

A Proof for Section 2

Throughout our proof, we presume without loss of generality that the elements in $\bar{\mathbf{x}} = (\bar{x}_1, \bar{x}_2, \dots, \bar{x}_d)$ are in descending order by their magnitude, i.e., $|\bar{x}_1| \geq |\bar{x}_2| \geq \dots \geq |\bar{x}_s|$ and $\bar{x}_i = 0$ for $s < i \leq d$. We also write $[n] := \{1, 2, \dots, n\}$ for brevity.

Recall that the partial hard thresholding algorithm with freedom parameter r proceeds as follows at the t -th iteration:

$$\begin{aligned} \mathbf{z}^t &= \mathbf{x}^{t-1} - \eta \nabla F(\mathbf{x}^{t-1}) \\ J^t &= S^{t-1} \cup \text{supp}(\nabla F(\mathbf{x}^{t-1}), r) \\ \mathbf{y}^t &= \text{HT}_k(\mathbf{z}_{J^t}^t) \\ S^t &= \text{supp}(\mathbf{y}^t) \\ \mathbf{x}^t &= \arg \min_{\text{supp}(\mathbf{x}) \subset S^t} F(\mathbf{x}) \end{aligned}$$

We first prove the results that appear in Section 3.

Lemma 8 (Restatement of Lemma 5). *Assume that $F(\mathbf{x})$ is ρ_{2k}^- -RSC and ρ_{2k}^+ -RSS. Consider the PHT(r) algorithm with $\eta < 1/\rho_{2k}^+$. Further assume that the sequence of $\{\mathbf{x}^t\}_{t \geq 0}$ satisfies*

$$\begin{aligned} \|\mathbf{x}^t - \bar{\mathbf{x}}\| &\leq \alpha \cdot \beta^t \|\mathbf{x}^0 - \bar{\mathbf{x}}\| + \psi_1, \\ \|\mathbf{x}^t - \bar{\mathbf{x}}\| &\leq \gamma \|\bar{\mathbf{x}}_{\bar{S}^t}\| + \psi_2, \end{aligned}$$

for positive $\alpha, \psi_1, \gamma, \psi_2$ and $0 < \beta < 1$. Suppose that at the n -th iteration ($n \geq 0$), S^n contains the indices of top p (in magnitude) elements of $\bar{\mathbf{x}}$. Then, for any integer $1 \leq q \leq s - p$, there exists an integer $\Delta \geq 1$ determined by

$$\sqrt{2} |\bar{x}_{p+q}| > \alpha \gamma \cdot \beta^{\Delta-1} \|\bar{\mathbf{x}}_{\{p+1, \dots, s\}}\| + \Psi$$

where

$$\Psi = \alpha \psi_2 + \psi_1 + \frac{1}{\rho_{2k}} \|\nabla_2 F(\bar{\mathbf{x}})\|,$$

such that $S^{n+\Delta}$ contains the indices of top $p+q$ elements of $\bar{\mathbf{x}}$ provided that $\Psi \leq \sqrt{2} \lambda \bar{x}_{\min}$ for some $\lambda \in (0, 1)$.

Proof. We aim at deriving a condition under which $[p+q] \subset S^{n+\Delta}$. To this end, it suffices to enforce

$$\min_{j \in [p+q]} |z_j^{n+\Delta}| > \max_{i \in \bar{S}} |z_i^{n+\Delta}|. \quad (7)$$

On one hand, for any $j \in [p+q]$,

$$\begin{aligned} |z_j^{n+\Delta}| &= \left| (\mathbf{x}^{n+\Delta-1} - \eta \nabla F(\mathbf{x}^{n+\Delta-1}))_j \right| \\ &\geq |\bar{x}_j| - \left| (\mathbf{x}^{n+\Delta-1} - \bar{\mathbf{x}} - \eta \nabla F(\mathbf{x}^{n+\Delta-1}))_j \right| \\ &\geq |\bar{x}_{p+q}| - \left| (\mathbf{x}^{n+\Delta-1} - \bar{\mathbf{x}} - \eta \nabla F(\mathbf{x}^{n+\Delta-1}))_j \right|. \end{aligned}$$

On the other hand, for all $i \in \bar{S}$,

$$|z_i^{n+\Delta}| = \left| (\mathbf{x}^{n+\Delta-1} - \bar{\mathbf{x}} - \eta \nabla F(\mathbf{x}^{n+\Delta-1}))_i \right|.$$

Hence, we know that to guarantee (7), it suffices to ensure for all $j \in [p+q]$ and $i \in \bar{S}$ that

$$|\bar{x}_{p+q}| > \left| (\mathbf{x}^{n+\Delta-1} - \bar{\mathbf{x}} - \eta \nabla F(\mathbf{x}^{n+\Delta-1}))_j \right| + \left| (\mathbf{x}^{n+\Delta-1} - \bar{\mathbf{x}} - \eta \nabla F(\mathbf{x}^{n+\Delta-1}))_i \right|.$$

Note that the right-hand side is upper bounded as follows:

$$\begin{aligned}
& \frac{1}{\sqrt{2}} \left| (\mathbf{x}^{n+\Delta-1} - \bar{\mathbf{x}} - \eta \nabla F(\mathbf{x}^{n+\Delta-1}))_j \right| + \frac{1}{\sqrt{2}} \left| (\mathbf{x}^{n+\Delta-1} - \bar{\mathbf{x}} - \eta \nabla F(\mathbf{x}^{n+\Delta-1}))_i \right| \\
& \leq \left\| (\mathbf{x}^{n+\Delta-1} - \bar{\mathbf{x}} - \eta \nabla F(\mathbf{x}^{n+\Delta-1}))_{\{j,i\}} \right\| \\
& \leq \left\| (\mathbf{x}^{n+\Delta-1} - \bar{\mathbf{x}} - \eta \nabla F(\mathbf{x}^{n+\Delta-1}) + \eta \nabla F(\bar{\mathbf{x}}))_{\{j,i\}} \right\| + \eta \left\| (\nabla F(\bar{\mathbf{x}}))_{\{j,i\}} \right\| \\
& \leq \phi_{2k} \left\| \mathbf{x}^{n+\Delta-1} - \bar{\mathbf{x}} \right\| + \eta \left\| \nabla_2 F(\bar{\mathbf{x}}) \right\| \\
& \leq \phi_{2k} \alpha \cdot \beta^{\Delta-1} \left\| \mathbf{x}^n - \bar{\mathbf{x}} \right\| + \phi \psi_1 + \eta \left\| \nabla_2 F(\bar{\mathbf{x}}) \right\|,
\end{aligned}$$

where ϕ_{2k} is given by Lemma 17. Note that $\phi_{2k} < 1$ whenever $0 < \eta < 1/\rho_{2k}^+$. Moreover,

$$\left\| \mathbf{x}^n - \bar{\mathbf{x}} \right\| \leq \gamma \left\| \bar{\mathbf{x}}_{\overline{S}^n} \right\| + \psi_2 \leq \gamma \left\| \bar{\mathbf{x}}_{[p]} \right\| + \psi_2 = \gamma \left\| \bar{\mathbf{x}}_{\{p+1, \dots, s\}} \right\| + \psi_2.$$

Put all the pieces together, we have

$$\begin{aligned}
& \frac{1}{\sqrt{2}} \left| (\mathbf{x}^{n+\Delta-1} - \bar{\mathbf{x}} - \eta \nabla F(\mathbf{x}^{n+\Delta-1}))_j \right| + \frac{1}{\sqrt{2}} \left| (\mathbf{x}^{n+\Delta-1} - \bar{\mathbf{x}} - \eta \nabla F(\mathbf{x}^{n+\Delta-1}))_i \right| \\
& \leq \alpha \gamma \cdot \beta^{\Delta-1} \left\| \bar{\mathbf{x}}_{\{p+1, \dots, s\}} \right\| + \alpha \psi_2 + \psi_1 + \eta \left\| \nabla_2 F(\bar{\mathbf{x}}) \right\| \\
& \leq \alpha \gamma \cdot \beta^{\Delta-1} \left\| \bar{\mathbf{x}}_{\{p+1, \dots, s\}} \right\| + \alpha \psi_2 + \psi_1 + \frac{1}{\rho_{2k}} \left\| \nabla_2 F(\bar{\mathbf{x}}) \right\|.
\end{aligned}$$

Therefore, when

$$\sqrt{2} |\bar{x}_{p+q}| > \alpha \gamma \cdot \beta^{\Delta-1} \left\| \bar{\mathbf{x}}_{\{p+1, \dots, s\}} \right\| + \alpha \psi_2 + \psi_1 + \frac{1}{\rho_{2k}} \left\| \nabla_2 F(\bar{\mathbf{x}}) \right\|,$$

we always have (7). Note that the above holds as far as $\Psi := \alpha \psi_2 + \psi_1 + \frac{1}{\rho_{2k}} \left\| \nabla_2 F(\bar{\mathbf{x}}) \right\|$ is strictly smaller than $\sqrt{2} |\bar{x}_s|$. \square

Theorem 9 (Restatement of Theorem 6). *Assume same conditions as in Lemma 5. Then PHT(r) successfully identifies the support of $\bar{\mathbf{x}}$ using $\left(\frac{\log 2}{2 \log(1/\beta)} + \frac{\log(\alpha \gamma / (1-\lambda))}{\log(1/\beta)} + 2 \right) s$ number of iterations.*

Proof. We partition the support set $S = [s]$ into K folds S_1, S_2, \dots, S_K , where each S_i is defined as follows:

$$S_i = \{s_{i-1} + 1, \dots, s_i\}, \forall 1 \leq i \leq K.$$

Here, $s_0 = 0$ and for all $1 \leq i \leq K$, the quantity s_i is inductively given by

$$s_i = \max \left\{ q : s_{i-1} + 1 \leq q \leq s \text{ and } |\bar{x}_q| > \frac{1}{\sqrt{2}} |\bar{x}_{s_{i-1}+1}| \right\}.$$

In this way, we note that for any two index sets S_i and S_j , $S_i \cap S_j = \emptyset$ if $i \neq j$. We also know by the definition of s_i that

$$|\bar{x}_{s_i+1}| \leq \frac{1}{\sqrt{2}} |\bar{x}_{s_{i-1}+1}|, \forall 1 \leq i \leq K-1. \quad (8)$$

Now we show that after a finite number of iterations, say n , the union of the S_i 's is contained in S^n , i.e., the support set of the iterate \mathbf{x}^n . To this end, we prove that for all $0 \leq i \leq K$,

$$\bigcup_{t=0}^i S_t \subset S^{n_0+n_1+\dots+n_i} \quad (9)$$

for some n_i 's given below. Above, $S_0 = \emptyset$.

