# Supplementary material for the paper: "Linear Convergence of Randomized Primal-Dual Coordinate Method for Large-scale Linear Constrained Convex Programming"

First of all, we have the following observations:

In algorithm RPDC, the indices i(k), k = 0, 1, 2, ... are random variables. After k iterations, RPDC method generates a random output  $(u^{k+1}, p^{k+1})$ . Recall the definition of filtration  $\mathcal{F}_k$  which is generated by the random variable i(0), i(1), ..., i(k), i.e.,

$$\mathcal{F}_k \stackrel{def}{=} \{i(0), i(1), \dots, i(k)\}, \mathcal{F}_k \subset \mathcal{F}_{k+1}.$$

Additionally,  $\mathcal{F} = (\mathcal{F}_k)_{k \in \mathbb{N}}$ ,  $\mathbb{E}_{\mathcal{F}_{k+1}} = \mathbb{E}(\cdot | \mathcal{F}_k)$  is the conditional expectation w.r.t.  $\mathcal{F}_k$  and the conditional expectation in term of i(k) given  $i(0), i(1), \ldots, i(k-1)$  as  $\mathbb{E}_{i(k)}$ .

Knowing  $\mathcal{F}_{k-1} = \{i(0), i(1), \dots, i(k-1)\}$ , we have:

$$\mathbb{E}_{i(k)}\langle \nabla_{i(k)}G(u^k), (u^k - u)_{i(k)}\rangle = \frac{1}{N}\langle \nabla G(u^k), u^k - u\rangle \ge \frac{1}{N} \big[ G(u^k) - G(u) \big], \tag{A.1}$$

$$\mathbb{E}_{i(k)} \left[ J_{i(k)}(u_{i(k)}^k) - J_{i(k)}(u_{i(k)}) \right] = \frac{1}{N} \left[ J(u^k) - J(u) \right], \tag{A.2}$$

and

$$\mathbb{E}_{i(k)}\langle q^k, A_{i(k)}(u^k - u)_{i(k)} \rangle = \frac{1}{N} \langle q^k, A(u^k - u) \rangle.$$
(A.3)

Secondly, reconsidering the point  $T(w^k) = (T_u(w^k), T_p(w^k))$  generated by one deterministic iteration of APP-AL (Cohen & Zhu, 1984) for given  $w^k$ ,

$$\begin{aligned} & \textbf{APP-AL} \\ & \left\{ \begin{array}{l} T_u(w^k) = \arg\min_{u \in \mathbf{U}} \langle \nabla G(u^k), u \rangle + J(u) + \langle q^k, Au \rangle + \frac{1}{\epsilon} D(u, u^k); \\ & T_p(w^k) = p^k + \gamma \left[ A T_u(w^k) - b \right], \end{array} \right. \end{aligned}$$

with  $q^k = p^k + \gamma(Au^k - b)$ , we have the following observations. The convex combination of  $u^k$  and  $T_u(w^k)$  provides the expected value of  $u^{k+1}$  as following.

$$\mathbb{E}_{i(k)}u^{k+1} = \frac{1}{N}T_u(w^k) + (1 - \frac{1}{N})u^k,$$
(A.4)

or

$$T_u(w^k) = N \mathbb{E}_{i(k)} u^{k+1} - (N-1)u^k.$$
(A.5)

Moreover, the point  $T(w^k)$  satisfies that: for any  $(u, p) \in \mathbf{U} \times \mathbf{R}^m$ ,

$$\begin{cases} \langle \nabla G(u^k), u - T_u(w^k) \rangle + J(u) - J(T_u(w^k)) + \langle q^k, A(u - T_u(w^k)) \rangle \\ + \frac{1}{\epsilon} \langle \nabla K(T_u(w^k)) - \nabla K(u^k), u - T_u(w^k) \rangle \geq 0, \\ \gamma \left[ A T_u(w^k) - b \right] = T_p(w^k) - p^k. \end{cases}$$
(A.6)

#### 1. Proof of Lemma 1

*Proof.* Take  $w' = w^*$  in (9), we have that

$$\begin{split} \Lambda(w,w^*) &= D(u^*,u) + \frac{\epsilon}{2N\rho} \|p - p^*\|^2 + \frac{\epsilon(N-1)}{N} [L(u,p) - L(u^*,p^*)] + \frac{\epsilon(N-2)\gamma}{2N} \|Au - b\|^2 \\ &= D(u^*,u) + \frac{\epsilon}{2N\rho} \|p - p^*\|^2 + \frac{\epsilon(N-1)}{N} [L(u,p^*) - L(u^*,p^*)] + \frac{\epsilon(N-1)}{N} \langle p - p^*, Au - b \rangle \\ &+ \frac{\epsilon(N-2)\gamma}{2N} \|Au - b\|^2. \end{split}$$
(A.7)

(i) Since  $L(u, p^*) - L(u^*, p^*) \ge 0$  and  $\frac{1}{2\gamma} \|p - p^*\|^2 + \frac{\gamma}{2} \|Au - b\|^2 + \langle p - p^*, Au - b \rangle \ge 0$ , (A.7) follows that

$$\Lambda(w,w^*) \ge D(u^*,u) + \frac{\epsilon}{2N\rho} \|p - p^*\|^2 - \frac{\epsilon(N-1)}{2N\gamma} \|p - p^*\|^2 - \frac{\epsilon\gamma}{2N} \|Au - b\|^2.$$

