

Supplementary Material of Tensor Robust Principal Component Analysis: Exact Recovery of Corrupted Low-Rank Tensors via Convex Optimization

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1 Structure of This Document

This document gives the detailed proof of Theorem 3.1 in the manuscript. Section 2 gives some other notations and properties which will be used in the proofs. Section 3 provides a way for the construction of the solution to the TRPCA problem, and Section 4 proves that the constructed solution is exact to the TRPCA problem. Section 5 gives the proofs of some lemmas which are used in Section 4.

Note that several parts of our proofs are quite different from RPCA [1]. For example, the proofs of Lemma 4.5 and many lemmas in Section 3. Our proofs further consider the properties of Fourier transformation and block circulant matrix.

2 Notations

Beyond the notations introduced in the paper, we need some other notations used in this document. The tensor spectral (or operator) norm of \mathcal{A} is defined as $\|\mathcal{A}\| = \|\bar{\mathcal{A}}\|$. The operator norm of an operator on tensor is defined as $\|\mathcal{L}\| = \sup_{\|\mathcal{A}\|_F=1} \|\mathcal{L}(\mathcal{A})\|_F$. The inner product of tensors has the following property

$$\langle \mathcal{A}, \mathcal{B} \rangle = \frac{1}{n_3} \langle \bar{\mathcal{A}}, \bar{\mathcal{B}} \rangle. \quad (1)$$

In this document, we define $\mathbf{e}_{ijk} = \mathbf{e}_i * \mathbf{e}_k * \mathbf{e}_j^*$ and the projection

$$\mathcal{P}_\Omega(\mathcal{Z}) = \sum_{ijk} \delta_{ijk} \mathcal{Z}_{ijk} \mathbf{e}_{ijk},$$

where $\delta_{ijk} = 1_{(i,j,k) \in \Omega}$, where $1_{(\cdot)}$ is the indicator function. Also Ω^c denotes the complement of Ω and $\mathcal{P}_{\Omega^\perp}$ is the projection onto Ω^c . Denote \mathcal{T} by the set

$$\mathcal{T} = \{\mathcal{U} * \mathcal{Y}^* + \mathcal{W} * \mathcal{V}^*, \mathcal{Y}, \mathcal{W} \in \mathbb{R}^{n \times r \times n_3}\}, \quad (2)$$

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and by \mathcal{T}^\perp its orthogonal complement. Then the projections onto \mathcal{T} and \mathcal{T}^\perp are respectively

$$\mathcal{P}_\mathcal{T}(\mathcal{Z}) = \mathcal{U} * \mathcal{U}^* * \mathcal{Z} + \mathcal{Z} * \mathcal{V} * \mathcal{V}^* - \mathcal{U} * \mathcal{U}^* * \mathcal{Z} * \mathcal{V} * \mathcal{V}^*,$$

$$\begin{aligned} \mathcal{P}_{\mathcal{T}^\perp}(\mathcal{Z}) &= \mathcal{Z} - \mathcal{P}_\mathcal{T}(\mathcal{Z}) \\ &= (\mathcal{I}_{n_1} - \mathcal{U} * \mathcal{U}^*) * \mathcal{Z} * (\mathcal{I}_{n_2} - \mathcal{V} * \mathcal{V}^*), \end{aligned}$$

where \mathcal{I}_n denotes the $n \times n \times n_3$ identity tensor. Note that $\mathcal{P}_\mathcal{T}$ is self-adjoint. So we have

$$\begin{aligned} &\|\mathcal{P}_\mathcal{T}(\mathbf{e}_{ijk})\|_F^2 \\ &= \langle \mathcal{P}_\mathcal{T}(\mathbf{e}_{ijk}), \mathbf{e}_{ijk} \rangle \\ &= \langle \mathcal{U} * \mathcal{U}^* * \mathbf{e}_{ijk} + \mathbf{e}_{ijk} * \mathcal{V} * \mathcal{V}^* - \mathcal{U} * \mathcal{U}^* * \mathbf{e}_{ijk} * \mathcal{V} * \mathcal{V}^*, \mathbf{e}_{ijk} \rangle. \end{aligned}$$

Note that

$$\begin{aligned} &\langle \mathcal{U} * \mathcal{U}^* * \mathbf{e}_{ijk}, \mathbf{e}_{ijk} \rangle \\ &= \langle \mathcal{U} * \mathcal{U}^* * \mathbf{e}_i * \mathbf{e}_k * \mathbf{e}_j^*, \mathbf{e}_i * \mathbf{e}_k * \mathbf{e}_j^* \rangle \\ &= \langle \mathcal{U}^* * \mathbf{e}_i, \mathcal{U} * \mathbf{e}_i * (\mathbf{e}_k * \mathbf{e}_j^* * \mathbf{e}_j * \mathbf{e}_k^*) \rangle \\ &= \langle \mathcal{U}^* * \mathbf{e}_i, \mathcal{U} * \mathbf{e}_i \rangle \\ &= \|\mathcal{U}^* * \mathbf{e}_i\|_F^2, \end{aligned}$$

where we use the fact that $\mathbf{e}_k * \mathbf{e}_j^* * \mathbf{e}_j * \mathbf{e}_k^* = \mathcal{I}_1$, which is the $1 \times 1 \times n_3$ identity tensor. Therefore, it is easy to see that

$$\begin{aligned} &\|\mathcal{P}_\mathcal{T}(\mathbf{e}_{ijk})\|_F^2 \\ &= \|\mathcal{U}^* * \mathbf{e}_i\|_F^2 + \|\mathcal{V}^* * \mathbf{e}_j\|_F^2 - \|\mathcal{U}^* * \mathbf{e}_i * \mathbf{e}_k * \mathbf{e}_j^* * \mathcal{V}\|_F^2 \\ &\leq \|\mathcal{U}^* * \mathbf{e}_i\|_F^2 + \|\mathcal{V}^* * \mathbf{e}_j\|_F^2 \\ &\leq \frac{\mu r (n_1 + n_2)}{n_1 n_2 n_3} \end{aligned} \quad (3)$$

$$= \frac{2\mu r}{nn_3}, \text{ when } n_1 = n_2 = n. \quad (4)$$

The following Tensor Incoherence Conditions will be used in the proofs

$$\max_{i=1, \dots, n_1} \|\mathcal{U}^* * \mathfrak{e}_i\|_F \leq \sqrt{\frac{\mu r}{n_1 n_3}}, \quad (5)$$

$$\max_{j=1, \dots, n_2} \|\mathcal{V}^* * \mathfrak{e}_j\|_F \leq \sqrt{\frac{\mu r}{n_2 n_3}}, \quad (6)$$

$$\|\mathcal{U} * \mathcal{V}^*\|_\infty \leq \sqrt{\frac{\mu r}{n_1 n_2 n_3^2}}. \quad (7)$$

3 Dual Certification

Theorem 3.1. (Subgradient of tensor nuclear norm) Let $\mathcal{A} \in \mathbb{R}^{n_1 \times n_2 \times n_3}$ with $\text{rank}_t(\mathcal{A}) = r$ and its skinny t -SVD be $\mathcal{A} = \mathcal{U} * \mathcal{S} * \mathcal{V}^*$. Denote $\partial \|\mathcal{A}\|_*$ be the set of the subgradients of the tensor nuclear norm at \mathcal{A} . We have $\mathcal{U} * \mathcal{V}^* + \mathcal{W} \in \partial \|\mathcal{A}\|_*$, where \mathcal{W} satisfies $\mathcal{U}^* * \mathcal{W} = \mathbf{0}$, $\mathcal{W} * \mathcal{V} = \mathbf{0}$ and $\|\mathcal{W}\| \leq 1$.

Proof. Let $\mathcal{G} \in \partial \|\mathcal{A}\|_*$. It is equivalent to the following statements [4]

$$\|\mathcal{A}\|_* = \langle \mathcal{G}, \mathcal{A} \rangle, \quad (8)$$

$$\|\mathcal{G}\| \leq 1. \quad (9)$$

So, to complete the proof, we only need to show that $\mathcal{G} = \mathcal{U} * \mathcal{V}^* + \mathcal{W}$, where $\mathcal{U}^* * \mathcal{W} = \mathbf{0}$, $\mathcal{W} * \mathcal{V} = \mathbf{0}$ and $\|\mathcal{W}\| \leq 1$, satisfies (8) and (9). First, we have

$$\begin{aligned} \langle \mathcal{G}, \mathcal{A} \rangle &= \langle \mathcal{U} * \mathcal{V}^* + \mathcal{W}, \mathcal{U} * \mathcal{S} * \mathcal{V}^* \rangle \\ &= \langle \mathcal{I}, \mathcal{S} \rangle + 0 = \frac{1}{n_3} \langle \bar{\mathcal{I}}, \bar{\mathcal{S}} \rangle \\ &= \frac{1}{n_3} \|\bar{\mathcal{A}}\|_* = \|\mathcal{A}\|_*. \end{aligned}$$

Also, (9) is obvious when considering the property of \mathcal{W} . \square

3.1 Dual Certificates

Lemma 3.2. Assume that $\|\mathcal{P}_\Omega \mathcal{P}_T\| < 1$. Then $(\mathcal{L}_0, \mathcal{S}_0)$ is the unique solution to the TRPCA problem if there is a pair (\mathcal{W}, F) obeying

$$\mathcal{U} * \mathcal{V}^* + \mathcal{W} = \lambda(\text{sgn}(\mathcal{S}_0) + F),$$

with $\mathcal{P}_T \mathcal{W} = \mathbf{0}$, $\|\mathcal{W}\| < 1$, $\mathcal{P}_\Omega F = \mathbf{0}$ and $\|F\|_\infty < 1$.

