

# TIP8419 - Tensor Algebra

## Homework 9

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### Alternating Least Squares (ALS) Algorithm

**Problem 1** For the third-order tensor  $\mathcal{X} \in \mathbb{C}^{I \times J \times K}$  provided in the file “cpd\_tensor.mat”, implement the plain-vanilla Alternating Least Squares (ALS) algorithm that estimates the factor matrices  $\mathbf{A} \in \mathbb{C}^{I \times R}$ ,  $\mathbf{B} \in \mathbb{C}^{J \times R}$  and  $\mathbf{C} \in \mathbb{C}^{K \times R}$  by solving the following problem

$$(\hat{\mathbf{A}}, \hat{\mathbf{B}}, \hat{\mathbf{C}}) = \min_{\mathbf{A}, \mathbf{B}, \mathbf{C}} \left\| \mathcal{X} - \sum_{r=1}^R \mathbf{a}_r \circ \mathbf{b}_r \circ \mathbf{c}_r \right\|_F^2,$$

where  $\mathbf{A} = [\mathbf{a}_1, \dots, \mathbf{a}_R]$ ,  $\mathbf{B} = [\mathbf{b}_1, \dots, \mathbf{b}_R]$ ,  $\mathbf{C} = [\mathbf{c}_1, \dots, \mathbf{c}_R]$ . Considering a successful run, compare the estimated matrices  $\hat{\mathbf{A}}$ ,  $\hat{\mathbf{B}}$  and  $\hat{\mathbf{C}}$  with the original ones. Explain the results.

Hint: An error measure at the  $i$ -th iteration can be calculated from the following formula:

$$e_{(i)} = \left\| [\mathcal{X}]_{(1)} - \hat{\mathbf{A}}_{(i)} (\hat{\mathbf{C}}_{(i)} \diamond \hat{\mathbf{B}}_{(i)})^T \right\|_F \quad (1)$$

The convergence at the  $i$ -th iteration can be declared when  $e_{(i-1)} - e_{(i)} < \delta$ , where  $\delta$  is a prescribed threshold value (e.g.  $\delta = 10^{-6}$ ).

**Problem 2** Assuming 1000 Monte Carlo experiments, generate a tensor  $\mathcal{X}_0 = \text{CPD}(\mathbf{A}, \mathbf{B}, \mathbf{C})$ , where  $\mathbf{A} \in \mathbb{C}^{I \times R}$ ,  $\mathbf{B} \in \mathbb{C}^{J \times R}$  and  $\mathbf{C} \in \mathbb{C}^{K \times R}$  have unit norm columns with elements randomly drawn from a Normal distribution, with  $R = 3$ . Let  $\mathcal{X} = \mathcal{X}_0 + \alpha \mathcal{V}$  be a noisy version of  $\mathcal{X}_0$ , where  $\mathcal{V}$  is the additive noise term, whose elements are also drawn from a Normal distribution. The parameter  $\alpha$  controls the power (variance) of the noise term, and is defined as a function of the signal to noise ratio (SNR), in dB, as follows

$$\text{SNR}_{\text{dB}} = 10 \log_{10} \left( \frac{\|\mathcal{X}_0\|_F^2}{\|\alpha \mathcal{V}\|_F^2} \right). \quad (2)$$

Assuming the SNR range  $[0, 5, 10, 15, 20, 25, 30]$  dB, find the estimates  $\hat{\mathbf{A}}$ ,  $\hat{\mathbf{B}}$  and  $\hat{\mathbf{C}}$  obtained with the ALS algorithm for  $(I, J, K) = (10, 4, 2)$ .

Let us define the normalized mean square error (NMSE) measure as follows

$$\text{NMSE}(\mathbf{Q}) = \frac{1}{1000} \sum_{i=1}^{1000} \frac{\|\hat{\mathbf{Q}}(i) - \mathbf{Q}(i)\|_F^2}{\|\mathbf{Q}(i)\|_F^2}, \quad (3)$$

where  $\mathbf{Q}(i)$  e  $\hat{\mathbf{Q}}(i)$  represent the original data matrix and the reconstructed one at the  $i$ th Monte Carlo experiment, respectively. For each SNR value, plot  $\text{NMSE}(\mathbf{A})$ ,  $\text{NMSE}(\mathbf{B})$  and  $\text{NMSE}(\mathbf{C})$  as a function of the SNR. Discuss the obtained results.