From Assumption 2, we have  $D(u^*, u) \ge \frac{\beta}{2} ||u - u^*||^2$ . Together with the fact  $Au^* = b$  and  $\rho < \frac{2\gamma}{2N-1}$ , above inequality follows that

$$\Lambda(w, w^*) \ge d_1 \|w - w^*\|^2$$

with 
$$d_1 = \min\left\{\frac{1}{2N}[N\beta - \epsilon\gamma\lambda_{\max}(A^{\top}A)], \frac{\epsilon}{4N\gamma}\right\}.$$

(ii) By Young's inequality, (A.7) follows that

$$\begin{split} \Lambda(w,w^*) &\leq D(u^*,u) + \frac{\epsilon}{2N\rho} \|p - p^*\|^2 + \frac{\epsilon(N-1)}{N} [L(u,p^*) - L(u^*,p^*)] \\ &+ \frac{\epsilon(N-1)}{N} [\frac{1}{2\gamma} \|p - p^*\|^2 + \frac{\gamma}{2} \|Au - b\|^2] + \frac{\epsilon(N-2)\gamma}{2N} \|Au - b\|^2. \end{split}$$

From Assumption 2, we have  $D(u^*, u) \leq \frac{B}{2} ||u - u^*||^2$ . Together with the fact  $Au^* = b$  and  $2\gamma > (2N - 1)\rho$ , above inequality follows that

$$\begin{split} \Lambda(w,w^*) &\leq d_2 \|w - w^*\|^2 + \frac{\epsilon(N-1)}{N} [L(u,p^*) - L(u^*,p^*)], \end{split}$$
 with  $d_2 &= \max\left\{ \frac{(4N-3)\epsilon}{(4N-2)N\rho}, \frac{NB + \epsilon(2N-3)\gamma\lambda_{\max}(A^\top A)}{2N} \right\}. \end{split}$ 

(iii) By the definition of  $\Lambda(w, w')$ , we have

$$\Lambda(w,w') \geq \frac{\epsilon(N-1)}{N} [L(u,p) - L(u^*,p^*)] + \frac{\epsilon(N-2)\gamma}{2N} ||Au - b||^2 \\
= \frac{\epsilon(N-1)}{N} [L(u,p) - L(u,p^*)] + \frac{\epsilon(N-1)}{N} [L(u,p^*) - L(u^*,p^*)] + \frac{\epsilon(N-2)\gamma}{2N} ||Au - b||^2 \\
\geq \frac{\epsilon(N-1)}{N} [L(u,p) - L(u,p^*)] + \frac{\epsilon(N-2)\gamma}{2N} ||Au - b||^2 \\
= \frac{\epsilon(N-1)}{N} \langle p - p^*, Au - b \rangle + \frac{\epsilon(N-2)\gamma}{2N} ||Au - b||^2 \\
\geq -d_3 ||p - p^*||^2,$$
(A.8)

with  $d_3 = \frac{\epsilon (N-1)^2}{2\gamma N(N-2)}$ .

#### 2. Proof of Lemma 2

*Proof.* Step 1: Estimate  $\frac{\epsilon}{N} \mathbb{E}_{i(k)} [L(u^{k+1}, q^k) - L(u, q^k)];$ For all  $u \in \mathbf{U}$ , the unique solution  $u^{k+1}$  of the primal problem of RPDC is characterized by the following variational inequality:

$$\begin{split} \langle \nabla_{i(k)} G(u^k), (u^{k+1} - u)_{i(k)} \rangle + J_{i(k)}(u^{k+1}_{i(k)}) - J_{i(k)}(u_{i(k)}) + \langle q^k, A_{i(k)}(u^{k+1} - u)_{i(k)} \rangle \\ + \frac{1}{\epsilon} \langle \nabla K(u^{k+1}) - \nabla K(u^k), u^{k+1} - u \rangle \le 0, \end{split}$$

which follows that

$$\langle \nabla_{i(k)} G(u^k), \left(u^k - u - (u^k - u^{k+1})\right)_{i(k)} \rangle + J_{i(k)}(u^k_{i(k)}) - J_{i(k)}(u_{i(k)}) - \left(J_{i(k)}(u^k_{i(k)}) - J_{i(k)}(u^{k+1}_{i(k)})\right) \\ + \langle q^k, A_{i(k)} \left(u^k - u - (u^k - u^{k+1})\right)_{i(k)} \rangle + \frac{1}{\epsilon} \langle \nabla K(u^{k+1}) - \nabla K(u^k), u^{k+1} - u \rangle \le 0.$$
(A.9)

Observing that for any separable mapping  $\psi(u) = \sum_{i=1}^{N} \psi_i(u_i)$ , we have  $\psi_{i(k)}(u_{i(k)}^k) - \psi_{i(k)}(u_{i(k)}^{k+1}) = \psi(u^k) - \psi(u^{k+1})$ . Therefore, (A.9) follows that

$$\langle \nabla_{i(k)} G(u^{k}), (u^{k} - u)_{i(k)} \rangle + J_{i(k)}(u^{k}_{i(k)}) - J_{i(k)}(u_{i(k)}) + \langle q^{k}, A_{i(k)}(u^{k} - u)_{i(k)} \rangle$$

$$\leq \langle \nabla G(u^{k}), u^{k} - u^{k+1} \rangle + J(u^{k}) - J(u^{k+1}) + \langle q^{k}, A(u^{k} - u^{k+1}) \rangle$$

$$+ \frac{1}{\epsilon} \langle \nabla K(u^{k+1}) - \nabla K(u^{k}), u - u^{k+1} \rangle.$$
(A.10)

Taking expectation with respect to i(k) on both side of (A.10), together the condition expectation (A.1)-(A.3), we get

$$\frac{1}{N} \left[ L(u^{k}, q^{k}) - L(u, q^{k}) \right] \leq \mathbb{E}_{i(k)} \left\{ \langle \nabla G(u^{k}), u^{k} - u^{k+1} \rangle + J(u^{k}) - J(u^{k+1}) + \langle q^{k}, A(u^{k} - u^{k+1}) \rangle + \frac{1}{\epsilon} \langle \nabla K(u^{k+1}) - \nabla K(u^{k}), u - u^{k+1} \rangle \right\}. \quad (A.11)$$

or

$$\frac{1}{N} \mathbb{E}_{i(k)} \left[ L(u^{k+1}, q^k) - L(u, q^k) \right] \leq \mathbb{E}_{i(k)} \left\{ \underbrace{\langle \nabla G(u^k), u^k - u^{k+1} \rangle}_{\mathfrak{a}_1} + J(u^k) - J(u^{k+1}) + \frac{\langle q^k, A(u^k - u^{k+1}) \rangle}{\mathfrak{a}_1} + \frac{1}{N} \left[ L(u^{k+1}, q^k) - L(u^k, q^k) \right] + \underbrace{\frac{1}{\epsilon} \langle \nabla K(u^{k+1}) - \nabla K(u^k), u - u^{k+1} \rangle}_{\mathfrak{a}_2} \right\}.$$
(A.12)

By the gradient Lipschitz of G, term  $a_1$  in (A.12) is bounded by

$$\mathfrak{a}_1 = \langle \nabla G(u^k), u^k - u^{k+1} \rangle \le G(u^k) - G(u^{k+1}) + \frac{B_G}{2} \| u^k - u^{k+1} \|^2.$$
(A.13)

The simple algebraic operation and Assumption 2 follows that

$$\mathfrak{a}_{2} = \frac{1}{\epsilon} \langle \nabla K(u^{k+1}) - \nabla K(u^{k}), u - u^{k+1} \rangle = \frac{1}{\epsilon} \left[ D(u, u^{k}) - D(u, u^{k+1}) - D(u^{k+1}, u^{k}) \right] \\ \leq \frac{1}{\epsilon} \left[ D(u, u^{k}) - D(u, u^{k+1}) \right] - \frac{\beta}{2\epsilon} \|u^{k} - u^{k+1}\|^{2}.$$
 (A.14)

Combining (A.12)-(A.14), we obtain that

$$\frac{\epsilon}{N} \mathbb{E}_{i(k)} \left[ L(u^{k+1}, q^k) - L(u, q^k) \right] \leq \left[ D(u, u^k) - \mathbb{E}_{i(k)} D(u, u^{k+1}) \right] + \mathbb{E}_{i(k)} \left\{ \frac{\epsilon(N-1)}{N} \underbrace{\left[ L(u^k, q^k) - L(u^{k+1}, q^k) \right]}_{\mathfrak{a}_3} - \frac{\beta - \epsilon B_G}{2} \| u^k - u^{k+1} \|^2 \right\}$$
(A.15)

Since  $p^{k+1} = p^k + \rho(Au^{k+1} - b)$  and  $q^k = p^k + \gamma(Au^k - b)$ , term  $\mathfrak{a}_3$  in (A.15) follows that

$$\begin{aligned} \mathfrak{a}_{3} &= L(u^{k},q^{k}) - L(u^{k+1},q^{k}) \\ &= L(u^{k},p^{k}) - L(u^{k+1},p^{k+1}) + \langle q^{k} - p^{k},Au^{k} - b \rangle + \langle p^{k+1} - q^{k},Au^{k+1} - b \rangle \\ &= L(u^{k},p^{k}) - L(u^{k+1},p^{k+1}) + \gamma \|Au^{k} - b\|^{2} + \rho \|Au^{k+1} - b\|^{2} - \gamma \langle Au^{k} - b,Au^{k+1} - b \rangle \\ &= L(u^{k},p^{k}) - L(u^{k+1},p^{k+1}) + \frac{\gamma}{2} \|Au^{k} - b\|^{2} + (\rho - \frac{\gamma}{2}) \|Au^{k+1} - b\|^{2} + \frac{\gamma}{2} \|A(u^{k} - u^{k+1})\|^{2} \\ &\leq L(u^{k},p^{k}) - L(u^{k+1},p^{k+1}) + \frac{\gamma}{2} \|Au^{k} - b\|^{2} + (\rho - \frac{\gamma}{2}) \|Au^{k+1} - b\|^{2} \\ &+ \frac{\gamma \lambda_{\max}(A^{\top}A)}{2} \|u^{k} - u^{k+1}\|^{2}. \end{aligned}$$
(A.16)

#### Combining (A.15)-(A.16), we have that

$$\frac{\epsilon}{N} \mathbb{E}_{i(k)} \left[ L(u^{k+1}, q^k) - L(u, q^k) \right] \leq \left[ D(u, u^k) - \mathbb{E}_{i(k)} D(u, u^{k+1}) \right] + \mathbb{E}_{i(k)} \left\{ \frac{\epsilon(N-1)}{N} \left[ L(u^k, p^k) - L(u^{k+1}, p^{k+1}) \right] - \frac{\beta - \epsilon \left[ B_G + \frac{N-1}{N} \gamma \lambda_{\max}(A^\top A) \right]}{2} \| u^k - u^{k+1} \|^2 + \frac{\epsilon \gamma(N-1)}{2N} \| A u^k - b \|^2 + \frac{\epsilon(2\rho - \gamma)(N-1)}{2N} \| A u^{k+1} - b \|^2 \right\}$$
(A.17)

