## **Input:** Response tensor $\mathcal{Y} \in \mathbb{R}^{d_1 \times \cdots \times d_K}$ , covariate matrices $X_k \in \mathbb{R}^{d_k \times p_k}$ for $k = 1, \dots, K$ , target Tucker rank $\mathbf{r} = (r_1, \dots, r_K)$ , link function f, infinity norm bound $\alpha$ **Output:** Low-rank estimation for the coefficient tensor $\mathcal{B} \in \mathbb{R}^{p_1 \times \cdots \times p_K}$ .

1: Calculate  $\check{\mathcal{B}} = \mathcal{Y} \times_1 \left[ (\boldsymbol{X}_1^T \boldsymbol{X}_1)^{-1} \boldsymbol{X}_1^T \right] \times_2 \cdots \times_K \left[ (\boldsymbol{X}_K^T \boldsymbol{X}_K)^{-1} \boldsymbol{X}_K^T \right].$ 

2: Initialize the iteration index t = 0. Initialize the core tensor  $C^{(0)}$  and factor matrices  $M_{\iota}^{(0)} \in \mathbb{R}^{p_k \times r_k}$  via

4:

5:

12: end while

rank-r Tucker approximation of  $\dot{\mathcal{B}}$ , in the least-square sense.

**Algorithm 1** Generalized tensor response regression with covariates on multiple modes

3: while the relative increase in objective function  $\mathcal{L}_{\mathcal{V}}(\mathcal{B})$  is less than the tolerance do Update iteration index  $t \leftarrow t + 1$ .

for k = 1 to K do

Obtain the factor matrix  $M_k^{(t+1)} \in \mathbb{R}^{p_k \times r_k}$  by solving  $d_k$  separate GLMs with link function f. 6:

Update the columns of  $M_k^{(t+1)}$  by Gram-Schmidt orthogonalization. end for

8: Obtain the core tensor  $\mathcal{C}^{(t+1)} \in \mathbb{R}^{r_1 \times \cdots \times r_K}$  by solving a GLM with  $\text{vec}(\mathcal{Y})$  as response,  $\bigcirc_{k=1}^K [X_k M_k^{(t)}]$ 9:

as covariates, and f as link function. Here  $\odot$  denotes the Khatri-Rao product of matrices.

Rescale the core tensor subject to the infinity norm constraint.

10:

Update  $\mathcal{B}^{(t+1)} \leftarrow \mathcal{C}^{(t+1)} \times_1 M_1^{(t+1)} \times_2 \cdots \times_K M_V^{(t+1)}$ . 11: