
Supplementary Material: Dual Averaging and Proximal Gradient Descent for Online Alternating Direction Multiplier Method

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In this supplementary material, we give the proofs for the theorems in the main text. We consider more general optimization problem:

$$\min_{x \in \mathcal{X}, y \in \mathcal{Y}} \frac{1}{T} \sum_{t=1}^T f(x, w_t) + \psi(y), \quad (\text{S-1a})$$

$$\text{s.t. } Ax + By - b = 0, \quad (\text{S-1b})$$

where $A \in \mathbb{R}^{l \times m}$, $B \in \mathbb{R}^{l \times d}$ and $b \in \mathbb{R}^l$. We solve this problem in an online manner using OPG-ADMM and RDA-ADMM techniques. The algorithm in the main text corresponds to the situation where $B = -I$ and $b = \mathbf{0}$.

Theorems 1, 2, 3 and 4 in the main text correspond to Theorems 7 (and Eq. (S-31)), 8, 9 and 5 in this supplementary material respectively.

A. Convergence rate of OPG-ADMM

Corresponding to the optimization problem (S-1), we consider the following generalized version of OPG-ADMM: Let $x_1 = \mathbf{0}$, $\lambda_1 = \mathbf{0}$, and $By_1 = b$ (we assume there exists y_1 that satisfies this equality for simplicity), and the update rule of the t -th step is given by

$$x_{t+1} = \operatorname{argmin}_{x \in \mathcal{X}} \left\{ g_t^\top x - \lambda_t^\top (Ax + By_t - b) + \frac{\rho}{2} \|Ax + By_t - b\|^2 + \frac{1}{2\eta_t} \|x - x_t\|_{G_t}^2 \right\}, \quad (\text{S-2a})$$

$$y_{t+1} = \operatorname{argmin}_{y \in \mathcal{Y}} \left\{ \psi(y) - \lambda_t^\top (Ax_{t+1} + By - b) + \frac{\rho}{2} \|Ax_{t+1} + By - b\|^2 \right\}, \quad (\text{S-2b})$$

$$\lambda_{t+1} = \lambda_t - \rho(Ax_{t+1} + By_{t+1} - b), \quad (\text{S-2c})$$

where $G_t = \gamma I - \eta_t \rho A^\top A$. Moreover we define

$$\tilde{\lambda}_t = \lambda_t - \rho(Ax_{t+1} + By_t - b). \quad (\text{S-3})$$

This method satisfies the following regret property.

Theorem 4. *For all $x^* \in \mathcal{X}$, $y^* \in \mathcal{Y}$ and $\lambda^* \in \mathbb{R}^l$, we have*

$$\begin{aligned} & \frac{1}{T} \sum_{t=1}^T (f(x_t, w_t) + \psi(y_t)) - \frac{1}{T} \sum_{t=1}^T (f(x^*, w_t) + \psi(y^*)) \\ & + \frac{1}{T} \sum_{t=1}^T \begin{pmatrix} -A^\top \tilde{\lambda}_t \\ -B^\top \tilde{\lambda}_t \\ Ax_t + By_t - b \end{pmatrix}^\top \begin{pmatrix} x_t - x^* \\ y_t - y^* \\ \tilde{\lambda}_t - \lambda^* \end{pmatrix} \end{aligned}$$

$$\begin{aligned}
 & + \frac{1}{T} \sum_{t=1}^T \frac{\|\lambda_t - \lambda_{t+1}\|^2}{2\rho} + \frac{\|\lambda_{T+1} - \lambda^*\|^2}{2\rho T} \\
 & \leq \frac{\|x^*\|_{G_1}^2}{2\eta_1 T} + \frac{1}{T} \sum_{t=2}^T \left(\frac{\gamma}{2\eta_t} - \frac{\gamma}{2\eta_{t-1}} - \frac{\sigma}{2} \right) \|x_t - x^*\|^2 + \frac{1}{T} \sum_{t=1}^T \frac{\eta_t}{2} \|g_t\|_{G_t^{-1}}^2 \\
 & \quad + \frac{\rho}{2T} \|b - By^*\|^2 + \frac{\|\lambda^*\|^2}{2\rho T} \\
 & \quad + \frac{1}{T} (\langle Ax_{T+1}, \lambda^* \rangle + \langle B(y^* - y_{T+1}), \lambda_{T+1} - \lambda^* \rangle - \langle By^* - b, \lambda^* \rangle).
 \end{aligned}$$

Proof. Let $f_t(x) := f(x, w_t)$. By the optimality of x_{t+1} and y_t , we have that, for all $x \in \mathcal{X}$ and $y \in \mathcal{Y}$,

$$\langle g_t - A^\top \tilde{\lambda}_t + G_t(x_{t+1} - x_t)/\eta_t, x - x_{t+1} \rangle \geq 0 \quad (\text{S-4})$$

and

$$\langle \nabla \psi(y_t) - B^\top \lambda_t, y - y_t \rangle \geq 0, \quad (\text{S-5})$$

where we used the relation $\nabla \psi(y_t) - B^\top \lambda_{t-1} + \rho B^\top (Ax_t + By_t - b) = \nabla \psi(y_t) - B^\top \lambda_t$. Using these inequality, for given $x^* \in \mathcal{X}$, $y^* \in \mathcal{Y}$ and λ^* , we have that

$$\begin{aligned}
 & f_t(x_t) + \psi(y_t) - f_t(x^*) - \psi(y^*) \\
 & \leq \langle g_t, x_t - x^* \rangle - \frac{\sigma}{2} \|x_t - x^*\|^2 + \langle \nabla \psi(y_t), y_t - y^* \rangle \\
 & = \langle g_t, x_{t+1} - x^* \rangle + \langle \nabla \psi(y_t), y_t - y^* \rangle + \langle g_t, x_t - x_{t+1} \rangle - \frac{\sigma}{2} \|x_t - x^*\|^2 \\
 & \leq \langle A^\top \tilde{\lambda}_t - G_t(x_{t+1} - x_t)/\eta_t, x_{t+1} - x^* \rangle + \langle B^\top \lambda_t, y_t - y^* \rangle + \langle g_t, x_t - x_{t+1} \rangle - \frac{\sigma}{2} \|x_t - x^*\|^2 \\
 & = \begin{pmatrix} -A^\top \tilde{\lambda}_t \\ -B^\top \tilde{\lambda}_t \end{pmatrix}^\top \begin{pmatrix} x^* - x_t \\ y^* - y_t \end{pmatrix} + \begin{pmatrix} x^* - x_{t+1} \\ y^* - y_t \end{pmatrix}^\top \begin{pmatrix} G_t(x_{t+1} - x_t)/\eta_t \\ -B^\top (\lambda_t - \tilde{\lambda}_t) \end{pmatrix} + \langle A^\top \tilde{\lambda}_t, x_{t+1} - x_t \rangle + \langle g_t, x_t - x_{t+1} \rangle \\
 & \quad - \frac{\sigma}{2} \|x_t - x^*\|^2 \\
 & = \begin{pmatrix} -A^\top \tilde{\lambda}_t \\ -B^\top \tilde{\lambda}_t \\ Ax_t + By_t - b \end{pmatrix}^\top \begin{pmatrix} x^* - x_t \\ y^* - y_t \\ \lambda^* - \tilde{\lambda}_t \end{pmatrix} + \begin{pmatrix} x^* - x_{t+1} \\ y^* - y_t \end{pmatrix}^\top \begin{pmatrix} G_t(x_{t+1} - x_t)/\eta_t \\ -B^\top (\lambda_t - \tilde{\lambda}_t) \\ \frac{\tilde{\lambda}_t - \lambda_t}{\rho} - A(x_t - x_{t+1}), \end{pmatrix} \\
 & \quad + \langle A^\top \tilde{\lambda}_t, x_{t+1} - x_t \rangle + \langle g_t, x_t - x_{t+1} \rangle - \frac{\sigma}{2} \|x_t - x^*\|^2,
 \end{aligned}$$

where, in the last line, we used

$$Ax_t + By_t - b = -\frac{\tilde{\lambda}_t - \lambda_t}{\rho} + A(x_t - x_{t+1})$$

by the definition of $\tilde{\lambda}_t$ and λ_t .

