Input: Noisy and incomplete data tensor \mathcal{Y}_{Ω} , rank r, resolution parameter H.

1: **for** $\pi \in \mathcal{H} = \{-1, \dots, -\frac{1}{u}, 0, \frac{1}{u}, \dots, 1\}$ **do** Random initialization of tensor factors $\mathbf{A}_k = [\mathbf{a}_1^{(k)}, \dots, \mathbf{a}_r^{(k)}] \in \mathbb{R}^{d_k \times r}$ for all $k \in [K]$.

for $k = 1, \ldots, K$ do Update A_k while holding others fixed:

$$\mathbf{A}_k \leftarrow \operatorname{arg\,min}_{\mathbf{A}_k \in \mathbb{R}^{d_k \times r}} \sum_{\omega \in \Omega} |\mathcal{Y}(\omega) - \pi| F(\mathcal{Z}(\omega) \operatorname{sgn}(\mathcal{Y}(\omega) - \pi)),$$

Output: Estimated signal tensor $\hat{\Theta} = \frac{1}{2H+1} \sum_{\pi \in \mathcal{H}} \operatorname{sgn}(\mathcal{Z}_{\pi}).$

end while 9:

4:

5:

6:

11: end for

9: **end while**
10: Return
$$\mathcal{Z}_{\pi} \leftarrow \sum_{s \in [r]} \boldsymbol{a}_{s}^{(1)} \otimes \cdots \otimes \boldsymbol{a}_{s}^{(K)}$$
.

Algorithm 1 Nonparametric tensor completion

where $F(\cdot)$ is the large-margin loss, and $\mathcal{Z} = \sum_{s \in [r]} a_s^{(1)} \otimes \cdots \otimes a_s^{(K)}$ is a rank-r tensor.

rs fixed:
$$|Y(x)-\pi|E(Z(x))\exp(Y(x)-\pi)$$

$$(0)-\pi)),$$