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Supplementary Materials

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8 In this section, we give proofs of three theorems in paper.
910 *Proof of theorem 3.1*

$$\begin{aligned}
& \sum_{i=1}^n \|\hat{\xi}_i - \gamma_i \times_1 \hat{\beta}_1 \times_2 \cdots \times_K \hat{\beta}_K\|_F^2 \\
&= \sum_{i=1}^n [\text{vec}(\hat{\xi}_i) - (\hat{\beta}_K \otimes \cdots \otimes \hat{\beta}_1) \text{vec}(\gamma_i)]^T [\text{vec}(\hat{\xi}_i) - (\hat{\beta}_K \otimes \cdots \otimes \hat{\beta}_1) \text{vec}(\gamma_i)] \quad (16) \\
&= \sum_{i=1}^n [\text{vec}(\hat{\xi}_i)^T \text{vec}(\hat{\xi}_i) - \text{vec}(\gamma_i)^T (\hat{\beta}_K \otimes \cdots \otimes \hat{\beta}_1) \text{vec}(\gamma_i) \\
&\quad - \text{vec}(\hat{\xi}_i)^T (\hat{\beta}_K \otimes \cdots \otimes \hat{\beta}_1)^T \text{vec}(\gamma_i) \\
&\quad + \text{vec}(\gamma_i)^T (\hat{\beta}_K \otimes \cdots \otimes \hat{\beta}_1)^T (\hat{\beta}_K \otimes \cdots \otimes \hat{\beta}_1) \text{vec}(\gamma_i)] \\
&= \sum_{i=1}^n \text{vec}(\hat{\xi}_i)^T \text{vec}(\hat{\xi}) - 2 \sum_{i=1}^n \text{vec}(\hat{\xi}_i)^T (\hat{\beta}_K \otimes \cdots \otimes \hat{\beta}_1) \text{vec}(\gamma_i) \\
&\quad + \sum_{i=1}^n \text{vec}(\gamma_i)^T \text{vec}(\gamma_i), \quad (17)
\end{aligned}$$

30 where equation (16) holds because of the fourth conclusion in *proposition 2.3* and equation
31 (17) holds due to $\hat{\beta}_k^T \hat{\beta}_k = I_{d_k}$, $k = 1, \dots, K$. Also, since the last term in equation (17) is
32 a constant, the minimization in equation (11) is equivalent to minimizing
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$$Q(\gamma) = \sum_{i=1}^n \text{vec}(\gamma_i)^T \text{vec}(\gamma_i) - 2 \sum_{i=1}^n \text{vec}(\hat{\xi}_i)^T (\hat{\beta}_K \otimes \cdots \otimes \hat{\beta}_1) \text{vec}(\gamma_i). \quad (18)$$

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35 Differentiate $Q(\gamma)$ with respect to $\text{vec}(\gamma_i)$
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$$\begin{aligned}
\frac{\partial Q(\gamma)}{\partial \text{vec}(\gamma_i)^T} &= 2\text{vec}(\gamma_i)^T - 2\text{vec}(\hat{\xi}_i)^T (\hat{\beta}_K \otimes \cdots \otimes \hat{\beta}_1) \\
&= 0.
\end{aligned}$$

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38 So we can obtain $\text{vec}(\gamma_i) = (\hat{\beta}_K \otimes \cdots \otimes \hat{\beta}_1)^T \text{vec}(\hat{\xi}_i)$. Then by the fourth conclusion in
39 *proposition 2.3*, we can get $\hat{\gamma}_i = \hat{\xi}_i \times_1 \hat{\beta}_1^T \times_2 \cdots \times_K \hat{\beta}_K^T$. ■.
4041 *Proof of theorem 3.2.* From Theorem 3.1, we know that
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$$\hat{\gamma}_i = \hat{\xi}_i \times_1 \hat{\beta}_1^T \times_2 \cdots \times_K \hat{\beta}_K^T$$

