

Supplementary Materials for Compressed Smooth Sparse Decomposition*

Appendices

Appendix A. Proof of Theorem 2.1.

We prove Theorem 2.1 by contradiction: Assume there exists two different decompositions for the same smooth plus sparse signal \mathbf{y} , i.e., $\mathbf{y} = \mathbf{m}_1 + \mathbf{a}_1 = \mathbf{m}_2 + \mathbf{a}_2$, where $\mathbf{m}_1 \neq \mathbf{m}_2$ and $\mathbf{a}_1 \neq \mathbf{a}_2$. Then,

$$\mathbf{m}_1 - \mathbf{m}_2 = -\mathbf{a}_1 + \mathbf{a}_2 \quad (3)$$

Since $\mathbf{m}_1 \neq \mathbf{m}_2$, we can normalize both side by $\|\mathbf{m}_1 - \mathbf{m}_2\|_2$ and denote $\tilde{\mathbf{m}} = (\mathbf{m}_1 - \mathbf{m}_2)/\|\mathbf{m}_1 - \mathbf{m}_2\|_2$ and $\tilde{\mathbf{a}} = (-\mathbf{a}_1 + \mathbf{a}_2)/\|\mathbf{m}_1 - \mathbf{m}_2\|_2$. Notice that $\tilde{\mathbf{m}}$ is in the column space of \mathbf{B} which is spanned by columns of the \mathbf{U} (recall that $\mathbf{B} = \mathbf{U}\Sigma\mathbf{V}^T$), i.e., $\tilde{\mathbf{m}} = \mathbf{U}\mathbf{x}$, where \mathbf{x} is the coefficient vector, $\mathbf{x} \in \mathbb{R}^r$. Since $\|\tilde{\mathbf{m}}\|_2 = 1$, we have $\|\mathbf{x}\|_2 = 1$. We can bound each element in $\tilde{\mathbf{m}}$ as follows:

$$|\tilde{m}_i| = \mathbf{e}_i^T \mathbf{U} \mathbf{x} \leq \|\mathbf{e}_i^T \mathbf{U}\|_2 \|\mathbf{x}\|_2 \leq \max_{i \in \{1, \dots, r\}} \|\mathbf{U}^T \mathbf{e}_i\|_2 \leq \sqrt{\frac{\mu(\mathbf{B})r}{n}} < \sqrt{\frac{1}{2ls}}, \quad \forall i \in \{1, \dots, n\}.$$

According to Definition 2.1, $\|\mathbf{a}_1\|_0 \leq ls$ and $\|\mathbf{a}_2\|_0 \leq ls$. Therefore, $\|\tilde{\mathbf{a}}\|_0 \leq 2ls$. Moreover, according to Eq. (3), we conclude that $\|\tilde{\mathbf{m}}\|_0 \leq 2ls$. Denote the support of $\tilde{\mathbf{m}}$ as $\mathbf{T}_{\tilde{\mathbf{m}}}$, we have,

$$\|\tilde{\mathbf{m}}\|_2 = \sqrt{\sum_{i \in \mathbf{T}_{\tilde{\mathbf{m}}}} |\tilde{m}_i|^2} < 1 \text{ which is a contradiction.} \quad \blacksquare$$

Appendix B. Proof of Theorem 2.2.

We prove the Theorem 2.2 with contradiction: Assume there exists another vector $\mathbf{y} = \mathbf{B}\boldsymbol{\theta} + \mathbf{B}_a\boldsymbol{\theta}_a \in MS_{r,s,\mu,l}$ such that $\mathbf{A}\mathbf{y} = \mathbf{b}$ and $\mathbf{y} \neq \mathbf{y}_0$. Then, $\mathbf{z} = \mathbf{y} - \mathbf{y}_0 = \mathbf{B}(\boldsymbol{\theta} - \boldsymbol{\theta}_0) + \mathbf{B}_a(\boldsymbol{\theta}_a - \boldsymbol{\theta}_{a0})$ is a nonzero vector. Since $\|\boldsymbol{\theta}_a - \boldsymbol{\theta}_{a0}\|_1 \leq \|\boldsymbol{\theta}_a\|_1 + \|\boldsymbol{\theta}_{a0}\|_1 \leq 2s$, by Definition 2.3, we have that $\mathbf{z} \in MS_{r,2s,\mu,l}$. Therefore, $0 = \|\mathbf{A}\mathbf{z}\|_2 \geq (1 - \delta_{r,2s,\mu,l})\|\mathbf{z}\|_2 > 0$, which is a contradiction. Notice that the proof is inspired by the proof of Lemma 3.1 in Candes and Tao (2005). \blacksquare

Appendix C. Proof of Theorem 2.3.

This proof is inspired by Baraniuk et al. (2008) and Tanner and Vary (2020).

We will first derive the RIC for a fixed subspace $MS_{r,\mathbf{T},\mu,l}$ of $MS_{r,s,\mu,l}$ when $\boldsymbol{\theta}_a$ is restricted in a fixed subspace \mathbf{T} with the fixed support such that the number of non-zero elements is s .

$$MS_{r,\mathbf{T},\mu,l} = \{\mathbf{y} \in \mathbb{R}^n \mid \mathbf{y} = \mathbf{B}\boldsymbol{\theta} + \mathbf{B}_a\boldsymbol{\theta}_a, \mathbf{B}_a \in \mathbb{R}^{n \times q}, l(\mathbf{B}_a) = l, \boldsymbol{\theta}_a \in \mathbf{T}, \mathbf{B} \in \mathbb{R}^{n \times r}, \mu(\mathbf{B}) = \mu < n(2rsl)^{-1}, \boldsymbol{\theta} \in \mathbb{R}^r\}.$$

Then, we use a covering argument that counts over all possible sparse subspaces \mathbf{T} with support less than or equal to s . Finally, we can derive the RIC for $MS_{r,s,\mu,l}$.

The following lemma describes the RIC for a fixed subspace $MS_{r,\mathbf{T},\mu,l}$ and is proved in Appendix E.

* No data ethics considerations are foreseen related to this paper.

Lemma 6.1. *RIC for a fixed subspace $MS_{r,T,\mu,l}$. Let $\mathbf{A} \in \mathbb{R}^{p \times n}$ be a matrix from the families described in Definition 3.5. Further, assume $\mu < n(2rsl)^{-1}$ and basis matrix \mathbf{B}_a for sparse signal component satisfies the RIP with RIC $\delta_{\mathbf{B}_{a,s}} \in (0,1)$, i.e., $\delta_{\mathbf{B}_{a,s}}$ is the smallest positive constant such that*

$$(1 - \delta_{\mathbf{B}_{a,s}})\|\boldsymbol{\theta}_a\|_2 \leq \|\mathbf{B}_a \boldsymbol{\theta}_a\|_2 \leq (1 + \delta_{\mathbf{B}_{a,s}})\|\boldsymbol{\theta}_a\|_2, \quad \forall \boldsymbol{\theta}_a \in \{\boldsymbol{\theta}_a \in \mathbb{R}^q \mid \|\boldsymbol{\theta}_a\|_0 \leq s\}.$$

For a given $\delta \in (0,1)$, there exists a constant $c_0 > 0$ depending only on δ , such that the RIC for $MS_{r,T,\mu,l}$ is upper bounded by δ with the probability of at least $1 - 2\left(\frac{24}{\delta}\tau_1\right)^r \left(\frac{24}{\delta}\tau_0\right)^s e^{-pc_0(\delta/2)}$,

where $\eta = \sqrt{\frac{\mu rsl}{n}}$, $\tau_0 = \frac{1}{\sqrt{(1-\delta_{\mathbf{B}_{a,s}})(1-\eta^2)}}$, $\tau_1 = \|\mathbf{B}^\dagger\|_2 \left(1 + \frac{1}{\sqrt{1-\eta^2}}\right)$, and $\mathbf{B}^\dagger = (\mathbf{B}^T \mathbf{B})^{-1} \mathbf{B}^T$. \blacksquare

Notice that for a fixed subspace $MS_{r,T,\mu,l}$, the RIP will fail with probability less than or equal to $2\left(\frac{24}{\delta}\tau_1\right)^r \left(\frac{24}{\delta}\tau_0\right)^s e^{-pc_0(\delta/2)}$. Since there are $\binom{n}{s} \leq \left(\frac{en}{s}\right)^s$ such subspaces, the probability to fail for $MS_{r,s,\mu,l}$, which is a combination of those $\binom{n}{s}$ subspaces, will be less than or equal to

$$\binom{n}{s} 2\left(\frac{24}{\delta}\tau_1\right)^r \left(\frac{24}{\delta}\tau_0\right)^s e^{-pc_0(\frac{\delta}{2})} \leq \exp\left(-c_0\left(\frac{\delta}{2}\right)p + \ln 2 + r \ln \frac{24}{\delta}\tau_1 + s\left(1 + \ln \frac{24}{\delta}\tau_0 + \ln \frac{n}{s}\right)\right).$$

Then, for any give δ , there exist $c_1, c_2 > 0$, such that the probability to fail for $MS_{r,s,\mu,l}$ is less than or equal to $\exp(-c_1 p)$, provided that $p \geq c_2 \left(\ln 2 + r \ln \frac{24}{\delta}\tau_1 + s\left(1 + \ln \frac{24}{\delta}\tau_0 + \ln \frac{n}{s}\right)\right)$, where $c_2 = \left[c_0\left(\frac{\delta}{2}\right) - c_1\right]^{-1}$. This finishes the proof. \blacksquare

Appendix D. Proof of Theorem 2.4.

The proof of Theorem 2.4 is inspired by Candes et al. (2006) and Tanner and Vary (2020). Assume that in Problem (1), ϵ_1 is properly chosen such that Problem (1) is feasible. In the following discussion, we will use $(\cdot)^*$ to denote the optimal solution of Problem (1) and $(\cdot)_0$ to denote the signal we wish to recover. Let $\mathbf{R} = \mathbf{X}^* - \mathbf{X}_0 = \mathbf{R}^m + \mathbf{R}^a$ where $\mathbf{R}^m = \mathbf{m} - \mathbf{m}_0 = \mathbf{B}\boldsymbol{\theta}^* - \mathbf{B}\boldsymbol{\theta}_0$ and $\mathbf{R}^a = \mathbf{B}_a \boldsymbol{\theta}_a^* - \mathbf{B}_a \boldsymbol{\theta}_{a0}$ are the residual of the smooth and sparse signal component, respectively.

Let $\mathbf{h} = \boldsymbol{\theta}_a^* - \boldsymbol{\theta}_{a0} = \mathbf{h}_{\mathbf{T}_0} + \mathbf{h}_{\mathbf{T}_0^c}$, where \mathbf{T}_0 is the support of $\boldsymbol{\theta}_{a0}$ and $\mathbf{h}_{\mathbf{T}_0}$ denotes the projection of \mathbf{h} onto \mathbf{T}_0 such that

$$\mathbf{h}_{\mathbf{T}_0}(t) = \begin{cases} t, & \text{if } t \in \mathbf{T}_0 \\ 0, & \text{otherwise} \end{cases}$$

and \mathbf{T}_0^c denotes the complementary set of \mathbf{T}_0 .

Since $\boldsymbol{\theta}_{a0}$ is feasible and $\boldsymbol{\theta}_a^*$ is the optimal solution of Problem (1), we must have $\|\boldsymbol{\theta}_a^*\|_1 \leq \|\boldsymbol{\theta}_{a0}\|_1$, which is equivalent to $\|\boldsymbol{\theta}_{a0} + \mathbf{h}_{\mathbf{T}_0} + \mathbf{h}_{\mathbf{T}_0^c}\|_1 \leq \|\boldsymbol{\theta}_{a0}\|_1$. Since \mathbf{T}_0 and \mathbf{T}_0^c are complementary to each other, we have $\|\boldsymbol{\theta}_{a0} + \mathbf{h}_{\mathbf{T}_0}\|_1 + \|\mathbf{h}_{\mathbf{T}_0^c}\|_1 \leq \|\boldsymbol{\theta}_{a0}\|_1$. Since $\|\boldsymbol{\theta}_{a0} + \mathbf{h}_{\mathbf{T}_0}\|_1 \geq \|\boldsymbol{\theta}_{a0}\|_1 - \|\mathbf{h}_{\mathbf{T}_0}\|_1$, we have $\|\boldsymbol{\theta}_{a0}\|_1 - \|\mathbf{h}_{\mathbf{T}_0}\|_1 + \|\mathbf{h}_{\mathbf{T}_0^c}\|_1 \leq \|\boldsymbol{\theta}_{a0}\|_1$. Hence $\|\mathbf{h}_{\mathbf{T}_0^c}\|_1 \leq \|\mathbf{h}_{\mathbf{T}_0}\|_1$. Since $\|\mathbf{h}_{\mathbf{T}_0}\|_1 \leq \sqrt{s}\|\mathbf{h}_{\mathbf{T}_0}\|_2$, we have

$$\|\mathbf{h}_{\mathbf{T}_0^c}\|_1 \leq \sqrt{s}\|\mathbf{h}_{\mathbf{T}_0}\|_2 \quad (4)$$

Similar to Candes et al. (2006), we order the elements of \mathbf{T}_0^c in decreasing order of their magnitude and enumerate \mathbf{T}_0^c as $v_1, \dots, v_{n-|\mathbf{T}_0|}$. Then, \mathbf{T}_0^c is divided into subsets \mathbf{T}_i^c of size M , where

$$\mathbf{T}_i^c = \{v_j: (i-1)M \leq j \leq iM\}.$$

Let $\mathbf{h}_{\mathbf{T}_i^c}$ be the projection of \mathbf{h} onto \mathbf{T}_i^c , we have

$$\begin{aligned} \|\mathbf{h}_{\mathbf{T}_i^c}\|_0 &\leq M, \quad \forall i \geq 1 \\ \mathbf{T}_i^c \cap \mathbf{T}_j^c &= \emptyset, \quad \forall i \neq j \\ \|\mathbf{h}_{\mathbf{T}_{i+1}^c}\|_2 &\leq \frac{1}{\sqrt{M}}\|\mathbf{h}_{\mathbf{T}_i^c}\|_1, \quad \forall i \geq 1 \end{aligned} \quad (5)$$

where the last inequality comes from the fact that \mathbf{T}_0^c is in decreasing order, such that

$$|\mathbf{h}_{\mathbf{T}_{i+1}^c}|_{(v)} \leq \frac{1}{M} \sum_{j \in \mathbf{T}_i^c} |\mathbf{h}_{\mathbf{T}_i^c}|_{(j)}. \quad \forall v \in \mathbf{T}_{i+1}^c$$

