

matrix case:

$Y = PQ$:

P deterministic (orthonormal) matrix ,

Q i.i.d noise entry.

$Y = [S1, S2]$

$S1$: row covariance $\rightarrow P^* P^T$

$S2$: column covariance $\rightarrow I$

Thought about SupCP

$Y = [D; \text{noise}, A2, A3]$;

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$T = [D; I, A2, A3]$ — deterministic (r-by-d2-by-d3)

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$A = T \times \text{noise factor}$

$\rightarrow A[1, :] = \text{linear combination of matrix slices}$

$A[2, :] = \text{linear}$

1 SupCP covariance

Consider the observation $\mathcal{Y} \in \mathbb{R}^{d \times d \times d}$, the covariance $\mathbf{X} \in \mathbb{R}^{d \times R}$. Recall the SupCP model,

$$\mathcal{Y} = [\mathbf{A}_1, \mathbf{A}_2, \mathbf{A}_3] + \mathcal{E}, \quad \mathbf{A}_1 = \mathbf{X}\mathbf{B} + \mathcal{E}',$$

where $\mathbf{A}_k \in \mathbb{R}^{d \times R}$, $\mathbf{B} \in \mathbb{R}^{p \times R}$ is the coefficient matrix, $\mathcal{E} \in \mathbb{R}^{d \times d \times d}$ has i.i.d. entries from $N(0, \sigma_e^2)$, and $\mathcal{E}' \in \mathbb{R}^{d \times R}$ has i.i.d. rows from $N(0, \Sigma)$.

Note that

$$\text{vec}(\mathcal{Y}) = [\mathbf{X}\mathbf{B} \odot \mathbf{A}_2 \odot \mathbf{A}_3] \mathbf{1}_R + [\mathcal{E}' \odot \mathbf{A}_2 \odot \mathbf{A}_3] \mathbf{1}_R + \text{vec}(\mathcal{E}'),$$

where \odot is the column-wise Kronecker product. Since \mathcal{E}' is independent with \mathcal{E} and $\text{cov}(\text{vec}(\mathcal{E})) = \mathbf{I}_{d^3}$, we only need to calculate $\text{cov}([\mathcal{E}' \odot \mathbf{A}_2 \odot \mathbf{A}_3] \mathbf{1}_R)$. Note that

$$[\mathcal{E}' \odot \mathbf{A}_2 \odot \mathbf{A}_3] \mathbf{1}_R = \begin{bmatrix} (\mathcal{E}'_1 \odot \mathbf{A}_2 \odot \mathbf{A}_3) \mathbf{1}_R \\ \vdots \\ (\mathcal{E}'_d \odot \mathbf{A}_2 \odot \mathbf{A}_3) \mathbf{1}_R \end{bmatrix}, \quad \text{and} \quad \mathcal{E}'_i \perp \mathcal{E}'_j, i \neq j \in [d],$$

where $\mathcal{E}'_i \in \mathbb{R}^{1 \times R}$ refers to the i -th row of \mathcal{E}' . Therefore, we know that $\text{cov}([\mathcal{E}' \odot \mathbf{A}_2 \odot \mathbf{A}_3] \mathbf{1}_R)$ is block-wise diagonal with diagonal elements $\text{cov}((\mathcal{E}'_i \odot \mathbf{A}_2 \odot \mathbf{A}_3) \mathbf{1}_R), i \in [d]$. Also notice that

$$(\mathcal{E}'_i \odot \mathbf{A}_2 \odot \mathbf{A}_3) \mathbf{1}_R = \sum_{k=1}^R \mathcal{E}'_{ik} \otimes \mathbf{A}_{2k} \otimes \mathbf{A}_{3k} = [\mathbf{A}_2 \odot \mathbf{A}_3] \mathcal{E}'_i^T.$$