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46 for every i , plugging $\gamma_i \triangleq \hat{\gamma}_i = \hat{\xi}_i \times_1 \hat{\beta}_1^T \times_2 \cdots \times_K \hat{\beta}_K^T$ into $\sum_{i=1}^n \|\hat{\xi}_i - \gamma_i \times_1 \hat{\beta}_1 \times_2 \cdots \times_K \hat{\beta}_K\|_F^2$,
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$$\begin{aligned}
 & \sum_{i=1}^n \|\hat{\boldsymbol{\xi}}_i - \hat{\boldsymbol{\xi}}_i \times {}_1\boldsymbol{\beta}_1^T \times {}_2\cdots \times {}_K\boldsymbol{\beta}_K^T \times {}_1\boldsymbol{\beta}_1 \times {}_2\cdots \times {}_K\boldsymbol{\beta}_K\|_F^2 \\
 &= \sum_{i=1}^n \|\hat{\boldsymbol{\xi}}_i - \hat{\boldsymbol{\xi}}_i \times {}_1(\boldsymbol{\beta}_1\boldsymbol{\beta}_1^T) \times {}_2\cdots \times {}_K(\boldsymbol{\beta}_K\boldsymbol{\beta}_K^T)\|_F^2 \\
 &= \sum_{i=1}^n [\text{vec}(\hat{\boldsymbol{\xi}}_i) - \text{vec}(\hat{\boldsymbol{\xi}}_i \times {}_1(\boldsymbol{\beta}_1\boldsymbol{\beta}_1^T) \times {}_2\cdots \times {}_K(\boldsymbol{\beta}_K\boldsymbol{\beta}_K^T))]^T \\
 &\quad \times [\text{vec}(\hat{\boldsymbol{\xi}}_i) - \text{vec}(\hat{\boldsymbol{\xi}}_i \times {}_1(\boldsymbol{\beta}_1\boldsymbol{\beta}_1^T) \times {}_2\cdots \times {}_K(\boldsymbol{\beta}_K\boldsymbol{\beta}_K^T))] \\
 &= \sum_{i=1}^n \text{vec}(\hat{\boldsymbol{\xi}}_i)^T \text{vec}(\hat{\boldsymbol{\xi}}_i) - \sum_{i=1}^n \text{vec}(\hat{\boldsymbol{\xi}}_i)^T ((\boldsymbol{\beta}_K\boldsymbol{\beta}_K^T) \otimes \cdots \otimes (\boldsymbol{\beta}_1\boldsymbol{\beta}_1^T)) \text{vec}(\hat{\boldsymbol{\xi}}_i) \\
 &\quad - \sum_{i=1}^n \text{vec}(\hat{\boldsymbol{\xi}}_i)^T ((\boldsymbol{\beta}_K\boldsymbol{\beta}_K^T) \otimes \cdots \otimes (\boldsymbol{\beta}_1\boldsymbol{\beta}_1^T)) \text{vec}(\hat{\boldsymbol{\xi}}_i) \\
 &\quad - \sum_{i=1}^n \text{vec}(\hat{\boldsymbol{\xi}}_i)^T ((\boldsymbol{\beta}_K\boldsymbol{\beta}_K^T) \otimes \cdots \otimes (\boldsymbol{\beta}_1\boldsymbol{\beta}_1^T)) \times ((\boldsymbol{\beta}_K\boldsymbol{\beta}_K^T) \otimes \cdots \otimes (\boldsymbol{\beta}_1\boldsymbol{\beta}_1^T)) \text{vec}(\hat{\boldsymbol{\xi}}_i) \\
 &= \sum_{i=1}^n \text{vec}(\hat{\boldsymbol{\xi}}_i)^T \text{vec}(\hat{\boldsymbol{\xi}}_i) - \sum_{i=1}^n \text{vec}(\hat{\boldsymbol{\xi}}_i)^T ((\boldsymbol{\beta}_K\boldsymbol{\beta}_K^T) \otimes \cdots \otimes (\boldsymbol{\beta}_1\boldsymbol{\beta}_1^T)) \text{vec}(\hat{\boldsymbol{\xi}}_i) \\
 &= \sum_{i=1}^n \|\hat{\boldsymbol{\xi}}_i\|_F^2 - \sum_{i=1}^n \|\hat{\boldsymbol{\xi}}_i \times {}_1\boldsymbol{\beta}_1^T \times {}_2\cdots \times {}_K\boldsymbol{\beta}_K^T\|_F^2.
 \end{aligned}$$

35 Hence the minimization of equation (11) is equivalent to the following optimizing problem
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$$\{\hat{\boldsymbol{\beta}}_k\}_{k=1}^K = \arg \max_{\boldsymbol{\beta}_k^T \boldsymbol{\beta}_k = I_{d_k}} \sum_{i=1}^n \|\hat{\boldsymbol{\xi}}_i \times {}_1\boldsymbol{\beta}_1^T \times {}_2\cdots \times {}_K\boldsymbol{\beta}_K^T\|_F^2,$$