Define $\mathbf{R}_{\mathbf{T}_i^c}^a = \mathbf{B}_a \mathbf{h}_{\mathbf{T}_i^c}$, $\mathbf{R}_{\mathbf{T}_0}^a = \mathbf{B}_a \mathbf{h}_{\mathbf{T}_0}$ and combine Eq. (4) and Eq. (5), we have

$$\begin{aligned} \sum_{j \geq 2} \|\mathbf{R}_{\mathbf{T}_j^c}^a\|_2 &\leq \sum_{j \geq 2} \sqrt{1 + \delta_{\mathbf{B}_a, M}} \|\mathbf{h}_{\mathbf{T}_j^c}\|_2 \stackrel{(a)}{\leq} \sum_{j \geq 1} \frac{\sqrt{1 + \delta_{\mathbf{B}_a, M}} \|\mathbf{h}_{\mathbf{T}_j^c}\|_1}{\sqrt{M}} = \frac{\sqrt{1 + \delta_{\mathbf{B}_a, M}} \|\mathbf{h}_{\mathbf{T}_0^c}\|_1}{\sqrt{M}} \\ &\stackrel{(b)}{\leq} \frac{\sqrt{s} \sqrt{1 + \delta_{\mathbf{B}_a, M}} \|\mathbf{h}_{\mathbf{T}_0}\|_2}{\sqrt{M}} \leq \frac{\sqrt{s} \sqrt{1 + \delta_{\mathbf{B}_a, M}} \|\mathbf{R}_{\mathbf{T}_0}^a\|_2}{\sqrt{M} \sqrt{1 - \delta_{\mathbf{B}_a, M}}}, \end{aligned}$$

where (a) follows Eq.(5), (b) follows Eq. (4) and the RIP property of \mathbf{B}_a is used since $\|\mathbf{h}_{\mathbf{T}_j^c}\|_0 \leq M$.

Denote $\gamma = \sqrt{\frac{1 + \delta_{\mathbf{B}_a, M+s}}{1 - \delta_{\mathbf{B}_a, M+s}}}$ (A tighter bound can be achieved by using $\sqrt{\frac{1 + \delta_{\mathbf{B}_a, M}}{1 - \delta_{\mathbf{B}_a, M}}}$. However, we adopt $\delta_{\mathbf{B}_a, M+s}$ instead of $\delta_{\mathbf{B}_a, M}$ for simplicity in the following proof), we have

$$\sum_{j \geq 2} \|\mathbf{R}_{\mathbf{T}_j^c}^a\|_2 \leq \sqrt{\frac{s}{M}} \gamma \|\mathbf{R}_{\mathbf{T}_0}^a\|_2 \quad (6)$$

Next, we derive the bound for \mathbf{R}^m and \mathbf{R}^a respectively.

Bound for \mathbf{R}^m :

$$\begin{aligned} \|\mathbf{A}\mathbf{R}^m\|_2^2 &= |\langle \mathbf{A}\mathbf{R}^m, \mathbf{A}(\mathbf{R} - \mathbf{R}^a) \rangle| \\ &= |\langle \mathbf{A}\mathbf{R}^m, \mathbf{A}(\mathbf{R} - \mathbf{R}^a) \rangle| \\ &= |\langle \mathbf{A}\mathbf{R}^m, \mathbf{A}\mathbf{R} \rangle + \langle \mathbf{A}\mathbf{R}^m, -\mathbf{A}\mathbf{R}^a \rangle| \\ &\leq |\langle \mathbf{A}\mathbf{R}^m, \mathbf{A}\mathbf{R} \rangle| + |\langle \mathbf{A}\mathbf{R}^m, -\mathbf{A}\mathbf{R}^a \rangle| \\ &= |\langle \mathbf{A}\mathbf{R}^m, \mathbf{A}\mathbf{R} \rangle| + \left| \langle \mathbf{A}\mathbf{R}^m, -\mathbf{A}(\mathbf{R}_{\mathbf{T}_0}^a + \mathbf{R}_{\mathbf{T}_1^c}^a + \sum_{j \geq 2} \mathbf{R}_{\mathbf{T}_j^c}^a) \rangle \right| \\ &\leq |\langle \mathbf{A}\mathbf{R}^m, \mathbf{A}\mathbf{R} \rangle| + \left| \langle \mathbf{A}\mathbf{R}^m, \mathbf{A}(\mathbf{R}_{\mathbf{T}_0}^a + \mathbf{R}_{\mathbf{T}_1^c}^a) \rangle \right| + \sum_{j \geq 2} |\langle \mathbf{A}\mathbf{R}^m, \mathbf{A}\mathbf{R}_{\mathbf{T}_j^c}^a \rangle|. \end{aligned} \quad (7)$$

In the following discussion, we will bound those terms respectively. According to Cauchy-Schwarz inequality, the first term can be bounded as

$$|\langle \mathbf{A}\mathbf{R}^m, \mathbf{A}\mathbf{R} \rangle| \leq \|\mathbf{A}\mathbf{R}^m\|_2 \|\mathbf{A}\mathbf{R}\|_2 \leq \sqrt{1 + \delta_{r,s,\mu,l} \epsilon_1} \|\mathbf{R}^m\|_2, \quad (8)$$

where the last inequality comes from the RIP property and the first constraint in Problem (1).

The third term can be bounded as follows: denote $\mathbf{z}_1 = \mathbf{R}^m / \|\mathbf{R}^m\|_2$ and $\mathbf{z}_2 = \mathbf{R}_{\mathbf{T}_j^c}^a / \|\mathbf{R}_{\mathbf{T}_j^c}^a\|_2$, we have

$$\begin{aligned} \frac{|\langle \mathbf{A}\mathbf{R}^m, \mathbf{A}\mathbf{R}_{\mathbf{T}_j^c}^a \rangle|}{\|\mathbf{R}^m\|_2 \|\mathbf{R}_{\mathbf{T}_j^c}^a\|_2} &= |\langle \mathbf{A}\mathbf{z}_1, \mathbf{A}\mathbf{z}_2 \rangle| = \frac{1}{4} |\|\mathbf{A}(\mathbf{z}_1 + \mathbf{z}_2)\|_2^2 - \|\mathbf{A}(\mathbf{z}_1 - \mathbf{z}_2)\|_2^2| \\ &\leq_{(a)} \frac{1}{4} \max \left\{ |(1 + \delta_{r,M,\mu,l})\|\mathbf{z}_1 + \mathbf{z}_2\|_2^2 - (1 - \delta_{r,M,\mu,l})\|\mathbf{z}_1 - \mathbf{z}_2\|_2^2|, \right. \\ &\quad \left. |(1 + \delta_{r,M,\mu,l})\|\mathbf{z}_1 - \mathbf{z}_2\|_2^2 - (1 - \delta_{r,M,\mu,l})\|\mathbf{z}_1 + \mathbf{z}_2\|_2^2| \right\} \\ &= |\delta_{r,M,\mu,l} + \langle \mathbf{z}_1, \mathbf{z}_2 \rangle| \leq_{(b)} \delta_{r,M,\mu,l} + \frac{\eta_1}{1 - \eta_1^2}, \end{aligned}$$

where $\eta_1 = \sqrt{\frac{\mu r M l}{n}}$; inequality (a) follows the RIP property since $\mathbf{z}_1 + \mathbf{z}_2 \in MS_{r,M,\mu,l}$ and $\mathbf{z}_1 - \mathbf{z}_2 \in MS_{r,M,\mu,l}$; inequality (b) follows Eq. (26) in the proof of Lemma 6.2 and $\|\mathbf{z}_1\|_2 = \|\mathbf{z}_2\|_2 = 1$, $\langle \mathbf{z}_1, \mathbf{z}_2 \rangle \leq \frac{\eta_1}{1 - \eta_1^2} \|\mathbf{z}_1\|_2 \|\mathbf{z}_2\|_2 = \frac{\eta_1}{1 - \eta_1^2}$, provided that $M \leq 2s$.

