7)$$

Therefore, for each k and $1 \leq i, j \leq n$ denote $v_i^{(k)} = (V_k)_{ii}$. We have $|v_i^{(k)} - \pi_i| \leq \sqrt{\frac{1}{\pi}} \beta^k$.

Let $\epsilon_k = \sqrt{\frac{1}{\pi}} \beta^k$, then $\|D_k^{-1}\| \leq \frac{1}{\pi}$ and since

$$\left| \frac{1}{v_i^{(k)}} - \frac{1}{v_i^{(k+1)}} \right| = \frac{|v_i^{(k)} - v_i^{(k+1)}|}{v_i^{(k)} v_i^{(k+1)}} \leq \frac{\sqrt{\frac{1}{\pi}} \beta^k (\beta + 1)}{\pi^2}$$

we have

$$\|\delta_k\| \leq \frac{\sqrt{\frac{1}{\pi}} \beta^k (\beta + 1)}{\pi^2}$$

So we just let $\kappa_\pi = \max \left\{ \frac{1}{\pi}, \frac{1 + \beta}{\pi^{2.5}} \right\} = \frac{1 + \beta}{\pi^{2.5}}$

It can be easily verified that $\|C_i\|_\pi \leq \kappa_\pi (\beta^i + k \beta^k)$. Thus, using Jensen inequality, we have

$$\|\mathbf{y}^{(k+1)} - A_\infty \mathbf{y}^{(k+1)}\|_\pi^2 \leq \frac{\kappa_\pi^2 \beta^{k+1}}{1 - \beta} (k + 2)^2 \|\mathbf{y}^{(0)}\|_\pi^2 + \frac{\kappa_\pi^2}{1 - \beta} \sum_{i=0}^k \beta^i (1 + k \beta^{k-i})^2 \|\Delta^{(k-i)}\|_\pi^2 \quad (8)$$

Notice that

$$\begin{aligned} \|\mathbf{x}^{(k+1)} - \mathbf{w}^{(k+1)}\|^2 &= \|(A - \mathbf{1}_n \pi^T)(\mathbf{x}^{(k)} - \mathbf{w}^{(k)}) - \alpha(I_n - \mathbf{1}_n \pi^T)(\mathbf{y}^{(k)} - A_\infty \mathbf{y}^{(k)})\|^2 \\ &\leq 2\beta^2 \|\mathbf{x}^{(k)} - \mathbf{w}^{(k)}\|^2 + 2\alpha^2 \|\mathbf{y}^{(k)} - A_\infty \mathbf{y}^{(k)}\|^2 \end{aligned} \quad (9)$$

here we use the fact that $\|I_n - \mathbf{1}_n \pi^T\|_\pi = 1$. Then we have

$$\begin{aligned} \|\Delta^{(k)}\|_\pi^2 &\leq 3L^2 (\|\mathbf{x}^{(k+1)} - \mathbf{w}^{(k+1)}\|_\pi^2 + \|\mathbf{w}^{(k+1)} - \mathbf{w}^{(k)}\|_\pi^2 + \|\mathbf{x}^{(k)} - \mathbf{w}^{(k)}\|_\pi^2) \\ &\leq 3L^2 (1 + 2\beta^2) \|\mathbf{x}^{(k)} - \mathbf{w}^{(k)}\|_\pi^2 + 3\alpha^2 L^2 \|A_\infty \mathbf{y}^{(k)}\|_\pi^2 + 6\alpha^2 L^2 \|\mathbf{y}^{(k)} - A_\infty \mathbf{y}^{(k)}\|_\pi^2 \end{aligned} \quad (10)$$

Plug (??) into (??), we should have

$$\begin{aligned} &\|\mathbf{y}^{(k+1)} - A_\infty \mathbf{y}^{(k+1)}\|_\pi^2 \\ &\leq \frac{\kappa_\pi^2 \beta^{k+1}}{1 - \beta} (k + 2)^2 \|\mathbf{y}^{(0)}\|_\pi^2 + \frac{3L^2 \kappa_\pi^2 (1 + 2\beta^2)}{1 - \beta} \sum_{i=0}^k \beta^i (1 + k \beta^{k-i})^2 \|\mathbf{x}^{(i)} - \mathbf{w}^{(i)}\|_\pi^2 \\ &\quad + \frac{3\alpha^2 L^2 \kappa_\pi^2}{1 - \beta} \sum_{i=0}^k \beta^i (1 + k \beta^{k-i})^2 \|A_\infty \mathbf{y}^{(i)}\|_\pi^2 + \frac{6\alpha^2 L^2 \kappa_\pi^2 (1 + 2\beta^2)}{1 - \beta} \sum_{i=0}^k \beta^i (1 + k \beta^{k-i})^2 \|\mathbf{y}^{(i)} - A_\infty \mathbf{y}^{(i)}\|_\pi^2 \end{aligned} \quad (11)$$

Sum up (??) and we obtain:

$$\begin{aligned} \sum_{k=0}^T \|\mathbf{y}^{(k+1)} - A_\infty \mathbf{y}^{(k+1)}\|_\pi^2 &\leq \frac{\kappa_\pi^2 (1 + \beta)}{(1 - \beta)^4} \|\mathbf{y}^{(0)}\|_\pi^2 + \frac{3L^2 \kappa_\pi^2 (1 + 2\beta^2)}{1 - \beta} \left(\frac{T}{1 - \beta} + D_2 + 1 \right) \sum_{k=0}^T \beta^k \|\mathbf{x}^{(k)} - \mathbf{w}^{(k)}\|_\pi^2 \\ &\quad + \frac{3\alpha^2 L^2 \kappa_\pi^2}{1 - \beta} \left(\frac{T}{1 - \beta} + D_2 + 1 \right) \sum_{k=0}^T \beta^k \|A_\infty \mathbf{y}^{(k)}\|_\pi^2 + \frac{6\alpha^2 L^2 \kappa_\pi^2 (1 + 2\beta^2)}{1 - \beta} \left(\frac{T}{1 - \beta} + D_2 + 1 \right) \sum_{k=0}^T \beta^k \|\mathbf{y}^{(k)} - A_\infty \mathbf{y}^{(k)}\|_\pi^2 \end{aligned} \quad (12)$$

Here is the detailed calculation process.

Since

$$\begin{aligned}
\sum_{k=0}^T k(k+1)\beta^k &= \beta \left(\sum_{k=1}^T \beta^{k+1} \right)^{(2)} \\
&= \beta \frac{2 - (T+1)(T+2)\beta^T + 2T(T+2)\beta^{T+1} - (T+1)(T+2)\beta^{T+2}}{(1-\beta)^3} \\
&\leq \beta \frac{2}{(1-\beta)^3} = D_1
\end{aligned} \tag{13}$$

And

$$\sum_{k=0}^T k\beta^k = \beta \left(\sum_{k=1}^T \beta^k \right)^{(1)} = \beta \frac{1 - (T+1)\beta^T + T\beta^{T+1}}{(1-\beta)^2} \leq \frac{\beta}{(1-\beta)^2} = D_2 \tag{14}$$

Then

$$\begin{aligned}
\sum_{k=0}^T k^2\beta^k &= \beta \frac{1 + \beta - (T+1)^2\beta^T + (2T^2 + 2T - 1)\beta^{T+1} - T(T+3)\beta^{T+2}}{(1-\beta)^3} \\
&\leq \beta \frac{1 + \beta}{(1-\beta)^3} = D_3
\end{aligned} \tag{15}$$