Step 2: Estimate  $\frac{\epsilon}{N}\mathbb{E}_{i(k)}\big[L(u^{k+1},p)-L(u^{k+1},q^k)\big]$ 

$$\begin{split} L(u^{k+1},p) - L(u^{k+1},q^k) &= \langle p - q^k, Au^{k+1} - b \rangle \\ &= \frac{1}{\rho} \langle p - p^k, p^{k+1} - p^k \rangle - \gamma \langle Au^k - b, Au^{k+1} - b \rangle \\ &= \frac{1}{2\rho} \left[ \|p - p^k\|^2 - \|p - p^{k+1}\|^2 + \|p^k - p^{k+1}\|^2 \right] - \gamma \langle Au^k - b, Au^{k+1} - b \rangle \\ &= \frac{1}{2\rho} \left[ \|p - p^k\|^2 - \|p - p^{k+1}\|^2 + \|p^k - p^{k+1}\|^2 \right] + \frac{\gamma}{2} \|A(u^k - u^{k+1})\|^2 \\ &- \frac{\gamma}{2} \|Au^k - b\|^2 - \frac{\gamma}{2} \|Au^{k+1} - b\|^2 \\ &= \frac{1}{2\rho} \left[ \|p - p^k\|^2 - \|p - p^{k+1}\|^2 \right] + \frac{\gamma}{2} \|A(u^k - u^{k+1})\|^2 \\ &- \frac{\gamma}{2} \|Au^k - b\|^2 + \frac{\rho - \gamma}{2} \|Au^{k+1} - b\|^2 \quad (\text{since } p^{k+1} = p^k + \rho(Au^{k+1} - b).) \\ &\leq \frac{1}{2\rho} \left[ \|p - p^k\|^2 - \|p - p^{k+1}\|^2 \right] + \frac{\gamma\lambda_{\max}(A^{\top}A)}{2} \|u^k - u^{k+1}\|^2 \\ &- \frac{\gamma}{2} \|Au^k - b\|^2 + \frac{\rho - \gamma}{2} \|Au^{k+1} - b\|^2 \end{split}$$
(A.18)

Multiply  $\frac{\epsilon}{N}$  on both side of above inequality, we obtain that:  $\forall p \in \mathbf{R}^m$ 

$$\frac{\epsilon}{N} \left[ L(u^{k+1}, p) - L(u^{k+1}, q^k) \right] \leq \frac{\epsilon}{2N\rho} \left[ \|p - p^k\|^2 - \|p - p^{k+1}\|^2 \right] + \frac{\epsilon \frac{1}{N} \gamma \lambda_{\max}(A^\top A)}{2} \|u^k - u^{k+1}\|^2 - \frac{\epsilon \gamma}{2N} \|Au^k - b\|^2 + \frac{\epsilon(\rho - \gamma)}{2N} \|Au^{k+1} - b\|^2.$$
(A.19)

Taking expectation with respect to i(k) on both side of inequality (A.19), we have

$$\frac{\epsilon}{N} \mathbb{E}_{i(k)} \left[ L(u^{k+1}, p) - L(u^{k+1}, q^k) \right] \leq \frac{\epsilon}{2N\rho} \left[ \|p - p^k\|^2 - \mathbb{E}_{i(k)} \|p - p^{k+1}\|^2 \right] + \frac{\epsilon \frac{1}{N} \gamma \lambda_{\max}(A^\top A)}{2} \mathbb{E}_{i(k)} \|u^k - u^{k+1}\|^2 - \frac{\epsilon \gamma}{2N} \|Au^k - b\|^2 + \frac{\epsilon(\rho - \gamma)}{2N} \mathbb{E}_{i(k)} \|Au^{k+1} - b\|^2.$$
(A.20)

Step 3: Estimate the variance of  $\Lambda(w^k, w)$ .

Summing inequalities (A.17) and (A.20), with  $d_4 = \frac{\min\left\{\frac{\beta - \epsilon[B_G + \gamma\lambda_{\max}(A^{\top}A)]}{2}, \frac{\epsilon[2\gamma - (2N-1)\rho]}{2N}\right\}}{\max\{N^2 + 2\gamma^2(N^2 + 2)\lambda_{\max}(A^{\top}A), 4\gamma^2\}}$ , we have that