Therefore,

$$|\langle \mathbf{A}\mathbf{R}^m, \mathbf{A}\mathbf{R}_{\mathbf{T}_j^c}^a \rangle| \leq \left(\delta_{r,M,\mu,l} + \frac{\eta_1}{1 - \eta_1^2} \right) \|\mathbf{R}^m\|_2 \|\mathbf{R}_{\mathbf{T}_j^c}^a\|_2. \quad (9)$$

Similarly, the second term can be bounded as

$$|\langle \mathbf{A}\mathbf{R}^m, \mathbf{A}(\mathbf{R}_{\mathbf{T}_0}^a + \mathbf{R}_{\mathbf{T}_1^c}^a) \rangle| \leq \left(\delta_{r,M+s,\mu,l} + \frac{\eta_2}{1 - \eta_2^2} \right) \|\mathbf{R}^m\|_2 \|\mathbf{R}_{\mathbf{T}_0}^a + \mathbf{R}_{\mathbf{T}_1^c}^a\|_2, \quad (10)$$

where $\eta_2 = \sqrt{\frac{\mu r (M+s) l}{n}}$, provided that $M \leq s$.

Plugging Eq. (8), Eq. (9) and Eq. (10) into Eq. (7), we have

$$\begin{aligned} \|\mathbf{A}\mathbf{R}^m\|_2^2 &\leq \|\mathbf{R}^m\|_2^2 \left(\sqrt{1 + \delta_{r,s,\mu,l} \epsilon_1} + \left(\delta_{r,M+s,\mu,l} + \frac{\eta_2}{1 - \eta_2^2} \right) \|\mathbf{R}_{\mathbf{T}_0}^a + \mathbf{R}_{\mathbf{T}_1^c}^a\|_2 + \left(\delta_{r,M,\mu,l} + \frac{\eta_1}{1 - \eta_1^2} \right) \|\mathbf{R}_{\mathbf{T}_j^c}^a\|_2 \right) \\ &\leq_{(a)} \|\mathbf{R}^m\|_2^2 \left(\sqrt{1 + \delta_{r,s,\mu,l} \epsilon_1} + \left(\delta_{r,M+s,\mu,l} + \frac{\eta_2}{1 - \eta_2^2} \right) \|\mathbf{R}_{\mathbf{T}_0}^a + \mathbf{R}_{\mathbf{T}_1^c}^a\|_2 + \left(\delta_{r,M,\mu,l} + \frac{\eta_1}{1 - \eta_1^2} \right) \sqrt{\frac{s}{M}} \gamma \|\mathbf{R}_{\mathbf{T}_0}^a\|_2 \right), \end{aligned}$$

where inequality (a) follows Eq. (6).

According to the RIP property, we have

$$\begin{aligned} (1 - \delta_{r,s,\mu,l}) \|\mathbf{R}^m\|_2^2 &\leq \|\mathbf{R}^m\|_2^2 \left(\sqrt{1 + \delta_{r,s,\mu,l} \epsilon_1} + \left(\delta_{r,M+s,\mu,l} + \frac{\eta_2}{1 - \eta_2^2} \right) \|\mathbf{R}_{\mathbf{T}_0}^a + \mathbf{R}_{\mathbf{T}_1^c}^a\|_2 + \left(\delta_{r,M,\mu,l} + \frac{\eta_1}{1 - \eta_1^2} \right) \sqrt{\frac{s}{M}} \gamma \|\mathbf{R}_{\mathbf{T}_0}^a\|_2 \right). \end{aligned}$$

Consequently, we have

$$\|R^m\|_2 \leq \frac{\sqrt{1 + \delta_{r,s,\mu,l}}\epsilon_1 + \left(\left(\delta_{r,M,\mu,l} + \frac{\eta_1}{1 - \eta_1^2} \right) \sqrt{\frac{s}{M}}\gamma^2 + \left(\delta_{r,M+s,\mu,l} + \frac{\eta_2}{1 - \eta_2^2} \right) \right) \|R_{T_0}^a + R_{T_1^c}^a\|_2}{(1 - \delta_{r,s,\mu,l})}. \quad (11)$$

where the inequality follows from:

$$\begin{aligned} \|R_{T_0}^a\|_2 &= \|B_a h_{T_0}\|_2 \leq_{(a)} \sqrt{1 + \delta_{B_a,s}} \|h_{T_0}\|_2 \leq_{(b)} \sqrt{1 + \delta_{B_a,s}} \|h_{T_0} + h_{T_1^c}\|_2 \\ &\leq_{(c)} \sqrt{\frac{1 + \delta_{B_a,s}}{1 - \delta_{B_a,M+s}}} \|B_a(h_{T_0} + h_{T_1^c})\|_2 \leq \gamma \|R_{T_0}^a + R_{T_1^c}^a\|_2, \end{aligned} \quad (12)$$

where inequalities (a) and (c) follow the RIP property of B_a and inequality (b) follows that $T_0 \cap T_1^c = \emptyset$.

Bound for R^a :

$$\begin{aligned} \|A(R_{T_0}^a + R_{T_1^c}^a)\|_2^2 &= |\langle A(R_{T_0}^a + R_{T_1^c}^a), A(R_{T_0}^a + R_{T_1^c}^a - R + R) \rangle| \\ &= |\langle A(R_{T_0}^a + R_{T_1^c}^a), AR \rangle| + |\langle A(R_{T_0}^a + R_{T_1^c}^a), -A(R^m + \sum_{j \geq 2} R_{T_j^c}^a) \rangle| \\ &\leq |\langle A(R_{T_0}^a + R_{T_1^c}^a), AR \rangle| + |\langle AR^m, A(R_{T_0}^a + R_{T_1^c}^a) \rangle| + \sum_{j \geq 2} |\langle A(R_{T_0}^a + R_{T_1^c}^a), AR_{T_j^c}^a \rangle| \end{aligned} \quad (13)$$

In the following discussion, we will bound those terms respectively. According to Cauchy-Schwarz inequality, the first term can be bounded as

$$|\langle A(R_{T_0}^a + R_{T_1^c}^a), AR \rangle| \leq \|A(R_{T_0}^a + R_{T_1^c}^a)\|_2 \|AR\|_2 \leq \sqrt{1 + \delta_{r,M+s,\mu,l}}\epsilon_1 \|R_{T_0}^a + R_{T_1^c}^a\|_2, \quad (14)$$

where the last inequality comes from the RIP property and the first constraint in Problem (1).