Therefore, we have

$$\text{cov}([\mathcal{E}'_i \odot \mathbf{A}_2 \odot \mathbf{A}_3] \mathbf{1}_R) = \text{cov}([\mathbf{A}_2 \odot \mathbf{A}_3] \mathcal{E}'_i^T) = [\mathbf{A}_2 \odot \mathbf{A}_3] \Sigma [\mathbf{A}_2 \odot \mathbf{A}_3]^T,$$

and thus the whole covariance matrix $\text{cov}(\text{vec}(\mathcal{Y}))$ is

$$\begin{aligned} \text{cov}(\text{vec}(\mathcal{Y})) &= \text{cov}(\text{vec}(\mathcal{E})) + \text{cov}([\mathcal{E}' \odot \mathbf{A}_2 \odot \mathbf{A}_3] \mathbf{1}_R) \\ &= \mathbf{I}_{d^3} + \begin{bmatrix} [\mathbf{A}_2 \odot \mathbf{A}_3] \Sigma [\mathbf{A}_2 \odot \mathbf{A}_3]^T & \mathbf{0} & \mathbf{0} \\ \mathbf{0} & \ddots & \mathbf{0} \\ \mathbf{0} & \mathbf{0} & [\mathbf{A}_2 \odot \mathbf{A}_3] \Sigma [\mathbf{A}_2 \odot \mathbf{A}_3]^T \end{bmatrix} \end{aligned}$$

2 SupCP performance

Write the second part in terms of normal tensor ensemble.

$Y = \text{MVN}(0, \Sigma_1, \Sigma_2, \Sigma_3)$, where Σ_k is mode- k

covariance. In this note, we consider the Gaussian data and all the matrices are full rank. Let $\mathcal{Y} \in \mathbb{R}^{d \times d \times d}$, $\mathbf{X}_k \in \mathbb{R}^{d \times p}$, $k \in [3]$. Consider the tucker rank $\mathbf{r} = (r, r, r)$ and CP rank R . The dimension of \mathbf{M}_k can be obtained by the context. What is Σ_k , $i=1,2,3$ in this case? Dimension? Intuition?

model 3: [B1X1, B2X2; B3X3]

vs. model 1: [A1, A2, A3].

provide {A1,A2,A3}~{in distribution same} [B1(X1^TX1)^{-1}X1,].

2.1 Without supervision

A1: d-by-r matrix, orthonormal column emsemble

B1X1: d-by-r matrix, orthonormal column emsemble (provided $p > r$)

Recall the STD and SupCP models without the supervision.

$$\begin{aligned} STD &: \mathcal{Y} = \mathcal{C} \times \{\mathbf{M}_1, \mathbf{M}_2, \mathbf{M}_3\} + \mathcal{E} \\ SupCP &: \mathcal{Y} = \llbracket \mathbf{A}_1, \mathbf{A}_2, \mathbf{A}_3 \rrbracket + \mathcal{E}. \end{aligned}$$

The fitted value $\text{vec}(\mathcal{Y})$ for STD and SupCP lie in

$$\begin{aligned} \mathcal{P}_{STD} &= \{C(\mathbf{M}_1 \otimes \mathbf{M}_2 \otimes \mathbf{M}_3) \mid \mathbf{M}_k \in \mathbb{R}^{d \times r}, \mathbf{M}_k^T \mathbf{M}_k = \mathbf{I}_r\}, \\ \mathcal{P}_{SupCP} &= \{C(\mathbf{A}_1 \odot \mathbf{A}_2 \odot \mathbf{A}_3) \mid \mathbf{A}_k \in \mathbb{R}^{d \times R}\}, \end{aligned}$$

respectively, where $C(\mathbf{X})$ refers to the column space of the matrix \mathbf{X} . Note that $\text{rank}(\mathbf{M}_1 \otimes \mathbf{M}_2 \otimes \mathbf{M}_3) = r^3$ and $\text{rank}(\mathbf{A}_1 \odot \mathbf{A}_2 \odot \mathbf{A}_3) = R$.

