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, maximum norm bound α **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 p_k separate GLMs with link function f. 6: Update the columns of $M_k^{(t+1)}$ by Gram-Schmidt orthogonalization.

8: end for

9:

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, $\bigotimes_{k=1}^K [X_k M_k^{(t)}]$

as covariates, and f as link function. Here \otimes denotes the kronecker product of matrices.

Rescale the core tensor subject to the maximum norm constraint.

10:

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