We pick $n_0 = 0$ and it is easy to verify that $S_0 \subset S^0$. Now suppose that (9) holds for $i-1$. That is, the index set of the top s_{i-1} elements of $\bar{\mathbf{x}}$ is contained in $S^{n_0+\dots+n_{i-1}}$. Due to Lemma 5, (9) holds for i as long as n_i satisfies

$$\sqrt{2} |\bar{x}_{s_i}| > \alpha \gamma \cdot \beta^{n_i-1} \left\| \bar{\mathbf{x}}_{\{s_{i-1}+1, \dots, s\}} \right\| + \Psi, \quad (10)$$

where Ψ is given in Lemma 5. Note that

$$\begin{aligned} \|\bar{\mathbf{x}}_{\{s_{i-1}+1, \dots, s\}}\|^2 &= \|\bar{\mathbf{x}}_{S_i}\|^2 + \dots + \|\bar{\mathbf{x}}_{S_K}\|^2 \\ &\leq (\bar{x}_{s_{i-1}+1})^2 |S_i| + \dots + (\bar{x}_{s_{r-1}+1})^2 |S_K| \\ &\leq (\bar{x}_{s_{i-1}+1})^2 (|S_i| + 2^{-1} |S_{i+1}| + \dots + 2^{i-K} |S_K|) \\ &< 2(\bar{x}_{s_i})^2 (|S_i| + 2^{-1} |S_{i+1}| + \dots + 2^{i-K} |S_K|), \end{aligned}$$

where the second inequality follows from (8) and the last inequality follows from the definition of q_i . Denote for simplicity

$$W_i := |S_i| + 2^{-1} |S_{i+1}| + \dots + 2^{i-K} |S_K|.$$

As we assumed $\Psi \leq \sqrt{2}\lambda\bar{x}_{\min}$, we get

$$\alpha\gamma \cdot \beta^{n_i-1} \|\bar{\mathbf{x}}_{\{s_{i-1}+1, \dots, s\}}\| + \Psi < \sqrt{2}\alpha\gamma |\bar{x}_{s_i}| \beta^{n_i-1} \sqrt{W_i} + \sqrt{2}\lambda |\bar{x}_{s_i}|.$$

Picking

$$n_i = \log_{1/\beta} \frac{\alpha\gamma\sqrt{W_i}}{1-\lambda} + 2$$

guarantees (10). It remains to calculate the total number of iterations. In fact, we have

$$\begin{aligned} t_{\max} &= n_0 + n_1 + \dots + n_K \\ &= \frac{1}{2\log(1/\beta)} \sum_{i=1}^K \log W_i + K \cdot \frac{\log(\alpha\gamma/(1-\lambda))}{\log(1/\beta)} + 2K \\ &\stackrel{\zeta_1}{\leq} \frac{K}{2\log(1/\beta)} \log \left(\frac{1}{K} \sum_{i=1}^K W_i \right) + \left(\frac{\log(\alpha\gamma/(1-\lambda))}{\log(1/\beta)} + 2 \right) K \\ &\stackrel{\zeta_2}{\leq} \frac{K}{2\log(1/\beta)} \log \left(\frac{2}{K} \sum_{i=1}^K |S_i| \right) + \left(\frac{\log(\alpha\gamma/(1-\lambda))}{\log(1/\beta)} + 2 \right) K \\ &= \frac{K}{2\log(1/\beta)} \log \frac{2s}{K} + \left(\frac{\log(\alpha\gamma/(1-\lambda))}{\log(1/\beta)} + 2 \right) K \\ &\stackrel{\zeta_3}{\leq} \left(\frac{\log 2}{2\log(1/\beta)} + \frac{\log(\alpha\gamma/(1-\lambda))}{\log(1/\beta)} + 2 \right) s. \end{aligned}$$

Above, ζ_1 immediately follows by observing that the logarithmic function is concave. ζ_2 uses the fact that after rearrangement, the coefficient of $|S_i|$ is $\sum_{j=0}^{i-1} 2^{-j}$ which is always smaller than 2. Finally, since the function $a \log(2s/a)$ is monotonically increasing with respect to a and $1 \leq a \leq s$, ζ_3 follows. \square

Lemma 10 (Restatement of Lemma 7). *Assume that $F(\mathbf{x})$ satisfies the properties of RSC and RSS at sparsity level $k + s + r$. Let $\rho^- := \rho_{k+s+r}^-$ and $\rho^+ := \rho_{k+s+r}^+$. Consider the support set $J^t = S^{t-1} \cup \text{supp}(\nabla F(\mathbf{x}^{t-1}), r)$. We have for any $0 < \theta \leq 1/\rho^+$,*

$$\|\bar{\mathbf{x}}_{J^t}\| \leq \nu(1 - \theta\rho^-) \|\mathbf{x}^{t-1} - \bar{\mathbf{x}}\| + \frac{\nu}{\rho^-} \|\nabla_{s+r} F(\bar{\mathbf{x}})\|,$$

where $\nu = \sqrt{s-r+2}$. In particular, picking $\theta = 1/\rho^+$ gives

$$\|\bar{\mathbf{x}}_{J^t}\| \leq \nu \left(1 - \frac{1}{\kappa} \right) \|\mathbf{x}^{t-1} - \bar{\mathbf{x}}\| + \frac{\nu}{\rho^-} \|\nabla_{s+r} F(\bar{\mathbf{x}})\|.$$

Proof. Let $T = \text{supp}(\nabla F(\mathbf{x}^{t-1}), r)$. Then $J^t = S^{t-1} \cup T$ and $S^{t-1} \cap T = \emptyset$. Since T contains the top r elements of $\nabla F(\mathbf{x}^{t-1})$, we have that each element in $T \setminus S$ is larger (in magnitude) than that in $S \setminus T$. In particular, we observe for $T \neq S$ that

$$\frac{1}{|T \setminus S|} \left\| (\nabla F(\mathbf{x}^{t-1}))_{T \setminus S} \right\|^2 \geq \frac{1}{|S \setminus T|} \left\| (\nabla F(\mathbf{x}^{t-1}))_{S \setminus T} \right\|^2,$$

which implies

$$\left\| (\nabla F(\mathbf{x}^{t-1}))_{T \setminus S} \right\| \geq \sqrt{\frac{r - |T \cap S|}{s - |T \cap S|}} \left\| (\nabla F(\mathbf{x}^{t-1}))_{S \setminus T} \right\| \geq \sqrt{\frac{1}{s - r + 1}} \left\| (\nabla F(\mathbf{x}^{t-1}))_{S \setminus T} \right\|.$$

Since $\nabla F(\mathbf{x}^{t-1})$ is supported on $\overline{S^{t-1}}$, the LHS reads as

$$\left\| (\nabla F(\mathbf{x}^{t-1}))_{T \setminus S} \right\| = \left\| (\nabla F(\mathbf{x}^{t-1}))_{T \setminus (S \cup S^{t-1})} \right\| = \frac{1}{\theta} \left\| (\mathbf{x}^{t-1} - \theta \nabla F(\mathbf{x}^{t-1}) - \bar{\mathbf{x}})_{T \setminus (S \cup S^{t-1})} \right\|.$$

Now we look at the RHS. It follows that

$$\begin{aligned} \left\| (\nabla F(\mathbf{x}^{t-1}))_{S \setminus T} \right\| &= \left\| (\nabla F(\mathbf{x}^{t-1}))_{S \setminus (T \cup S^{t-1})} \right\| \\ &= \frac{1}{\theta} \left\| (\mathbf{x}^{t-1} - \theta \nabla F(\mathbf{x}^{t-1}) - \bar{\mathbf{x}})_{S \setminus (T \cup S^{t-1})} + \bar{\mathbf{x}}_{S \setminus (T \cup S^{t-1})} \right\| \\ &\geq \frac{1}{\theta} \left\| \bar{\mathbf{x}}_{S \setminus (T \cup S^{t-1})} \right\| - \frac{1}{\theta} \left\| (\mathbf{x}^{t-1} - \theta \nabla F(\mathbf{x}^{t-1}) - \bar{\mathbf{x}})_{S \setminus (T \cup S^{t-1})} \right\|. \end{aligned}$$