$$\Lambda(w^{k},w) - \mathbb{E}_{i(k)}\Lambda(w^{k+1},w) \\
\geq \mathbb{E}_{i(k)}\left\{\frac{\epsilon}{N}\left[L(u^{k+1},p) - L(u,q^{k})\right] + \frac{\beta - \epsilon[B_{G} + \gamma\lambda_{\max}(A^{\top}A)]}{2} \|u^{k} - u^{k+1}\|^{2} + \frac{\epsilon[2\gamma - (2N-1)\rho]}{2N}\|Au^{k+1} - b\|^{2}\right\} \\
\geq \mathbb{E}_{i(k)}\left\{\frac{\epsilon}{N}\left[L(u^{k+1},p) - L(u,q^{k})\right] + d_{4}\left[\left(N^{2} + 2\gamma^{2}(N^{2} + 2)\lambda_{\max}(A^{\top}A)\right) \|u^{k} - u^{k+1}\|^{2} + 4\gamma^{2}\|Au^{k+1} - b\|^{2}\right]\right\} \\
\geq \mathbb{E}_{i(k)}\left\{\frac{\epsilon}{N}\left[L(u^{k+1},p) - L(u,q^{k})\right] + d_{4}\left[\left(1 + 2\gamma^{2}\lambda_{\max}(A^{\top}A)\right)N^{2}\|u^{k} - u^{k+1}\|^{2} + 4\gamma^{2}\|A(u^{k} - u^{k+1})\|^{2} + \|Au^{k+1} - b\|^{2}\right]\right\} \\
\geq \mathbb{E}_{i(k)}\left\{\frac{\epsilon}{N}\left[L(u^{k+1},p) - L(u,q^{k})\right] + d_{4}\left[\left(1 + 2\gamma^{2}\lambda_{\max}(A^{\top}A)\right)N^{2}\|u^{k} - u^{k+1}\|^{2} + 2\gamma^{2}\|Au^{k} - b\|^{2}\right]\right\} \\
= \frac{\epsilon}{N}\mathbb{E}_{i(k)}\left[L(u^{k+1},p) - L(u,q^{k})\right] + d_{4}\left[\left(1 + 2\gamma^{2}\lambda_{\max}(A^{\top}A)\right)N^{2}\mathbb{E}_{i(k)}\|u^{k} - u^{k+1}\|^{2} + 2\gamma^{2}\|Au^{k} - b\|^{2}\right]. \tag{A.21}$$

By Jensen's inequality, (A.21) follows that

$$\Lambda(w^{k}, w) - \mathbb{E}_{i(k)}\Lambda(w^{k+1}, w) \geq \frac{\epsilon}{N} \mathbb{E}_{i(k)} \left[ L(u^{k+1}, p) - L(u, q^{k}) \right] \\
+ d_{4} \left[ \left( 1 + 2\gamma^{2}\lambda_{\max}(A^{\top}A) \right) N^{2} \| u^{k} - \mathbb{E}_{i(k)}u^{k+1} \|^{2} + 2\gamma^{2} \| Au^{k} - b \|^{2} \right] (A.22)$$

Since  $\mathbb{E}_{i(k)} u^{k+1} - u^k = \frac{1}{N} [T_u(w^k) - u^k]$  in (A.4), (A.22) yields that

$$\Lambda(w^{k}, w) - \mathbb{E}_{i(k)}\Lambda(w^{k+1}, w) \geq \frac{\epsilon}{N} \mathbb{E}_{i(k)} \left[ L(u^{k+1}, p) - L(u, q^{k}) \right] \\
+ d_{4} \left[ \left( 1 + 2\gamma^{2}\lambda_{\max}(A^{\top}A) \right) \|u^{k} - T_{u}(w^{k})\|^{2} + 2\gamma^{2} \|Au^{k} - b\|^{2} \right]. \quad (A.23)$$

Since  $\lambda_{\max}(A^{\top}A) \|u^k - T_u(w^k)\|^2 \ge \|A[u^k - T_u(w^k)]\|^2$  and  $T_p(w^k) - p^k = \gamma[AT_u(w^k) - b]$ , (A.23) follows that

$$\begin{split} \Lambda(w^{k},w) - \mathbb{E}_{i(k)}\Lambda(w^{k+1},w) &\geq \frac{\epsilon}{N}\mathbb{E}_{i(k)}\left[L(u^{k+1},p) - L(u,q^{k})\right] \\ &\quad + d_{4}\left[\|u^{k} - T_{u}(w^{k})\|^{2} + 2\gamma^{2}\|A[u^{k} - T_{u}(w^{k})]\|^{2} + 2\gamma^{2}\|Au^{k} - b\|^{2}\right] \\ &\geq \frac{\epsilon}{N}\mathbb{E}_{i(k)}\left[L(u^{k+1},p) - L(u,q^{k})\right] + d_{4}\left[\|u^{k} - T_{u}(w^{k})\|^{2} + \gamma^{2}\|AT_{u}(w^{k}) - b\|^{2}\right] \\ &\geq \frac{\epsilon}{N}\mathbb{E}_{i(k)}\left[L(u^{k+1},p) - L(u,q^{k})\right] + d_{4}\|w^{k} - T(w^{k})\|^{2}. \end{split}$$

Then we have the result of Lemma 2.

## 3. Proof of Theorem 1 (Almost surely convergence)

Proof.

(i) Take  $w = w^*$  in Lemma 2, we have

$$\Lambda(w^k, w^*) \ge \mathbb{E}_{i(k)}\Lambda(w^{k+1}, w^*) + \frac{\epsilon}{N}\mathbb{E}_{i(k)}\left[L(u^{k+1}, p^*) - L(u^*, q^k)\right] + d_4\|w^k - T(w^k)\|^2.$$
(A.24)

Observe that  $L(u^{k+1}, p^*) - L(u^*, q^k) \ge 0$ . From statement (i) of Lemma 1, we have that  $\Lambda(w^k, w^*)$  is nonnegative. By the Robbins-Siegmund Lemma (Robbins & Siegmund, 1971), we obtain that  $\lim_{k \to +\infty} \Lambda(w^k, w^*)$  almost surely exists,

$$\sum_{k=0}^{+\infty} \|w^k - T(w^k)\|^2 < +\infty \text{ a.s..}$$

- (ii) Since  $\lim_{k \to +\infty} \Lambda(w^k, w^*)$  almost surely exists, thus  $\Lambda(w^k, w^*)$  is almost surely bounded. Thanks statement (i) of Lemma 1, it implies the sequences  $\{w^k\}$  is almost surely bounded.
- (iii) From statement (i) we have that

$$\lim_{k \to \infty} \|w^k - T(w^k)\| = 0 \quad \text{a.s.}.$$

By variational inequality system (A.6), we have that any cluster point of a realization sequence generated by RPDC almost surely is a saddle point of Lagrangian for (P).