Then

$$\begin{aligned}
\sum_{k=i}^T \beta^i (1 + k\beta^{k-i})^2 &= \sum_{k=i}^T (\beta^i + 2k\beta^k + k^2\beta^{(2k-i)}) \\
&= (T-i+1)\beta^i + \beta^i \sum_{s=0}^{T-i} s\beta^s + \beta^i \sum_{s=0}^{T-i} \beta^s \\
&\leq \beta^i (T+1-i + D_2 + \frac{i}{1-\beta}) \leq \beta^i (\frac{T}{1-\beta} + D_2 + 1)
\end{aligned} \tag{16}$$

Therefore, we have Choose $\alpha \leq ?$, then we can move the last term to left and we obtain the GCE lemma \square

3.2 Estimate of Consensus Error

[LLY: Xinyi, please complete it following the proof in Lemma B.6 in my paper]

Lemma 2. When $\alpha \leq \frac{(1-\beta)^2}{12L\kappa_\pi(1+2\beta^2)(\frac{T}{1-\beta} + D_2 + 1)\|I_n - \mathbf{1}_n\pi^T\|_\pi}$, we have:

$$\sum_{k=0}^{T+1} \|\mathbf{w}^{(k)} - \mathbf{x}^{(k)}\|^2 \leq \frac{4\alpha^2 \|I_n - \mathbf{1}_n\pi^T\|_\pi^2 \kappa_\pi^2 (1+\beta)}{(1-\beta)^6} \|\mathbf{y}^{(0)}\|_\pi^2 + \frac{12\alpha^4 L^2 \kappa_\pi^2 \|I_n - \mathbf{1}_n\pi^T\|_\pi^2}{(1-\beta)^3} (\frac{T}{1-\beta} + D_2 + 1) \sum_{k=0}^T \|A_\infty \mathbf{y}^{(k)}\|_\pi^2 \tag{17}$$

Proof. $\mathbf{w}^{(k)}$ or $w^{(k)}$ should be the true model parameter. Left multiply A_∞ on each side of (??), we have

$$\mathbf{w}^{(k+1)} = \mathbf{w}^{(k)} - \alpha A_\infty \mathbf{y}^{(k)} \tag{18}$$

To estimate consensus error, we unfold the iteration:

$$\begin{aligned}
\mathbf{x}^{(k+1)} - \mathbf{w}^{(k+1)} &= (A - \mathbf{1}_n \pi^T)(\mathbf{x}^{(k)} - \mathbf{w}^{(k)}) - \alpha(I_n - \mathbf{1}_n \pi^T)(\mathbf{y}^{(k)} - A_\infty \mathbf{y}^{(k)}) \\
&= \dots \\
&= -\alpha \sum_{i=0}^k (A - \mathbf{1}_n \pi^T)^i (I_n - \mathbf{1}_n \pi^T)(\mathbf{y}^{(k-i)} - A_\infty \mathbf{y}^{(k-i)})
\end{aligned} \tag{19}$$

Then we can use Jensen inequality

$$\|\mathbf{x}^{(k+1)} - \mathbf{w}^{(k+1)}\|^2 \leq \alpha^2 \sum_{i=0}^k \frac{\beta^i}{1-\beta} \|I_n - \mathbf{1}_n \pi^T\|_\pi^2 \|\mathbf{y}^{(k-i)} - A_\infty \mathbf{y}^{(k-i)}\|_\pi^2 \tag{20}$$