Hence,

$$\begin{aligned} &\left\| \bar{\mathbf{x}}_{\overline{S^t}} \right\| \\ &= \left\| \bar{\mathbf{x}}_{S \setminus (T \cup S^{t-1})} \right\| \\ &\leq \sqrt{s - r + 1} \left\| (\mathbf{x}^{t-1} - \theta \nabla F(\mathbf{x}^{t-1}) - \bar{\mathbf{x}})_{T \setminus (S \cup S^{t-1})} \right\| + \left\| (\mathbf{x}^{t-1} - \theta \nabla F(\mathbf{x}^{t-1}) - \bar{\mathbf{x}})_{S \setminus (T \cup S^{t-1})} \right\| \\ &\leq \sqrt{s - r + 1} \left\| (\mathbf{x}^{t-1} - \theta \nabla F(\mathbf{x}^{t-1}) - \bar{\mathbf{x}})_{T \setminus S} \right\| + \left\| (\mathbf{x}^{t-1} - \theta \nabla F(\mathbf{x}^{t-1}) - \bar{\mathbf{x}})_{S \setminus T} \right\| \\ &\leq \nu \left\| (\mathbf{x}^{t-1} - \theta \nabla F(\mathbf{x}^{t-1}) - \bar{\mathbf{x}})_{T \Delta S} \right\| \\ &\leq \nu \left\| (\mathbf{x}^{t-1} - \theta \nabla F(\mathbf{x}^{t-1}) - \bar{\mathbf{x}} + \theta \nabla F(\bar{\mathbf{x}}))_{T \Delta S} \right\| + \nu \theta \left\| (\nabla F(\bar{\mathbf{x}}))_{T \Delta S} \right\| \\ &\leq \nu \phi_{k+s+r} \left\| \mathbf{x}^{t-1} - \bar{\mathbf{x}} \right\| + \nu \theta \left\| (\nabla F(\bar{\mathbf{x}}))_{T \Delta S} \right\|, \end{aligned}$$

where $\nu = \sqrt{s - r + 2}$ and the last inequality uses Lemma 18. For any $0 < \theta \leq 1/\rho^+$, we have

$$\left\| \bar{\mathbf{x}}_{\overline{S^t}} \right\| \leq \nu(1 - \theta m) \left\| \mathbf{x}^{t-1} - \bar{\mathbf{x}} \right\| + \frac{\nu}{\rho^-} \left\| \nabla_{s+r} F(\bar{\mathbf{x}}) \right\|.$$

□

A.1 Proof of Prop. 2

Proof. Recall that we set $k = s$. Using Lemma 11, we have

$$F(\mathbf{x}^t) - F(\bar{\mathbf{x}}) \leq \mu_t (F(\mathbf{x}^{t-1}) - F(\bar{\mathbf{x}})),$$

where $\mu_t = 1 - 2\rho_{2s}^- \eta(1 - \eta\rho_{2s}^+) \cdot \frac{|S^t \setminus S^{t-1}|}{|S^t \setminus S^{t-1}| + |S \setminus S^{t-1}|}$. Now combining this with Prop. 21, we have

$$\left\| \mathbf{x}^t - \bar{\mathbf{x}} \right\| \leq \sqrt{2\kappa} \sqrt{\mu_1 \mu_2 \dots \mu_t} \left\| \mathbf{x}^0 - \bar{\mathbf{x}} \right\| + \frac{3}{\rho_{2s}^-} \left\| \nabla_{2s} F(\bar{\mathbf{x}}) \right\|.$$

Note that before the algorithm terminates, $1 \leq |S^t \setminus S^{t-1}| \leq r$. Hence,

$$\mu_t \leq 1 - \frac{2\eta\rho_{2s}^-(1 - \eta\rho_{2s}^+)}{1 + s} =: \mu.$$

It then follows that

$$\left\| \mathbf{x}^t - \bar{\mathbf{x}} \right\| \leq \sqrt{2\kappa} (\sqrt{\mu})^t \left\| \mathbf{x}^0 - \bar{\mathbf{x}} \right\| + \frac{3}{\eta} \left\| \nabla_{2s} F(\bar{\mathbf{x}}) \right\|. \quad (11)$$

Lemma 19 tells us

$$\left\| \mathbf{x}^t - \bar{\mathbf{x}} \right\| \leq \kappa \left\| \bar{\mathbf{x}}_{\overline{S^t}} \right\| + \frac{1}{\eta} \left\| \nabla_s F(\bar{\mathbf{x}}) \right\|. \quad (12)$$

Hence, in light of Lemma 5 and Theorem 6, we obtain that PHT(r) recovers the support using at most

$$t_{\max} = \left(\frac{\log 2}{\log(1/\mu)} + \frac{\log(2\kappa)}{\log(1/\mu)} + \frac{2 \log(\kappa/(1-\lambda))}{\log(1/\mu)} + 2 \right) \|\bar{\mathbf{x}}\|_0$$

iterations. Note that picking $\eta = O(1/\rho_{2s}^+)$, we have $\mu = O(1 - \frac{1}{\kappa})$ and $\log(1/\mu) = O(1/\kappa)$. This gives the $O(s\kappa \log \kappa)$ bound. \square

Lemma 11. *Consider the PHT(r) algorithm. Suppose that $F(\mathbf{x})$ is ρ_{k+s}^- -RSC and ρ_{2k}^+ -RSS. Using the parameter $k = s$ and $\eta < 1/\rho_{2s}^+$, we have*

$$F(\mathbf{x}^t) - F(\bar{\mathbf{x}}) \leq \mu_t (F(\mathbf{x}^{t-1}) - F(\bar{\mathbf{x}})),$$

where $\mu_t = 1 - 2\eta\rho_{2s}^-(1 - \eta\rho_{2s}^+) \cdot \frac{|S^t \setminus S^{t-1}|}{|S^t \setminus S^{t-1}| + |S \setminus S^{t-1}|}$.

Proof. Using the RSS property, we have

$$\begin{aligned} F(\mathbf{z}_{S^t}^t) - F(\mathbf{x}^{t-1}) &\leq \langle \nabla F(\mathbf{x}^{t-1}), \mathbf{z}_{S^t}^t - \mathbf{x}^{t-1} \rangle + \frac{\rho_{2s}^+}{2} \|\mathbf{z}_{S^t}^t - \mathbf{x}^{t-1}\|^2 \\ &\stackrel{\zeta_1}{\leq} \left\langle \nabla_{S^t \setminus S^{t-1}} F(\mathbf{x}^{t-1}), \mathbf{z}_{S^t \setminus S^{t-1}}^t \right\rangle + \frac{\rho_{2s}^+}{2} \left(\|\mathbf{z}_{S^t \setminus S^{t-1}}^t\|^2 \right. \\ &\quad \left. + \|\mathbf{z}_{S^t \cap S^{t-1}}^t - \mathbf{x}_{S^t \cap S^{t-1}}^{t-1}\|^2 + \|\mathbf{x}_{S^{t-1} \setminus S^t}^{t-1}\|^2 \right) \\ &\stackrel{\zeta_2}{\leq} \left\langle \nabla_{S^t \setminus S^{t-1}} F(\mathbf{x}^{t-1}), \mathbf{z}_{S^t \setminus S^{t-1}}^t \right\rangle + \rho_{2s}^+ \|\mathbf{z}_{S^t \setminus S^{t-1}}^t\|^2 \\ &\stackrel{\zeta_3}{\leq} -\eta(1 - \eta\rho_{2s}^+) \|\nabla_{S^t \setminus S^{t-1}} F(\mathbf{x}^{t-1})\|^2. \end{aligned}$$

Above, we observe that $\nabla F(\mathbf{x}^{t-1})$ is supported on $\overline{S^{t-1}}$ and we simply decompose the support set $S^t \cup S^{t-1}$ into three mutually disjoint sets, and hence ζ_1 holds. To see why ζ_2 holds, we note that for any set $\Omega \subset S^{t-1}$, $\mathbf{z}_{\Omega}^t = \mathbf{x}_{\Omega}^{t-1}$. Hence, $\mathbf{z}_{S^t \cap S^{t-1}}^t = \mathbf{x}_{S^t \cap S^{t-1}}^{t-1}$. Moreover, since $\mathbf{x}_{S^{t-1} \setminus S^t}^{t-1} = \mathbf{z}_{S^{t-1} \setminus S^t}^t$ and any element in $\mathbf{z}_{S^t \setminus S^{t-1}}^t$ is not larger than that in $\mathbf{z}_{S^t \setminus S^{t-1}}^t$ (recall that S^t is obtained by hard thresholding), we have $\|\mathbf{x}_{S^{t-1} \setminus S^t}^{t-1}\| \leq \|\mathbf{z}_{S^t \setminus S^{t-1}}^t\|$ where we use the fact that $|S^t \setminus S^t| = |S^t \setminus S^{t-1}|$. Therefore, ζ_2 holds. Finally, we write $\mathbf{z}_{S^t \setminus S^{t-1}}^t = -\eta \nabla_{S^t \setminus S^{t-1}} F(\mathbf{x}^{t-1})$ and obtain ζ_3 .