The third term can be bounded as follows: denote $z_2 = R_{T_j^c}^a / \|R_{T_j^c}^a\|_2, j \geq 2$, and $z_3 = (R_{T_0}^a + R_{T_1^c}^a) / \|R_{T_0}^a + R_{T_1^c}^a\|_2$, we have

$$\begin{aligned} &\frac{|\langle A(R_{T_0}^a + R_{T_1^c}^a), AR_{T_j^c}^a \rangle|}{\|R_{T_0}^a + R_{T_1^c}^a\|_2 \|R_{T_j^c}^a\|_2} = |\langle Az_3, Az_2 \rangle| \\ &= \frac{1}{4} |\|A(z_3 + z_2)\|_2^2 - \|A(z_3 - z_2)\|_2^2| \\ &\leq_{(a)} \frac{1}{4} \max \left\{ |(1 + \delta_{r,2M+s,\mu,l})\|z_3 + z_2\|_2^2 - (1 - \delta_{r,2M+s,\mu,l})\|z_3 - z_2\|_2^2|, \right. \\ &\quad \left. |(1 + \delta_{r,M+2s,\mu,l})\|z_3 - z_2\|_2^2 - (1 - \delta_{r,2M+s,\mu,l})\|z_3 + z_2\|_2^2| \right\} \\ &= |\delta_{r,2M+s,\mu,l} + \langle z_3, z_2 \rangle| \\ &\leq_{(b)} \delta_{r,2M+s,\mu,l}, \end{aligned}$$

where inequality (a) follows the RIP property since $z_3 + z_2 \in MS_{r,2M+s,\mu,l}$ and $z_3 - z_2 \in MS_{r,2M+s,\mu,l}$; inequality (b) comes from that $T_i^c \cap T_j^c = \emptyset, \forall i \neq j$ and $T_0 \cap T_i^c = \emptyset, \forall i$.

Therefore,

$$|\langle \mathbf{A}\mathbf{R}^m, \mathbf{A}\mathbf{R}_{\mathbf{T}_j^c}^a \rangle| \leq \delta_{r,2M+s,\mu,l} \|\mathbf{R}_{\mathbf{T}_0}^a + \mathbf{R}_{\mathbf{T}_1^c}^a\|_2 \|\mathbf{R}_{\mathbf{T}_j^c}^a\|_2. \quad (15)$$

Plugging Eq. (10), Eq. (14) and Eq. (15) into Eq. (13), we have

$$\begin{aligned} & \left\| \mathbf{A}(\mathbf{R}_{\mathbf{T}_0}^a + \mathbf{R}_{\mathbf{T}_1^c}^a) \right\|_2^2 \\ & \leq \|\mathbf{R}_{\mathbf{T}_0}^a + \mathbf{R}_{\mathbf{T}_1^c}^a\|_2 \left(\sqrt{1 + \delta_{r,M+s,\mu,l}\epsilon_1} + \left(\delta_{r,M+s,\mu,l} + \frac{\eta_2}{1-\eta_2^2} \right) \|\mathbf{R}^m\|_2 + \delta_{r,2M+s,\mu,l} \sum_{j \geq 2} \|\mathbf{R}_{\mathbf{T}_j^c}^a\|_2 \right) \\ & \leq_{(a)} \|\mathbf{R}_{\mathbf{T}_0}^a + \mathbf{R}_{\mathbf{T}_1^c}^a\|_2 \left(\sqrt{1 + \delta_{r,M+s,\mu,l}\epsilon_1} + \left(\delta_{r,M+s,\mu,l} + \frac{\eta_2}{1-\eta_2^2} \right) \|\mathbf{R}^m\|_2 + \delta_{r,2M+s,\mu,l} \sqrt{\frac{s}{M}} \gamma \|\mathbf{R}_{\mathbf{T}_0}^a\|_2 \right) \\ & \leq_{(b)} \|\mathbf{R}_{\mathbf{T}_0}^a + \mathbf{R}_{\mathbf{T}_1^c}^a\|_2 \left(\sqrt{1 + \delta_{r,M+s,\mu,l}\epsilon_1} + \left(\delta_{r,M+s,\mu,l} + \frac{\eta_2}{1-\eta_2^2} \right) \|\mathbf{R}^m\|_2 + \delta_{r,2M+s,\mu,l} \sqrt{\frac{s}{M}} \gamma^2 \|\mathbf{R}_{\mathbf{T}_0}^a + \mathbf{R}_{\mathbf{T}_1^c}^a\|_2 \right), \end{aligned}$$

where inequalities (a) and (b) follows the same argument as deriving Eq. (11).

According to the RIP property, we have

$$\begin{aligned} & (1 - \delta_{r,M+s,\mu,l}) \|\mathbf{R}_{\mathbf{T}_0}^a + \mathbf{R}_{\mathbf{T}_1^c}^a\|_2^2 \\ & \leq \|\mathbf{R}_{\mathbf{T}_0}^a + \mathbf{R}_{\mathbf{T}_1^c}^a\|_2 \left(\sqrt{1 + \delta_{r,M+s,\mu,l}\epsilon_1} + \left(\delta_{r,M+s,\mu,l} + \frac{\eta_2}{1-\eta_2^2} \right) \|\mathbf{R}^m\|_2 + \delta_{r,2M+s,\mu,l} \sqrt{\frac{s}{M}} \gamma^2 \|\mathbf{R}_{\mathbf{T}_0}^a\|_2 \right). \end{aligned}$$

Consequently, we have

$$\begin{aligned} & \|\mathbf{R}_{\mathbf{T}_0}^a + \mathbf{R}_{\mathbf{T}_1^c}^a\|_2 \\ & \leq \frac{\left(\sqrt{1 + \delta_{r,M+s,\mu,l}\epsilon_1} + \left(\delta_{r,M+s,\mu,l} + \frac{\eta_2}{1-\eta_2^2} \right) \|\mathbf{R}^m\|_2 + \delta_{r,2M+s,\mu,l} \sqrt{\frac{s}{M}} \gamma^2 \|\mathbf{R}_{\mathbf{T}_0}^a + \mathbf{R}_{\mathbf{T}_1^c}^a\|_2 \right)}{(1 - \delta_{r,M+s,\mu,l})} \quad (16) \end{aligned}$$

Notice that Eq. (11) and Eq. (16) still hold if we relax $\delta_{r,M+s,\mu,l}$, $\delta_{r,s,\mu,l}$ to $\delta_{r,2M+s,\mu,l}$. For simplicity, here we replace $\delta_{r,M+s,\mu,l}$, $\delta_{r,s,\mu,l}$ with $\delta_{r,2M+s,\mu,l}$ in the following derivation.

Plugging Eq. (11) into Eq. (16), and let $x \equiv \|\mathbf{R}_{\mathbf{T}_0}^a + \mathbf{R}_{\mathbf{T}_1^c}^a\|_2$, $y \equiv \|\mathbf{R}^m\|_2$, we have

$$\left(D_1 - \frac{B_1 B_2}{D_2} - C_1 \right) x \leq A_1 \epsilon_1 + \frac{B_1}{D_2} A_2 \epsilon_1 \quad (17)$$

where

$$A_1 = \sqrt{1 + \delta_{r,2M+s,\mu,l}\epsilon_1}, B_1 = \left(\delta_{r,2M+s,\mu,l} + \frac{\eta_2}{1-\eta_2^2} \right), C_1 = \delta_{r,2M+s,\mu,l} \sqrt{\frac{s}{M}} \gamma^2, D_1 = (1 - \delta_{r,2M+s,\mu,l}),$$

and

$$\begin{aligned} A_2 &= \sqrt{1 + \delta_{r,2M+s,\mu,l}\epsilon_1}, B_2 = \left(\delta_{r,2M+s,\mu,l} + \frac{\eta_1}{1-\eta_1^2} \right) \sqrt{\frac{s}{M}} \gamma^2 + \left(\delta_{r,2M+s,\mu,l} + \frac{\eta_2}{1-\eta_2^2} \right), D_2 \\ &= (1 - \delta_{r,2M+s,\mu,l}). \end{aligned}$$