Lemma 3.3. Assume that $\|\mathcal{P}_\Omega \mathcal{P}_T\| \leq \frac{1}{2}$ and $\lambda < \frac{1}{\sqrt{n_3}}$. Then $(\mathcal{L}_0, \mathcal{S}_0)$ is the unique solution to the TRPCA problem if there is a pair (\mathcal{W}, F) obeying

$$(\mathcal{U} * \mathcal{V}^* + \mathcal{W}) = \lambda(\text{sgn}(\mathcal{S}_0) + F + \mathcal{P}_\Omega \mathcal{D}),$$

with $\mathcal{P}_T \mathcal{W} = \mathbf{0}$, $\|\mathcal{W}\| \leq \frac{1}{2}$, $\mathcal{P}_\Omega F = \mathbf{0}$ and $\|F\|_\infty \leq \frac{1}{2}$, and $\|\mathcal{P}_\Omega \mathcal{D}\|_F \leq \frac{1}{4}$.

Lemma 3.3 implies that it suffices to produce a dual certificate \mathcal{W} obeying

$$\begin{cases} \mathcal{W} \in T^\perp, \\ \|\mathcal{W}\| < \frac{1}{2}, \\ \|\mathcal{P}_\Omega(\mathcal{U} * \mathcal{V}^* + \mathcal{W} - \lambda \text{sgn}(\mathcal{S}_0))\|_F \leq \frac{\lambda}{4}, \\ \|\mathcal{P}_{\Omega^\perp}(\mathcal{U} * \mathcal{V}^* + \mathcal{W})\|_\infty < \frac{\lambda}{2}. \end{cases} \quad (10)$$

3.2 Dual Certification via the Golfing Scheme

Before we introduce our construction, our model assumes that $\Omega \sim \text{Ber}(\rho)$, or equivalently that $\Omega^c \sim \text{Ber}(1 - \rho)$. Now the distribution of Ω^c is the same as that of $\Omega^c = \Omega_1 \cup \Omega_2 \cup \dots \cup \Omega_{j_0}$, where each Ω_j follows the Bernoulli model with parameter q , which satisfies

$$\mathbb{P}((i, j, k) \in \Omega) = \mathbb{P}(\text{Bin}(j_0, q) = 0) = (1 - q)^{j_0}, \quad (11)$$

so that the two models are the same if $\rho = (1 - q)^{j_0}$. Note that because of overlaps between the Ω_j 's, $q \geq (1 - \rho)/j_0$.

Now, we construct a dual certificate

$$\mathcal{W} = \mathcal{W}^{\mathcal{L}} + \mathcal{W}^{\mathcal{S}}, \quad (12)$$

where each component is as follows:

1. Construction of $\mathcal{W}^{\mathcal{L}}$ via the Golfing Scheme. Let $j_0 = 2 \log(nn_3)$ and Ω_j , $j = 1, \dots, j_0$, be defined as previously described so that $\Omega^c = \cup_{1 \leq j \leq j_0} \Omega_j$. Then define

$$\mathcal{W}^{\mathcal{L}} = \mathcal{P}_{T^\perp} \mathcal{Y}_{j_0}, \quad (13)$$

where

$$\mathcal{Y}_j = \mathcal{Y}_{j-1} + q^{-1} \mathcal{P}_{\Omega_j} \mathcal{P}_T (\mathcal{U} * \mathcal{V}^* - \mathcal{Y}_{j-1}), \quad \mathcal{Y}_0 = \mathbf{0}.$$

2. Construction of $\mathcal{W}^{\mathcal{S}}$ via the Method of Least Squares. Assume that $\|\mathcal{P}_\Omega \mathcal{P}_T\| < 1/2$. Then, $\|\mathcal{P}_\Omega \mathcal{P}_T \mathcal{P}_\Omega\| < 1/4$, and thus, the operator $\mathcal{P}_\Omega - \mathcal{P}_\Omega \mathcal{P}_T \mathcal{P}_\Omega$ mapping Ω on to itself is invertible; we denote its inverse by $(\mathcal{P}_\Omega - \mathcal{P}_\Omega \mathcal{P}_T \mathcal{P}_\Omega)^{-1}$. We then set

$$\mathcal{W}^{\mathcal{S}} = \lambda \mathcal{P}_{T^\perp} (\mathcal{P}_\Omega - \mathcal{P}_\Omega \mathcal{P}_T \mathcal{P}_\Omega)^{-1} \text{sgn}(\mathcal{S}_0). \quad (14)$$

This is equivalent to

$$\mathcal{W}^{\mathcal{S}} = \lambda \mathcal{P}_{T^\perp} \sum_{k \geq 0} (\mathcal{P}_\Omega \mathcal{P}_T \mathcal{P}_\Omega)^k \text{sgn}(\mathcal{S}_0). \quad (15)$$

Since both $\mathcal{W}^{\mathcal{L}}$ and $\mathcal{W}^{\mathcal{S}}$ belong to T^\perp and $\mathcal{P}_\Omega \mathcal{W}^{\mathcal{S}} = \lambda \mathcal{P}_\Omega (\mathcal{I} - \mathcal{P}_T) (\mathcal{P}_\Omega - \mathcal{P}_\Omega \mathcal{P}_T \mathcal{P}_\Omega)^{-1} \text{sgn}(\mathcal{S}_0) = \lambda \text{sgn}(\mathcal{S}_0)$, we will establish that $\mathcal{W}^{\mathcal{L}} + \mathcal{W}^{\mathcal{S}}$ is a valid dual certificate if it obeys

$$\begin{cases} \|\mathcal{W}^{\mathcal{L}} + \mathcal{W}^{\mathcal{S}}\| < \frac{1}{2}, \\ \|\mathcal{P}_\Omega(\mathcal{U} * \mathcal{V}^* + \mathcal{W}^{\mathcal{L}})\|_F \leq \frac{\lambda}{4}, \\ \|\mathcal{P}_{\Omega^\perp}(\mathcal{U} * \mathcal{V}^* + \mathcal{W}^{\mathcal{L}} + \mathcal{W}^{\mathcal{S}})\|_\infty < \frac{\lambda}{2}. \end{cases} \quad (16)$$

This can be done by using the following two key lemmas:

Lemma 3.4. Assume that $\Omega \sim \text{Ber}(\rho)$ with parameter $\rho \leq \rho_s$ for some $\rho_s > 0$. Set $j_0 = 2\lceil \log(nn_3) \rceil$ (use $\log(n_{(1)}n_3)$ for the tensors of rectangular frontal slice). Then, the tensor $\mathcal{W}^{\mathcal{L}}$ obeys

- (a) $\|\mathcal{W}^{\mathcal{L}}\| < \frac{1}{4}$,
- (b) $\|\mathcal{P}_\Omega(\mathcal{U} * \mathcal{V}^* + \mathcal{W}^{\mathcal{L}})\|_F < \frac{\lambda}{4}$,
- (c) $\|\mathcal{P}_{\Omega^\perp}(\mathcal{U} * \mathcal{V}^* + \mathcal{W}^{\mathcal{L}})\|_\infty < \frac{\lambda}{4}$.

Lemma 3.5. Assume that \mathcal{S}_0 is supported on a set Ω sampled as in Lemma 3.4, and that the signs of \mathcal{S}_0 are independent and identically distributed symmetric (and independent of Ω). Then, the tensor $\mathcal{W}^{\mathcal{S}}$ (14) obeys

- (a) $\|\mathcal{W}^{\mathcal{S}}\| < \frac{1}{4}$,
- (b) $\|\mathcal{P}_{\Omega^\perp} \mathcal{W}^{\mathcal{S}}\|_\infty < \frac{\lambda}{4}$.

4 Proofs of Dual Certification

4.1 Preliminaries

Lemma 4.1. Suppose $\Omega \sim \text{Ber}(\rho)$. Then with high probability,

$$\|\mathcal{P}_T - \rho^{-1} \mathcal{P}_T \mathcal{P}_\Omega \mathcal{P}_T\| \leq \epsilon, \quad (17)$$

provided that $\rho \geq C_0 \epsilon^{-2} (\mu r \log(nn_3)) / (nn_3)$ for some numerical constant $C_0 > 0$. For the tensor of rectangular frontal slice, we need $\rho \geq C_0 \epsilon^{-2} (\mu r \log(n_{(1)}n_3)) / (n_{(2)}n_3)$.