Since \mathbf{x}^t is a minimizer of $F(\mathbf{x})$ over the support set S^t , it immediately follows that

$$F(\mathbf{x}^t) - F(\mathbf{x}^{t-1}) \leq F(\mathbf{z}_{S^t}^t) - F(\mathbf{x}^{t-1}) \leq -\eta(1 - \eta\rho_{2s}^+) \|\nabla_{S^t \setminus S^{t-1}} F(\mathbf{x}^{t-1})\|^2.$$

Now we invoke Lemma 12 and pick $\eta \leq 1/\rho_{2s}^+$,

$$F(\mathbf{x}^t) - F(\mathbf{x}^{t-1}) \leq -2m\eta(1 - \eta\rho_{2s}^+) \cdot \frac{|S^t \setminus S^{t-1}|}{|S^t \setminus S^{t-1}| + |S \setminus S^{t-1}|} (F(\mathbf{x}^{t-1}) - F(\bar{\mathbf{x}})),$$

which gives

$$F(\mathbf{x}^t) - F(\bar{\mathbf{x}}) \leq \mu_t (F(\mathbf{x}^{t-1}) - F(\bar{\mathbf{x}})),$$

where $\mu_t = 1 - 2\eta\rho_{2s}^-(1 - \eta\rho_{2s}^+) \cdot \frac{|S^t \setminus S^{t-1}|}{|S^t \setminus S^{t-1}| + |S \setminus S^{t-1}|}$. \square

Lemma 12. *Consider the PHT(r) algorithm and assume $F(\mathbf{x})$ is ρ_{k+s}^- -RSC. Then for all $t \geq 1$,*

$$\|\nabla_{S^t \setminus S^{t-1}} F(\mathbf{x}^{t-1})\|^2 \geq 2\rho_{k+s}^- \delta_t (F(\mathbf{x}^{t-1}) - F(\bar{\mathbf{x}})),$$

where

$$\delta_t = \frac{|S^t \setminus S^{t-1}|}{|S^t \setminus S^{t-1}| + |S \setminus S^{t-1}|}.$$

Proof. The lemma holds clearly for either $S^t = S^{t-1}$ or $F(\mathbf{x}^t) \leq F(\bar{\mathbf{x}})$. Hence, in the following we only prove the result by assuming $S^t \neq S^{t-1}$ and $F(\mathbf{x}^t) > F(\bar{\mathbf{x}})$. Due to the RSC property, we have

$$F(\bar{\mathbf{x}}) - F(\mathbf{x}^{t-1}) - \langle \nabla F(\mathbf{x}^{t-1}), \bar{\mathbf{x}} - \mathbf{x}^{t-1} \rangle \geq \frac{\rho_{k+s}^-}{2} \|\bar{\mathbf{x}} - \mathbf{x}^{t-1}\|^2,$$

which implies

$$\begin{aligned} \langle \nabla F(\mathbf{x}^{t-1}), -\bar{\mathbf{x}} \rangle &\geq \frac{\rho_{k+s}^-}{2} \|\bar{\mathbf{x}} - \mathbf{x}^{t-1}\|^2 + F(\mathbf{x}^{t-1}) - F(\bar{\mathbf{x}}) \\ &\geq \sqrt{2\rho_{k+s}^-} \|\bar{\mathbf{x}} - \mathbf{x}^{t-1}\| \sqrt{F(\mathbf{x}^{t-1}) - F(\bar{\mathbf{x}})}. \end{aligned}$$

By invoking Lemma 13 with $\mathbf{u} = \nabla F(\mathbf{x}^{t-1})$ and $\mathbf{z} = -\bar{\mathbf{x}}$ therein, we have

$$\begin{aligned} \langle \nabla F(\mathbf{x}^{t-1}), -\bar{\mathbf{x}} \rangle &\leq \sqrt{\frac{|S \setminus S^{t-1}|}{|S^t \setminus S^{t-1}|} + 1} \|\nabla_{S^t \setminus S^{t-1}} F(\mathbf{x}^{t-1})\| \cdot \|\bar{\mathbf{x}}_{S^t \setminus S^{t-1}}\| \\ &= \sqrt{\frac{|S \setminus S^{t-1}|}{|S^t \setminus S^{t-1}|} + 1} \|\nabla_{S^t \setminus S^{t-1}} F(\mathbf{x}^{t-1})\| \cdot \|(\bar{\mathbf{x}} - \mathbf{x}^t)_{S^t \setminus S^{t-1}}\| \\ &\leq \sqrt{\frac{|S \setminus S^{t-1}|}{|S^t \setminus S^{t-1}|} + 1} \|\nabla_{S^t \setminus S^{t-1}} F(\mathbf{x}^{t-1})\| \cdot \|\bar{\mathbf{x}} - \mathbf{x}^t\|. \end{aligned}$$

It is worth mentioning that the first inequality above holds because $\nabla F(\mathbf{x}^{t-1})$ is supported on $\overline{S^{t-1}}$ and $S^t \setminus S^{t-1}$ contains the $|S^t \setminus S^{t-1}|$ number of largest (in magnitude) elements of $\nabla F(\mathbf{x}^{t-1})$. Therefore, we obtain the result. \square

Lemma 13 (Lemma 1 in [28]). *Let \mathbf{u} and \mathbf{z} be two distinct vectors and let $W = \text{supp}(\mathbf{u}) \cap \text{supp}(\mathbf{z})$. Also, let U be the support set of the top r (in magnitude) elements in \mathbf{u} . Then, the following holds for all $r \geq 1$:*

$$\langle \mathbf{u}, \mathbf{z} \rangle \leq \sqrt{\left\lceil \frac{|W|}{r} \right\rceil} \|\mathbf{u}_U\| \cdot \|\mathbf{z}_W\|.$$

A.2 Proof of Theorem 3

Proof. Let $\rho^- := \rho_{2s+r}^-$ and $\rho^+ := \rho_{2s+r}^+$. Let $\phi := \phi_{2s+r} = 1 - \eta\rho^-$ be the quantity given in Lemma 17. Using Lemma 14, we obtain

$$\|\mathbf{x}^t - \bar{\mathbf{x}}\| \leq \left(\sqrt{2}\phi\kappa + \nu(\kappa - 1) \right) \|\mathbf{x}^{t-1} - \bar{\mathbf{x}}\| + \frac{2\nu + 4}{\rho^-} \|\nabla_{s+r} F(\bar{\mathbf{x}})\|,$$

where $\nu = \sqrt{s - r + 2}$. We need to ensure that the convergence coefficient is smaller than 1. Consider $\eta = \eta'/\rho^+$ with $\eta' \in (0, 1]$ for which $\phi = 1 - \eta'/\kappa$. It follows that

$$\sqrt{2}\phi\kappa + \nu(\kappa - 1) = \sqrt{2}(\kappa - \eta') + \nu(\kappa - 1) \leq (\sqrt{2} + \nu)(\kappa - \eta').$$

Hence, when we pick $1 - \frac{1}{\sqrt{2} + \nu} < \eta' \leq 1$, and the condition number satisfies

$$\kappa < \eta' + \frac{1}{\sqrt{2} + \nu},$$

the sequence of $\mathbf{x}^t - \bar{\mathbf{x}}$ contracts. On the other hand, using Lemma 19 we get

$$\|\mathbf{x}^t - \bar{\mathbf{x}}\| \leq \kappa \|\bar{\mathbf{x}}_{S^t}\| + \frac{1}{\rho^-} \|\nabla_s F(\bar{\mathbf{x}})\|.$$

Hence, applying Lemma 5 and Theorem 6 we obtain the result. \square

Lemma 14. Consider the PHT(r) algorithm with $k = s$. Suppose that $F(\mathbf{x})$ is ρ_{2s+r}^- -RSC and ρ_{2s+r}^+ -RSS. Further suppose that $\kappa < 2$. Let the step size $\eta \leq 1/\rho_{2s+r}^+$. Then it holds that

$$\|\mathbf{x}^t - \bar{\mathbf{x}}\| \leq \left(\sqrt{2}\phi\kappa + \nu(\kappa - 1) \right) \|\mathbf{x}^{t-1} - \bar{\mathbf{x}}\| + \frac{2\nu + 4}{\rho_{2s+r}^-} \|\nabla_{s+r} F(\bar{\mathbf{x}})\|,$$

where $\phi = 1 - \eta\rho_{2s+r}^-$ and $\nu = \sqrt{s - r + 2}$.

Proof. Consider the vector $\mathbf{z}_{J^t}^t$. It is easy to see that $J^t \setminus S^t$ contains the r smallest elements of $\mathbf{z}_{J^t}^t$. Hence, for any subset $T \subset J^t$ such that $|T| \geq r$, we have

$$\|\mathbf{z}_{J^t \setminus S^t}^t\| \leq \|\mathbf{z}_T^t\|.$$

In particular, we choose $T = J^t \setminus S$ and obtain

$$\|\mathbf{z}_{J^t \setminus S^t}^t\| \leq \|\mathbf{z}_{J^t \setminus S}^t\|.$$

Eliminating the common contribution from $J^t \setminus (S^t \cup S)$ gives

$$\|\mathbf{z}_{J^t \cap S^t \setminus S^t}^t\| \leq \|\mathbf{z}_{J^t \cap S^t \setminus S}^t\|. \quad (13)$$

The LHS of (13) reads as

$$\begin{aligned} \|\mathbf{z}_{J^t \cap S^t \setminus S^t}^t\| &= \|(\mathbf{x}^{t-1} - \eta\nabla F(\mathbf{x}^{t-1}) - \bar{\mathbf{x}})_{J^t \cap S^t \setminus S^t} + \bar{\mathbf{x}}_{J^t \setminus S^t}\| \\ &\geq \|\bar{\mathbf{x}}_{J^t \setminus S^t}\| - \|(\mathbf{x}^{t-1} - \eta\nabla F(\mathbf{x}^{t-1}) - \bar{\mathbf{x}})_{J^t \cap S^t \setminus S^t}\|, \end{aligned}$$

while the RHS (13) is given by

$$\|\mathbf{z}_{J^t \cap S^t \setminus S}^t\| = \|(\mathbf{x}^{t-1} - \eta\nabla F(\mathbf{x}^{t-1}) - \bar{\mathbf{x}})_{J^t \cap S^t \setminus S}\|.$$

Hence, we have

$$\begin{aligned} \|\bar{\mathbf{x}}_{J^t \setminus S^t}\| &\leq \|(\mathbf{x}^{t-1} - \eta\nabla F(\mathbf{x}^{t-1}) - \bar{\mathbf{x}})_{J^t \cap S^t \setminus S^t}\| + \|(\mathbf{x}^{t-1} - \eta\nabla F(\mathbf{x}^{t-1}) - \bar{\mathbf{x}})_{J^t \cap S^t \setminus S}\| \\ &\leq \sqrt{2} \|(\mathbf{x}^{t-1} - \eta\nabla F(\mathbf{x}^{t-1}) - \bar{\mathbf{x}})_{J^t}\| \\ &\leq \sqrt{2}\phi_{2s+r} \|\mathbf{x}^{t-1} - \bar{\mathbf{x}}\| + \sqrt{2}\eta \|\nabla_{k+r} F(\bar{\mathbf{x}})\|, \end{aligned}$$

where we use Lemma 18 for the last inequality and $\phi_{2s+r} = 1 - \eta\rho_{2s+r}^-$ for $\eta \leq 1/\rho_{2s+r}^+$. On the other hand, Lemma 7 shows that

$$\|\bar{\mathbf{x}}_{\bar{J}^t}\| \leq \nu \left(1 - \frac{1}{\kappa} \right) \|\mathbf{x}^{t-1} - \bar{\mathbf{x}}\| + \frac{\nu}{\rho_{2s+r}^-} \|\nabla_{s+r} F(\bar{\mathbf{x}})\|,$$

where $\nu = \sqrt{s - r + 2}$. The fact $\bar{S}^t = (J^t \setminus S^t) \cup \bar{J}^t$ implies

$$\begin{aligned} \|\bar{\mathbf{x}}_{\bar{S}^t}\| &\leq \|\bar{\mathbf{x}}_{J^t \setminus S^t}\| + \|\bar{\mathbf{x}}_{\bar{J}^t}\| \\ &\leq \left(\sqrt{2}\phi_{2s+r} + \nu \left(1 - \frac{1}{\kappa} \right) \right) \|\mathbf{x}^{t-1} - \bar{\mathbf{x}}\| + \left(\sqrt{2}\eta + \frac{\nu}{\rho_{2s+r}^-} \right) \|\nabla_{k+r} F(\bar{\mathbf{x}})\|. \end{aligned}$$

Next, we invoke Lemma 19 to get

$$\|\mathbf{x}^t - \bar{\mathbf{x}}\| \leq \kappa \|\bar{\mathbf{x}}_{\bar{S}^t}\| + \frac{1}{\rho_{2s+r}^-} \|\nabla_k F(\bar{\mathbf{x}})\|.$$

Therefore,

$$\begin{aligned} \|\mathbf{x}^t - \bar{\mathbf{x}}\| &\leq \left(\sqrt{2}\phi_{2s+r}\kappa + \nu(\kappa - 1) \right) \|\mathbf{x}^{t-1} - \bar{\mathbf{x}}\| + \left(\sqrt{2}\eta\kappa + \frac{\nu\kappa}{\rho_{2s+r}^-} + \frac{1}{\rho_{2s+r}^-} \right) \|\nabla_{s+r} F(\bar{\mathbf{x}})\| \\ &\leq \left(\sqrt{2}\phi_{2s+r}\kappa + \nu(\kappa - 1) \right) \|\mathbf{x}^{t-1} - \bar{\mathbf{x}}\| + \frac{2\nu + 4}{\rho_{2s+r}^-} \|\nabla_{s+r} F(\bar{\mathbf{x}})\|, \end{aligned}$$

where we use the assumption that $\kappa < 2$ and $\eta \leq 1/\rho_{2s+r}^+ < 1/\rho_{2s+r}^-$ for the last inequality. \square

A.3 Proof of Theorem 4

Proof. Using Lemma 15, we have

$$F(\mathbf{x}^t) - F(\bar{\mathbf{x}}) \leq \mu (F(\mathbf{x}^{t-1}) - F(\bar{\mathbf{x}})),$$

where

$$\mu = 1 - \frac{\eta \rho_{2k}^- (1 - \eta \rho_{2k}^+)}{2}.$$

Now Prop. 21 suggests that

$$\|\mathbf{x}^t - \bar{\mathbf{x}}\| \leq \sqrt{2\kappa} (\sqrt{\mu})^t \|\mathbf{x}^0 - \bar{\mathbf{x}}\| + \frac{3}{\rho_{2k}^-} \|\nabla_{k+s} F(\bar{\mathbf{x}})\|,$$

and Lemma 19 implies

$$\|\mathbf{x}^t - \bar{\mathbf{x}}\| \leq \kappa \|\bar{\mathbf{x}}_{S^t}\| + \frac{1}{\rho_{2k}^-} \|\nabla_k F(\bar{\mathbf{x}})\|.$$

Combining these with Lemma 5 and Theorem 6 we complete the proof. \square

Lemma 15. Consider the PHT(r) algorithm. Suppose that $F(\mathbf{x})$ is ρ_{2k}^- -RSC and ρ_{2k}^+ -RSS, and let $\kappa = \rho_{2k}^+ / \rho_{2k}^-$ be the condition number. Picking the step size $0 < \eta < 1 / \rho_{2k}^+$ and the sparsity parameter $k \geq s + \left(1 + \frac{4}{\eta^2 (\rho_{2k}^-)^2}\right) \min\{r, s\}$, then we have

$$F(\mathbf{x}^t) - F(\mathbf{x}^{t-1}) \leq -\frac{\eta \rho_{2k}^- (1 - \eta \rho_{2k}^+)}{2} (F(\mathbf{x}^{t-1}) - F(\bar{\mathbf{x}})).$$

Proof. Using Lemma 16 we obtain

$$F(\mathbf{x}^t) - F(\mathbf{x}^{t-1}) \leq -\frac{1 - \eta \rho_{2k}^+}{2\eta} \|\mathbf{z}_{S^t}^t - \mathbf{x}^{t-1}\|^2.$$

Note that for the right-hand side, we may expand it as follows:

$$\begin{aligned} \|\mathbf{z}_{S^t}^t - \mathbf{x}^{t-1}\|^2 &= \|\mathbf{x}_{S^t}^{t-1} - \mathbf{x}^{t-1} - \eta \nabla_{S^t} F(\mathbf{x}^{t-1})\|^2 \\ &= \left\| -\mathbf{x}_{S^t \setminus S^{t-1}}^{t-1} - \eta \nabla_{S^t \setminus S^{t-1}} F(\mathbf{x}^{t-1}) \right\|^2 \\ &= \left\| \mathbf{x}_{S^t \setminus S^{t-1}}^{t-1} \right\|^2 + \eta^2 \left\| \nabla_{S^t \setminus S^{t-1}} F(\mathbf{x}^{t-1}) \right\|^2, \end{aligned}$$

where we use the fact that \mathbf{x}^{t-1} is supported on S^{t-1} and $\nabla F(\mathbf{x}^{t-1})$ is support on $\overline{S^{t-1}}$ for the second equality, and the third one follows in that the support sets are disjoint. It then follows quickly that

$$F(\mathbf{x}^t) - F(\mathbf{x}^{t-1}) \leq -\frac{(1 - \eta \rho_{2k}^+) \eta}{2} \left\| \nabla_{S^t \setminus S^{t-1}} F(\mathbf{x}^{t-1}) \right\|^2.$$

It remains to lower bound the right-hand side in terms of $F(\mathbf{x}^{t-1}) - F(\bar{\mathbf{x}})$. In fact, in the following, we show that

$$\left\| \nabla_{S^t \setminus S^{t-1}} F(\mathbf{x}^{t-1}) \right\|^2 \geq \rho_{2k}^- (F(\mathbf{x}^{t-1}) - F(\bar{\mathbf{x}})). \quad (14)$$

This suggests

$$F(\mathbf{x}^t) - F(\mathbf{x}^{t-1}) \leq -\frac{\eta \rho_{2k}^- (1 - \eta \rho_{2k}^+)}{2} (F(\mathbf{x}^{t-1}) - F(\bar{\mathbf{x}}))$$

which completes the proof. In the sequel, we prove the inequality (14) by discussing the size of the support set $S^t \setminus S^{t-1}$.

First, we consider $r \geq s$. Then it is possible that $|S^t \setminus S^{t-1}| \geq s$.