Here we require $D_1 - \frac{B_1 B_2}{D_2} - C_1 > 0$ in Eq. (17) and let $M = s$, which is

$$1 - \gamma^2 \alpha_1 \alpha_2 - \alpha_2^2 - ((1 + \alpha_1 + \alpha_2) \gamma^2 + 2\alpha_2 + 2) \delta_{r,3s,\mu,l} > 0. \quad (18)$$

Let $a = (1 + \alpha_1 + \alpha_2)\gamma^2 + 2\alpha_2 + 2$ and $c = 1 - \gamma^2\alpha_1\alpha_2 - \alpha_2^2$, where $\alpha_1 = \frac{\eta}{1-\eta^2}$, $\alpha_2 = \frac{\sqrt{2}\eta}{1-2\eta^2}$, $\gamma = \sqrt{\frac{1+\delta_{B_a,2s}}{1-\delta_{B_a,2s}}}$, $\eta = \sqrt{\frac{\mu rsl}{n}}$. If $c > 0$, then, there exist a $\delta_{r,3s,\mu,l} > 0$ such that Eq. (18) is valid. Then the denominator $-a\delta_{r,3s,\mu,l} + c > 0 \forall \delta_{r,3s,\mu,l} \in (0, c/a)$. Consequently,

$$\left\| \mathbf{R}_{T_0}^a + \mathbf{R}_{T_1^c}^a \right\|_2 \leq \frac{(1 + \alpha_2)\sqrt{1 + \delta_{r,3s,\mu,l}}}{c - a\delta_{r,3s,\mu,l}} \epsilon_1.$$

Notice that from Eq. (6) and Eq. (12), we have

$$\sum_{j \geq 2} \left\| \mathbf{R}_{T_j^c}^a \right\|_2 \leq \gamma \left\| \mathbf{R}_{T_0}^a \right\|_2 \leq \gamma^2 \left\| \mathbf{R}_{T_0}^a + \mathbf{R}_{T_1^c}^a \right\|_2$$

Therefore, $\left\| \mathbf{R}^a \right\|_2 \leq \left\| \mathbf{R}_{T_0}^a + \mathbf{R}_{T_1^c}^a \right\|_2 + \sum_{j \geq 2} \left\| \mathbf{R}_{T_j^c}^a \right\|_2 \leq C_a \epsilon_1$, where

$$C_a = \frac{(1 + \gamma^2)(1 + \alpha_2)\sqrt{1 + \delta_{r,3s,\mu,l}}}{c - a\delta_{r,3s,\mu,l}}.$$

Similarly, we can bound $\left\| \mathbf{R}^m \right\|_2$ as $\left\| \mathbf{R}^m \right\|_2 \leq C_m \epsilon_1$, where

$$C_m = \frac{\sqrt{1 + \delta_{r,3s,\mu,l}} + \left(\delta_{r,3s,\mu,l} + \frac{\gamma^2}{1 + \gamma^2} \alpha_1 + \frac{1}{1 + \gamma^2} \alpha_2 \right) C_a}{(1 - \delta_{r,3s,\mu,l})}.$$

Appendix E. Proof of Lemma 6.1.

In this section, we will provide the proof for Lemma 6.1. By linearity of the measurement matrix \mathbf{A} , without loss of generality, it is enough to prove this lemma when $\|\mathbf{y}\|_2 = 1$. The proof mainly has two steps. First, the bounds for $\boldsymbol{\theta}_a$ and $\boldsymbol{\theta}$ are derived and a finite set of points to approximate the set $MS_{r,T,\mu,l}$ to any accuracy in norm 2 sense can be found. Then, the concentration inequality can be applied through a union bound. This is a common approach in compressive sensing literature (Baraniuk et al., 2008; Tanner & Vary, 2020).

To derive the upper bounds for $\boldsymbol{\theta}_a$ and $\boldsymbol{\theta}$, we first derive the upper bounds for the signal $\mathbf{m} = \mathbf{B}\boldsymbol{\theta}$ and $\mathbf{a} = \mathbf{B}_a\boldsymbol{\theta}_a$, which are given in the following lemma.

Lemma 6.2. *The smooth signal component \mathbf{m} and sparse signal component \mathbf{a} of the signal \mathbf{y} in $MS_{r,T,\mu,l}$ with $\mu < \frac{n}{2rsl}$ can be bounded as follows*

$$\|\mathbf{m}\|_2 = \|\mathbf{B}\boldsymbol{\theta}\|_2 \leq \frac{\|\mathbf{y}\|_2}{\sqrt{1 - \eta^2}}, \quad (19)$$

$$\|\mathbf{a}\|_2 = \|\mathbf{B}_a\boldsymbol{\theta}_a\|_2 \leq \frac{\|\mathbf{y}\|_2}{\sqrt{1 - \eta^2}}, \quad (20)$$

where $\eta = \sqrt{\frac{\mu rsl}{n}}$. ■

The proof is presented in Appendix F.

According to the RIP for \mathbf{B}_a , we have

$$\sqrt{(1 - \delta_{\mathbf{B}_a, s})} \|\boldsymbol{\theta}_a\|_2 \leq \|\mathbf{B}_a \boldsymbol{\theta}_a\|_2 \leq \sqrt{(1 + \delta_{\mathbf{B}_a, s})} \|\boldsymbol{\theta}_a\|_2. \quad (21)$$

Combine Eq. (20) and Eq. (21), we have

$$\|\boldsymbol{\theta}_a\|_2 \leq \frac{1}{\sqrt{(1 - \delta_{\mathbf{B}_a, s})}} \|\mathbf{B}_a \boldsymbol{\theta}_a\|_2 \leq \frac{\|\mathbf{y}\|_2}{\sqrt{(1 - \delta_{\mathbf{B}_a, s})(1 - \eta^2)}} = \frac{1}{\sqrt{(1 - \delta_{\mathbf{B}_a, s})(1 - \eta^2)}}.$$

Denote $\tau_0 = \frac{1}{\sqrt{(1 - \delta_{\mathbf{B}_a, s})(1 - \eta^2)}}$, we have $\|\boldsymbol{\theta}_a\|_2 \leq \tau_0$. Recall that $\mathbf{y} = \mathbf{B}\boldsymbol{\theta} + \mathbf{B}_a \boldsymbol{\theta}_a$, we have $\boldsymbol{\theta} =$

$\mathbf{B}^\dagger(\mathbf{y} - \mathbf{B}_a \boldsymbol{\theta}_a)$, where $\mathbf{B}^\dagger = (\mathbf{B}^\top \mathbf{B})^{-1} \mathbf{B}^\top$. Therefore, according to triangle inequality and Cauchy-Schwarz inequality, we have

$$\|\boldsymbol{\theta}\|_2 \leq \|\mathbf{B}^\dagger\|_2 (\|\mathbf{y}\|_2 + \|\mathbf{B}_a \boldsymbol{\theta}_a\|_2) \leq \|\mathbf{B}^\dagger\|_2 \left(1 + \frac{1}{\sqrt{1 - \eta^2}}\right).$$

Denote $\tau_1 = \|\mathbf{B}^\dagger\|_2 \left(1 + \frac{1}{\sqrt{1 - \eta^2}}\right)$, we have $\|\boldsymbol{\theta}\|_2 \leq \tau_1$.