Sum up, we have

$$\begin{aligned}
\sum_{k=0}^{T+1} \|\mathbf{w}^{(k)} - \mathbf{x}^{(k)}\|^2 &\leq \alpha^2 \|I_n - \mathbf{1}_n \pi^T\|_\pi^2 \sum_{k=0}^T \sum_{i=0}^k \frac{\beta^{k-i}}{1-\beta} \|\mathbf{y}^{(i)} - A_\infty \mathbf{y}^{(i)}\|_\pi^2 \\
&\leq \frac{\alpha^2 \|I_n - \mathbf{1}_n \pi^T\|_\pi^2}{(1-\beta)^2} \sum_{k=0}^T \|\mathbf{y}^{(k)} - A_\infty \mathbf{y}^{(k)}\|_\pi^2 \\
&\leq \frac{\alpha^2 \|I_n - \mathbf{1}_n \pi^T\|_\pi^2}{(1-\beta)^2} \left(\frac{2\kappa_\pi^2(1+\beta)}{(1-\beta)^4} \|\mathbf{y}^{(0)}\|_\pi^2 + \frac{6L^2\kappa_\pi^2(1+2\beta^2)}{(1-\beta)} \left(\frac{T}{1-\beta} + D_2 + 1 \right) \sum_{k=0}^T \|\mathbf{x}^{(k)} - \mathbf{w}^{(k)}\|_\pi^2 \right. \\
&\quad \left. + \frac{6\alpha^2 L^2 \kappa_\pi^2}{1-\beta} \left(\frac{T}{1-\beta} + D_2 + 1 \right) \sum_{k=0}^T \|A_\infty \mathbf{y}^{(k)}\|_\pi^2 \right)
\end{aligned} \tag{21}$$

□

3.3 Descent Lemma

Lemma 3. When $\alpha \leq ?$, we have:

$$f(w^{(k+1)}) \leq f(w^{(k)}) - \frac{\alpha}{4c_{k,1}} \|\pi^T \mathbf{y}^{(k)}\|^2 - \frac{\alpha}{2c_{k,1}} \|\mathbf{1}_n^T \nabla f(\mathbf{w}^{(k)})\|^2 + \frac{\alpha L^2 c_{k,2}}{n c_{k,1}} \|\mathbf{x}^{(k)} - \mathbf{w}^{(k)}\|^2 + \frac{\alpha b^2 c_{k,2}}{c_{k,1}} \tag{22}$$

Define $c_{k,1} = \sum_{i=1}^n \frac{\pi_i}{D_{k,i}}$, $c_{k,2} = \sum_{i=1}^n \frac{\pi_i^2}{D_{k,i}^2}$, then we have

$$\begin{aligned}
&\|\pi^T \mathbf{y}^{(k)} - \frac{c_{k,1}}{n} \mathbf{1}_n^T \nabla f(\mathbf{w}^{(k)})\|^2 \\
&= \left\| \sum_{i=1}^n \frac{\pi_i}{D_{k,i}} (\nabla f(x_i^{(k)}) - \frac{1}{n} \mathbf{1}_n^T \nabla f(\mathbf{w}^{(k)})) \right\|^2 \\
&\leq 2 \left\| \sum_{i=1}^n \frac{\pi_i}{D_{k,i}} (\nabla f_i(x_i^{(k)}) - \nabla f(x_i^{(k)})) \right\|^2 + 2 \left\| \sum_{i=1}^n \frac{\pi_i}{D_{k,i}} \sum_{j=1}^n \frac{1}{n} (\nabla f(x_j^{(k)}) - \nabla f(w_j^{(k)})) \right\|^2 \\
&= 2 \left\| \sum_{i=1}^n \frac{\pi_i}{D_{k,i}} (\nabla f_i(x_i^{(k)}) - \nabla f(x_i^{(k)})) \right\|^2 + \frac{2c_{k,2}}{n^2} \|\mathbf{1}_n^T (\nabla f(\mathbf{x}^{(k)}) - \nabla f(\mathbf{w}^{(k)}))\|^2 \\
&\leq 2b^2 c_{k,2} + \frac{2L^2 c_{k,2}}{n} \|\mathbf{x}^{(k)} - \mathbf{w}^{(k)}\|^2
\end{aligned} \tag{23}$$