Corollary 4.2. Assume that $\Omega \sim \text{Ber}(\rho)$, then $\|\mathcal{P}_\Omega \mathcal{P}_T\|^2 \leq \rho + \epsilon$, provided that $1 - \rho \geq C \epsilon^{-2} (\mu r \log(nn_3)) / (nn_3)$, where C is as in Lemma 4.1. For the tensor with frontal slice, the modification is as in Lemma 4.1.

Note that this corollary shows that $\|\mathcal{P}_\Omega \mathcal{P}_T\| \leq 1/2$, provided $|\Omega|$ is not too large.

Lemma 4.3. Suppose that $\mathcal{Z} \in \mathcal{T}$ is a fixed tensor, and $\Omega \sim \text{Ber}(\rho)$. Then, with high probability,

$$\|\mathcal{Z} - \rho^{-1} \mathcal{P}_T \mathcal{P}_\Omega \mathcal{Z}\|_\infty \leq \epsilon \|\mathcal{Z}\|_\infty, \quad (18)$$

provided that $\rho \geq C_0 \epsilon^{-2} (\mu r \log(nn_3)) / (nn_3)$ (for the tensor of rectangular frontal slice, $\rho \geq C_0 \epsilon^{-2} (\mu r \log(n_{(1)}n_3)) / (n_{(2)}n_3)$) for some numerical constant $C_0 > 0$.

Lemma 4.4. Suppose \mathcal{Z} is fixed, and $\Omega \sim \text{Ber}(\rho)$. Then, with high probability,

$$\|(\mathcal{I} - \rho^{-1} \mathcal{P}_\Omega) \mathcal{Z}\| \leq \sqrt{\frac{C_0 nn_3 \log(nn_3)}{\rho}} \|\mathcal{Z}\|_\infty, \quad (19)$$

for some numerical constant $C_0 > 0$ provided that $\rho \geq C_0 \log(nn_3) / (nn_3)$ (or $\rho \geq C_0 \log(n_{(1)}n_3) / (n_{(2)}n_3)$ for the tensors with rectangular frontal slice).

4.2 Proof of Lemma 3.4

Proof. We first introduce some notations. Set $\mathcal{Z}_j = \mathcal{U} * \mathcal{V}^* - \mathcal{P}_T \mathcal{Y}_j$ obeying

$$\mathcal{Z}_j = (\mathcal{P}_T - q^{-1} \mathcal{P}_T \mathcal{P}_{\Omega_j} \mathcal{P}_T) \mathcal{Z}_{j-1}.$$

So $\mathcal{Z}_j \in \mathcal{T}$ for all $j \geq 0$. Also, note that when

$$q \geq C_0 \epsilon^{-2} \frac{\mu r \log(nn_3)}{nn_3}, \quad (20)$$

or for the tensors with rectangular frontal slices $q \geq C_0 \epsilon^{-2} \frac{\mu r \log(n_{(1)}n_3)}{n_{(2)}n_3}$, we have

$$\|\mathcal{Z}_j\|_\infty \leq \epsilon \|\mathcal{Z}_{j-1}\|_\infty \leq \epsilon^j \|\mathcal{U} * \mathcal{V}^*\|_\infty, \quad (21)$$

by Lemma 4.3 and

$$\|\mathcal{Z}_j\|_F \leq \epsilon \|\mathcal{Z}_{j-1}\|_F \leq \epsilon^j \|\mathcal{U} * \mathcal{V}^*\|_F \leq \epsilon^j \sqrt{r}. \quad (22)$$

We assume $\epsilon \leq e^{-1}$.

1. Proof of (a). Note that $\mathcal{Y}_{j_0} = \sum_j q^{-1} \mathcal{P}_{\Omega_j} \mathcal{Z}_{j-1}$. We have

$$\begin{aligned} \|\mathcal{W}^{\mathcal{L}}\| &= \|\mathcal{P}_{T^\perp} \mathcal{Y}_{j_0}\| \leq \sum_j \|q^{-1} \mathcal{P}_{T^\perp} \mathcal{P}_{\Omega_j} \mathcal{Z}_{j-1}\| \\ &\leq \sum_j \|\mathcal{P}_{T^\perp} (q^{-1} \mathcal{P}_{\Omega_j} \mathcal{Z}_{j-1} - \mathcal{Z}_{j-1})\| \\ &\leq \sum_j \|q^{-1} \mathcal{P}_{\Omega_j} \mathcal{Z}_{j-1} - \mathcal{Z}_{j-1}\| \\ &\leq C'_0 \sqrt{\frac{nn_3 \log(nn_3)}{q}} \sum_j \|\mathcal{Z}_{j-1}\|_\infty \\ &\leq C'_0 \sqrt{\frac{nn_3 \log(nn_3)}{q}} \sum_j \epsilon^{j-1} \|\mathcal{U} * \mathcal{V}^*\|_\infty \\ &\leq C'_0 (1 - \epsilon)^{-1} \sqrt{\frac{nn_3 \log(nn_3)}{q}} \|\mathcal{U} * \mathcal{V}^*\|_\infty. \end{aligned}$$

The fourth step is from Lemma 4.4 and the fifth is from (21). Now by using (20) and (7), we have

$$\|\mathcal{W}^{\mathcal{L}}\| \leq C' \epsilon,$$

for some numerical constant C' .

2. Proof of (b). Since $\mathcal{P}_\Omega \mathcal{Y}_{j_0} = 0$, $\mathcal{P}_\Omega(\mathcal{U} * \mathcal{V}^* + \mathcal{P}_{T^\perp} \mathcal{Y}_{j_0}) = \mathcal{P}_\Omega(\mathcal{U} * \mathcal{V}^* - \mathcal{P}_T \mathcal{Y}_{j_0}) = \mathcal{P}_\Omega(\mathcal{Z}_{j_0})$, and it follows from (22) that

$$\|\mathcal{Z}_{j_0}\|_F \leq \epsilon^{j_0} \|\mathcal{U} * \mathcal{V}^*\|_F \leq \epsilon^{j_0} \sqrt{r}.$$

Since $\epsilon \leq e^{-1}$ and $j_0 \geq 2 \log(nn_3)$, $\epsilon^{j_0} \leq (nn_3)^{-2}$ and this proves the claim.

3. Proof of (c). We have $\mathcal{U} * \mathcal{V}^* + \mathcal{W}^{\mathcal{L}} = \mathcal{Z}_{j_0} + \mathcal{Y}_{j_0}$ and know that \mathcal{Y}_{j_0} is supported on Ω^c . Therefore, since $\|\mathcal{Z}_{j_0}\|_F \leq \lambda/8$.

We only need to show that $\|\mathcal{Y}_{j_0}\|_\infty \leq \lambda/8$. Indeed,

$$\begin{aligned}\|\mathcal{Y}_{j_0}\|_\infty &\leq q^{-1} \sum_j \|\mathcal{P}_{\Omega_j} \mathcal{Z}_{j-1}\|_\infty \\ &\leq q^{-1} \sum_j \|\mathcal{Z}_{j-1}\|_\infty \\ &\leq q^{-1} \sum_j \epsilon^{j-1} \|\mathcal{U} * \mathcal{V}^*\|_\infty.\end{aligned}$$

Since $\|\mathcal{U} * \mathcal{V}^*\|_\infty \leq \sqrt{\frac{\mu r}{n^2 n_3^2}}$, this gives

$$\|\mathcal{Y}_{j_0}\|_\infty \leq C' \frac{\epsilon^2}{\sqrt{\mu r (\log(nn_3))^2}},$$

for some numerical constant C' whenever q obeys (20). Since $\lambda = 1/\sqrt{nn_3}$, $\|\mathcal{Y}_{j_0}\|_\infty \leq \lambda/8$ if

$$\epsilon \leq C \left(\frac{\mu r (\log(nn_3))^2}{nn_3} \right)^{1/4}.$$

The claim is proved by using (20), (7) and sufficiently small ϵ (provided that ρ_r is sufficiently small. Note that everything is consistent since $C_0 \epsilon^{-2} \frac{\mu r \log(nn_3)}{nn_3} < 1$. \square

4.3 Proof of Lemma 3.5

Proof. We denote $\mathcal{M} = \text{sgn}(\mathcal{S}_0)$ distributed as

$$\mathcal{M}_{ijk} = \begin{cases} 1, & \text{w.p. } \rho/2, \\ 0, & \text{w.p. } 1 - \rho, \\ -1, & \text{w.p. } \rho/2. \end{cases} \quad (23)$$

Note that for any $\sigma > 0$, $\{\|\mathcal{P}_\Omega \mathcal{P}_T\| \leq \sigma\}$ holds with high probability provided that ρ is sufficiently small, see Corollary 4.2.

1. Proof of (a). By construction,

$$\begin{aligned}\mathcal{W}^{\mathcal{S}} &= \lambda \mathcal{P}_{T^\perp} \mathcal{M} + \lambda \mathcal{P}_{T^\perp} \sum_{k \geq 1} (\mathcal{P}_\Omega \mathcal{P}_T \mathcal{P}_\Omega)^k \mathcal{M} \\ &:= \mathcal{P}_{T^\perp} \mathcal{W}_0^{\mathcal{S}} + \mathcal{P}_{T^\perp} \mathcal{W}_1^{\mathcal{S}}.\end{aligned}$$

Note that $\|\mathcal{P}_{T^\perp} \mathcal{W}_0^{\mathcal{S}}\| \leq \|\mathcal{W}_0^{\mathcal{S}}\| = \lambda \|\mathcal{M}\|$ and $\|\mathcal{P}_{T^\perp} \mathcal{W}_1^{\mathcal{S}}\| \leq \|\mathcal{W}_1^{\mathcal{S}}\| = \lambda \|\mathcal{R}(\mathcal{M})\|$, where $\mathcal{R} = \sum_{k \geq 1} (\mathcal{P}_\Omega \mathcal{P}_T \mathcal{P}_\Omega)^k$. Now, we will respectively show that $\lambda \|\mathcal{M}\|$ and $\lambda \|\mathcal{R}(\mathcal{M})\|$ are small enough when ρ is sufficiently small for $\lambda = 1/\sqrt{nn_3}$. Therefore, $\|\mathcal{W}^{\mathcal{S}}\| \leq 1/4$.

1) Bound $\|\mathcal{M}\|$.

Lemma 4.5. *For the Bernoulli sign variable \mathcal{M} defined in (23), there exists a function $\varphi(\rho)$ satisfying $\lim_{\rho \rightarrow 0^+} \varphi(\rho) = 0$, such that the following statement holds with with large probability,*

$$\|\mathcal{M}\| \leq \varphi(\rho) \sqrt{nn_3}. \quad (24)$$

The proof has three steps.

Step 1: Approximation. We first introduce some notations.

Let \mathbf{f}_i^* be the i -th row of \mathbf{F}_{n_3} , and $\mathbf{M}^H = \begin{bmatrix} \mathbf{M}_1^H \\ \mathbf{M}_2^H \\ \vdots \\ \mathbf{M}_n^H \end{bmatrix} \in \mathbb{R}^{n n_3 \times n}$

be a matrix unfolded by \mathcal{M} , where $\mathbf{M}_i^H \in \mathbb{R}^{n_3 \times n}$ is the i -th horizontal slice of \mathcal{M} , i.e., $[\mathbf{M}_i^H]_{kj} = \mathcal{M}_{ikj}$. Consider that $\bar{\mathcal{M}} = \text{fft}(\mathcal{M}, [], 3)$, we have

$$\bar{\mathbf{M}}_i = \begin{bmatrix} \mathbf{f}_i^* \mathbf{M}_1^H \\ \mathbf{f}_i^* \mathbf{M}_2^H \\ \vdots \\ \mathbf{f}_i^* \mathbf{M}_n^H \end{bmatrix}, \quad (25)$$

where $\bar{\mathbf{M}}_i \in \mathbb{R}^{n \times n}$ is the i -th frontal slice of $\bar{\mathcal{M}}$. Note that

$$\|\mathcal{M}\| = \|\bar{\mathcal{M}}\| = \max_{i=1, \dots, n_3} \|\bar{\mathbf{M}}_i\|. \quad (26)$$

Let N be the $1/2$ -net for \mathbb{S}^{n-1} of size at most 5^n (see Lemma 5.2 in [3]). Then Lemma 5.3 in [3] gives

$$\|\bar{\mathbf{M}}_i\| \leq 2 \max_{\mathbf{x} \in N} \|\bar{\mathbf{M}}_i \mathbf{x}\|_2. \quad (27)$$

So we consider to bound $\|\bar{\mathbf{M}}_i \mathbf{x}\|_2$.

Step 2: Concentration. We can express $\|\bar{\mathbf{M}}_i \mathbf{x}\|_2^2$ as a sum of independent random variables

$$\|\bar{\mathbf{M}}_i \mathbf{x}\|_2^2 = \sum_{j=1}^n (\mathbf{f}_i^* \mathbf{M}_j^H \mathbf{x})^2 := \sum_{j=1}^n z_j^2, \quad (28)$$

where $z_j = \langle \mathbf{M}_j^H, \mathbf{f}_i \mathbf{x}^* \rangle$, $j = 1, \dots, n$, are independent sub-gaussian random variables with $\mathbb{E} z_j^2 = \rho \|\mathbf{f}_i \mathbf{x}^*\|_F^2 = \rho n_3$. Using (23), we have

$$|[\mathbf{M}_j^H]_{kl}| = \begin{cases} 1, & \text{w.p. } \rho, \\ 0, & \text{w.p. } 1 - \rho. \end{cases}$$

Thus, the sub-gaussian norm of $[\mathbf{M}_j^H]_{kl}$, denoted as $\|\cdot\|_{\psi_2}$, is

$$\begin{aligned}\|[\mathbf{M}_j^H]_{kl}\|_{\psi_2} &= \sup_{p \geq 1} p^{-\frac{1}{2}} (\mathbb{E} |[\mathbf{M}_j^H]_{kl}|^p)^{\frac{1}{p}} \\ &= \sup_{p \geq 1} p^{-\frac{1}{2}} \rho^{\frac{1}{p}}.\end{aligned}$$

Define the function $\phi(x) = x^{-\frac{1}{2}} \rho^{\frac{1}{x}}$ on $[1, +\infty)$. The only stationary point occurs at $x^* = \log \rho^{-2}$. Thus,

$$\begin{aligned}\phi(x) &\leq \max(\phi(1), \phi(x^*)) \\ &= \max\left(\rho, (\log \rho^{-2})^{-\frac{1}{2}} \rho^{\frac{1}{\log \rho^{-2}}}\right) \\ &:= \psi(\rho).\end{aligned} \quad (29)$$

Therefore, $\|[\mathbf{M}_j^H]_{kl}\|_{\psi_2} \leq \psi(\rho)$. Consider that z_j is a sum of independent centered sub-gaussian random variables $[\mathbf{M}_j^H]_{kl}$'s, by using Lemma 5.9 in [3], we have $\|z_j\|_{\psi_2}^2 \leq c_1(\psi(\rho))^2 n_3$, where c_1 is an absolute constant. Therefore, by Remark 5.18 and Lemma 5.14 in [3], $z_j^2 - \rho n_3$ are independent centered sub-exponential random variables with $\|z_j^2 - \rho n_3\|_{\psi_1} \leq 2\|z_j\|_{\psi_2}^2 \leq 4\|z_j\|_{\psi_2}^2 \leq 4c_1(\psi(\rho))^2 n_3$.

Now, we use an exponential deviation inequality, Corollary 5.17 in [3], to control the sum (28). We have

$$\begin{aligned} & \mathbb{P}(|\|\bar{\mathbf{M}}_i \mathbf{x}\|_2^2 - \rho n n_3| \geq tn) \\ &= \mathbb{P}\left(\left|\sum_{j=1}^n (z_j^2 - \rho n_3)\right| \geq tn\right) \\ &\leq 2 \exp\left(-c_2 n \min\left(\left(\frac{t}{4c_1(\psi(\rho))^2 n_3}\right)^2, \frac{t}{4c_1(\psi(\rho))^2 n_3}\right)\right), \end{aligned}$$

where $c_2 > 0$. Let $t = c_3(\psi(\rho))^2 n_3$ for some absolute constant c_3 , we have

$$\begin{aligned} & \mathbb{P}(|\|\bar{\mathbf{M}}_i \mathbf{x}\|_2^2 - \rho n n_3| \geq c_3(\psi(\rho))^2 n n_3) \\ &\leq 2 \exp\left(-c_2 n \min\left(\left(\frac{c_3}{4c_1}\right)^2, \frac{c_3}{4c_1}\right)\right). \end{aligned}$$

Step 3 Union bound. Taking the union bound over all \mathbf{x} in the net N of cardinality $|N| \leq 5^n$, we obtain

$$\begin{aligned} & \mathbb{P}\left(\left|\max_{\mathbf{x} \in N} \|\bar{\mathbf{M}}_i \mathbf{x}\|_2^2 - \rho n n_3\right| \geq c_3(\psi(\rho))^2 n n_3\right) \\ &\leq 2 \cdot 5^n \cdot \exp\left(-c_2 n \min\left(\left(\frac{c_3}{4c_1}\right)^2, \frac{c_3}{4c_1}\right)\right). \end{aligned}$$

Furthermore, taking the union bound over all $i = 1, \dots, n_3$, we have

$$\begin{aligned} & \mathbb{P}\left(\max_i \left|\max_{\mathbf{x} \in N} \|\bar{\mathbf{M}}_i \mathbf{x}\|_2^2 - \rho n n_3\right| \geq c_3(\psi(\rho))^2 n n_3\right) \\ &\leq 2 \cdot 5^n \cdot n_3 \cdot \exp\left(-c_2 n \min\left(\left(\frac{c_3}{4c_1}\right)^2, \frac{c_3}{4c_1}\right)\right). \end{aligned}$$

This implies that, with high probability (when the constant c_3 is large enough),

$$\max_i \max_{\mathbf{x} \in N} \|\bar{\mathbf{M}}_i \mathbf{x}\|_2^2 \leq (\rho + c_3(\psi(\rho))^2) n n_3. \quad (30)$$

Let $\varphi(\rho) = 2\sqrt{\rho + c_3(\psi(\rho))^2}$ and it satisfies $\lim_{\rho \rightarrow 0^+} \varphi(\rho) = 0$ by using (29). The proof is completed by further combining (26), (27) and (30).