#### 4. Proof of Theorem 2 (Expected primal suboptimality and expected feasibility)

Proof.

(i) Let  $h(w, w') = \Lambda(w, w') + \frac{d_3}{d_1}\Lambda(w, w^*)$ . By statement (i) and (iii) in Lemma 1, we have  $h(w, w') \ge 0$ . From Lemma 2, we obtain that

$$\mathbb{E}_{i(k)}\frac{\epsilon}{N}\left[L(u^{k+1}, p) - L(u, q^k)\right] \le \Lambda(w^k, w) - \mathbb{E}_{i(k)}\Lambda(w^{k+1}, w)$$

Taking expectation with respect to  $\mathcal{F}_t$ , t > k for above inequality, we obtain that

$$\frac{\epsilon}{N} \mathbb{E}_{\mathcal{F}_t} \left[ L(u^{k+1}, p) - L(u, q^k) \right] \le \mathbb{E}_{\mathcal{F}_t} [\Lambda(w^k, w) - \Lambda(w^{k+1}, w)].$$
(A.25)

Take  $w = w^*$  in (A.25), we obtain

$$0 \le \mathbb{E}_{\mathcal{F}_t}[\Lambda(w^k, w^*) - \Lambda(w^{k+1}, w^*)].$$
(A.26)

By the combination of (A.25) and (A.26), it follows

$$\frac{\epsilon}{N} \mathbb{E}_{\mathcal{F}_t} \left[ L(u^{k+1}, p) - L(u, q^k) \right] \le \mathbb{E}_{\mathcal{F}_t} [h(w^k, w) - h(w^{k+1}, w)]$$
(A.27)

From the definition of  $\bar{u}_t$  and  $\bar{p}_t$ , we have  $\bar{u}_t \in \mathbf{U}$  and  $\bar{p}_t \in \mathbf{R}^m$ . From the convexity of set  $\mathbf{U}, \mathbf{R}^m$  and the function L(u', p) - L(u, p') is convex in u' and linear in p', for all  $u \in \mathbf{U}$  and  $p \in \mathbf{R}^m$ , we have that

$$\mathbb{E}_{\mathcal{F}_{t}}\left[L(\bar{u}_{t}, p) - L(u, \bar{p}_{t})\right] \leq \mathbb{E}_{\mathcal{F}_{t}}\frac{1}{t+1}\sum_{k=0}^{t}\left[L(u^{k+1}, p) - L(u, q^{k})\right] \leq \frac{Nh(w^{0}, w)}{\epsilon(t+1)}.$$
(A.28)

(ii) If  $\mathbb{E}_{\mathcal{F}_t} \|A\bar{u}_t - b\| = 0$ , statement (ii) is obviously. Otherwise,  $\mathbb{E}_{\mathcal{F}_t} \|A\bar{u}_t - b\| \neq 0$  i.e., there is set  $\mathbb{W}$  such that  $\mathbb{P}\{\omega \in \mathbb{W} | \|A\bar{u}_t - b\| \neq 0\} > 0$ . Let  $\hat{p}$  be a random vector:

$$\hat{p}(\omega) = \begin{cases} 0 & \omega \notin \mathbb{W} \\ \frac{M\left(A\bar{u}_t - b\right)}{\|A\bar{u}_t - b\|} & \omega \in \mathbb{W}. \end{cases}$$
(A.29)

Noted that for  $\omega \notin \mathbb{W}$ , we have  $\hat{p}(\omega) = 0$  and  $||A\bar{u}_t - b|| = 0$ . Thus

$$\langle \hat{p}(\omega), A\bar{u}_t - b \rangle = M \|A\bar{u}_t - b\| = 0.$$
(A.30)

Otherwise, for  $\omega \in \mathbb{W}$ , we have that

$$\langle \hat{p}(\omega), A\bar{u}_t - b \rangle = M \|A\bar{u}_t - b\|.$$
(A.31)

Together (A.30) and (A.31), we have

$$\langle \hat{p}, A\bar{u}_t - b \rangle = M \|A\bar{u}_t - b\| \tag{A.32}$$

Moreover, since  $Au^* = b$ , we have

$$L(\bar{u}_t, \hat{p}) - L(u^*, \bar{p}_t) = F(\bar{u}_t) + \langle \hat{p}, A\bar{u}_t - b \rangle - F(u^*) = F(\bar{u}_t) - F(u^*) + M ||A\bar{u}_t - b||.$$
(A.33)

Moreover, by taking  $u = \bar{u}_t$  in the right hand side of saddle point inequality, we have

$$F(\bar{u}_t) - F(u^*) \ge -\langle p^*, A\bar{u}_t - b \rangle \ge - \|p^*\| \|A\bar{u}_t - b\|.$$
(A.34)

Combine (A.33) and (A.34), we have that

$$||A\bar{u}_t - b|| \le \frac{L(\bar{u}_t, \hat{p}) - L(u^*, \bar{p}_t)}{(M - ||p^*||)}$$