Using L-smoothness inequality, we have

$$\begin{aligned}
f(w^{(k+1)}) &\leq f(w^{(k)}) - \alpha \langle \pi^T \mathbf{y}^{(k)}, \frac{1}{n} \mathbf{1}_n^T \nabla f(\mathbf{w}^k) \rangle + \frac{\alpha^2 L}{2} \|\pi^T \mathbf{y}^{(k)}\|^2 \\
&= f(w^{(k)}) - \frac{\alpha}{c_{k,1}} \langle \pi^T D_k^{-1} \nabla f(\mathbf{x}^{(k)}), \frac{c_{k,1}}{n} \mathbf{1}_n^T \nabla f(\mathbf{w}^{(k)}) \rangle + \frac{\alpha^2 L}{2} \|\pi^T D_k^{-1} \nabla f(\mathbf{x}^{(k)})\|^2 \\
&= f(w^{(k)}) - \frac{\alpha}{2c_{k,1}} \|\pi^T D_k^{-1} \nabla f(\mathbf{x}^{(k)})\|^2 - \frac{\alpha c_{k,1}}{2n^2} \|\mathbf{1}_n^T \nabla f(\mathbf{w}^k)\|^2 \\
&\quad + \frac{\alpha}{2c_{k,1}} \|\pi^T \mathbf{y}^{(k)} - \frac{c_{k,1}}{n} \mathbf{1}_n^T \nabla f(\mathbf{w}^{(k)})\|^2 + \frac{\alpha^2 L}{2} \|\pi^T D_k^{-1} \nabla f(\mathbf{x}^{(k)})\|^2 \\
&\stackrel{(a)}{\leq} f(w^{(k)}) - \frac{\alpha - c_{k,1} \alpha^2 L}{2c_{k,1}} \|\pi^T D_k^{-1} \nabla f(\mathbf{x}^{(k)})\|^2 - \frac{\alpha c_{k,1}}{2n^2} \|\mathbf{1}_n^T \nabla f(\mathbf{w}^k)\|^2 \\
&\quad + \frac{\alpha}{2c_{k,1}} (2b^2 c_{k,2} + \frac{2L^2 c_{k,2}}{n} \|\mathbf{x}^{(k)} - \mathbf{w}^{(k)}\|^2) \\
&\stackrel{(b)}{\leq} f(w^{(k)}) - \frac{\alpha}{4c_{k,1}} \|\pi^T \mathbf{y}^{(k)}\|^2 - \frac{\alpha c_{k,1}}{2n^2} \|\mathbf{1}_n^T \nabla f(\mathbf{w}^k)\|^2 + \frac{\alpha L^2 c_{k,2}}{nc_{k,1}} \|\mathbf{x}^{(k)} - \mathbf{w}^{(k)}\|^2 \\
&\quad + \frac{\alpha b^2 c_{k,2}}{c_{k,1}}
\end{aligned} \tag{24}$$

(a) uses (??), (b) holds when $\alpha \leq \frac{1}{2c_{k,1}L}$.

3.4 Main Theorem

[LLY: Please complete it following the proof of Theorem 4 in my paper.]

Theorem 1. *When*

Since $|v_i^{(k)} - \pi_i| \leq \sqrt{\frac{1}{\pi}} \beta^k$, we have $n \frac{\bar{\pi}}{\bar{\pi} + \sqrt{\frac{1}{\pi}} \beta} \leq c_{k,i} \leq n$. And notice that $c_{k,2} \leq c_{k,1}^2 \leq nc_{k,2}$

$$\begin{aligned}
\|\mathbf{1}_n^T \nabla f(\mathbf{w}^k)\|^2 &\leq \frac{2n^2}{\alpha c_{k,1}} (f(w^{(k)}) - f(w^{(k+1)})) - \frac{n^2}{2c_{k,1}^2} \|\pi^T \mathbf{y}^{(k)}\|^2 + \frac{2nL^2 c_{k,2}}{c_{k,1}^2} \|\mathbf{x}^{(k)} - \mathbf{w}^{(k)}\|^2 + \frac{2b^2 n^2 c_{k,2}}{c_{k,1}^2} \\
&\leq \frac{2n(\bar{\pi} + \sqrt{\frac{1}{\pi}} \beta)}{\alpha \bar{\pi}} (f(w^{(k)}) - f(w^{(k+1)})) - \frac{1}{2} \|\pi^T \mathbf{y}^{(k)}\|^2 + 2nL^2 \|\mathbf{x}^{(k)} - \mathbf{w}^{(k)}\|^2 + 2b^2 n^2 \quad (25)
\end{aligned}$$