Case 1. $|S^t \setminus S^{t-1}| \geq s$. Using the RSC property, we have

$$\begin{aligned}
& \frac{\rho_{2k}^-}{2} \|\bar{\mathbf{x}} - \mathbf{x}^{t-1}\|^2 \\
& \leq F(\bar{\mathbf{x}}) - F(\mathbf{x}^{t-1}) - \langle \nabla F(\mathbf{x}^{t-1}), \bar{\mathbf{x}} - \mathbf{x}^{t-1} \rangle \\
& \leq F(\bar{\mathbf{x}}) - F(\mathbf{x}^{t-1}) + \frac{\rho_{2k}^-}{2} \|\bar{\mathbf{x}} - \mathbf{x}^{t-1}\|^2 + \frac{1}{2\rho_{2k}^-} \|\nabla_{S \cup S^{t-1}} F(\mathbf{x}^{t-1})\|^2 \\
& = F(\bar{\mathbf{x}}) - F(\mathbf{x}^{t-1}) + \frac{\rho_{2k}^-}{2} \|\bar{\mathbf{x}} - \mathbf{x}^{t-1}\|^2 + \frac{1}{2\rho_{2k}^-} \|\nabla_{S \setminus S^{t-1}} F(\mathbf{x}^{t-1})\|^2.
\end{aligned}$$

Therefore, we get

$$\|\nabla_{S \setminus S^{t-1}} F(\mathbf{x}^{t-1})\|^2 \geq 2\rho_{2k}^- (F(\mathbf{x}^{t-1}) - F(\bar{\mathbf{x}})).$$

Recall that $S^t \setminus S^{t-1}$ contains the largest elements of $\mathbf{z}_{S^{t-1}}^t$. Hence, for any support set $T \subset \overline{S^{t-1}}$ with $|T| \leq |S^t \setminus S^{t-1}|$, we have

$$\|\mathbf{z}_T^t\| \leq \|\mathbf{z}_{S^t \setminus S^{t-1}}^t\|.$$

In particular, we can choose $T = S \setminus S^{t-1}$ as we assumed that $|S^t \setminus S^{t-1}| \geq s \geq |T|$. Then it holds that

$$\|\mathbf{z}_{S^t \setminus S^{t-1}}^t\|^2 \geq \|\mathbf{z}_{S \setminus S^{t-1}}^t\|^2.$$

Note that for the left-hand side, $\mathbf{z}_{S^t \setminus S^{t-1}}^t = -\eta \nabla_{S^t \setminus S^{t-1}} F(\mathbf{x}^{t-1})$ while for the right-hand side, it is exactly equal to $-\eta \nabla_{S \setminus S^{t-1}} F(\mathbf{x}^{t-1})$. This completes the proof of the first case.

Case 2. $|S^t \setminus S^{t-1}| < s \leq r$. The proof of this part is more involved. We still begin with the RSC property, which gives

$$\begin{aligned}
\frac{\rho_{2k}^-}{2} \|\bar{\mathbf{x}} - \mathbf{x}^{t-1}\|^2 & \leq F(\bar{\mathbf{x}}) - F(\mathbf{x}^{t-1}) - \langle \nabla F(\mathbf{x}^{t-1}), \bar{\mathbf{x}} - \mathbf{x}^{t-1} \rangle \\
& \leq F(\bar{\mathbf{x}}) - F(\mathbf{x}^{t-1}) + \frac{\rho_{2k}^-}{4} \|\bar{\mathbf{x}} - \mathbf{x}^{t-1}\|^2 + \frac{1}{\rho_{2k}^-} \|\nabla_{S \cup S^{t-1}} F(\mathbf{x}^{t-1})\|^2 \\
& = F(\bar{\mathbf{x}}) - F(\mathbf{x}^{t-1}) + \frac{\rho_{2k}^-}{4} \|\bar{\mathbf{x}} - \mathbf{x}^{t-1}\|^2 + \frac{1}{\rho_{2k}^-} \|\nabla_{S \setminus S^{t-1}} F(\mathbf{x}^{t-1})\|^2 \\
& = F(\bar{\mathbf{x}}) - F(\mathbf{x}^{t-1}) + \frac{\rho_{2k}^-}{4} \|\bar{\mathbf{x}} - \mathbf{x}^{t-1}\|^2 + \frac{1}{\rho_{2k}^-} \|\nabla_{S \setminus (S^t \cup S^{t-1})} F(\mathbf{x}^{t-1})\|^2 \\
& \quad + \frac{1}{\rho_{2k}^-} \|\nabla_{(S^t \setminus S^{t-1}) \cap S} F(\mathbf{x}^{t-1})\|^2 \\
& \leq F(\bar{\mathbf{x}}) - F(\mathbf{x}^{t-1}) + \frac{\rho_{2k}^-}{4} \|\bar{\mathbf{x}} - \mathbf{x}^{t-1}\|^2 + \frac{1}{\rho_{2k}^-} \|\nabla_{S \setminus (S^t \cup S^{t-1})} F(\mathbf{x}^{t-1})\|^2 \\
& \quad + \frac{1}{\rho_{2k}^-} \|\nabla_{S^t \setminus S^{t-1}} F(\mathbf{x}^{t-1})\|^2. \tag{15}
\end{aligned}$$

Note that the last term is retained for deduction. What we need to show is a proper bound of the term $\|\nabla_{S \setminus (S^t \cup S^{t-1})} F(\mathbf{x}^{t-1})\|^2$ above. First, we observe that

$$\mathbf{z}_{S \setminus (S^t \cup S^{t-1})}^t = -\eta \nabla_{S \setminus (S^t \cup S^{t-1})} F(\mathbf{x}^{t-1}).$$

Next, we compare the elements of $S \setminus (S^t \cup S^{t-1})$ to those in $(S^t \cap S^{t-1}) \setminus S$. For convenience, we denote $T = J^t \setminus (S^{t-1} \cup S^t)$. Since S^t contains the k largest elements of $\mathbf{z}_{J^t}^t$, those of $(S^t \cap S^{t-1}) \setminus S$ are larger than those in T . On the other hand, recall that elements in $J^t \setminus S^{t-1}$ are larger than those in \bar{J}^t due to the partial hard thresholding. Since T is a subset of $J^t \setminus S^{t-1}$, we have that T is larger

than $\overline{J^t}$. Consequently, elements in $(S^t \cap S^{t-1}) \setminus S$ are larger than those in $T \cup \overline{J^t} = \overline{S^{t-1} \cup S^t}$. This suggests that

$$\frac{\left\| \mathbf{z}_{S \setminus (S^t \cup S^{t-1})}^t \right\|^2}{|S \setminus (S^t \cup S^{t-1})|} \leq \frac{\left\| \mathbf{z}_{(S^t \cap S^{t-1}) \setminus S}^t \right\|^2}{|(S^t \cap S^{t-1}) \setminus S|}.$$

Note that $|S^t \setminus S^{t-1}| < s$ implies $|(S^t \cap S^{t-1}) \setminus S| \geq k - 2s$. Therefore,

$$\begin{aligned} \eta^2 \left\| \nabla_{S \setminus (S^t \cup S^{t-1})} F(\mathbf{x}^{t-1}) \right\|^2 &\leq \frac{s}{k-2s} \left\| \mathbf{x}_{(S^t \cap S^{t-1}) \setminus S}^{t-1} - \eta \nabla_{(S^t \cap S^{t-1}) \setminus S} F(\mathbf{x}^{t-1}) \right\|^2 \\ &= \frac{s}{k-2s} \left\| \mathbf{x}_{(S^t \cap S^{t-1}) \setminus S}^{t-1} \right\|^2 \\ &= \frac{s}{k-2s} \left\| (\mathbf{x}^{t-1} - \bar{\mathbf{x}})_{(S^t \cap S^{t-1}) \setminus S} \right\|^2 \\ &\leq \frac{s}{k-2s} \left\| \mathbf{x}^{t-1} - \bar{\mathbf{x}} \right\|^2. \end{aligned}$$

Plugging the above into (15), we obtain

$$\begin{aligned} \frac{\rho_{2k}^-}{2} \left\| \bar{\mathbf{x}} - \mathbf{x}^{t-1} \right\|^2 &\leq F(\bar{\mathbf{x}}) - F(\mathbf{x}^{t-1}) + \frac{\rho_{2k}^-}{4} \left\| \bar{\mathbf{x}} - \mathbf{x}^{t-1} \right\|^2 + \frac{s}{(k-2s)\eta^2 \rho_{2k}^-} \left\| \bar{\mathbf{x}} - \mathbf{x}^{t-1} \right\|^2 \\ &\quad + \frac{1}{\rho_{2k}^-} \left\| \nabla_{S^t \setminus S^{t-1}} F(\mathbf{x}^{t-1}) \right\|^2. \end{aligned}$$

Picking $k \geq 2s + \frac{4s}{\eta^2 (\rho_{2k}^-)^2}$ gives

$$\frac{\rho_{2k}^-}{2} \left\| \bar{\mathbf{x}} - \mathbf{x}^{t-1} \right\|^2 \leq F(\bar{\mathbf{x}}) - F(\mathbf{x}^{t-1}) + \frac{\rho_{2k}^-}{2} \left\| \bar{\mathbf{x}} - \mathbf{x}^{t-1} \right\|^2 + \frac{1}{\rho_{2k}^-} \left\| \nabla_{S^t \setminus S^{t-1}} F(\mathbf{x}^{t-1}) \right\|^2,$$

which is exactly the claim (14).

Now we consider the parameter setting $r < s$. In this case, $|S^t \setminus S^{t-1}|$ cannot be greater than s . In fact, like we have done for Case 2, we can show that

$$\eta^2 \left\| \nabla_{S \setminus (S^t \cup S^{t-1})} F(\mathbf{x}^{t-1}) \right\|^2 \leq \frac{r}{k-r-s} \left\| \mathbf{x}^{t-1} - \bar{\mathbf{x}} \right\|^2.$$

Plugging the above into (15), we obtain

$$\begin{aligned} \frac{\rho_{2k}^-}{2} \left\| \bar{\mathbf{x}} - \mathbf{x}^{t-1} \right\|^2 &\leq F(\bar{\mathbf{x}}) - F(\mathbf{x}^{t-1}) + \frac{\rho_{2k}^-}{4} \left\| \bar{\mathbf{x}} - \mathbf{x}^{t-1} \right\|^2 + \frac{r}{(k-r-s)\eta^2 \rho_{2k}^-} \left\| \bar{\mathbf{x}} - \mathbf{x}^{t-1} \right\|^2 \\ &\quad + \frac{1}{\rho_{2k}^-} \left\| \nabla_{S^t \setminus S^{t-1}} F(\mathbf{x}^{t-1}) \right\|^2. \end{aligned}$$

Using $k \geq s + r + \frac{4r}{\eta^2 (\rho_{2k}^-)^2}$ we prove (14).