42 which completes the proof of the theorem. ■.
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46 *Proof of theorem 3.3.* By Theorem 3.2, $\boldsymbol{\beta}_1, \dots, \boldsymbol{\beta}_K$ maximize
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$$\sum_{i=1}^n \|\hat{\boldsymbol{\xi}}_i \times {}_1\boldsymbol{\beta}_1^T \times {}_2\cdots \times {}_K\boldsymbol{\beta}_K^T\|_F^2. \tag{19}$$

51 Let $\mathbf{Y}_i = \hat{\boldsymbol{\xi}}_i \times {}_1\boldsymbol{\beta}_1^T \times {}_2\cdots \times {}_K\boldsymbol{\beta}_K^T$, then $\mathbf{Y}_{i(k)} = \boldsymbol{\beta}_k^T \hat{\boldsymbol{\xi}}_{i(k)} (\boldsymbol{\beta}_K \otimes \cdots \otimes \boldsymbol{\beta}_{k+1} \otimes \boldsymbol{\beta}_{k-1} \cdots \otimes \boldsymbol{\beta}_1)$,
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so we rewrite the above formula (19) as

$$\begin{aligned} & \sum_{i=1}^n \|\hat{\boldsymbol{\xi}}_i \times_1 \boldsymbol{\beta}_1^T \times_2 \cdots \times_K \boldsymbol{\beta}_K^T\|_F^2 = \sum_{i=1}^n \|\mathbf{Y}_i\|_F^2 \\ &= \sum_{i=1}^n \text{Trace}(\mathbf{Y}_{i(k)} \mathbf{Y}_{i(k)}^T) \\ &= \sum_{i=1}^n \text{Trace}(\boldsymbol{\beta}_k^T \hat{\boldsymbol{\xi}}_{i(k)} (\boldsymbol{\beta}_K \otimes \cdots \otimes \boldsymbol{\beta}_{k+1} \otimes \boldsymbol{\beta}_{k-1} \otimes \cdots \otimes \boldsymbol{\beta}_1) \\ &\quad \times (\boldsymbol{\beta}_K \otimes \cdots \otimes \boldsymbol{\beta}_{k+1} \otimes \boldsymbol{\beta}_{k-1} \otimes \cdots \otimes \boldsymbol{\beta}_1)^T \hat{\boldsymbol{\xi}}_{i(k)}^T \boldsymbol{\beta}_k) \\ &= \text{Trace}\left(\boldsymbol{\beta}_k^T \sum_{i=1}^n \hat{\boldsymbol{\xi}}_{i(k)} ((\boldsymbol{\beta}_K \boldsymbol{\beta}_K^T) \otimes \cdots \otimes (\boldsymbol{\beta}_{k+1} \boldsymbol{\beta}_{k+1}^T) \otimes (\boldsymbol{\beta}_{k-1} \boldsymbol{\beta}_{k-1}^T) \otimes \cdots \otimes (\boldsymbol{\beta}_1 \boldsymbol{\beta}_1^T)) \hat{\boldsymbol{\xi}}_{i(k)}^T \boldsymbol{\beta}_k\right) \\ &\triangleq \text{Trace}(\boldsymbol{\beta}_k^T M_k \boldsymbol{\beta}_k), \end{aligned}$$

where $M_k \triangleq \sum_{i=1}^n \hat{\boldsymbol{\xi}}_{i(k)} ((\boldsymbol{\beta}_K \boldsymbol{\beta}_K^T) \otimes \cdots \otimes (\boldsymbol{\beta}_{k+1} \boldsymbol{\beta}_{k+1}^T) \otimes (\boldsymbol{\beta}_{k-1} \boldsymbol{\beta}_{k-1}^T) \otimes \cdots \otimes (\boldsymbol{\beta}_1 \boldsymbol{\beta}_1^T)) \hat{\boldsymbol{\xi}}_{i(k)}^T$.

Hence, given $\boldsymbol{\beta}_1, \dots, \boldsymbol{\beta}_{k-1}, \boldsymbol{\beta}_{k+1}, \dots, \boldsymbol{\beta}_K$, the maximum of $\sum_{i=1}^n \|\hat{\boldsymbol{\xi}}_i \times_1 \boldsymbol{\beta}_1^T \times_2 \cdots \times_K \boldsymbol{\beta}_K^T\|_F^2 = \text{Trace}(\boldsymbol{\beta}_k^T M_k \boldsymbol{\beta}_k)$ is obtained. Hence, $\boldsymbol{\beta} \in \mathbb{R}^{p_k \times d_k}$ consists of the d_k eigenvectors of the matrix M_k corresponding to the largest d_k eigenvalues. ■