Now we bound $\langle x^* - x_{t+1}, G_t(x_{t+1} - x_t)/\eta_t \rangle$. By Lemma 10, this can be bounded as

$$\langle x^* - x_{t+1}, G_t(x_{t+1} - x_t)/\eta_t \rangle = -\frac{\|x_{t+1} - x^*\|_{G_t}^2}{2\eta_t} + \frac{\|x_t - x^*\|_{G_t}^2}{2\eta_t} - \frac{\|x_{t+1} - x_t\|_{G_t}^2}{2\eta_t}.$$

On the other hand,

$$\langle g_t, x_t - x_{t+1} \rangle \leq \|g_t\|_{G_t^{-1}} \|x_t - x_{t+1}\|_{G_t} \leq \frac{\eta_t}{2} \|g_t\|_{G_t^{-1}}^2 + \frac{1}{2\eta_t} \|x_t - x_{t+1}\|_{G_t}^2.$$

Combining these two inequalities, we have

$$\langle x^* - x_{t+1}, G_t(x_{t+1} - x_t)/\eta_t \rangle + \langle g_t, x_t - x_{t+1} \rangle \leq -\frac{\|x_{t+1} - x^*\|_{G_t}^2}{2\eta_t} + \frac{\|x_t - x^*\|_{G_t}^2}{2\eta_t} + \frac{\eta_t}{2} \|g_t\|_{G_t^{-1}}^2.$$

This and Lemma 11 gives

$$\begin{aligned}
 & \begin{pmatrix} x^* - x_{t+1} \\ y^* - y_t \\ \lambda^* - \tilde{\lambda}_t \end{pmatrix}^\top \begin{pmatrix} G_t(x_{t+1} - x_t)/\eta_t \\ -B^\top(\lambda_t - \tilde{\lambda}_t) \\ \frac{\tilde{\lambda}_t - \lambda_t}{\rho} - A(x_t - x_{t+1}) \end{pmatrix} + \langle A^\top \tilde{\lambda}_t, x_{t+1} - x_t \rangle + \langle g_t, x_t - x_{t+1} \rangle, \\
 &= \langle x^* - x_{t+1}, G_t(x_{t+1} - x_t)/\eta_t \rangle + \langle g_t, x_t - x_{t+1} \rangle + \langle \lambda^*, A(x_{t+1} - x_t) \rangle \\
 & \quad + \begin{pmatrix} y^* - y_t \\ \lambda^* - \tilde{\lambda}_t \end{pmatrix}^\top \begin{pmatrix} -B^\top(\lambda_t - \tilde{\lambda}_t) \\ \frac{\tilde{\lambda}_t - \lambda_t}{\rho} \end{pmatrix} \\
 & \leq \frac{\|x_t - x^*\|_{G_t}^2}{2\eta_t} - \frac{\|x_{t+1} - x^*\|_{G_t}^2}{2\eta_t} + \langle \lambda^*, A(x_{t+1} - x_t) \rangle + \frac{\eta_t}{2} \|g_t\|_{G_t^{-1}}^2 \\
 & \quad + \frac{\rho}{2} \|y_t - y^*\|_{B^\top B}^2 - \frac{\rho}{2} \|y_{t+1} - y^*\|_{B^\top B}^2 + \frac{\|\lambda_t - \lambda^*\|^2}{2\rho} - \frac{\|\lambda_{t+1} - \lambda^*\|^2}{2\rho} \\
 & \quad + \langle B(y^* - y_{t+1}), \lambda_{t+1} - \lambda^* \rangle - \langle B(y^* - y_t), \lambda_t - \lambda^* \rangle \\
 & \quad - \frac{\|\lambda_t - \lambda_{t+1}\|^2}{2\rho}. \tag{S-6}
 \end{aligned}$$

Now summing up this bound for $t = 1, \dots, T$, then we have

$$\begin{aligned}
 & \sum_{t=1}^T (f_t(x_t) + \psi(y_t)) - \sum_{t=1}^T (f_t(x^*) - \psi(y^*)) \\
 & \leq \sum_{t=1}^T \begin{pmatrix} -A^\top \tilde{\lambda}_t \\ -B^\top \tilde{\lambda}_t \\ Ax_t + By_t - b \end{pmatrix}^\top \begin{pmatrix} x^* - x_t \\ y^* - y_t \\ \lambda^* - \tilde{\lambda}_t \end{pmatrix} \\
 & \quad + \frac{\|x_1 - x^*\|_{G_1}^2}{2\eta_1} + \sum_{t=2}^T \left(\frac{\|x_t - x^*\|_{G_t}^2}{2\eta_t} - \frac{\|x_t - x^*\|_{G_{t-1}}^2}{2\eta_{t-1}} \right) + \langle \lambda^*, A(x_{T+1} - x_1) \rangle + \sum_{t=1}^T \frac{\eta_t}{2} \|g_t\|_{G_t^{-1}}^2 \\
 & \quad + \frac{\rho}{2} \|y_1 - y^*\|_{B^\top B}^2 + \frac{\|\lambda_1 - \lambda^*\|^2}{2\rho} - \frac{\|\lambda_{T+1} - \lambda^*\|^2}{2\rho} \\
 & \quad + \langle B(y^* - y_{T+1}), \lambda_{T+1} - \lambda^* \rangle - \langle B(y^* - y_1), \lambda_1 - \lambda^* \rangle \\
 & \quad - \sum_{t=1}^T \frac{\|\lambda_t - \lambda_{t+1}\|^2}{2\rho} - \sum_{t=1}^T \frac{\sigma}{2} \|x_t - x^*\|^2.
 \end{aligned}$$

Now since $G_t = \gamma I - \eta_t \rho A^\top A$, we have that

$$\frac{\|x_t - x^*\|_{G_t}^2}{2\eta_t} - \frac{\|x_t - x^*\|_{G_{t-1}}^2}{2\eta_{t-1}} = \left(\frac{\gamma}{2\eta_t} - \frac{\gamma}{2\eta_{t-1}} \right) \|x_t - x^*\|^2.$$

This and the initial settings of x_1, y_1, λ_1 give the assertion. \square

Here we simplify Theorem 4. First note that by Eq. (S-5)

$$\langle y^* - y_{T+1}, B^\top(\lambda_{T+1} - \lambda^*) \rangle \leq \langle y^* - y_{T+1}, \nabla \psi(y_{T+1}) - B^\top \lambda^* \rangle.$$

Since the diameters of \mathcal{X} and \mathcal{Y} are bounded by R , $\|g_t\| \leq G$, and the subgradient of ψ is bounded by L_ψ , if $\|B^\top \lambda^*\| \leq L_\psi$ and B is invertible, there exists a constant K depending on $R, B, \rho, \lambda^*, L_\psi, \eta_1$ such that the bound shown in Theorem 4 can be further bounded as

$$\frac{1}{T} \sum_{t=1}^T (f(x_t, w_t) + \psi(y_t)) - \frac{1}{T} \sum_{t=1}^T (f(x^*, w_t) + \psi(y^*))$$

$$\begin{aligned}
 & + \frac{1}{T} \sum_{t=1}^T \begin{pmatrix} A^\top \tilde{\lambda}_t \\ B^\top \tilde{\lambda}_t \\ Ax_t + By_t - b \end{pmatrix}^\top \begin{pmatrix} x^* - x_t \\ y^* - y_t \\ \tilde{\lambda}_t - \lambda^* \end{pmatrix} \\
 & + \frac{1}{T} \sum_{t=1}^T \frac{\|\lambda_t - \lambda_{t+1}\|^2}{2\rho} + \frac{\|\lambda_{T+1} - \lambda^*\|^2}{2\rho T} \\
 & \leq \frac{1}{T} \sum_{t=2}^T \max \left\{ \frac{\gamma}{2\eta_t} - \frac{\gamma}{2\eta_{t-1}} - \frac{\sigma}{2}, 0 \right\} R^2 + \frac{1}{T} \sum_{t=1}^T \frac{\eta_t}{2} G^2 + \frac{K}{T}
 \end{aligned} \tag{S-7}$$

$$=:\Xi_T. \tag{S-8}$$

Theorem 5. Suppose ψ is Lipschitz continuous with a constant L_ψ , i.e., $|\psi(y) - \psi(y')| \leq L_\psi \|y - y'\|$ ($\forall y, y' \in \mathcal{Y}$), and B is invertible. We utilize $y'_t := B^{-1}(b - Ax_t)$ as an estimator of y at the t -th step, and let $\bar{y}'_t := \frac{1}{t} \sum_{\tau=1}^t y'_\tau$. Then, for all $x^* \in \mathcal{X}, y^* \in \mathcal{Y}$ such that $Ax^* + By^* - b = 0$, there exists a constant K depending on $R, A, B, L_\psi, \rho, \eta_1$ such that

$$\begin{aligned}
 & \mathbb{E}_{w_{1:T}} [(f(\bar{x}_T, w_T) + \psi(\bar{y}'_T)) - (f(x^*, w_T) + \psi(y^*))] + \frac{\rho}{2} \mathbb{E}_{w_{1:T}} [\|A\bar{x}_{T+1} + B\bar{y}'_{T+1} - b\|^2] \\
 & \leq \frac{1}{T} \sum_{t=2}^T \max \left\{ \frac{\gamma}{2\eta_t} - \frac{\gamma}{2\eta_{t-1}} - \frac{\sigma}{2}, 0 \right\} R^2 + \frac{1}{T} \sum_{t=1}^T \frac{\eta_t}{2} G^2 + \frac{K}{T}.
 \end{aligned}$$