2) Bound $\|\mathcal{R}(\mathcal{M})\|$.

For simplicity, let $\mathcal{Z} = \mathcal{R}(\mathcal{M})$. We have

$$\|\mathcal{Z}\| = \|\bar{\mathcal{Z}}\| = \sup_{\mathbf{x} \in \mathbb{S}^{n n_3 - 1}} \|\bar{\mathcal{Z}} \mathbf{x}\|_2. \quad (31)$$

The optimal \mathbf{x} to (31) is an eigenvector of $\bar{\mathcal{Z}}^* \bar{\mathcal{Z}}$. Since $\bar{\mathcal{Z}}$ is a block diagonal matrix, the optimal \mathbf{x} has a block sparse structure, i.e., $\mathbf{x} \in B = \{\mathbf{x} \in \mathbb{R}^{n n_3} | \mathbf{x} = [\mathbf{x}_1^\top, \dots, \mathbf{x}_i^\top \cdots, \mathbf{x}_{n_3}^\top]^\top, \text{ with } \mathbf{x}_i \in \mathbb{R}^n, \text{ and there exists } j \text{ such that } \mathbf{x}_j \neq \mathbf{0} \text{ and } \mathbf{x}_i = \mathbf{0}, i \neq j\}$. Note that $\|\mathbf{x}\|_2 = \|\mathbf{x}_j\|_2 = 1$. Let N be the $1/2$ -net for \mathbb{S}^{n-1} of size at most 5^n (see Lemma 5.2 in [3]). Then the $1/2$ -net, denoted as N' , for B has the size at most $n_3 \cdot 5^n$. We have

$$\begin{aligned} \|\mathcal{R}(\mathcal{M})\| &= \|\text{bdiag}(\overline{\mathcal{R}(\mathcal{M})})\| \\ &= \sup_{\mathbf{x}, \mathbf{y} \in B} \langle \mathbf{x}, \text{bdiag}(\overline{\mathcal{R}(\mathcal{M})}) \mathbf{y} \rangle \\ &= \sup_{\mathbf{x}, \mathbf{y} \in B} \langle \mathbf{x} \mathbf{y}^*, \text{bdiag}(\overline{\mathcal{R}(\mathcal{M})}) \rangle \\ &= \sup_{\mathbf{x}, \mathbf{y} \in B} \langle \text{bdiag}^*(\mathbf{x} \mathbf{y}^*), \overline{\mathcal{R}(\mathcal{M})} \rangle, \end{aligned}$$

where bdiag^* , the joint operator of bdiag , maps the block diagonal matrix $\mathbf{x} \mathbf{y}^*$ to a tensor of size $n \times n \times n_3$. Let $\mathcal{Z}' = \text{bdiag}^*(\mathbf{x} \mathbf{y}^*)$ and $\mathcal{Z} = \text{ifft}(\mathcal{Z}', \cdot, 3)$. We have

$$\begin{aligned} \|\mathcal{R}(\mathcal{M})\| &= \sup_{\mathbf{x}, \mathbf{y} \in B} \langle \mathcal{Z}', \overline{\mathcal{R}(\mathcal{M})} \rangle \\ &= \sup_{\mathbf{x}, \mathbf{y} \in B} n_3 \langle \mathcal{Z}, \mathcal{R}(\mathcal{M}) \rangle \\ &= \sup_{\mathbf{x}, \mathbf{y} \in B} n_3 \langle \mathcal{R}(\mathcal{Z}), \mathcal{M} \rangle \\ &\leq \sup_{\mathbf{x}, \mathbf{y} \in N'} 4n_3 \langle \mathcal{R}(\mathcal{Z}), \mathcal{M} \rangle. \end{aligned}$$

For a fixed pair (\mathbf{x}, \mathbf{y}) of unit-normed vectors, define the random variable

$$X(\mathbf{x}, \mathbf{y}) = \langle 4n_3 \mathcal{R}(\mathcal{Z}), \mathcal{M} \rangle.$$

Conditional on $\Omega = \text{supp}(\mathcal{M})$, the signs of \mathcal{M} are independent and identically distributed symmetric and Hoeffding's inequality gives

$$\mathbb{P}(|X(\mathbf{x}, \mathbf{y})| > t | \Omega) \leq 2 \exp\left(\frac{-2t^2}{\|4n_3 \mathcal{R}(\mathcal{Z})\|_F^2}\right).$$

Note that $\|4n_3 \mathcal{R}(\mathcal{Z})\|_F \leq 4n_3 \|\mathcal{R}\| \|\mathcal{Z}\|_F = 4\sqrt{n_3} \|\mathcal{R}\| \|\mathcal{Z}'\|_F = 4\sqrt{n_3} \|\mathcal{R}\|$. Therefore, we have

$$\mathbb{P}\left(\sup_{\mathbf{x}, \mathbf{y} \in N'} |X(\mathbf{x}, \mathbf{y})| > t | \Omega\right) \leq 2|N'|^2 \exp\left(-\frac{t^2}{8n_3 \|\mathcal{R}\|^2}\right).$$

Hence,

$$\mathbb{P}(\|\mathcal{R}(\mathcal{M})\| > t | \Omega) \leq 2|N'|^2 \exp\left(-\frac{t^2}{8n_3 \|\mathcal{R}\|^2}\right).$$

On the event $\{\|\mathcal{P}_\Omega \mathcal{P}_T\| \leq \sigma\}$,

$$\|\mathcal{R}\| \leq \sum_{k \geq 1} \sigma^{2k} = \frac{\sigma^2}{1 - \sigma^2},$$

and, therefore, unconditionally,

$$\begin{aligned} & \mathbb{P}(\|\mathcal{R}(\mathcal{M})\| > t) \\ & \leq 2|N'|^2 \exp\left(-\frac{\gamma^2 t^2}{8n_3}\right) + \mathbb{P}(\|\mathcal{P}_\Omega \mathcal{P}_T\| \geq \sigma), \quad \gamma = \frac{1 - \sigma^2}{2\sigma^2} \\ & = 2n_3^2 \cdot 5^{2n} \exp\left(-\frac{\gamma^2 t^2}{8n_3}\right) + \mathbb{P}(\|\mathcal{P}_\Omega \mathcal{P}_T\| \geq \sigma). \end{aligned}$$

Let $t = c\sqrt{nn_3}$, where c can be a small absolute constant. Then the above inequality implies that $\|\mathcal{R}(\mathcal{M})\| \leq t$ with high probability.

2. Proof of (b) Observe that

$$\mathcal{P}_{\Omega^\perp} \mathcal{W}^{\mathcal{S}} = -\lambda \mathcal{P}_{\Omega^\perp} \mathcal{P}_T (\mathcal{P}_\Omega - \mathcal{P}_\Omega \mathcal{P}_T \mathcal{P}_\Omega)^{-1} \mathcal{M}.$$

Now for $(i, j, k) \in \Omega^c$, $\mathcal{W}_{ijk}^{\mathcal{S}} = \langle \mathcal{W}^{\mathcal{S}}, \mathbf{e}_{ijk} \rangle$, and we have $\mathcal{W}_{ijk}^{\mathcal{S}} = \lambda \langle \mathcal{Q}(i, j, k), \mathcal{M} \rangle$, where $\mathcal{Q}(i, j, k)$ is the tensor $-(\mathcal{P}_\Omega - \mathcal{P}_\Omega \mathcal{P}_T \mathcal{P}_\Omega)^{-1} \mathcal{P}_\Omega \mathcal{P}_T (\mathbf{e}_{ijk})$. Conditional on $\Omega = \text{supp}(\mathcal{M})$, the signs of \mathcal{M} are independent and identically distributed symmetric, and the Hoeffding's inequality gives

$$\mathbb{P}(|\mathcal{W}_{ijk}^{\mathcal{S}}| > t\lambda|\Omega) \leq 2 \exp\left(-\frac{2t^2}{\|\mathcal{Q}(i, j, k)\|_F^2}\right),$$

and

$$\begin{aligned} & \mathbb{P}(\sup_{i,j,k} |\mathcal{W}_{ijk}^{\mathcal{S}}| > t\lambda/n_3|\Omega) \\ & \leq 2n^2 n_3 \exp\left(-\frac{2t^2}{\sup_{i,j,k} \|\mathcal{Q}(i, j, k)\|_F^2}\right). \end{aligned}$$