Overall, we find that picking $k \geq s + \left(1 + \frac{4}{\eta^2 (\rho_{2k}^-)^2}\right) \min\{r, s\}$ always guarantees the result. \square

Lemma 16. Consider the PHT(r) algorithm. Suppose that $F(\mathbf{x})$ is ρ_{2k}^+ -RSS. We have

$$F(\mathbf{x}^t) - F(\mathbf{x}^{t-1}) \leq -\frac{1 - \eta \rho_{2k}^+}{2\eta} \left\| \mathbf{z}_{S^t}^t - \mathbf{x}^{t-1} \right\|^2.$$

Proof. We partition \mathbf{z}^t into four disjoint parts: $S^{t-1} \setminus S^t$, $S^{t-1} \cap S^t$, $S^t \setminus S^{t-1}$ and $\overline{J^t}$. It then follows that

$$\begin{aligned} \left\| \mathbf{z}_{S^t}^t - \mathbf{z}^t \right\|^2 &= \left\| \mathbf{z}_{S^{t-1} \setminus S^t}^t \right\|^2 + \left\| \mathbf{z}_{\overline{J^t}}^t \right\|^2 \\ &\leq \left\| \mathbf{z}_{S^t \setminus S^{t-1}}^t \right\|^2 + \left\| \mathbf{z}_{\overline{J^t}}^t \right\|^2 \\ &= \left\| \mathbf{z}_{S^{t-1}}^t \right\|^2 \\ &= \eta^2 \left\| \nabla F(\mathbf{x}^{t-1}) \right\|^2. \end{aligned}$$

On the other hand, the LHS reads as

$$\begin{aligned}\|\mathbf{z}_{S^t}^t - \mathbf{z}^t\|^2 &= \|\mathbf{z}_{S^t}^t - \mathbf{x}^{t-1} + \eta \nabla F(\mathbf{x}^{t-1})\|^2 \\ &= \|\mathbf{z}_{S^t}^t - \mathbf{x}^{t-1}\|^2 + \eta^2 \|\nabla F(\mathbf{x}^{t-1})\|^2 + 2\eta \langle \nabla F(\mathbf{x}^{t-1}), \mathbf{z}_{S^t}^t - \mathbf{x}^{t-1} \rangle.\end{aligned}$$

Hence,

$$\langle \nabla F(\mathbf{x}^{t-1}), \mathbf{z}_{S^t}^t - \mathbf{x}^{t-1} \rangle \leq -\frac{1}{2\eta} \|\mathbf{z}_{S^t}^t - \mathbf{x}^{t-1}\|^2.$$

Using the RSS property, we have

$$\begin{aligned}F(\mathbf{x}^t) - F(\mathbf{x}^{t-1}) &\leq F(\mathbf{y}^t) - F(\mathbf{x}^{t-1}) \\ &= F(\mathbf{z}_{S^t}^t) - F(\mathbf{x}^{t-1}) \\ &\leq \langle \nabla F(\mathbf{x}^{t-1}), \mathbf{z}_{S^t}^t - \mathbf{x}^{t-1} \rangle + \frac{\rho_{2k}^+}{2} \|\mathbf{z}_{S^t}^t - \mathbf{x}^{t-1}\|^2 \\ &\leq -\frac{1 - \eta \rho_{2k}^+}{2\eta} \|\mathbf{z}_{S^t}^t - \mathbf{x}^{t-1}\|^2.\end{aligned}$$

□

B Technical Lemmas

Lemma 17. *Suppose that $F(\mathbf{x})$ is ρ_K^- -RSC and ρ_K^+ -RSS for some sparsity level $K > 0$. Then for all $\theta \in \mathbb{R}$, all vectors $\mathbf{x}, \mathbf{x}' \in \mathbb{R}^d$ and for any Hessian matrix \mathbf{H} of $F(\mathbf{x})$, we have*

$$|\langle \mathbf{x}, (\mathbf{I} - \theta \mathbf{H}) \mathbf{x}' \rangle| \leq \phi_K \|\mathbf{x}\| \cdot \|\mathbf{x}'\|,$$

provided that $|\text{supp}(\mathbf{x}) \cup \text{supp}(\mathbf{x}')| \leq K$, and

$$\|((\mathbf{I} - \theta \mathbf{H}) \mathbf{x})_S\| \leq \phi_K \|\mathbf{x}\|, \quad \text{if } |S \cup \text{supp}(\mathbf{x})| \leq K,$$

where

$$\phi_K = \max \{ |\theta \rho_K^- - 1|, |\theta \rho_K^+ - 1| \}.$$

Proof. Since \mathbf{H} is a Hessian matrix, we always have a decomposition $\mathbf{H} = \mathbf{A}^\top \mathbf{A}$ for some matrix \mathbf{A} . Denote $T = \text{supp}(\mathbf{x}) \cup \text{supp}(\mathbf{x}')$. By simple algebra, we have

$$\begin{aligned}|\langle \mathbf{x}, (\mathbf{I} - \theta \mathbf{H}) \mathbf{x}' \rangle| &= |\langle \mathbf{x}, \mathbf{x}' \rangle - \theta \langle \mathbf{A} \mathbf{x}, \mathbf{A} \mathbf{x}' \rangle| \\ &\stackrel{\zeta_1}{\leq} |\langle \mathbf{x}, \mathbf{x}' \rangle - \theta \langle \mathbf{A}_T \mathbf{x}, \mathbf{A}_T \mathbf{x}' \rangle| \\ &= \left| \left\langle \mathbf{x}, (\mathbf{I} - \theta \mathbf{A}_T^\top \mathbf{A}_T) \mathbf{x}' \right\rangle \right| \\ &\leq \left\| \mathbf{I} - \theta \mathbf{A}_T^\top \mathbf{A}_T \right\| \cdot \|\mathbf{x}\| \cdot \|\mathbf{x}'\| \\ &\stackrel{\zeta_2}{\leq} \max \{ |\theta \rho_K^- - 1|, |\theta \rho_K^+ - 1| \} \cdot \|\mathbf{x}\| \cdot \|\mathbf{x}'\|.\end{aligned}$$

Here, ζ_1 follows from the fact that $\text{supp}(\mathbf{x}) \cup \text{supp}(\mathbf{y}) = T$ and ζ_2 holds because the RSC and RSS properties imply that the singular values of any Hessian matrix restricted on an K -sparse support set are lower and upper bounded by ρ_K^- and ρ_K^+ , respectively.

For some index set S subject to $|S \cup \text{supp}(\mathbf{x})| \leq K$, let $\mathbf{x}' = ((\mathbf{I} - \theta \mathbf{H}) \mathbf{x})_S$. We immediately obtain

$$\|\mathbf{x}'\|^2 = \langle \mathbf{x}', (\mathbf{I} - \theta \mathbf{H}) \mathbf{x} \rangle \leq \phi_K \|\mathbf{x}'\| \cdot \|\mathbf{x}\|,$$

indicating

$$\|((\mathbf{I} - \theta \mathbf{H}) \mathbf{x})_S\| \leq \phi_K \|\mathbf{x}\|.$$

□

Lemma 18. Suppose that $F(\mathbf{x})$ is ρ_K^- -RSC and ρ_K^+ -RSS for some sparsity level $K > 0$. For all vectors $\mathbf{x}, \mathbf{x}' \in \mathbb{R}^d$ and support set T such that $|\text{supp}(\mathbf{x} - \mathbf{x}') \cup T| \leq K$, the following holds for all $\theta \in \mathbb{R}$:

$$\|(\mathbf{x} - \mathbf{x}' - \theta \nabla F(\mathbf{x}) + \theta \nabla F(\mathbf{x}'))_T\| \leq \phi_K \|\mathbf{x} - \mathbf{x}'\|,$$

where ϕ_K is given in Lemma 17.

Proof. In fact, for any two vectors \mathbf{x} and \mathbf{x}' , there always exists a quantity $t \in [0, 1]$, such that

$$\nabla F(\mathbf{x}) - \nabla F(\mathbf{x}') = \nabla^2 F(t\mathbf{x} + (1-t)\mathbf{x}')(\mathbf{x} - \mathbf{x}').$$

Let $\mathbf{H} = \nabla^2 F(t\mathbf{x} + (1-t)\mathbf{x}')$. We write

$$\begin{aligned} & \|(\mathbf{x} - \mathbf{x}' - \theta \nabla F(\mathbf{x}) + \theta \nabla F(\mathbf{x}'))_T\| \\ &= \|(\mathbf{x} - \mathbf{x}' - \theta \mathbf{H}(\mathbf{x} - \mathbf{x}'))_T\| \\ &= \|((\mathbf{I} - \theta \mathbf{H})(\mathbf{x} - \mathbf{x}'))_T\| \\ &\leq \phi_K \|\mathbf{x} - \mathbf{x}'\|, \end{aligned}$$

where the last inequality applies Lemma 17. \square

Lemma 19. Suppose that $F(\mathbf{x})$ is ρ_K^- -RSC and ρ_K^+ -RSS for some sparsity level $K > 0$. Let $\kappa := \rho_K^+/\rho_K^-$. For all vectors $\mathbf{x}, \mathbf{x}' \in \mathbb{R}^d$ with $|\text{supp}(\mathbf{x}) \cup \text{supp}(\mathbf{x}')| \leq K$, we have

$$\begin{aligned} \|\mathbf{x} - \mathbf{x}'\| &\leq \kappa \|\mathbf{x}'_T\| + \frac{1}{\rho_K^-} \|(\nabla F(\mathbf{x}) - \nabla F(\mathbf{x}'))_T\|, \\ \|(\mathbf{x} - \mathbf{x}')_T\| &\leq \left(1 - \frac{1}{\kappa}\right) \|\mathbf{x} - \mathbf{x}'\| + \frac{1}{\rho_K^-} \|(\nabla F(\mathbf{x}) - \nabla F(\mathbf{x}'))_T\|. \end{aligned}$$

where T is the support set of \mathbf{x} .