Proof. Note that

$$B(y'_t - y_t) = b - Ax_t - By_t = (\lambda_t - \lambda_{t-1})/\rho. \tag{S-9}$$

Thus, noting $y'_1 = y_1$ by the initialization, we have

$$B(\bar{y}'_T - \bar{y}_T) = \frac{1}{\rho T} (\lambda_T - \lambda_1). \tag{S-10}$$

Since $b - Ax_t - By_t = (\lambda_t - \lambda_{t-1})/\rho$, the Lagrangian part in the statement of Theorem 4 can be bounded as

$$\begin{aligned}
 & \begin{pmatrix} A^\top \tilde{\lambda}_t \\ B^\top \tilde{\lambda}_t \\ Ax_t + By_t - b \end{pmatrix}^\top \begin{pmatrix} x^* - x_t \\ y^* - y_t \\ \tilde{\lambda}_t - \lambda^* \end{pmatrix} = \begin{pmatrix} A^\top \lambda^* \\ B^\top \lambda^* \\ Ax^* + By^* - b \end{pmatrix}^\top \begin{pmatrix} x^* - x_t \\ y^* - y_t \\ \tilde{\lambda}_t - \lambda^* \end{pmatrix} \\
 & = \langle \lambda^*, A(x^* - x_t) + B(y^* - y_t) \rangle = \langle \lambda^*, b - Ax_t - By_t \rangle \\
 & = \langle \lambda^*, (\lambda_t - \lambda_{t-1})/\rho \rangle.
 \end{aligned}$$

Therefore

$$\begin{aligned}
 \frac{1}{T} \sum_{t=1}^T \begin{pmatrix} A^\top \tilde{\lambda}_t \\ B^\top \tilde{\lambda}_t \\ Ax_t + By_t - b \end{pmatrix}^\top \begin{pmatrix} x^* - x_t \\ y^* - y_t \\ \tilde{\lambda}_t - \lambda^* \end{pmatrix} & = \langle \lambda^*, A(x^* - \bar{x}_T) + B(y^* - \bar{y}_T) \rangle \\
 & = \frac{1}{T\rho} \langle \lambda^*, \lambda_T - \lambda_1 \rangle,
 \end{aligned} \tag{S-11}$$

where we again used $Ax_1 + By_1 - b = 0$.

Eq. (S-10) gives

$$\begin{aligned}
 f(\bar{x}_T, w_T) + \psi(\bar{y}'_T) & \leq f(\bar{x}_T, w_T) + \psi(\bar{y}_T) + \langle \nabla \psi(\bar{y}'_T), \bar{y}'_T - \bar{y}_T \rangle \\
 & \leq f(\bar{x}_T, w_T) + \psi(\bar{y}_T) + \frac{1}{T\rho} \langle \nabla \psi(\bar{y}'_T), B^{-1}(\lambda_T - \lambda_1) \rangle.
 \end{aligned}$$

Thus if we set

$$\lambda^* = B^{-\top} \nabla \psi(\bar{y}'_T),$$

(note that $\|B^\top \lambda^*\| \leq L_\psi$) then we have

$$f(\bar{x}_T, w_T) + \psi(\bar{y}'_T) \leq f(\bar{x}_T, w_T) + \psi(\bar{y}_T) + \frac{1}{T\rho} \langle \lambda^*, \lambda_T - \lambda_1 \rangle.$$

Since $\{w_t\}_{t=1}^T$ is independently identically distributed, Jensen's inequality yields

$$\begin{aligned} & \mathbb{E}_{w_{1:T}}[(f(\bar{x}_T, w_T) + \psi(\bar{y}'_T)) - (f(x^*, w_T) + \psi(y^*))] \\ & \leq \mathbb{E}_{w_{1:T}}[f(\bar{x}_T, w_T) + \psi(\bar{y}_T) - (f(x^*, w_T) + \psi(y^*)) + \frac{1}{T\rho} \langle \lambda^*, \lambda_T - \lambda_1 \rangle] \\ & \leq \mathbb{E}_{w_{1:T}} \left[\frac{1}{T} \sum_{t=1}^T (f(x_t, w_T) + \psi(y_t)) - (f(x^*, w_T) + \psi(y^*)) + \frac{1}{T\rho} \langle \lambda^*, \lambda_T - \lambda_1 \rangle \right] \\ & = \mathbb{E}_{w_{1:T}} \left[\frac{1}{T} \sum_{t=1}^T (f(x_t, w_t) + \psi(y_t)) - (f(x^*, w_t) + \psi(y^*)) \right. \\ & \quad \left. + \frac{1}{T} \sum_{t=1}^T \begin{pmatrix} A^\top \tilde{\lambda}_t \\ B^\top \tilde{\lambda}_t \\ Ax_t + By_t - b \end{pmatrix}^\top \begin{pmatrix} x^* - x_t \\ y^* - y_t \\ \tilde{\lambda}_t - \lambda^* \end{pmatrix} \right]. \end{aligned} \quad (\text{S-12})$$

By combining this with Theorem 4 (and Eq. (S-8)), we have:

$$\begin{aligned} & \mathbb{E}_{w_{1:T}}[(f(\bar{x}_T, w_T) + \psi(\bar{y}'_T)) - (f(x^*, w_T) + \psi(y^*))] + \mathbb{E}_{w_{1:T}} \left[\frac{1}{T} \sum_{t=1}^T \frac{\|\lambda_t - \lambda_{t+1}\|^2}{2\rho} \right] \\ & \leq \Xi_T. \end{aligned} \quad (\text{S-13})$$

Finally we lower bound the second term of the LHS of Eq. (S-9). Remind that $\lambda_t - \lambda_{t+1} = \rho(Ax_{t+1} + By_{t+1} - b)$ (Eq. (S-10)). Thus

$$\begin{aligned} \mathbb{E}_{w_{1:T}} \left[\frac{1}{T} \sum_{t=1}^T \frac{\|\lambda_t - \lambda_{t+1}\|^2}{2\rho} \right] & = \mathbb{E}_{w_{1:T}} \left[\frac{\rho}{2T} \sum_{t=1}^T \|Ax_{t+1} + By_{t+1} - b\|^2 \right] \\ & = \frac{T+1}{T} \mathbb{E}_{w_{1:T}} \left[\frac{\rho}{2(T+1)} \sum_{t=1}^{T+1} \|Ax_t + By_t - b\|^2 \right] \quad (\because \text{the definition of } x_1, y_1) \\ & \geq \frac{\rho}{2} \mathbb{E}_{w_{1:T}} \left[\left\| \frac{1}{T+1} \sum_{t=1}^{T+1} (Ax_t + By_t - b) \right\|^2 \right] \\ & = \frac{\rho}{2} \mathbb{E}_{w_{1:T}} \left[\|A\bar{x}_{T+1} + B\bar{y}_{T+1} - b\|^2 \right]. \end{aligned} \quad (\text{S-14})$$

This gives the assertion. \square

Now substituting $\eta_t = \eta_0/\sqrt{t}$, Theorem 5 gives that, for all $\sigma \geq 0$,

$$\begin{aligned} & \mathbb{E}_{w_{1:T}}[(f(\bar{x}_T, w_T) + \psi(\bar{y}'_T)) - (f(x^*, w_T) + \psi(y^*))] \\ & \leq \frac{\gamma}{T} \sum_{t=2}^T \left(\frac{1}{4\eta_0\sqrt{t-1}} \right) R^2 + \frac{1}{T} \sum_{t=1}^T \frac{\eta_0}{2\sqrt{t}} G^2 + \frac{K}{T} \\ & \leq \frac{\gamma}{2\eta_0\sqrt{T}} R^2 + \frac{\eta_0}{\sqrt{T}} G^2 + \frac{K}{T} \\ & \leq \frac{C_2}{\sqrt{T}}, \end{aligned} \quad (\text{S-15})$$

where C_2 is a constant depending on $R, G, A, B, L_\psi, \rho, \eta_0, \gamma$. Moreover if $\sigma > 0$, by letting $\eta_t = \frac{\gamma}{\sigma t}$, we have that

$$\begin{aligned} & \mathbb{E}_{w_{1:T}}[(f(\bar{x}_T, w_T) + \psi(\bar{y}_T)) - (f(x^*, w_T) + \psi(y^*))] \\ & \leq C_2' \frac{\log(T)}{T}, \end{aligned} \quad (\text{S-16})$$

C_2' is a constant depending on $R, G, A, B, L_\psi, \rho, \eta_0, \gamma, \sigma$.

Theorem 6. Suppose ψ is Lipschitz continuous with a constant L_ψ , i.e., $|\psi(y) - \psi(y')| \leq L_\psi \|y - y'\|$ ($\forall y, y' \in \mathcal{Y}$), and A is invertible. We utilize $x'_t := A^{-1}(b - By_t)$ instead of x_t at the t -th step, and let $\bar{x}'_t := \frac{1}{t} \sum_{\tau=1}^t x'_\tau$. Then, for all $x^* \in \mathcal{X}, y^* \in \mathcal{Y}$ such that $Ax^* + By^* - b = 0$, there exists a constant K depending on R, G, A, B, ρ, η_1 such that

$$\begin{aligned} & \mathbb{E}_{w_{1:T}}[(f(\bar{x}'_T, w_T) + \psi(\bar{y}_T)) - (f(x^*, w_T) + \psi(y^*))] \\ & \leq \frac{1}{T} \sum_{t=2}^T \left(\frac{\gamma}{2\eta_t} - \frac{\gamma}{2\eta_{t-1}} - \frac{\sigma}{2} \right) R^2 + \frac{1}{T} \sum_{t=1}^T \frac{\eta_t}{2} G^2 + \frac{K}{T}. \end{aligned}$$

In particular, for $\eta_t = \eta_0/\sqrt{t}$, the RHS is further bounded as

$$\begin{aligned} & \mathbb{E}_{w_{1:T}}[(f(\bar{x}_T, w_T) + \psi(\bar{y}_T)) - (f(x^*, w_T) + \psi(y^*))] \\ & \leq \frac{C_2}{\sqrt{T}}, \end{aligned} \quad (\text{S-17})$$

where C_2 is a constant depending on $R, G, A, B, L_\psi, \rho, \eta_0, \gamma$. Moreover, if $\sigma > 0$, by letting $\eta_t = \gamma/(\sigma t)$, we have that

$$\begin{aligned} & \mathbb{E}_{w_{1:T}}[(f(\bar{x}_T, w_T) + \psi(\bar{y}_T)) - (f(x^*, w_T) + \psi(y^*))] \\ & \leq C_2' \frac{\log(T)}{T}, \end{aligned} \quad (\text{S-18})$$

where C_2' is a constant depending on $R, G, A, B, L_\psi, \rho, \eta_0, \gamma, \sigma$.

Proof. Eq. (S-10) gives

$$\begin{aligned} f(\bar{x}'_T, w_T) + \psi(\bar{y}_T) & \leq f(\bar{x}'_T, w_T) + \psi(\bar{y}_T) + \langle \nabla_x f(x, w_T)|_{x=\bar{x}'_T}, \bar{x}'_T - \bar{x}_T \rangle \\ & = f(\bar{x}_T, w_T) + \psi(\bar{y}_T) + \langle \nabla_x f(x, w_T)|_{x=\bar{x}'_T}, A^{-1}(b - B\bar{y} - A\bar{x}_T) \rangle \\ & = f(\bar{x}_T, w_T) + \psi(\bar{y}_T) + \frac{1}{T\rho} \langle \nabla_x f(x, w_T)|_{x=\bar{x}'_T}, A^{-1}(\lambda_T - \lambda_1) \rangle. \end{aligned}$$

Thus if we set

$$\lambda^* = A^{-\top} \nabla_x f(x, w_T)|_{x=\bar{x}'_T},$$

(note that $\|A^{-\top} \nabla_x f(x, w_T)|_{x=\bar{x}'_T}\| \leq \|A^{-1}\|G$), then we have

$$f(\bar{x}'_T, w_T) + \psi(\bar{y}_T) \leq f(\bar{x}_T, w_T) + \psi(\bar{y}_T) + \frac{1}{T\rho} \langle \lambda^*, \lambda_T - \lambda_1 \rangle.$$

With the same reasoning as Eq. (S-12), we have

$$\begin{aligned} & \mathbb{E}_{w_{1:T}}[(f(\bar{x}'_T, w_T) + \psi(\bar{y}_T)) - (f(x^*, w_T) + \psi(y^*))] \\ & \leq \mathbb{E}_{w_{1:T}} \left[\frac{1}{T} \sum_{t=1}^T (f(x_t, w_t) + \psi(y_t)) - (f(x^*, w_t) + \psi(y^*)) \right. \\ & \quad \left. + \frac{1}{T} \sum_{t=1}^T \begin{pmatrix} A^\top \tilde{\lambda}_t \\ B^\top \tilde{\lambda}_t \end{pmatrix}^\top \begin{pmatrix} x^* - x_t \\ y^* - y_t \\ \tilde{\lambda}_t - \lambda^* \end{pmatrix} \right]. \end{aligned}$$

By combining this with Theorem 4 (and Eq. (S-8)), we have:

$$\begin{aligned} & \mathbb{E}_{w_{1:T}} [(f(\bar{x}'_T, w_T) + \psi(\bar{y}_T)) - (f(x^*, w_T) + \psi(y^*))] \\ & \leq \Xi_T. \end{aligned} \quad (\text{S-19})$$

This gives the assertion. \square

B. Convergence rate of RDA-ADMM

We define the generalized RDA-ADMM corresponding to (S-1) as follows: Let $x_1 = \mathbf{0}$, $\lambda_1 = \mathbf{0}$, and $By_1 = b - Ax_1$, and the update of the t -th step is given by

$$x_{t+1} = \operatorname{argmin}_{x \in \mathcal{X}} \left\{ \bar{g}_t^\top x - \bar{\lambda}_t^\top Ax + \frac{\rho}{2t} \|Ax\|^2 + \rho(A\bar{x}_t + B\bar{y}_t - b)^\top Ax + \frac{1}{2\eta_t} \|x\|_{G_t}^2 \right\}, \quad (\text{S-20a})$$

$$y_{t+1} = \operatorname{argmin}_{y \in \mathcal{Y}} \left\{ \psi(y) - \lambda_t^\top (Ax_{t+1} + By - b) + \frac{\rho}{2} \|Ax_{t+1} + By - b\|^2 \right\}, \quad (\text{S-20b})$$

$$\lambda_{t+1} = \lambda_t - \rho(Ax_{t+1} + By_{t+1} - b), \quad (\text{S-20c})$$

where $G_t = \gamma I - \frac{\rho\eta_t}{t} A^\top A$. As in the analysis of OPG-ADMM, we define

$$\tilde{\lambda}_t = \lambda_t - \rho(Ax_{t+1} + By_t - b). \quad (\text{S-21})$$

Moreover we suppose that η_t/t is non-increasing.

Theorem 7. *For all $x^* \in \mathcal{X}$, $y^* \in \mathcal{Y}$ and $\lambda^* \in \mathbb{R}^l$, we have*

$$\begin{aligned} & \frac{1}{T} \sum_{t=1}^T (f(x_t, w_t) + \psi(y_t)) - \frac{1}{T} \sum_{t=1}^T (f(x^*, w_t) + \psi(y^*)) \\ & + \frac{1}{T} \sum_{t=1}^T \begin{pmatrix} A^\top \tilde{\lambda}_t \\ B^\top \tilde{\lambda}_t \\ Ax_t + By_t - b \end{pmatrix}^\top \begin{pmatrix} x^* - x_t \\ y^* - y_t \\ \tilde{\lambda}_t - \lambda^* \end{pmatrix} + \frac{1}{T} \sum_{t=1}^T \frac{\|\lambda_t - \lambda_{t+1}\|^2}{2\rho} \\ & \leq \frac{1}{T} \sum_{t=2}^T \frac{\eta_{t-1}}{2(t-1)} \|g_t\|_{G_{t-1}^{-1}}^2 + \frac{\eta_1}{T} \|g_1\|_{G_1^{-1}}^2 + \frac{1}{\eta_T} \|x^*\|_{G_T} \\ & + \frac{\rho}{2T} \|b - By^*\|^2 + \frac{\|\lambda^*\|^2}{2\rho T} \\ & + \frac{1}{T} (\langle y^* - y_{T+1}, B^\top (\lambda_{T+1} - \lambda^*) \rangle - \langle By^* - b, \lambda^* \rangle) \\ & + \frac{\rho}{2T} \|A(x_{T+1} - x^*)\|^2 + \frac{1}{T} \langle \lambda^*, Ax_{T+1} \rangle. \end{aligned}$$