By using (4), we have

$$\begin{aligned} \|\mathcal{P}_\Omega \mathcal{P}_T (\mathbf{e}_{ijk})\|_F & \leq \|\mathcal{P}_\Omega \mathcal{P}_T\| \|\mathcal{P}_T (\mathbf{e}_{ijk})\|_F \\ & \leq \sigma \sqrt{\frac{2\mu r}{nn_3}}, \end{aligned}$$

on the event $\{\|\mathcal{P}_\Omega \mathcal{P}_T\| \leq \sigma\}$. On the same event, we have $\|(\mathcal{P}_\Omega - \mathcal{P}_\Omega \mathcal{P}_T \mathcal{P}_\Omega)^{-1}\| \leq (1 - \sigma^2)^{-1}$ and thus $\|\mathcal{Q}(i, j, k)\|_F^2 \leq \frac{2\sigma^2}{(1 - \sigma^2)^2} \frac{\mu r}{nn_3}$. Then, unconditionally,

$$\begin{aligned} & \mathbb{P}\left(\sup_{i,j,k} |\mathcal{W}_{ijk}^{\mathcal{S}}| > t\lambda\right) \\ & \leq 2n^2 n_3 \exp\left(-\frac{nn_3 \gamma^2 t^2}{\mu r}\right) + \mathbb{P}(\|\mathcal{P}_\Omega \mathcal{P}_T\| \geq \sigma), \end{aligned}$$

where $\gamma = \frac{(1 - \sigma^2)^2}{2\sigma^2}$. This proves the claim when $\mu r < \rho_r' nn_3 \log(nn_3)^{-1}$ and ρ_r' is sufficiently small. \square

5 Proofs of Some Lemmas

Lemma 5.1. [2] Consider a finite sequence $\{\mathbf{Z}_k\}$ of independent, random $n_1 \times n_2$ matrices that satisfy the assumption $\mathbb{E}\mathbf{Z}_k = \mathbf{0}$ and $\|\mathbf{Z}_k\| \leq R$ almost surely. Let $\sigma^2 =$

$\max\{\|\sum_k \mathbb{E}[\mathbf{Z}_k \mathbf{Z}_k^*]\|, \max\{\|\sum_k \mathbb{E}[\mathbf{Z}_k^* \mathbf{Z}_k]\|\}$. Then, for any $t \geq 0$, we have

$$\begin{aligned} \mathbb{P}\left[\left\|\sum_k \mathbf{Z}_k\right\| \geq t\right] & \leq (n_1 + n_2) \exp\left(-\frac{t^2}{2\sigma^2 + \frac{2}{3}Rt}\right) \\ & \leq (n_1 + n_2) \exp\left(-\frac{3t^2}{8\sigma^2}\right), \text{ for } t \leq \frac{\sigma^2}{R}. \end{aligned} \quad (32)$$

Or, for any $c > 0$, we have

$$\left\|\sum_k \mathbf{Z}_k\right\| \geq 2\sqrt{c\sigma^2 \log(n_1 + n_2)} + cB \log(n_1 + n_2), \quad (34)$$

with probability at least $1 - (n_1 + n_2)^{1-c}$.

5.1 Proof of Lemma 3.2

Proof. For any $\mathcal{H} \neq \mathbf{0}$, $(\mathcal{L}_0 + \mathcal{H}, \mathcal{S}_0 - \mathcal{H})$ is also a feasible solution. We show that its objective is larger than that at $(\mathcal{L}_0, \mathcal{S}_0)$, hence proving that $(\mathcal{L}_0, \mathcal{S}_0)$ is the unique solution. To do this, let $\mathcal{U} * \mathcal{V}^* + \mathcal{W}_0$ be an arbitrary subgradient of the tensor nuclear norm at \mathcal{L}_0 , and $\text{sgn}(\mathcal{S}_0) + \mathcal{F}_0$ be an arbitrary subgradient of the ℓ_1 -norm at \mathcal{S}_0 . Then we have

$$\begin{aligned} & \|\mathcal{L}_0 + \mathcal{H}\|_* + \lambda \|\mathcal{S}_0 - \mathcal{H}\|_1 \\ & \geq \|\mathcal{L}_0\|_* + \lambda \|\mathcal{S}_0\|_1 + \langle \mathcal{U} * \mathcal{V}^* + \mathcal{W}_0, \mathcal{H} \rangle \\ & \quad - \lambda \langle \text{sgn}(\mathcal{S}_0) + \mathcal{F}_0, \mathcal{H} \rangle. \end{aligned}$$

Now pick \mathcal{W}_0 such that $\langle \mathcal{W}_0, \mathcal{H} \rangle = \|\mathcal{P}_{T^\perp} \mathcal{H}\|_*$ and $\langle \mathcal{F}_0, \mathcal{H} \rangle = -\|\mathcal{P}_{\Omega^\perp} \mathcal{H}\|$. We have

$$\begin{aligned} & \|\mathcal{L}_0 + \mathcal{H}\|_* + \lambda \|\mathcal{S}_0 - \mathcal{H}\|_1 \\ & \geq \|\mathcal{L}_0\|_* + \lambda \|\mathcal{S}_0\|_1 + \|\mathcal{P}_{T^\perp} \mathcal{H}\|_* + \lambda \|\mathcal{P}_{\Omega^\perp} \mathcal{H}\|_1 \\ & \quad + \langle \mathcal{U} * \mathcal{V}^* - \lambda \text{sgn}(\mathcal{S}_0), \mathcal{H} \rangle. \end{aligned}$$

By assumption

$$\begin{aligned} & |\langle \mathcal{U} * \mathcal{V}^* - \lambda \text{sgn}(\mathcal{S}_0), \mathcal{H} \rangle| \\ & \leq |\langle \mathcal{W}, \mathcal{H} \rangle| + \lambda |\langle \mathcal{F}, \mathcal{H} \rangle| \\ & \leq \beta (\|\mathcal{P}_{T^\perp} \mathcal{H}\|_* + \lambda \|\mathcal{P}_{\Omega^\perp} \mathcal{H}\|_1), \end{aligned}$$

where $\beta = \max(\|\mathcal{W}\|, \|\mathcal{F}\|_\infty) < 1$. Thus

$$\begin{aligned} & \|\mathcal{L}_0 + \mathcal{H}\|_* + \lambda \|\mathcal{S}_0 - \mathcal{H}\|_1 \\ & \geq \|\mathcal{L}_0\|_* + \lambda \|\mathcal{S}_0\|_1 + (1 - \beta) (\|\mathcal{P}_{T^\perp} \mathcal{H}\|_* + \lambda \|\mathcal{P}_{\Omega^\perp} \mathcal{H}\|_1). \end{aligned}$$

Note that $\|\mathcal{P}_\Omega \mathcal{P}_T\| < 1$. This is equivalent to $\Omega \cap T = \{\mathbf{0}\}$. Thus $\|\mathcal{P}_{T^\perp} \mathcal{H}\|_* + \lambda \|\mathcal{P}_{\Omega^\perp} \mathcal{H}\|_1 > 0$ unless $\mathcal{H} = \mathbf{0}$. \square

5.2 Proof of Lemma 3.3

Proof. Following the proof of Lemma 3.2, we have

$$\begin{aligned}
& \|\mathcal{L}_0 + \mathcal{H}\|_* + \lambda \|\mathcal{S}_0 - \mathcal{H}\|_1 \\
& \geq \|\mathcal{L}_0\|_* + \lambda \|\mathcal{S}_0\|_1 + \frac{1}{2} (\|\mathcal{P}_{T^\perp} \mathcal{H}\|_* + \lambda \|\mathcal{P}_{\Omega^\perp} \mathcal{H}\|_1) \\
& \quad - \lambda \langle \mathcal{P}_\Omega \mathcal{D}, \mathcal{H} \rangle \\
& \geq \|\mathcal{L}_0\|_* + \lambda \|\mathcal{S}_0\|_1 + \frac{1}{2} (\|\mathcal{P}_{T^\perp} \mathcal{H}\|_* + \lambda \|\mathcal{P}_{\Omega^\perp} \mathcal{H}\|_1) \\
& \quad - \frac{\lambda}{4} \|\mathcal{P}_\Omega \mathcal{H}\|_F.
\end{aligned}$$