Since we have derived the bounds for $\boldsymbol{\theta}_a$ and $\boldsymbol{\theta}$, the covering number of $MS_{r, \mathbf{T}, \mu, l}$ is given by the following lemma whose proof is in Appendix G.

Lemma 6.3. *There exists a set $\mathbf{Q} \in MS_{r, \mathbf{T}, \mu, l}$, such that for all $\mathbf{y} \in MS_{r, \mathbf{T}, \mu, l}$, with $\|\mathbf{y}\|_2 = 1$ we have $\min_{\mathbf{q} \in \mathbf{Q}} \|\mathbf{q} - \mathbf{y}\|_2 \leq \frac{\delta}{4}$, and $|\mathbf{Q}| \leq \left(\frac{24}{\delta} \tau_1\right)^r \left(\frac{24}{\delta} \tau_0\right)^s$, where $|\mathbf{Q}|$ is its cardinality. \blacksquare*

Next, we will prove the main result by applying the concentration inequality (Definition 3.5. (ii)) with union bound. Let $\epsilon = \delta/2$,

$$\left(1 - \frac{\delta}{2}\right) \|\mathbf{q}\|_2^2 \leq \|\mathbf{A}\mathbf{q}\|_2^2 \leq \left(1 + \frac{\delta}{2}\right) \|\mathbf{q}\|_2^2 \quad \forall \mathbf{q} \in \mathbf{Q}, \quad (22)$$

with probability greater than $1 - 2|\mathbf{Q}|e^{-pc_0(\delta/2)}$.

Since $\delta \in (0, 1)$, we have $1 - \frac{\delta}{2} \leq \sqrt{1 - \frac{\delta}{2}}$ and $\sqrt{1 + \frac{\delta}{2}} \leq 1 + \frac{\delta}{2}$. Then Eq. (22) can be written as

$$\left(1 - \frac{\delta}{2}\right) \|\mathbf{q}\|_2 \leq \|\mathbf{A}\mathbf{q}\|_2 \leq \left(1 + \frac{\delta}{2}\right) \|\mathbf{q}\|_2 \quad \forall \mathbf{q} \in \mathbf{Q}, \quad (23)$$

with probability greater than $1 - 2|\mathbf{Q}|e^{-pc_0(\delta/2)}$.

By the triangle inequality, we have

$$\|\mathbf{A}\mathbf{y}\|_2 \leq \|\mathbf{A}(\mathbf{y} - \mathbf{q})\|_2 + \|\mathbf{A}\mathbf{q}\|_2. \quad (24)$$

Define

$$U = \max_{\mathbf{y} \in MS_{r, \mathbf{T}, \mu, l}, \|\mathbf{y}\|_2 = 1} \|\mathbf{A}\mathbf{y}\|_2, \quad (25)$$

which is attainable since $MS_{r, \mathbf{T}, \mu, l}$ is closed.

Combine Eq. (23) and Eq. (24), we have $\forall \mathbf{y} \in MS_{r, \mathbf{T}, \mu, l}$, with $\|\mathbf{y}\|_2 = 1$, there exist a $\mathbf{q} \in MS_{r, \mathbf{T}, \mu, l}$, such that $\|\mathbf{A}\mathbf{y}\|_2 \leq \|\mathbf{A}(\mathbf{y} - \mathbf{q})\|_2 + \left(1 + \frac{\delta}{2}\right) \|\mathbf{q}\|_2$, with probability greater than $1 - 2|\mathbf{Q}|e^{-pc_0(\delta/2)}$.

Since $\mathbf{q} - \mathbf{y} \in MS_{r, \mathbf{T}, \mu, l}$, if $\mathbf{q} - \mathbf{y} = \mathbf{0}$, we have $\|\mathbf{A}\mathbf{y}\|_2 \leq \left(1 + \frac{\delta}{2}\right) \|\mathbf{q}\|_2 = 1 + \frac{\delta}{2}$.

If $\mathbf{q} - \mathbf{y} \neq \mathbf{0}$, we have $\|\mathbf{A}\mathbf{y}\|_2 \leq \left\| \mathbf{A} \frac{(\mathbf{q}-\mathbf{y})}{\|\mathbf{q}-\mathbf{y}\|_2} \right\|_2 \|\mathbf{q} - \mathbf{y}\|_2 + \left(1 + \frac{\delta}{2}\right) \|\mathbf{q}\|_2 \leq \frac{\delta}{4}U + 1 + \frac{\delta}{2}$.

Notice that the second inequality comes from Eq.(25) combined with Q being a $\frac{\delta}{4}$ covering of $MS_{r,T,\mu,l}$. In summary, we have $\|\mathbf{A}\mathbf{y}\|_2 \leq \frac{\delta}{4}U + 1 + \frac{\delta}{2}$.

Since U is attainable, according to Eq. (25), we have $U \leq \frac{\delta}{4}U + 1 + \frac{\delta}{2}$. Consequently, we have $U \leq 1 + \frac{3}{4-\delta}\delta \leq 1 + \delta$, since $\delta < 1$. Therefore, $\|\mathbf{A}\mathbf{y}\|_2 \leq 1 + \delta$, with probability greater than $1 - 2|\mathbf{Q}|e^{-pc_0(\delta/2)}$.

Similarly, we can prove that $\|\mathbf{A}\mathbf{y}\|_2 \geq 1 - \delta$ with probability greater than $1 - 2|\mathbf{Q}|e^{-pc_0(\delta/2)}$.

Finally, according to Lemma 6.3, we have that

$$1 - 2|\mathbf{Q}|e^{-pc_0(\frac{\delta}{2})} \geq 1 - 2\left(\frac{24}{\delta}\tau_1\right)^r \left(\frac{24}{\delta}\tau_0\right)^s e^{-pc_0(\frac{\delta}{2})}.$$

This finishes the proof. ■

Appendix F. Proof of Lemma 6.2.

To prove the result, we first derive a nontrivial upper bound for the inner produce between \mathbf{m} and \mathbf{a} .

Let $\mathbf{B} = \mathbf{U}\mathbf{\Sigma}\mathbf{V}^T$ be the reduced SVD of \mathbf{B} , then

$$\begin{aligned} |\mathbf{m}^T \mathbf{a}| &= |\boldsymbol{\theta}^T \mathbf{B}^T \mathbf{a}| = |\boldsymbol{\theta}^T \mathbf{V} \mathbf{\Sigma} \mathbf{U}^T \mathbf{a}| = \left| \boldsymbol{\theta}^T \mathbf{V} \mathbf{\Sigma} \mathbf{U}^T \sum_i^n a_i \mathbf{e}_i \right| = \left| \boldsymbol{\theta}^T \mathbf{V} \mathbf{\Sigma} \sum_i^n a_i \mathbf{U}^T \mathbf{e}_i \right| \\ &\leq \|\boldsymbol{\theta}^T \mathbf{V} \mathbf{\Sigma}\|_2 \left\| \sum_i^n a_i \mathbf{U}^T \mathbf{e}_i \right\|_2 \\ &\leq \|\boldsymbol{\theta}^T \mathbf{V} \mathbf{\Sigma}\|_2 \sum_i^n |a_i| \|\mathbf{U}^T \mathbf{e}_i\|_2 \\ &\leq \|\boldsymbol{\theta}^T \mathbf{V} \mathbf{\Sigma}\|_2 \sum_i^n |a_i| \max_{j \in \{1, \dots, r\}} \|\mathbf{U}^T \mathbf{e}_j\|_2 \\ &\leq_{(a)} \|\boldsymbol{\theta}^T \mathbf{V} \mathbf{\Sigma} \mathbf{U}^T\|_2 \|\mathbf{a}\|_1 \sqrt{\frac{\mu r}{n}} \\ &\leq_{(b)} \sqrt{\frac{\mu r s l}{n}} \|\mathbf{m}\|_2 \|\mathbf{a}\|_2. \end{aligned} \tag{26}$$