Proof. First observe that

$$\begin{aligned} \sum_{t=1}^T (f_t(x_t) + \psi(y_t) - f_t(x^*) - \psi(y^*)) & \leq \sum_{t=1}^T (\langle g_t, x_t - x^* \rangle + \langle \nabla \psi(y_t), y_t - y^* \rangle) \\ & = \sum_{t=1}^T (\langle g_t, x_t - x^* \rangle + \langle B^\top \lambda_t, y_t - y^* \rangle) \\ & = \sum_{t=1}^T (\langle g_t, x_t \rangle + \langle \lambda_t, B(y_t - y^*) \rangle) - T \langle \bar{g}_T, x^* \rangle. \end{aligned} \quad (\text{S-22})$$

Here we define ϱ_t and V_t as follows:

$$\begin{aligned}\varrho_t(g, x) &= \langle g, x \rangle + \langle \bar{\lambda}_t, Ax \rangle - \frac{\rho}{2t} \|Ax\|^2 - \rho(A\bar{x}_t + B\bar{y}_t - b)^\top Ax - \frac{1}{2\eta_t} \|x\|_{G_t}^2, \\ V_t(g) &= \max_{x \in \mathcal{X}} \{\varrho_t(g, x)\}.\end{aligned}$$

Obviously $V_t(-\bar{g}_t) = \varrho_t(-\bar{g}_t, x_{t+1})$ by the update rule of x_t . In particular $\langle x' - x_t, \nabla_x \varrho_{t-1}(-\bar{g}_{t-1}, x)|_{x=x_t} \rangle \leq 0$ for all $x' \in \mathcal{X}$. Because $\varrho_{t-1}(-\bar{g}_{t-1}, \cdot)$ is a concave quadratic function, we have

$$\begin{aligned}& \varrho_{t-1}(-\bar{g}_{t-1}, x_{t+1}) \\ & \leq \varrho_{t-1}(-\bar{g}_{t-1}, x_t) + \langle x_{t+1} - x_t, \nabla_x \varrho_{t-1}(-\bar{g}_{t-1}, x)|_{x=x_t} \rangle + \frac{1}{2} (x_{t+1} - x_t)^\top (\nabla_x^\top \nabla_x \varrho_{t-1}(-\bar{g}_{t-1}, x)|_{x=x_t}) (x_{t+1} - x_t) \\ & \leq \varrho_{t-1}(-\bar{g}_{t-1}, x_t) + \frac{1}{2} (x_{t+1} - x_t)^\top (\nabla_x^\top \nabla_x \varrho_{t-1}(-\bar{g}_{t-1}, x)|_{x=x_t}) (x_{t+1} - x_t).\end{aligned}$$

Here note that $\nabla_x^\top \nabla_x \varrho_{t-1}(-\bar{g}_{t-1}, x)|_{x=x_t} = -\frac{\rho}{t-1} A^\top A - \frac{G_{t-1}}{\eta_{t-1}}$. Therefore

$$\begin{aligned}\varrho_{t-1}(-\bar{g}_{t-1}, x_{t+1}) & \leq \varrho_{t-1}(-\bar{g}_{t-1}, x_t) - \frac{\rho}{2(t-1)} \|A(x_{t+1} - x_t)\|^2 - \frac{1}{2\eta_{t-1}} \|x_{t+1} - x_t\|_{G_{t-1}}^2 \\ & = V_{t-1}(-\bar{g}_{t-1}) - \frac{\rho}{2(t-1)} \|A(x_{t+1} - x_t)\|^2 - \frac{1}{2\eta_{t-1}} \|x_{t+1} - x_t\|_{G_{t-1}}^2.\end{aligned}$$

Using this inequality, we can compare $V_t(-\bar{g}_t)$ to $V_{t-1}(-\bar{g}_{t-1})$:

$$\begin{aligned}tV_t(-\bar{g}_t) &= t\varrho_t(-\bar{g}_t, x_{t+1}) \\ &\leq (t-1)\varrho_{t-1}(-\bar{g}_{t-1}, x_{t+1}) + \langle -g_t, x_{t+1} \rangle + \langle \lambda_t, Ax_{t+1} \rangle - \rho\langle Ax_{t+1}, Ax_t + By_t - b \rangle\end{aligned}\quad (\text{S-23})$$

$$\begin{aligned}& \leq (t-1)V_{t-1}(-\bar{g}_{t-1}) - \frac{\rho}{2} \|A(x_{t+1} - x_t)\|^2 - \frac{t-1}{2\eta_{t-1}} \|x_{t+1} - x_t\|_{G_{t-1}}^2 \\ & \quad - \langle g_t, x_t \rangle + \langle g_t, x_t - x_{t+1} \rangle + \langle \lambda_t, Ax_{t+1} \rangle - \rho\langle Ax_{t+1}, Ax_t + By_t - b \rangle.\end{aligned}\quad (\text{S-24})$$

Since $\langle g_t, x_t - x_{t+1} \rangle \leq \|g_t\|_{G_{t-1}^{-1}} \|x_t - x_{t+1}\|_{G_{t-1}} \leq \frac{\eta_{t-1}}{2(t-1)} \|g_t\|_{G_{t-1}^{-1}}^2 + \frac{t-1}{2\eta_{t-1}} \|x_t - x_{t+1}\|_{G_{t-1}}^2$, the RHS is further bounded by

$$\begin{aligned}tV_t(-\bar{g}_t) & \leq (t-1)V_{t-1}(-\bar{g}_{t-1}) - \frac{\rho}{2} \|A(x_{t+1} - x_t)\|^2 + \frac{\eta_{t-1}}{2(t-1)} \|g_t\|_{G_t^{-1}}^2 \\ & \quad - \langle g_t, x_t \rangle + \langle \lambda_t, Ax_{t+1} \rangle - \rho\langle Ax_{t+1}, Ax_t + By_t - b \rangle.\end{aligned}$$

Moreover

$$\begin{aligned}V_1(-g_1) &= \langle -g_1, x_2 \rangle + \langle \lambda_1, Ax_2 \rangle - \frac{\rho}{2} \|Ax_2\|^2 - \rho\langle Ax_1 + By_1 - b, Ax_2 \rangle - \frac{1}{2\eta_1} \|x_2\|_{G_1}^2 \\ & \leq \langle -g_1, x_1 \rangle + \langle \lambda_1, Ax_2 \rangle - \frac{\rho}{2} \|A(x_1 - x_2)\|^2 - \rho\langle Ax_1 + By_1 - b, Ax_2 \rangle \\ & \quad - \frac{1}{2\eta_1} \|x_2 + \eta_1 G_1^{-1} g_1\|_{G_1}^2 + \frac{\eta_1}{2} \|g_1\|_{G_1^{-1}}^2 + \langle g_1, x_1 \rangle - \rho\langle Ax_2, Ax_1 \rangle + \frac{\rho}{2} \|Ax_1\|^2.\end{aligned}$$

Summing up $t = 1, \dots, T$, we have

$$\begin{aligned}TV_T(-\bar{g}_T) & \leq -\frac{\rho}{2} \sum_{t=1}^T \|A(x_{t+1} - x_t)\|^2 + \sum_{t=2}^T \frac{\eta_{t-1}}{2(t-1)} \|g_t\|_{G_t^{-1}}^2 \\ & \quad - \sum_{t=1}^T \langle g_t, x_t \rangle + \sum_{t=1}^T \langle \lambda_t, Ax_{t+1} \rangle - \sum_{t=1}^T \rho\langle Ax_{t+1}, Ax_t + By_t - b \rangle \\ & \quad + \frac{\eta_1}{2} \|g_1\|_{G_1^{-1}}^2 + \langle g_1, x_1 \rangle - \rho\langle Ax_2, Ax_1 \rangle + \frac{\rho}{2} \|Ax_1\|^2.\end{aligned}\quad (\text{S-25})$$