On the other hand,

$$\begin{aligned}
\|\mathcal{P}_\Omega \mathcal{H}\|_F & \leq \|\mathcal{P}_\Omega \mathcal{P}_T \mathcal{H}\|_F + \|\mathcal{P}_\Omega \mathcal{P}_{T^\perp} \mathcal{H}\|_F \\
& = \frac{1}{2} \|\mathcal{H}\|_F + \|\mathcal{P}_{T^\perp} \mathcal{H}\|_F \\
& \leq \frac{1}{2} \|\mathcal{P}_\Omega \mathcal{H}\|_F + \frac{1}{2} \|\mathcal{P}_{\Omega^\perp} \mathcal{H}\|_F + \|\mathcal{P}_{T^\perp} \mathcal{H}\|_F.
\end{aligned}$$

Thus

$$\begin{aligned}
\|\mathcal{P}_\Omega \mathcal{H}\|_F & \leq \|\mathcal{P}_{\Omega^\perp} \mathcal{H}\|_F + 2 \|\mathcal{P}_{T^\perp} \mathcal{H}\|_F \\
& \leq \|\mathcal{P}_{\Omega^\perp} \mathcal{H}\|_1 + 2\sqrt{n_3} \|\mathcal{P}_{T^\perp} \mathcal{H}\|_*.
\end{aligned}$$

In conclusion,

$$\begin{aligned}
& \|\mathcal{L}_0 + \mathcal{H}\|_* + \lambda \|\mathcal{S}_0 - \mathcal{H}\|_1 \\
& \geq \|\mathcal{L}_0\|_* + \lambda \|\mathcal{S}_0\|_1 + \frac{1}{2} (1 - \lambda\sqrt{n_3}) \|\mathcal{P}_{T^\perp} \mathcal{H}\|_* \\
& \quad + \frac{\lambda}{4} \|\mathcal{P}_{\Omega^\perp} \mathcal{H}\|_1,
\end{aligned}$$

where the last two terms are strictly positive when $\mathcal{H} \neq \mathbf{0}$. \square

5.3 Proof of Lemma 4.1

Proof. For any tensor \mathcal{Z} , we can write

$$\begin{aligned}
& (\rho^{-1} \mathcal{P}_T \mathcal{P}_\Omega \mathcal{P}_T - \mathcal{P}_T) \mathcal{Z} \\
& = \sum_{ijk} (\rho^{-1} \delta_{ijk} - 1) \langle \mathbf{e}_{ijk}, \mathcal{P}_T \mathcal{Z} \rangle \mathcal{P}_T(\mathbf{e}_{ijk}) \\
& := \sum_{ijk} \mathcal{H}_{ijk}(\mathcal{Z})
\end{aligned}$$

where $\mathcal{H}_{ijk} : \mathbb{R}^{n \times n \times n_3} \rightarrow \mathbb{R}^{n \times n \times n_3}$ is a self-adjoint random operator with $\mathbb{E}[\mathcal{H}_{ijk}] = \mathbf{0}$. Define the matrix operator $\bar{\mathcal{H}}_{ijk} : \mathbb{B} \rightarrow \mathbb{B}$, where $\mathbb{B} = \{\bar{\mathcal{B}} : \mathcal{B} \in \mathbb{R}^{n \times n \times n_3}\}$ denotes the set consists of block diagonal matrices with the blocks as the frontal slices of $\bar{\mathcal{B}}$, as

$$\bar{\mathcal{H}}_{ijk}(\bar{\mathcal{Z}}) = (\rho^{-1} \delta_{ijk} - 1) \langle \mathbf{e}_{ijk}, \mathcal{P}_T(\mathcal{Z}) \rangle \text{bdiag}(\overline{\mathcal{P}_T(\mathbf{e}_{ijk})}).$$

By the above definitions, we have $\|\mathcal{H}_{ijk}\| = \|\bar{\mathcal{H}}_{ijk}\|$ and $\|\sum_{ijk} \mathcal{H}_{ijk}\| = \|\sum_{ijk} \bar{\mathcal{H}}_{ijk}\|$. Also $\bar{\mathcal{H}}_{ijk}$ is self-adjoint

and $\mathbb{E}[\bar{\mathcal{H}}_{ijk}] = \mathbf{0}$. To prove the result by the non-commutative Bernstein inequality, we need to bound $\|\bar{\mathcal{H}}_{ijk}\|$ and $\left\| \sum_{ijk} \mathbb{E}[\bar{\mathcal{H}}_{ijk}^2] \right\|$. First, we have

$$\begin{aligned}
\|\bar{\mathcal{H}}_{ijk}\| & = \sup_{\|\bar{\mathcal{Z}}\|_F=1} \|\bar{\mathcal{H}}_{ijk}(\bar{\mathcal{Z}})\|_F \\
& \leq \sup_{\|\bar{\mathcal{Z}}\|_F=1} \rho^{-1} \|\mathcal{P}_T(\mathbf{e}_{ijk})\|_F \|\text{bdiag}(\overline{\mathcal{P}_T(\mathbf{e}_{ijk})})\|_F \|\bar{\mathcal{Z}}\|_F \\
& = \sup_{\|\bar{\mathcal{Z}}\|_F=1} \rho^{-1} \|\mathcal{P}_T(\mathbf{e}_{ijk})\|_F^2 \|\bar{\mathcal{Z}}\|_F \\
& \leq \frac{2\mu r}{nn_3\rho},
\end{aligned}$$

where the last inequality uses (4). On the other hand, by direct computation, we have $\bar{\mathcal{H}}_{ijk}^2(\bar{\mathcal{Z}}) = (\rho^{-1} \delta_{ijk} - 1)^2 \langle \mathbf{e}_{ijk}, \mathcal{P}_T(\mathcal{Z}) \rangle \langle \mathbf{e}_{ijk}, \mathcal{P}_T(\mathbf{e}_{ijk}) \rangle \text{bdiag}(\overline{\mathcal{P}_T(\mathbf{e}_{ijk})})$. Note that $\mathbb{E}[(\rho^{-1} \delta_{ijk} - 1)^2] \leq \rho^{-1}$. We have

$$\begin{aligned}
& \left\| \sum_{ijk} \mathbb{E}[\bar{\mathcal{H}}_{ijk}^2(\bar{\mathcal{Z}})] \right\|_F \\
& \leq \rho^{-1} \left\| \sum_{ijk} \langle \mathbf{e}_{ijk}, \mathcal{P}_T(\mathcal{Z}) \rangle \langle \mathbf{e}_{ijk}, \mathcal{P}_T(\mathbf{e}_{ijk}) \rangle \text{bdiag}(\overline{\mathcal{P}_T(\mathbf{e}_{ijk})}) \right\|_F \\
& \leq \rho^{-1} \sqrt{n_3} \|\mathcal{P}_T(\mathbf{e}_{ijk})\|_F^2 \left\| \sum_{ijk} \langle \mathbf{e}_{ijk}, \mathcal{P}_T(\mathcal{Z}) \rangle \right\|_F \\
& = \rho^{-1} \sqrt{n_3} \|\mathcal{P}_T(\mathbf{e}_{ijk})\|_F^2 \|\mathcal{P}_T(\mathcal{Z})\|_F \\
& \leq \rho^{-1} \sqrt{n_3} \|\mathcal{P}_T(\mathbf{e}_{ijk})\|_F^2 \|\mathcal{Z}\|_F \\
& = \rho^{-1} \|\mathcal{P}_T(\mathbf{e}_{ijk})\|_F^2 \|\bar{\mathcal{Z}}\|_F \\
& \leq \frac{2\mu r}{nn_3\rho} \|\bar{\mathcal{Z}}\|_F.
\end{aligned}$$

This implies $\left\| \sum_{ijk} \mathbb{E}[\bar{\mathcal{H}}_{ijk}^2] \right\| \leq \frac{2\mu r}{nn_3\rho}$. Let $\epsilon \leq 1$. By Lemma 5.1, we have

$$\begin{aligned}
& \mathbb{P}[\|\rho^{-1} \mathcal{P}_T \mathcal{P}_\Omega \mathcal{P}_T - \mathcal{P}_T\| > \epsilon] \\
& = \mathbb{P}\left[\left\| \sum_{ijk} \mathcal{H}_{ijk} \right\| > \epsilon\right] \\
& = \mathbb{P}\left[\left\| \sum_{ijk} \bar{\mathcal{H}}_{ijk} \right\| > \epsilon\right] \\
& \leq 2nn_3 \exp\left(-\frac{3}{8} \cdot \frac{\epsilon^2}{2\mu r/(nn_3\rho)}\right) \\
& \leq 2(nn_3)^{1-\frac{3}{16}C_0},
\end{aligned}$$

where the last inequality uses $\rho \geq C_0 \epsilon^{-2} \mu r \log(nn_3)/(nn_3)$. Thus, $\|\rho^{-1} \mathcal{P}_T \mathcal{P}_\Omega \mathcal{P}_T - \mathcal{P}_T\| \leq \epsilon$ holds with high probability for some numerical constant C_0 . \square