Take expectation on both side of above inequality, we have that

$$\mathbb{E}_{\mathcal{F}_{t}} \| A \bar{u}_{t} - b \| \leq \frac{\mathbb{E}_{\mathcal{F}_{t}} [L(\bar{u}_{t}, \hat{p}) - L(u^{*}, \bar{p}_{t})]}{(M - \|p^{*}\|)} \leq \mathbb{E}_{\mathcal{F}_{t}} \frac{Nh(w^{0}, (u^{*}, \hat{p}))}{(M - \|p^{*}\|) \epsilon(t+1)} \qquad (by (i))$$

$$\leq \mathbb{E}_{\mathcal{F}_{t}} \frac{Nd_{5}}{(M - \|p^{*}\|) \epsilon(t+1)} \qquad (A.35)$$

where  $d_5 = \sup_{\|p\| < M} h(w^0, (u^*, p)).$ 

(iii) Again from (A.33), (A.34) and statement (ii), statement (iii) is coming.

 $\geq d_4[\|w^k - T(w^k)\|^2 + \frac{\epsilon}{N}[L(u^k, p^*) - L(u^*, p^*)].$ 

#### 5. Proof of Lemma 3

Proof.

- (i) This statement directly follows from the definition of  $\phi(w, w^*)$  and statement (i) in Lemma 1.
- (ii) This statement directly follows from the definition of  $\phi(w, w^*)$  and statement (ii) in Lemma 1.
- (iii) By the definition of  $\phi(w, w^*)$ , we have that.

$$\phi(w^{k}, w^{*}) - \mathbb{E}_{i(k)}\phi(w^{k+1}, w^{*}) = \Lambda(w^{k}, w^{*}) - \mathbb{E}_{i(k)}\left\{\Lambda(w^{k+1}, w^{*}) + \frac{\epsilon}{N}[L(u^{k}, p^{*}) - L(u^{*}, p^{*})] - \frac{\epsilon}{N}[L(u^{k+1}, p^{*}) - L(u^{*}, p^{*})]\right\}$$
  

$$\geq \Lambda(w^{k}, w^{*}) - \mathbb{E}_{i(k)}\left\{\Lambda(w^{k+1}, w^{*}) + \frac{\epsilon}{N}[L(u^{k}, p^{*}) - L(u^{*}, p^{*})] - \frac{\epsilon}{N}[L(u^{k+1}, p^{*}) - L(u^{*}, q^{k})]\right\}$$

(by the definition of saddle point.)

(by Lemma 2)

# 6. Proof of Theorem 3 (Global strong metric subregularity of H(w) implies linear convergence of RPDC)

*Proof.* Considering the reference point  $T(w^k)$  associated with given point  $w^k$ , we have that

$$\begin{cases} 0 \in \nabla G(u^k) + \partial J(T_u(w^k)) + A^{\top} q^k + \frac{1}{\epsilon} \left[ \nabla K(T_u(w^k)) - \nabla K(u^k) \right] + \mathcal{N}_{\mathbf{U}}(T_u(w^k)) \\ 0 = b - AT_u(w^k) + \frac{1}{\gamma} \left[ T_p(w^k) - p^k \right] \end{cases}$$
(A.36)

Thus

$$v(T(w^k)) = \begin{pmatrix} \nabla G(T_u(w^k)) - \nabla G(u^k) + A^\top (T_p(w^k) - q^k) + \frac{1}{\epsilon} \left[ \nabla K(u^k) - \nabla K(T_u(w^k)) \right] \\ \frac{1}{\gamma} \left[ p^k - T_p(w^k) \right] \end{pmatrix} \in H(T(w^k)).$$

From Assumption 1 and 2, there is  $\delta > 0$  such that

$$\|v(T(w^k))\|^2 \le \delta \|w^k - T(w^k)\|^2.$$
(A.37)

Since H(w) is global strong metric subregular at  $w^*$  for 0, then

$$\|T(w^{k}) - w^{*}\| \le \mathfrak{c}dist(0, H(T(w^{k}))) \le \mathfrak{c}\|v(T(w^{k}))\| \le \mathfrak{c}\sqrt{\delta}\|w^{k} - T(w^{k})\|.$$
(A.38)

Since  $||w^k - w^*|| \le ||T(w^k) - w^*|| + ||w^k - T(w^k)||$ , we have

$$\|w^{k} - w^{*}\| \le (\mathfrak{c}\sqrt{\delta} + 1)\|w^{k} - T(w^{k})\|.$$
(A.39)

From statement (iii) of Lemma 3, we have that

$$\begin{split} \phi(w^{k},w^{*}) - \mathbb{E}_{i(k)}\phi(w^{k+1},w^{*}) &\geq d_{4}\|w^{k} - T(w^{k})\|^{2} + \frac{\epsilon}{N}[L(u^{k},p^{*}) - L(u^{*},p^{*})] \\ &\geq \frac{d_{4}}{(\mathfrak{c}\sqrt{\delta}+1)^{2}}\|w^{k} - w^{*}\|^{2} + \frac{\epsilon}{N}[L(u^{k},p^{*}) - L(u^{*},p^{*})] \quad \text{(by (A.39))} \\ &\geq \delta'\{d_{2}\|w^{k} - w^{*}\|^{2} + \epsilon[L(u^{k},p^{*}) - L(u^{*},p^{*})]\} \\ &\geq \delta'\phi(w^{k},w^{*}). \quad \text{(by (i) of Lemma 3)} \quad (A.40) \end{split}$$

where  $\delta' = \min\{\frac{d_4}{\max\{d_2(\mathfrak{c}\sqrt{\delta}+1)^2, d_4+1\}}, \frac{1}{N+1}\} < 1$ . It follows that

$$\mathbb{E}_{i(k)}\phi(w^{k+1}, w^*) \le \alpha \phi(w^k, w^*).$$
(A.41)

where  $\alpha = 1 - \delta' \in (0, 1)$ . Taking expectation with respect to  $\mathcal{F}_{k+1}$  for above inequality, we obtain that

$$\mathbb{E}_{\mathcal{F}_{k+1}}\phi(w^{k+1}, w^*) \le \alpha^{k+1}\phi(w^0, w^*).$$
(A.42)