Using this inequality, we observe that

$$\begin{aligned}
 & - \langle T\bar{g}_T, x^* \rangle \\
 = & T\varrho_T(-\bar{g}_T, x^*) - \langle T\bar{\lambda}_T, Ax^* \rangle + \frac{\rho}{2}\|Ax^*\|^2 + T\rho\langle Ax^*, A\bar{x}_T + B\bar{y}_T - b \rangle + \frac{T}{\eta_T}\|x^*\|_{G_T}^2 \\
 \leq & TV_T(-\bar{g}_T) - \langle T\bar{\lambda}_T, Ax^* \rangle + \frac{\rho}{2}\|Ax^*\|^2 + T\rho\langle Ax^*, A\bar{x}_T + B\bar{y}_T - b \rangle + \frac{T}{\eta_T}\|x^*\|_{G_T}^2 \\
 \leq & -\frac{\rho}{2}\sum_{t=1}^T \|A(x_{t+1} - x_t)\|^2 - \sum_{t=2}^T \frac{\eta_{t-1}}{2(t-1)} \|g_t\|_{G_t^{-1}}^2 \\
 & - \sum_{t=1}^T \langle g_t, x_t \rangle + \sum_{t=1}^T \langle \lambda_t, Ax_{t+1} \rangle - \sum_{t=1}^T \rho\langle Ax_{t+1}, Ax_t + By_t - b \rangle \\
 & + \frac{\eta_1}{2}\|g_1\|_{G_1^{-1}}^2 + \langle g_1, x_1 \rangle - \rho\langle Ax_2, Ax_1 \rangle + \frac{\rho}{2}\|Ax_1\|^2 \\
 & - \langle T\bar{\lambda}_T, Ax^* \rangle + \frac{\rho}{2}\|Ax^*\|^2 + T\rho\langle Ax^*, A\bar{x}_T + B\bar{y}_T - b \rangle + \frac{T}{\eta_T}\|x^*\|_{G_T}^2 \quad (\because \text{Eq. (S-25)}) \\
 = & -\frac{\rho}{2}\sum_{t=1}^T \|A(x_{t+1} - x_t)\|^2 + \sum_{t=2}^T \frac{\eta_{t-1}}{2(t-1)} \|g_t\|_{G_t^{-1}}^2 \\
 & - \sum_{t=1}^T \langle g_t, x_t \rangle + \sum_{t=1}^T \langle \lambda_t - \rho(Ax_t + By_t - b), A(x_{t+1} - x^*) \rangle + \frac{\rho}{2}\|Ax^*\|^2 + \frac{T}{\eta_T}\|x^*\|_{G_T}^2 \\
 & + \frac{\eta_1}{2}\|g_1\|_{G_1^{-1}}^2 + \langle g_1, x_1 \rangle - \rho\langle Ax_2, Ax_1 \rangle + \frac{\rho}{2}\|Ax_1\|^2.
 \end{aligned}$$

Substituting this inequality into the RHS of Eq. (S-22), we obtain

$$\begin{aligned}
 & \sum_{t=1}^T (\langle g_t, x_t \rangle + \langle \lambda_t, B(y_t - y^*) \rangle) - T\langle \bar{g}_T, x^* \rangle \\
 \leq & \sum_{t=1}^T \langle \lambda_t, B(y_t - y^*) \rangle + \sum_{t=1}^T \langle \lambda_t - \rho(Ax_t + By_t - b), A(x_{t+1} - x^*) \rangle \\
 & - \frac{\rho}{2}\sum_{t=1}^T \|A(x_{t+1} - x_t)\|^2 + \sum_{t=2}^T \frac{\eta_{t-1}}{2(t-1)} \|g_t\|_{G_t^{-1}}^2 \\
 & + \frac{\rho}{2}\|Ax^*\|^2 + \frac{T}{\eta_T}\|x^*\|_{G_T}^2 \\
 & + \frac{\eta_1}{2}\|g_1\|_{G_1^{-1}}^2 + \langle g_1, x_1 \rangle - \rho\langle Ax_2, Ax_1 \rangle + \frac{\rho}{2}\|Ax_1\|^2. \tag{S-26}
 \end{aligned}$$

From now on, we bound the first two terms of the RHS: (i) $\sum_{t=1}^T \langle \lambda_t, B(y_t - y^*) \rangle$ and (ii) $\sum_{t=1}^T \langle \lambda_t - \rho(Ax_t + By_t - b), A(x_{t+1} - x^*) \rangle$.

(i) *Evaluating* $\sum_{t=1}^T \langle \lambda_t, B(y_t - y^*) \rangle$. We have

$$\langle \lambda_t, B(y_t - y^*) \rangle = \langle \tilde{\lambda}_t, B(y_t - y^*) \rangle + \langle B(y_t - y^*), \lambda_t - \tilde{\lambda}_t \rangle. \tag{S-27}$$

(ii) *Evaluating* $\sum_{t=1}^T \langle \lambda_t - \rho(Ax_t + By_t - b), A(x_{t+1} - x^*) \rangle$. By the definition of $\tilde{\lambda}_t$, we have

$$\begin{aligned}
 & \langle \lambda_t - \rho(Ax_t + By_t - b), A(x_{t+1} - x^*) \rangle = \langle \tilde{\lambda}_t + \rho A(x_{t+1} - x_t), A(x_{t+1} - x^*) \rangle \\
 = & \langle \tilde{\lambda}_t, A(x_{t+1} - x^*) \rangle + \langle \tilde{\lambda}_t - \lambda_t^*, A(x_{t+1} - x_t) \rangle + \langle \lambda_t^*, A(x_{t+1} - x_t) \rangle \\
 & + \rho\langle A(x_{t+1} - x_t), A(x_{t+1} - x^*) \rangle
 \end{aligned}$$

$$\begin{aligned}
 &\leq \langle \tilde{\lambda}_t, A(x_t - x^*) \rangle + \langle \tilde{\lambda}_t - \lambda^*, A(x_{t+1} - x_t) \rangle + \langle \lambda^*, A(x_{t+1} - x_t) \rangle \\
 &\quad + \frac{\rho}{2} \|A(x_{t+1} - x^*)\|^2 - \frac{\rho}{2} \|A(x_t - x^*)\|^2 + \frac{\rho}{2} \|A(x_{t+1} - x_t)\|^2 \quad (\because \text{Lemma 10}). \tag{S-28}
 \end{aligned}$$

Here by substituting the relation $A(x_{t+1} - x_t) = \frac{\tilde{\lambda}_t - \lambda_t}{\rho} - (Ax_t + By_t - b)$ to the second term in the RHS, we have

$$\begin{aligned}
 &\langle \lambda_t - \rho(Ax_t + By_t - b), A(x_{t+1} - x^*) \rangle \\
 &= \langle \tilde{\lambda}_t, A(x_t - x^*) \rangle + \left\langle \tilde{\lambda}_t - \lambda^*, \frac{\tilde{\lambda}_t - \lambda_t}{\rho} - (Ax_t + By_t - b) \right\rangle + \langle \lambda^*, A(x_{t+1} - x_t) \rangle \\
 &\quad + \frac{\rho}{2} \|A(x_{t+1} - x^*)\|^2 - \frac{\rho}{2} \|A(x_t - x^*)\|^2 + \frac{\rho}{2} \|A(x_{t+1} - x_t)\|^2.
 \end{aligned}$$

Substituting Eqs. (S-27),(S-28) into the RHS of Eq. (S-30), we have that

$$\begin{aligned}
 &\sum_{t=1}^T (\langle g_t, x_t \rangle + \langle \lambda_t, B(y_t - y^*) \rangle) - T \langle \bar{g}_T, x^* \rangle \\
 &\leq \sum_{t=1}^T \begin{pmatrix} -A^\top \tilde{\lambda}_t \\ -B^\top \tilde{\lambda}_t \\ Ax_t + By_t - b \end{pmatrix}^\top \begin{pmatrix} x^* - x_t \\ y^* - y_t \\ \lambda^* - \tilde{\lambda}_t \end{pmatrix} + \sum_{t=1}^T \begin{pmatrix} y^* - y_t \\ \lambda^* - \tilde{\lambda}_t \end{pmatrix}^\top \begin{pmatrix} -B^\top (\lambda_t - \tilde{\lambda}_t) \\ \frac{\tilde{\lambda}_t - \lambda_t}{\rho} \end{pmatrix} \\
 &\quad + \langle \lambda^*, A(x_{T+1} - x_1) \rangle + \frac{\rho}{2} \|A(x_{T+1} - x^*)\|^2 - \frac{\rho}{2} \|A(x_1 - x^*)\|^2 + \sum_{t=2}^T \frac{\eta_{t-1}}{2(t-1)} \|g_t\|_{G_t^{-1}}^2 \\
 &\quad + \frac{\rho}{2} \|Ax^*\|^2 + \frac{T}{\eta_T} \|x^*\|_{G_T}^2 + \frac{\eta_1}{2} \|g_1\|_{G_1^{-1}}^2 + \langle g_1, x_1 \rangle - \rho \langle Ax_2, Ax_1 \rangle + \frac{\rho}{2} \|Ax_1\|^2. \tag{S-29}
 \end{aligned}$$

