

# Thought about SupCP

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## 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  $\mathcal{N}(0, \Sigma)$ .

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

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.

## 2.1 Without supervision

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

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

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

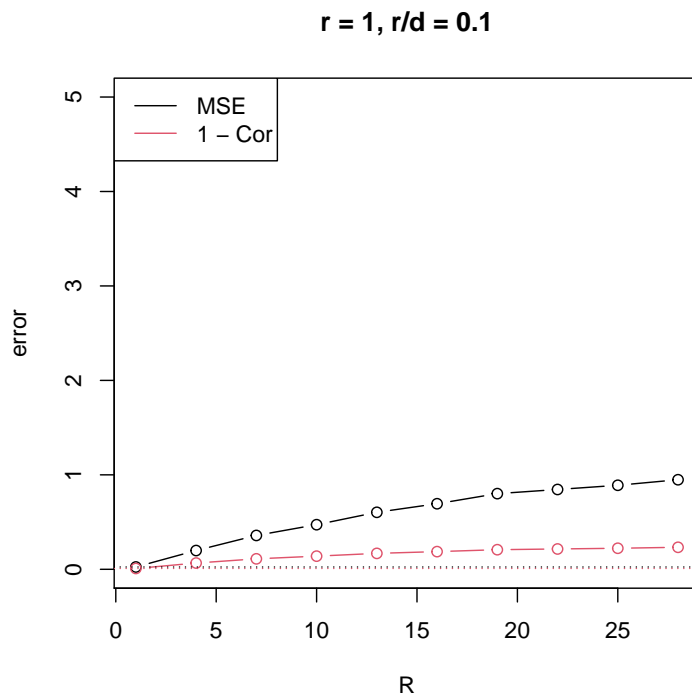


Figure 1: Error of CP model when data is generated from Tucker model. The dashed lines are errors of the STD estimation. Here we consider  $\mathbf{r} = (1, 1, 1)$  with  $d = 10$ .

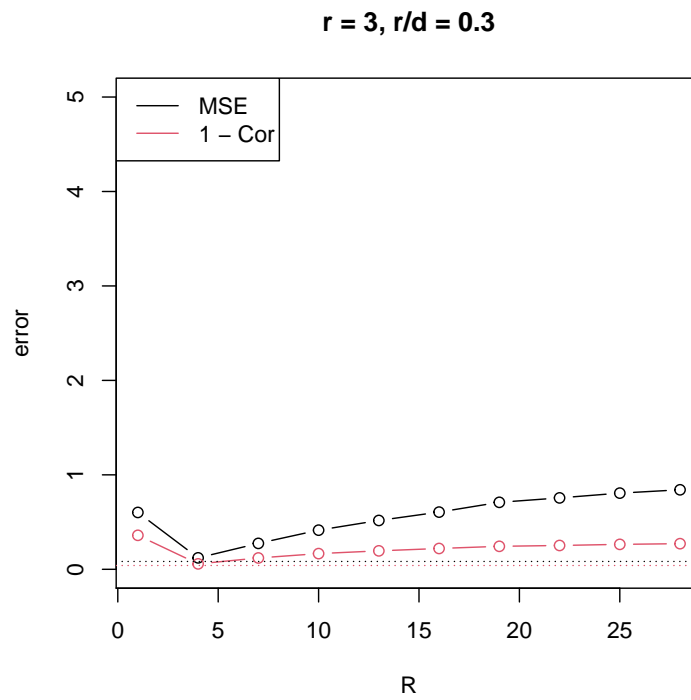


Figure 2: Error of CP model when data is generated from Tucker model. The dashed lines are errors of the STD estimation. Here we consider  $\mathbf{r} = (3, 3, 3)$  with  $d = 10$ .

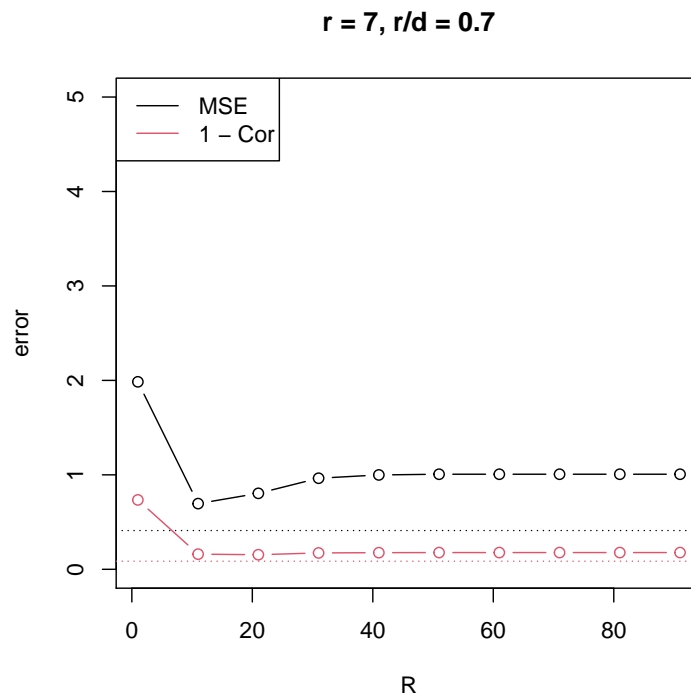


Figure 3: Error of CP model when data is generated from Tucker model. The dashed lines are errors of the STD estimation. Here we consider  $\mathbf{r} = (7, 7, 7)$  with  $d = 10$ .

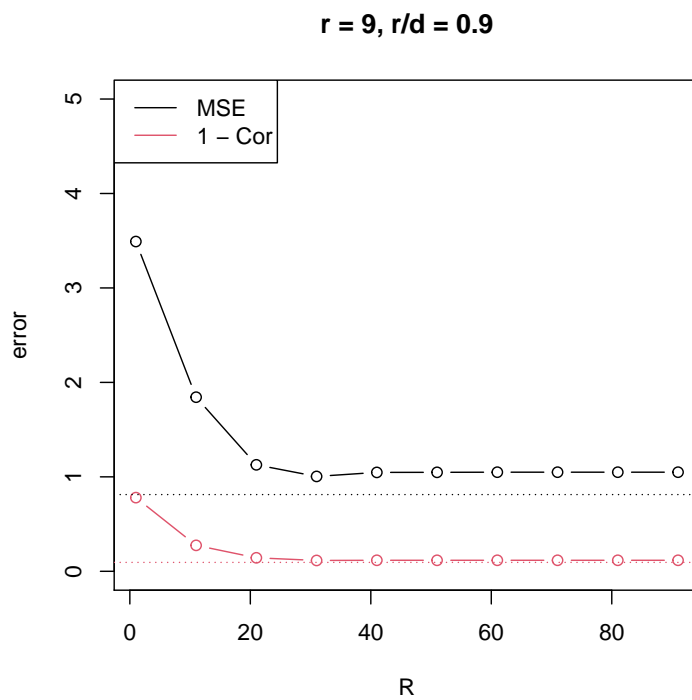


Figure 4: Error of CP model when data is generated from Tucker model. The dashed lines are errors of the STD estimation. Here we consider  $\mathbf{r} = (9, 9, 9)$  with  $d = 10$ .