where inequality (a) follows $\|\boldsymbol{\theta}^T \mathbf{V} \mathbf{\Sigma}\|_2 = \|\boldsymbol{\theta}^T \mathbf{V} \mathbf{\Sigma} \mathbf{U}^T\|_2$ since $\mathbf{U}^T \mathbf{U} = \mathbf{I}$ and Definition 2.2. Inequality (b)

follows from $\|\mathbf{a}\|_1 \leq \sqrt{ls} \|\mathbf{a}\|_2$.

Let $\eta = \sqrt{\frac{\mu r s l}{n}}$, since $\mu < \frac{n}{r s l}$, we have $\eta < 1$ and

$$|\mathbf{m}^T \mathbf{a}| = \frac{|\|\mathbf{y}\|_2^2 - \|\mathbf{m}\|_2^2 - \|\mathbf{a}\|_2^2|}{2} \leq \eta \|\mathbf{m}\|_2 \|\mathbf{a}\|_2.$$

Therefore, we have

$$\|\mathbf{m}\|_2^2 + \|\mathbf{a}\|_2^2 - \|\mathbf{y}\|_2^2 \leq 2\eta\|\mathbf{m}\|_2\|\mathbf{a}\|_2.$$

By completing the square, we have

$$(\|\mathbf{m}\|_2 + \eta\|\mathbf{a}\|_2)^2 + (1 - \eta^2)\|\mathbf{a}\|_2^2 - \|\mathbf{y}\|_2^2 \leq 0.$$

Since $(\|\mathbf{m}\|_2 + \eta\|\mathbf{a}\|_2)^2 \geq 0$, we have

$$\|\mathbf{a}\|_2 \leq \frac{1}{\sqrt{(1 - \eta^2)}}\|\mathbf{y}\|_2.$$

Similarly, we can derive that

$$\|\mathbf{m}\|_2 \leq \frac{1}{\sqrt{(1 - \eta^2)}}\|\mathbf{y}\|_2.$$

This finishes the proof. ■

Appendix G. Proof of Lemma 6.3.

We first state results for the covering number of a set (Vershynin, 2018): The covering number of a smallest ϵ -net for a unit l_2 norm ball in d dimensional space is $(3/\epsilon)^d$.

Let $\mathbf{M} = \{\mathbf{m} \in \mathbb{R}^n | \mathbf{m} = \mathbf{B}\boldsymbol{\theta}, \boldsymbol{\theta} \in \mathbb{R}^r, \|\boldsymbol{\theta}\|_2 \leq \tau_1, \mu(\mathbf{B}) = \mu\}$ and $\mathbf{S} = \{\mathbf{a} \in \mathbb{R}^n | \mathbf{a} = \mathbf{B}_a\boldsymbol{\theta}_a, \boldsymbol{\theta}_a \in \mathbf{T}, \|\boldsymbol{\theta}_a\|_2 \leq \tau_0, l(\mathbf{B}_a) = l\}$. There exists two finite $\frac{\delta}{8}$ covering sets of \mathbf{M} and \mathbf{S} , which are $\mathbf{Q}_\mathbf{M} \subseteq \mathbf{M}$ and $\mathbf{Q}_\mathbf{S} \subseteq \mathbf{S}$.

For all $\mathbf{q}_\mathbf{M} \in \mathbf{Q}_\mathbf{M}$, and for all $\mathbf{m} \in \mathbf{M}$, we have,

$$\min_{\mathbf{q}_\mathbf{M} \in \mathbf{Q}_\mathbf{M}} \|\mathbf{m} - \mathbf{q}_\mathbf{M}\|_2 \leq \frac{\delta}{8};$$

For all $\mathbf{q}_\mathbf{S} \in \mathbf{Q}_\mathbf{S}$, and for all $\mathbf{a} \in \mathbf{S}$, we have,

$$\min_{\mathbf{q}_\mathbf{S} \in \mathbf{Q}_\mathbf{S}} \|\mathbf{a} - \mathbf{q}_\mathbf{S}\|_2 \leq \frac{\delta}{8}.$$

According to (i) and (ii), we have $|\mathbf{Q}_\mathbf{M}| \leq \left(\frac{24}{\delta}\tau_1\right)^r$ and $|\mathbf{Q}_\mathbf{S}| \leq \left(\frac{24}{\delta}\tau_0\right)^s$.

Define $\mathbf{Q}_{\mathbf{MS}} = \{\mathbf{q}_\mathbf{M} + \mathbf{q}_\mathbf{S} | \mathbf{q}_\mathbf{M} \in \mathbf{Q}_\mathbf{M}, \mathbf{q}_\mathbf{S} \in \mathbf{Q}_\mathbf{S}\} \subseteq MS_{r,\mathbf{T},\mu,l}$. Then $\forall \mathbf{y} \in MS_{r,\mathbf{T},\mu,l}$, there exists a pair $\mathbf{q}_{\mathbf{MS}} = \mathbf{q}_\mathbf{M} + \mathbf{q}_\mathbf{S} \in MS_{r,\mathbf{T},\mu,l}$, such that

$$\|\mathbf{q}_{\mathbf{MS}} - \mathbf{y}\|_2 = \|\mathbf{q}_\mathbf{M} - \mathbf{m} + \mathbf{q}_\mathbf{S} - \mathbf{a}\|_2 \leq \|\mathbf{q}_\mathbf{M} - \mathbf{m}\|_2 + \|\mathbf{q}_\mathbf{S} - \mathbf{a}\|_2 \leq \frac{\delta}{4}.$$

Therefore, $\mathbf{Q}_{\mathbf{MS}}$ is a $\delta/4$ covering of $MS_{r,\mathbf{T},\mu,l}$ and $|\mathbf{Q}_{\mathbf{MS}}| \leq \left(\frac{24}{\delta}\tau_1\right)^r \left(\frac{24}{\delta}\tau_0\right)^s$. This finishes the proof. ■

References

- Baraniuk, R., Davenport, M., DeVore, R., & Wakin, M. (2008). A simple proof of the restricted isometry property for random matrices. *Constructive Approximation*, 28(3), 253-263.
- Candes, E. J., Romberg, J. K., & Tao, T. (2006). Stable signal recovery from incomplete and inaccurate measurements. *Communications on Pure and Applied Mathematics: A Journal Issued by the Courant Institute of Mathematical Sciences*, 59(8), 1207-1223.

- Candes, E. J., & Tao, T. (2005). Decoding by linear programming. *IEEE Transactions on Information Theory*, 51(12), 4203-4215.
- Tanner, J., & Vary, S. (2020). Compressed sensing of low-rank plus sparse matrices. *arXiv preprint arXiv:2007.09457*.
- Vershynin, R. (2018). High-dimensional probability: An introduction with applications in data science (Vol. 47). Cambridge University Press, Cambridge.