Finally we bound the second term of the RHS using Lemma 11, we obtain

$$\begin{aligned}
 &\sum_{t=1}^T (\langle g_t, x_t \rangle + \langle \lambda_t, B(y_t - y^*) \rangle) - T \langle \bar{g}_T, x^* \rangle \\
 &\leq \sum_{t=1}^T \begin{pmatrix} -A^\top \tilde{\lambda}_t \\ -B^\top \tilde{\lambda}_t \\ Ax_t + By_t - b \end{pmatrix}^\top \begin{pmatrix} x^* - x_t \\ y^* - y_t \\ \lambda^* - \tilde{\lambda}_t \end{pmatrix} \\
 &\quad + \frac{\rho}{2} \|y_1 - y^*\|_{B^\top B}^2 + \frac{\|\lambda_1 - \lambda^*\|^2}{2\rho} \\
 &\quad + \langle B(y^* - y_{T+1}), \lambda_{T+1} - \lambda^* \rangle - \langle B(y^* - y_1), \lambda_1 - \lambda^* \rangle - \sum_{t=1}^T \frac{\|\lambda_t - \lambda_{t+1}\|^2}{2\rho} \\
 &\quad + \langle \lambda^*, A(x_{T+1} - x_1) \rangle \\
 &\quad + \frac{\rho}{2} \|A(x_{T+1} - x^*)\|^2 - \frac{\rho}{2} \|A(x_1 - x^*)\|^2 + \sum_{t=2}^T \frac{\eta_{t-1}}{2(t-1)} \|g_t\|_{G_t^{-1}}^2 \\
 &\quad + \frac{\rho}{2} \|Ax^*\|^2 + \frac{T}{\eta_T} \|x^*\|_{G_T}^2 + \frac{\eta_1}{2} \|g_1\|_{G_1^{-1}}^2 + \langle g_1, x_1 \rangle - \rho \langle Ax_2, Ax_1 \rangle + \frac{\rho}{2} \|Ax_1\|^2. \tag{S-30}
 \end{aligned}$$

This, Eq. (S-22) and the initial settings of x_1, y_1, λ_1 give the assertion. \square

Here we simplify Theorem 7. First note that by Eq. (S-5)

$$\langle y^* - y_{T+1}, B^\top (\lambda_{T+1} - \lambda^*) \rangle \leq \langle y^* - y_{T+1}, \nabla \psi(y_{T+1}) - B^\top \lambda^* \rangle.$$

Since the diameters of \mathcal{X} and \mathcal{Y} are bounded by R , $\|g_t\| \leq G$, and the subgradient of ψ is bounded by L_ψ , if B is invertible, then there exists a constant K depending on $R, G, L_\psi, \rho, A, B, \eta_1, \lambda^*$ such that the bound shown in Theorem 7 can be further bounded as

$$\begin{aligned} & \frac{1}{T} \sum_{t=1}^T (f(x_t, w_t) + \psi(y_t)) - \frac{1}{T} \sum_{t=1}^T (f(x^*, w_t) + \psi(y^*)) \\ & - \langle \lambda^*, A\bar{x}_T + B\bar{y}_T - (Ax^* + By^*) \rangle + \frac{1}{T} \sum_{t=1}^T \frac{\|\lambda_t - \lambda_{t+1}\|^2}{2\rho} \\ & \leq \frac{1}{T} \sum_{t=2}^T \frac{\eta_{t-1}}{2(t-1)} G^2 + \frac{\gamma}{\eta_T} R^2 + \frac{K}{T}, \end{aligned} \quad (\text{S-31})$$

where we used Eq. (S-11).

Theorem 8. Suppose ψ is Lipschitz continuous with a constant L_ψ , i.e., $|\psi(y) - \psi(y')| \leq L_\psi \|y - y'\|$ ($\forall y, y' \in \mathcal{Y}$), and B is invertible. We utilize $y'_t := B^{-1}(b - Ax_t)$ as an estimator of y at the t -th step, and let $\bar{y}'_t := \frac{1}{t} \sum_{\tau=1}^t y'_\tau$. Then, for all $x^* \in \mathcal{X}, y^* \in \mathcal{Y}$ such that $Ax^* + By^* - b = 0$, there exists a constant K depending on $R, G, L_\psi, \rho, A, B, \eta_1$ such that

$$\begin{aligned} & \mathbb{E}_{w_{1:T}} [(f(\bar{x}_T, w_T) + \psi(\bar{y}'_T)) - (f(x^*, w_T) + \psi(y^*))] \\ & \leq \frac{1}{T} \sum_{t=2}^T \frac{\eta_{t-1}}{2(t-1)} G^2 + \frac{\gamma}{\eta_T} R^2 + \frac{K}{T}. \end{aligned}$$

Proof. Using $\eta_t = \eta_0 \sqrt{t}$ and Theorem 7 instead of $\eta_t = \eta_0 / \sqrt{t}$ and Theorem 4 respectively, the same proof as Theorem 5 yields the assertion. \square

Substituting $\eta_t = \eta_0 \sqrt{t}$, the bound in Theorem 8 can be simplified as

$$\begin{aligned} & \mathbb{E}_{w_{1:T}} [(f(\bar{x}_T, w_T) + \psi(\bar{y}'_T)) - (f(x^*, w_T) + \psi(y^*))] \\ & \leq \frac{1}{T} \sum_{t=2}^T \frac{\eta_0 \sqrt{t-1}}{2(t-1)} G^2 + \frac{\gamma}{\eta_0 \sqrt{T}} R^2 + \frac{K}{T} \\ & \leq \frac{\eta_0}{\sqrt{T}} G^2 + \frac{\gamma}{\eta_0 \sqrt{T}} R^2 + \frac{K}{T} \\ & \leq \frac{C_2}{\sqrt{T}}, \end{aligned}$$

where C_2 is a constant depending on $R, G, L_\psi, \rho, \eta_0, A, B, G, \gamma$.

Theorem 9. Suppose ψ is Lipschitz continuous with a constant L_ψ , i.e., $|\psi(y) - \psi(y')| \leq L_\psi \|y - y'\|$ ($\forall y, y' \in \mathcal{Y}$), and A is invertible. We utilize $x'_t := A^{-1}(b - By_t)$ instead of x_t at the t -th step, and let $\bar{x}'_t := \frac{1}{t} \sum_{\tau=1}^t x'_\tau$. Then, for all $x^* \in \mathcal{X}, y^* \in \mathcal{Y}$ such that $Ax^* + By^* - b = 0$, there exists a constant K depending on R, G, A, B, ρ, η_1 such that

$$\begin{aligned} & \mathbb{E}_{w_{1:T}} [(f(\bar{x}'_T, w_T) + \psi(\bar{y}_T)) - (f(x^*, w_T) + \psi(y^*))] \\ & \leq \frac{1}{T} \sum_{t=2}^T \frac{\eta_{t-1}}{2(t-1)} G^2 + \frac{\gamma}{\eta_T} R^2 + \frac{K}{T}. \end{aligned}$$

In particular, for $\eta_t = \eta_0 \sqrt{t}$, the RHS is further bounded as

$$\mathbb{E}_{w_{1:T}} [(f(\bar{x}'_T, w_T) + \psi(\bar{y}'_T)) - (f(x^*, w_T) + \psi(y^*))] \leq \frac{C_2}{\sqrt{T}}, \quad (\text{S-32})$$

where C_2 is a constant depending on $R, G, A, B, L_\psi, \rho, \eta_0, \gamma$.