It then follows

$$\begin{aligned}
\sum_{k=0}^{T+1} \|\mathbf{1}_n^T \nabla f(\mathbf{w}^k)\|^2 &\leq \frac{2n(\bar{\pi} + \sqrt{\frac{1}{\pi}}\beta)}{\alpha\bar{\pi}} \triangle + 2nL^2 \sum_{k=0}^{T+1} \|\mathbf{x}^{(k)} - \mathbf{w}^{(k)}\|^2 + 2b^2n^2(T+2) \\
&\leq \frac{2n(\bar{\pi} + \sqrt{\frac{1}{\pi}}\beta)}{\alpha\bar{\pi}} \triangle + 2nL^2 \left(\frac{4\alpha^2 \|I_n - \mathbf{1}_n \pi^T\|_\pi^2 \kappa_\pi^2 (1+\beta)}{(1-\beta)^6} \|\mathbf{y}^{(0)}\|_\pi^2 + \right. \\
&\quad \left. \frac{12\alpha^4 L^2 \kappa_\pi^2 \|I_n - \mathbf{1}_n \pi^T\|_\pi^2}{(1-\beta)^3} \left(\frac{T}{1-\beta} + D_2 + 1 \right) \sum_{k=0}^{T+1} \|A_\infty \mathbf{y}^{(k)}\|_\pi^2 \right) + 2b^2n^2(T+2) - \frac{1}{2} \sum_{k=0}^{T+1} \|\pi^T \mathbf{y}^{(k)}\|^2 \\
&= \frac{2n(\bar{\pi} + \sqrt{\frac{1}{\pi}}\beta)}{\alpha\bar{\pi}} \triangle + 2nL^2 \frac{4\alpha^2 \|I_n - \mathbf{1}_n \pi^T\|_\pi^2 \kappa_\pi^2 (1+\beta)}{(1-\beta)^6} \|\mathbf{y}^{(0)}\|_\pi^2 \\
&\quad - \left(\frac{1}{2} - \frac{12\alpha^4 L^2 \kappa_\pi^2 \|I_n - \mathbf{1}_n \pi^T\|_\pi^2}{(1-\beta)^3} \left(\frac{T}{1-\beta} + D_2 + 1 \right) \right) \sum_{k=0}^{T+1} \|\pi^T \mathbf{y}^{(k)}\|^2 + 2b^2n^2(T+2) \\
&\leq \frac{2n(\bar{\pi} + \sqrt{\frac{1}{\pi}}\beta)}{\alpha\bar{\pi}} \triangle + 2nL^2 \frac{4\alpha^2 \|I_n - \mathbf{1}_n \pi^T\|_\pi^2 \kappa_\pi^2 (1+\beta)}{(1-\beta)^6} \|\mathbf{y}^{(0)}\|_\pi^2 + 2b^2n^2(T+2) \tag{26}
\end{aligned}$$

Let

Then

$$\frac{\sum_{k=0}^{T+1} \|\mathbf{1}_n^T \nabla f(\mathbf{w}^k)\|^2}{T+2} \leq \frac{2n(\bar{\pi} + \sqrt{\frac{1}{\pi}}\beta)}{\alpha\bar{\pi}(T+2)} \triangle + 2nL^2 \frac{4\alpha^2 \|I_n - \mathbf{1}_n \pi^T\|_\pi^2 \kappa_\pi^2 (1+\beta)}{(1-\beta)^6(T+2)} \|\mathbf{y}^{(0)}\|_\pi^2 + 2b^2n^2 \tag{27}$$