1. If the true signal is generated from STD, the fitted value for STD model is

$$\text{vec}(\hat{\mathcal{Y}}_{STD}) \in C(\hat{\mathbf{M}}_1 \otimes \hat{\mathbf{M}}_2 \otimes \hat{\mathbf{M}}_3), \quad \text{for some} \quad \hat{\mathbf{M}}_k^T \hat{\mathbf{M}}_k = \mathbf{I}_r.$$

If $R \leq r^3$, the space \mathcal{P}_{SupCP} may not cover the best estimation from the true model, $\text{vec}(\hat{\mathcal{Y}}_{STD})$. Because $\text{vec}(\hat{\mathcal{Y}}_{STD})$ is a combination of r^3 bases of \mathbb{R}^3 , and the $\hat{\mathcal{Y}}_{SupCP} \in \mathcal{P}_{SupCP}$ is a combination of R bases of \mathbb{R}^3 .

If $R > r^3$, we can expect the space \mathcal{P}_{SupCP} may cover the best estimation $\text{vec}(\hat{\mathcal{Y}}_{STD})$.

See the following figures for numerical results.

References

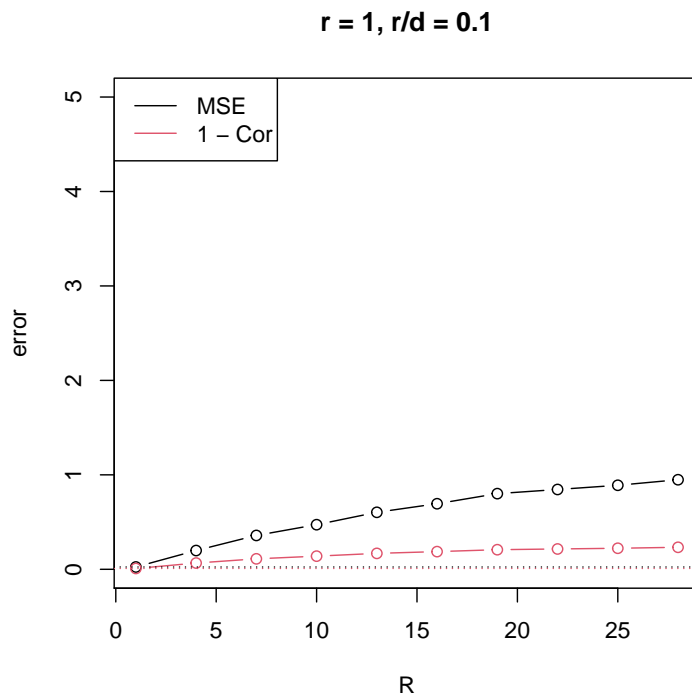


Figure 1: Error of CP model when data is generated from Tucker model. The dashed lines are errors of the STD estimation. Here we consider $\mathbf{r} = (1, 1, 1)$ with $d = 10$.

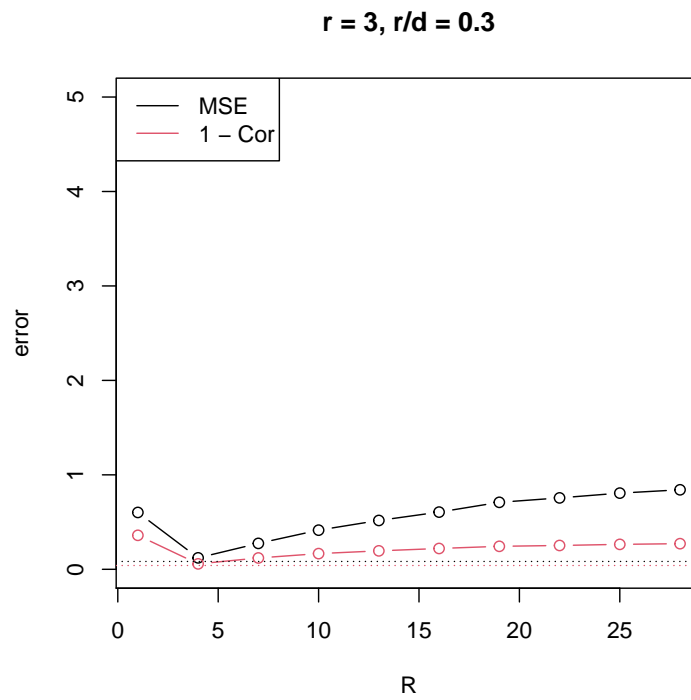


Figure 2: Error of CP model when data is generated from Tucker model. The dashed lines are errors of the STD estimation. Here we consider $\mathbf{r} = (3, 3, 3)$ with $d = 10$.