# 7. Proof of Corollary 1 (R-linear rate of the sequence $\{\mathbb{E}_{\mathcal{F}_k}w^k\}$ )

*Proof.* By statement (i) in Lemma 3, we have that  $\phi(w, w^*) \ge d_1 ||w - w^*||^2$ . By Theorem 3, we have that  $\mathbb{E}_{\mathcal{F}_k} \phi(w^k, w^*) \le \alpha^k \phi(w^0, w^*)$ .

Then we have that

$$\mathbb{E}_{\mathcal{F}_k} \| w^k - w^* \|^2 \le \frac{\alpha^k \phi(w^0, w^*)}{d_1}.$$

By convexity of  $\|\cdot\|^2$  and Jensen's inequality, we obtain that

$$\|\mathbb{E}_{\mathcal{F}_k}w^k - w^*\| \leq \hat{M}(\sqrt{lpha})^k \quad ext{with } \hat{M} = \sqrt{rac{\phi(w^0, w^*)}{d_1}}.$$

This shows that the sequence  $\{\mathbb{E}_{\mathcal{F}_k} w^k\}$  converges to the desired saddle point  $w^*$  at R-linear rate; i.e.,

$$\lim_{k \to \infty} \sup \sqrt[k]{\|\mathbb{E}_{\mathcal{F}_k} w^k - w^*\|} = \sqrt{\alpha} < 1.$$

#### 8. Proof of Proposition 1

*Proof.* By the piecewise linear of H(w) and Zheng and Ng (Zheng & Ng, 2014), we have that H(w) is global metric subregular at  $w^*$  for 0. Since Q is positive-definite, then problem (SVM) has unique solution  $u^*$ . Hence, to show H(w) is global strongly metric subregular, we need to prove uniqueness of the Lagrangian multiplier for (SVM). Suppose their are two multipliers p and p', thus we have

$$\begin{cases} 0 \in Qu^* - \mathbf{1}_n + py + \mathcal{N}_{[0,c]^n}(u^*) \\ 0 \in Qu^* - \mathbf{1}_n + p'y + \mathcal{N}_{[0,c]^n}(u^*) \end{cases}$$

Since there exists at least one component  $u_i^*$  of optimal solution  $u^*$  satisfies  $0 < u_i^* < c$ , then  $\xi_i = \mathcal{N}_{[0,c]}(u_i^*) = 0$ . Thus, we have that

$$\begin{cases} Q_i u^* - 1 + y_i p = 0\\ Q_i u^* - 1 + y_i p' = 0 \end{cases}$$
(A.43)

We conclude that p = p'. Therefore H(w) is global strongly metric subregular.

### 9. Proof of Proposition 2

**Proof.** By the piecewise linear of H(w) and Zheng and Ng (Zheng & Ng, 2014), we have that H(w) is global metric subregular at  $w^*$  for 0. Since  $\Sigma$  is positive-definite, then problem (MLP) has unique solution  $u^*$ . Hence, to show H(w) is global strongly metric subregular, we need to prove uniqueness of the Lagrangian multiplier for (MLP). Suppose their are two pare of multipliers  $(p_1, p_2)$  and  $(p'_1, p'_2)$ , thus we have

$$\begin{cases} 0 \in \Sigma u^* + \lambda \partial \|u^*\|_1 + p_1 \mu + p_2 \mathbf{1}_n \\ 0 \in \Sigma u^* + \lambda \partial \|u^*\|_1 + p'_1 \mu + p'_2 \mathbf{1}_n \end{cases}$$

Since  $u_i^* \neq 0$ ,  $u_i^* \neq 0$ , thus  $\xi_i = \partial |u_i^*|$  and  $\xi_j = \partial |u_j^*|$  are single valued and we have

$$\begin{cases} \Sigma_{i}u^{*} + \lambda\xi_{i} + \mu_{i}p_{1} + p_{2} = 0\\ \Sigma_{i}u^{*} + \lambda\xi_{i} + \mu_{i}p_{1}' + p_{2}' = 0 \end{cases}$$
(A.44)

$$\begin{cases} \Sigma_{j}u^{*} + \lambda\xi_{j} + \mu_{j}p_{1} + p_{2} = 0\\ \Sigma_{j}u^{*} + \lambda\xi_{j} + \mu_{j}p_{1}' + p_{2}' = 0 \end{cases}$$
(A.45)

It follows that

$$\begin{cases} \mu_i(p_1 - p'_1) + p_2 - p'_2 = 0\\ \mu_j(p_1 - p'_1) + p_2 - p'_2 = 0 \end{cases}$$
(A.46)

Since  $\mu_i \neq \mu_j$ , we conclude that  $p_1 = p'_1$  and  $p_2 = p'_2$ . Therefore H(w) is global strongly metric subregular.

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