5.4 Proof of Corollary 4.2

Proof. From Lemma 4.1, we have

$$\|\mathcal{P}_T - (1 - \rho)^{-1} \mathcal{P}_T \mathcal{P}_\Omega^\perp \mathcal{P}_T\| \leq \epsilon,$$

provided that $1 - \rho \geq C_0 \epsilon^{-2} (\mu r \log(nn_3))/n$. Note that $\mathcal{I} = \mathcal{P}_\Omega + \mathcal{P}_\Omega^\perp$, we have

$$\|\mathcal{P}_T - (1 - \rho)^{-1} \mathcal{P}_T \mathcal{P}_\Omega^\perp \mathcal{P}_T\| = (1 - \rho)^{-1} (\mathcal{P}_T \mathcal{P}_\Omega \mathcal{P}_T - \rho \mathcal{P}_T).$$

Then, by the triangular inequality

$$\|\mathcal{P}_T \mathcal{P}_\Omega \mathcal{P}_T\| \leq \epsilon(1 - \rho) + \rho \|\mathcal{P}_T\| = \rho + \epsilon(1 - \rho).$$

The proof is completed by using $\|\mathcal{P}_\Omega \mathcal{P}_T\|^2 = \|\mathcal{P}_T \mathcal{P}_\Omega \mathcal{P}_T\|$. \square

5.5 Proof of Lemma 4.3

Proof. For any tensor $\mathcal{Z} \in \mathcal{T}$, we write

$$\rho^{-1} \mathcal{P}_T \mathcal{P}_\Omega(\mathcal{Z}) = \sum_{ijk} \rho^{-1} \delta_{ijk} z_{ijk} \mathcal{P}_T(\mathbf{e}_{ijk}).$$

The (a, b, c) -th entry of $\rho^{-1} \mathcal{P}_T \mathcal{P}_\Omega(\mathcal{Z}) - \mathcal{Z}$ can be written as a sum of independent random variables, i.e.,

$$\begin{aligned} & \langle \rho^{-1} \mathcal{P}_T \mathcal{P}_\Omega(\mathcal{Z}) - \mathcal{Z}, \mathbf{e}_{abc} \rangle \\ &= \sum_{ijk} (\rho^{-1} \delta_{ijk} - 1) z_{ijk} \langle \mathcal{P}_T(\mathbf{e}_{ijk}), \mathbf{e}_{abc} \rangle \\ &:= \sum_{ijk} t_{ijk}, \end{aligned}$$

where t_{ijk} 's are independent and $\mathbb{E}(t_{ijk}) = 0$. Now we bound $|t_{ijk}|$ and $|\sum_{ijk} \mathbb{E}[t_{ijk}^2]|$. First

$$\begin{aligned} & |t_{ijk}| \\ & \leq \rho^{-1} \|\mathcal{Z}\|_\infty \|\mathcal{P}_T(\mathbf{e}_{ijk})\|_F \|\mathcal{P}_T(\mathbf{e}_{abc})\|_F \\ & \leq \frac{2\mu r}{nn_3 \rho} \|\mathcal{Z}\|_\infty. \end{aligned}$$

Second, we have

$$\begin{aligned} & \left| \sum_{ijk} \mathbb{E}[t_{ijk}^2] \right| \\ & \leq \rho^{-1} \|\mathcal{Z}\|_\infty^2 \sum_{ijk} \langle \mathcal{P}_T(\mathbf{e}_{ijk}), \mathbf{e}_{abc} \rangle^2 \\ & = \rho^{-1} \|\mathcal{Z}\|_\infty^2 \sum_{ijk} \langle \mathbf{e}_{ijk}, \mathcal{P}_T(\mathbf{e}_{abc}) \rangle^2 \\ & = \rho^{-1} \|\mathcal{Z}\|_\infty^2 \|\mathcal{P}_T(\mathbf{e}_{abc})\|_F^2 \\ & \leq \frac{2\mu r}{nn_3 \rho} \|\mathcal{Z}\|_\infty^2. \end{aligned}$$

Let $\epsilon \leq 1$. By Lemma 5.1, we have

$$\begin{aligned} & \mathbb{P} [|[\rho^{-1} \mathcal{P}_T \mathcal{P}_\Omega(\mathcal{Z}) - \mathcal{Z}]_{abc}| > \epsilon \|\mathcal{Z}\|_\infty] \\ &= \mathbb{P} \left[\left| \sum_{ijk} t_{ijk} \right| > \epsilon \|\mathcal{Z}\|_\infty \right] \\ & \leq 2 \exp \left(-\frac{3}{8} \cdot \frac{\epsilon^2 \|\mathcal{Z}\|_\infty^2}{2\mu r \|\mathcal{Z}\|_\infty^2 / (nn_3 \rho)} \right) \\ & \leq 2 (nn_3)^{-\frac{3}{16} C_0}, \end{aligned}$$

where the last inequality uses $\rho \geq C_0 \epsilon^{-2} \mu r \log(nn_3)/(nn_3)$. Thus, $\|\rho^{-1} \mathcal{P}_T \mathcal{P}_\Omega(\mathcal{Z}) - \mathcal{Z}\|_\infty \leq \epsilon \|\mathcal{Z}\|_\infty$ holds with high probability for some numerical constant C_0 . \square

5.6 Proof of Lemma 4.4

Proof. Denote the tensor $\mathcal{H}_{ijk} = (1 - \rho^{-1} \delta_{ijk}) z_{ijk} \mathbf{e}_{ijk}$. Then we have

$$(\mathcal{I} - \rho^{-1} \mathcal{P}_\Omega) \mathcal{Z} = \sum_{ijk} \mathcal{H}_{ijk}.$$

Note that δ_{ijk} 's are independent random scalars. Thus, \mathcal{H}_{ijk} 's are independent random tensors and $\bar{\mathcal{H}}_{ijk}$'s are independent random matrices. Observe that $\mathbb{E}[\bar{\mathcal{H}}_{ijk}] = \mathbf{0}$ and $\|\bar{\mathcal{H}}_{ijk}\| \leq \rho^{-1} \|\mathcal{Z}\|_\infty$. We have

$$\begin{aligned} & \left\| \sum_{ijk} \mathbb{E}[\bar{\mathcal{H}}_{ijk}^* \bar{\mathcal{H}}_{ijk}] \right\| \\ &= \left\| \sum_{ijk} \mathbb{E}[\mathcal{H}_{ijk}^* * \mathcal{H}_{ijk}] \right\| \\ &= \left\| \sum_{ijk} \mathbb{E}[(1 - \rho^{-1} \delta_{ijk})^2] z_{ijk}^2 (\mathbf{e}_j * \mathbf{e}_j^*) \right\| \\ &= \left\| \frac{1 - \rho}{\rho} \sum_{ijk} z_{ijk}^2 (\mathbf{e}_j * \mathbf{e}_j^*) \right\| \\ & \leq \frac{nn_3}{\rho} \|\mathcal{Z}\|_\infty^2. \end{aligned}$$

A similar calculation yields $\left\| \sum_{ijk} \mathbb{E}[\bar{\mathcal{H}}_{ijk}^* \bar{\mathcal{H}}_{ijk}] \right\| \leq \rho^{-1} nn_3 \|\mathcal{Z}\|_\infty^2$. Let $t = \sqrt{C_0 nn_3 \log(nn_3)/\rho} \|\mathcal{Z}\|_\infty$. When

$\rho \geq C_0 \log(nn_3)/(nn_3)$, we apply Lemma 5.1 and obtain

$$\begin{aligned}
& \mathbb{P} \left[\|(\mathcal{I} - \rho^{-1} \mathcal{P}_\Omega) \mathcal{Z}\| > t \right] \\
&= \mathbb{P} \left[\left\| \sum_{ijk} \mathcal{H}_{ijk} \right\| > t \right] \\
&= \mathbb{P} \left[\left\| \sum_{ijk} \bar{H}_{ijk} \right\| > t \right] \\
&\leq 2nn_3 \exp \left(-\frac{3}{8} \cdot \frac{C_0 nn_3 \log(nn_3) \|\mathcal{Z}\|_\infty^2 / \rho}{nn_3 \|\mathcal{Z}\|_\infty^2 / \rho} \right) \\
&\leq 2(nn_3)^{1-\frac{3}{8}C_0}.
\end{aligned}$$

Thus, $\|(\mathcal{I} - \rho^{-1} \mathcal{P}_\Omega) \mathcal{Z}\| > t$ holds with high probability for some numerical constant C_0 . \square

References

- [1] E. J. Candès, X. D. Li, Y. Ma, and J. Wright. Robust principal component analysis? *Journal of the ACM*, 58(3), 2011. [1](#)
- [2] J. A. Tropp. User-friendly tail bounds for sums of random matrices. *Foundations of Computational Mathematics*, 12(4):389–434, 2012. [6](#)
- [3] R. Vershynin. Introduction to the non-asymptotic analysis of random matrices. *arXiv preprint arXiv:1011.3027*, 2010. [4](#), [5](#)
- [4] G. A. Watson. Characterization of the subdifferential of some matrix norms. *Linear Algebra and its Applications*, 170:33–45, 1992. [2](#)