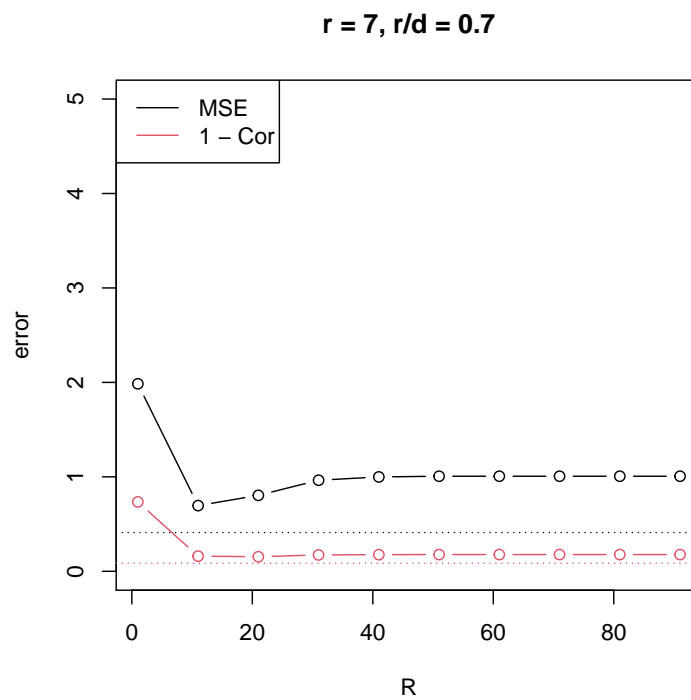


Figure 3: Error of CP model when data is generated from Tucker model. The dashed lines are errors of the STD estimation. Here we consider $\mathbf{r} = (7, 7, 7)$ with $d = 10$.

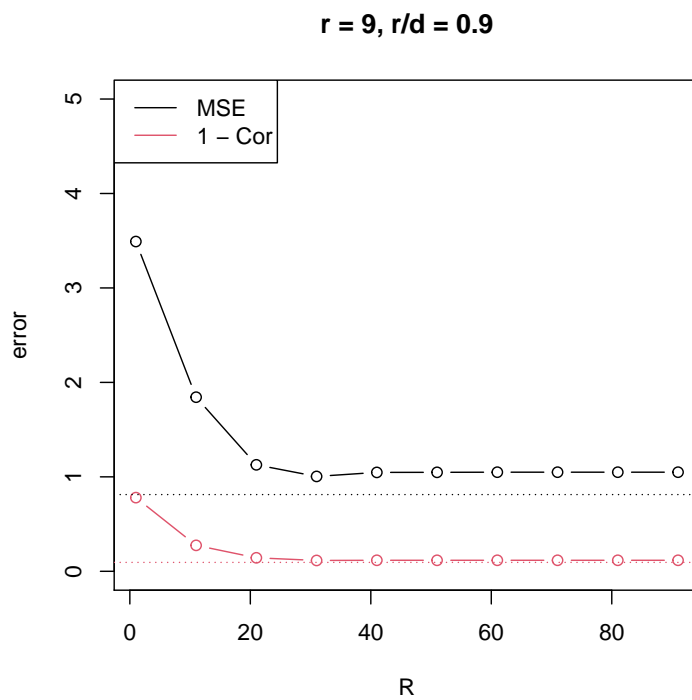


Figure 4: Error of CP model when data is generated from Tucker model. The dashed lines are errors of the STD estimation. Here we consider $\mathbf{r} = (9, 9, 9)$ with $d = 10$.