Proof. Using $\eta_t = \eta_0 \sqrt{t}$ and Theorem 7 instead of $\eta_t = \eta_0 / \sqrt{t}$ and Theorem 4 respectively, the same proof as Theorem 6 yields the assertion. \square

C. Auxiliary Lemmas

Lemma 10. For all symmetric matrix H , we have

$$(a-b)^\top H(c-b) = \frac{1}{2}\|a-b\|_H^2 - \frac{1}{2}\|a-c\|_H^2 + \frac{1}{2}\|c-b\|_H^2. \quad (\text{S-33})$$

Proof.

$$\begin{aligned} (a-b)^\top H(c-b) &= \left(a - \frac{c+b}{2} + \frac{c+b}{2} - b\right)^\top H(c-b) \\ &= \left(\frac{a-c}{2} + \frac{a-b}{2}\right)^\top H(c-b) + \left(\frac{c-b}{2}\right)^\top H(c-b) \\ &= \left(\frac{a-c}{2} + \frac{a-b}{2}\right)^\top H\{(a-b) - (a-c)\} + \left(\frac{c-b}{2}\right)^\top H(c-b) \\ &= \frac{1}{2}\|a-b\|_H^2 - \frac{1}{2}\|a-c\|_H^2 + \frac{1}{2}\|c-b\|_H^2. \end{aligned}$$

□

Lemma 11. Under the update rule (S-2) or (S-20), we have the following bound:

$$\begin{aligned} &\begin{pmatrix} y^* - y_t \\ \lambda^* - \tilde{\lambda}_t \end{pmatrix}^\top \begin{pmatrix} -B^\top(\lambda_t - \tilde{\lambda}_t) \\ \frac{\tilde{\lambda}_t - \lambda_t}{\rho} \end{pmatrix} \\ &\leq \frac{\rho}{2}\|y_t - y^*\|_{B^\top B}^2 - \frac{\rho}{2}\|y_{t+1} - y^*\|_{B^\top B}^2 + \frac{\|\lambda_t - \lambda^*\|^2}{2\rho} - \frac{\|\lambda_{t+1} - \lambda^*\|^2}{2\rho} \\ &\quad + \langle B(y^* - y_{t+1}), \lambda_{t+1} - \lambda^* \rangle - \langle B(y^* - y_t), \lambda_t - \lambda^* \rangle \\ &\quad - \frac{\|\lambda_t - \lambda_{t+1}\|^2}{2\rho}. \end{aligned} \quad (\text{S-34})$$

Proof. Since $\tilde{\lambda}_t = \lambda_{t+1} - \rho B(y_t - y_{t+1})$, we have

$$\begin{aligned} &\langle y^* - y_t, -B^\top(\lambda_t - \tilde{\lambda}_t) \rangle \\ &= \langle y^* - y_t, -B^\top(\lambda_t - \lambda_{t+1}) \rangle + \langle y^* - y_t, \rho B^\top B(y_{t+1} - y_t) \rangle \\ &\leq \langle y^*, B^\top(\lambda_{t+1} - \lambda_t) \rangle - \langle y_t, B^\top(\lambda_{t+1} - \lambda_t) \rangle \\ &\quad + \frac{\rho}{2}\|y_t - y^*\|_{B^\top B}^2 - \frac{\rho}{2}\|y_{t+1} - y^*\|_{B^\top B}^2 + \frac{\rho}{2}\|y_{t+1} - y_t\|_{B^\top B}^2, \end{aligned} \quad (\text{S-35})$$

where we used Lemma 10 in the last inequality.

On the other hand, by Lemma 10, we have

$$\begin{aligned} &\langle \lambda^* - \tilde{\lambda}_t, (\tilde{\lambda}_t - \lambda_t)/\rho \rangle \\ &= \underbrace{-\frac{\|\tilde{\lambda}_t - \lambda^*\|^2}{2\rho}}_{(i)} + \frac{\|\lambda_t - \lambda^*\|^2}{2\rho} - \underbrace{\frac{\|\tilde{\lambda}_t - \lambda_t\|^2}{2\rho}}_{(ii)}. \end{aligned} \quad (\text{S-36})$$

The first term (i) in Eq. (S-36) can be evaluated as

$$\begin{aligned} -\frac{\|\tilde{\lambda}_t - \lambda^*\|^2}{2\rho} &= -\frac{\|\tilde{\lambda}_t - \lambda_{t+1}\|^2}{2\rho} - \frac{\langle \tilde{\lambda}_t - \lambda_{t+1}, \lambda_{t+1} - \lambda^* \rangle}{\rho} - \frac{\|\lambda_{t+1} - \lambda^*\|^2}{2\rho} \\ &= -\frac{\|\tilde{\lambda}_t - \lambda_{t+1}\|^2}{2\rho} - \frac{\langle \rho B(y_{t+1} - y_t), \lambda_{t+1} - \lambda^* \rangle}{\rho} - \frac{\|\lambda_{t+1} - \lambda^*\|^2}{2\rho}. \end{aligned} \quad (\text{S-37})$$

Since $\tilde{\lambda}_t - \lambda_{t+1} = \rho B(y_{t+1} - y_t)$, the third term (ii) in Eq. (S-36) can be evaluated as

$$\begin{aligned} -\frac{\|\tilde{\lambda}_t - \lambda_t\|^2}{2\rho} &= -\frac{\|\tilde{\lambda}_t - \lambda_{t+1}\|^2}{2\rho} - \frac{1}{\rho} \langle \tilde{\lambda}_t - \lambda_{t+1}, \lambda_{t+1} - \lambda_t \rangle - \frac{\|\lambda_t - \lambda_{t+1}\|^2}{2\rho} \\ &= -\frac{\rho \|B(y_{t+1} - y_t)\|^2}{2} - \frac{1}{\rho} \langle \rho B(y_{t+1} - y_t), \lambda_{t+1} - \lambda_t \rangle - \frac{\|\lambda_t - \lambda_{t+1}\|^2}{2\rho}. \end{aligned} \quad (\text{S-38})$$

Here note that, by Eq. (S-5), we have

$$\langle B(y_{t+1} - y_t), \lambda_{t+1} - \lambda_t \rangle \geq \langle y_{t+1} - y_t, \nabla\psi(y_{t+1}) - \nabla\psi(y_t) \rangle \geq 0,$$

by the monotonicity of subgradients. Therefore the RHS of Eq. (S-38) is further bounded as

$$-\frac{\|\tilde{\lambda}_t - \lambda_t\|^2}{2\rho} \leq -\frac{\rho \|y_{t+1} - y_t\|_{B^\top B}^2}{2} - \frac{\|\lambda_t - \lambda_{t+1}\|^2}{2\rho}. \quad (\text{S-39})$$

Combining Eqs. (S-36), (S-37), (S-39), then

$$\begin{aligned} &\langle \lambda^* - \tilde{\lambda}_t, (\tilde{\lambda}_t - \lambda_t) / \rho \rangle \\ &\leq -\frac{\rho}{2} \|y_{t+1} - y_t\|_{B^\top B}^2 - \frac{\|\lambda_t - \lambda_{t+1}\|^2}{2\rho} + \frac{\|\lambda_t - \lambda^*\|^2}{2\rho} \\ &\quad - \frac{\|\tilde{\lambda}_t - \lambda_{t+1}\|^2}{2\rho} - \langle B(y_{t+1} - y_t), \lambda_{t+1} - \lambda^* \rangle - \frac{\|\lambda_{t+1} - \lambda^*\|^2}{2\rho} \end{aligned} \quad (\text{S-40})$$

Finally we combine Eqs. (S-40) and (S-35) so that we obtain

$$\begin{aligned} &\begin{pmatrix} y^* - y_t \\ \lambda^* - \tilde{\lambda}_t \end{pmatrix}^\top \begin{pmatrix} -B^\top(\lambda_t - \tilde{\lambda}_t) \\ \frac{\tilde{\lambda}_t - \lambda_t}{\rho} \end{pmatrix} \\ &\leq \langle y^*, B^\top(\lambda_{t+1} - \lambda_t) \rangle - \langle y_t, B^\top(\lambda_{t+1} - \lambda_t) \rangle \\ &\quad + \frac{\rho}{2} \|y_t - y^*\|_{B^\top B}^2 - \frac{\rho}{2} \|y_{t+1} - y^*\|_{B^\top B}^2 + \frac{\rho}{2} \|y_{t+1} - y_t\|_{B^\top B}^2, \\ &\quad - \frac{\rho}{2} \|y_{t+1} - y_t\|_{B^\top B}^2 - \frac{\|\lambda_t - \lambda_{t+1}\|^2}{2\rho} + \frac{\|\lambda_t - \lambda^*\|^2}{2\rho} \\ &\quad - \frac{\|\tilde{\lambda}_t - \lambda_{t+1}\|^2}{2\rho} - \langle B(y_{t+1} - y_t), \lambda_{t+1} - \lambda^* \rangle - \frac{\|\lambda_{t+1} - \lambda^*\|^2}{2\rho} \\ &\leq \frac{\rho}{2} \|y_t - y^*\|_{B^\top B}^2 - \frac{\rho}{2} \|y_{t+1} - y^*\|_{B^\top B}^2 + \frac{\|\lambda_t - \lambda^*\|^2}{2\rho} - \frac{\|\lambda_{t+1} - \lambda^*\|^2}{2\rho} \\ &\quad + \langle B(y^* - y_{t+1}), \lambda_{t+1} - \lambda^* \rangle - \langle B(y^* - y_t), \lambda_t - \lambda^* \rangle \\ &\quad - \frac{\|\lambda_t - \lambda_{t+1}\|^2}{2\rho}. \end{aligned}$$

This gives the assertion. \square