Proof. We begin with bounding the ℓ_2 -norm of the difference of \mathbf{x} and \mathbf{x}' . Let $\Omega = \text{supp}(\mathbf{x}')$. For any positive scalar $\theta \in \mathbb{R}$ we have

$$\begin{aligned} \|(\mathbf{x} - \mathbf{x}')_T\|^2 &= \langle \mathbf{x} - \mathbf{x}' - \theta \nabla F(\mathbf{x}) + \theta \nabla F(\mathbf{x}'), (\mathbf{x} - \mathbf{x}')_T \rangle \\ &\quad + \theta \langle \nabla F(\mathbf{x}) - \nabla F(\mathbf{x}'), (\mathbf{x} - \mathbf{x}')_T \rangle \\ &\leq \|(\mathbf{x} - \mathbf{x}' - \theta \nabla F(\mathbf{x}) + \theta \nabla F(\mathbf{x}'))_T\| \cdot \|(\mathbf{x} - \mathbf{x}')_T\| \\ &\quad + \theta \|(\nabla F(\mathbf{x}) - \nabla F(\mathbf{x}'))_T\| \cdot \|(\mathbf{x} - \mathbf{x}')_T\| \\ &\leq \|\mathbf{x} - \mathbf{x}' - \theta(\nabla F(\mathbf{x}))_{T \cup \Omega} + \theta(\nabla F(\mathbf{x}'))_{T \cup \Omega}\| \cdot \|(\mathbf{x} - \mathbf{x}')_T\| \\ &\quad + \theta \|(\nabla F(\mathbf{x}) - \nabla F(\mathbf{x}'))_T\| \cdot \|(\mathbf{x} - \mathbf{x}')_T\| \\ &\leq \phi_K \|\mathbf{x} - \mathbf{x}'\| \cdot \|(\mathbf{x} - \mathbf{x}')_T\| + \theta \|(\nabla F(\mathbf{x}) - \nabla F(\mathbf{x}'))_T\| \cdot \|(\mathbf{x} - \mathbf{x}')_T\|, \end{aligned}$$

where we recall that ϕ_K is given in Lemma 17. Dividing both sides by $\|(\mathbf{x} - \mathbf{x}')_T\|$ gives

$$\|(\mathbf{x} - \mathbf{x}')_T\| \leq \phi_K \|\mathbf{x} - \mathbf{x}'\| + \theta \|(\nabla F(\mathbf{x}) - \nabla F(\mathbf{x}'))_T\|.$$

On the other hand,

$$\begin{aligned} \|\mathbf{x} - \mathbf{x}'\| &\leq \|(\mathbf{x} - \mathbf{x}')_T\| + \|(\mathbf{x} - \mathbf{x}')_{\bar{T}}\| \\ &\leq \phi_K \|\mathbf{x} - \mathbf{x}'\| + \theta \|(\nabla F(\mathbf{x}) - \nabla F(\mathbf{x}'))_T\| + \|\mathbf{x}'_{\bar{T}}\|. \end{aligned}$$

Hence, we have

$$\|\mathbf{x} - \mathbf{x}'\| \leq \frac{1}{1 - \phi_K} \|\mathbf{x}'_{\bar{T}}\| + \frac{\theta}{1 - \phi_K} \|(\nabla F(\mathbf{x}) - \nabla F(\mathbf{x}'))_T\|.$$

Picking $\theta = 1/\rho_K^+$, we have $\phi_K = 1 - \frac{1}{\kappa}$. Plugging these into the above and noting that $\rho_K^+ \geq \rho_K^-$ complete the proof. \square

Lemma 20. Suppose that $F(\mathbf{x})$ is ρ_K^- -RSC. Then for any vectors \mathbf{x} and \mathbf{x}' with $\|\mathbf{x} - \mathbf{x}'\|_0 \leq K$, the following holds:

$$\|\mathbf{x} - \mathbf{x}'\| \leq \sqrt{\frac{2 \max\{F(\mathbf{x}) - F(\mathbf{x}'), 0\}}{\rho_K^-}} + \frac{2 \|(\nabla F(\mathbf{x}'))_T\|}{\rho_K^-},$$

where $T = \text{supp}(\mathbf{x} - \mathbf{x}')$.

Proof. The RSC property immediately implies

$$\begin{aligned} F(\mathbf{x}) - F(\mathbf{x}') &\geq \langle \nabla F(\mathbf{x}'), \mathbf{x} - \mathbf{x}' \rangle + \frac{\rho_K^-}{2} \|\mathbf{x} - \mathbf{x}'\|^2 \\ &\geq -\|\nabla_T F(\mathbf{x}')\| \cdot \|\mathbf{x} - \mathbf{x}'\| + \frac{\rho_K^-}{2} \|\mathbf{x} - \mathbf{x}'\|^2. \end{aligned}$$

Discussing the sign of $F(\mathbf{x}) - F(\mathbf{x}')$ and solving the above quadratic inequality completes the proof. \square

Proposition 21. Suppose that $F(\mathbf{x})$ is ρ_{k+s}^- -RSC and ρ_{2k}^+ -RSS. Let $\kappa := \rho_{2k}^+ / \rho_{k+s}^-$. Suppose that for all $t \geq 1$, \mathbf{x}^t is k -sparse and the following holds:

$$F(\mathbf{x}^t) - F(\bar{\mathbf{x}}) \leq \mu_t (F(\mathbf{x}^{t-1}) - F(\bar{\mathbf{x}})) + \tau,$$

where $0 < \mu_t < \mu < 1$ for some $\mu, \tau \geq 0$ and $\bar{\mathbf{x}}$ is an arbitrary s -sparse signal. Then,

$$\|\mathbf{x}^t - \bar{\mathbf{x}}\| \leq \sqrt{2\kappa}(\sqrt{\mu_1\mu_2 \dots \mu_t}) \|\mathbf{x}^0 - \bar{\mathbf{x}}\| + \frac{3}{\rho_{k+s}^-} \|\nabla_{k+s} F(\bar{\mathbf{x}})\| + \sqrt{\frac{2\tau}{\rho_{k+s}^- (1-\mu)}}.$$

Proof. The RSS property implies that

$$\begin{aligned} F(\mathbf{x}^0) - F(\bar{\mathbf{x}}) &\leq \langle \nabla F(\bar{\mathbf{x}}), \mathbf{x}^0 - \bar{\mathbf{x}} \rangle + \frac{\rho_{2k}^+}{2} \|\mathbf{x}^0 - \bar{\mathbf{x}}\|^2 \\ &\leq \frac{\rho_{2k}^+}{2} \|\mathbf{x}^0 - \bar{\mathbf{x}}\|^2 + \frac{1}{2\rho_{2k}^+} \|\nabla_{k+s} F(\bar{\mathbf{x}})\|^2 + \frac{\rho_{2k}^+}{2} \|\mathbf{x}^0 - \bar{\mathbf{x}}\|^2 \\ &\leq \rho_{2k}^+ \|\mathbf{x}^0 - \bar{\mathbf{x}}\|^2 + \frac{1}{2\rho_{2k}^+} \|\nabla_{k+s} F(\bar{\mathbf{x}})\|^2. \end{aligned}$$

Denote $\mu_{1:t} = \mu_1 \mu_2 \dots \mu_t$. We obtain

$$F(\mathbf{x}^t) - F(\bar{\mathbf{x}}) \leq \mu_{1:t} \rho^+ \|\mathbf{x}^0 - \bar{\mathbf{x}}\|^2 + \frac{1}{2\rho_{2k}^+} \|\nabla_{k+s} F(\bar{\mathbf{x}})\|^2 + \frac{\tau}{1-\mu}.$$

By Lemma 20, we have

$$\begin{aligned} &\|\mathbf{x}^t - \bar{\mathbf{x}}\| \\ &\leq \sqrt{\frac{2}{\rho_{k+s}^-}} \sqrt{\mu_{1:t} \rho_{2k}^+ \|\mathbf{x}^0 - \bar{\mathbf{x}}\|^2 + \frac{1}{2\rho_{2k}^+} \|\nabla_{k+s} F(\bar{\mathbf{x}})\|^2} + \frac{\tau}{1-\mu} + \frac{2}{\rho_{k+s}^-} \|\nabla_{k+s} F(\bar{\mathbf{x}})\| \\ &\leq \sqrt{2\kappa}(\sqrt{\mu_{1:t}}) \|\mathbf{x}^0 - \bar{\mathbf{x}}\| + \sqrt{\frac{1}{\rho_{k+s}^- \rho_{2k}^+}} \|\nabla_{k+s} F(\bar{\mathbf{x}})\| + \frac{2}{\rho_{k+s}^-} \|\nabla_{k+s} F(\bar{\mathbf{x}})\| + \sqrt{\frac{2\tau}{\rho_{k+s}^- (1-\mu)}} \\ &\leq \sqrt{2\kappa}(\sqrt{\mu_{1:t}}) \|\mathbf{x}^0 - \bar{\mathbf{x}}\| + \frac{3}{\rho_{k+s}^-} \|\nabla_{k+s} F(\bar{\mathbf{x}})\| + \sqrt{\frac{2\tau}{\rho_{k+s}^- (1-\mu)}}. \end{aligned}$$

\square