

F.M. Faulstich **Math 6590: Homework assignment 4 Due:** Wednesday 04/10/2024.

Exercise 1 Consider the CP decomposition of $\mathbf{X} \in \mathbb{R}^{n_1 \times \dots \times n_d}$, i.e., the representation of \mathbf{X} as a sum of elementary tensors

$$\mathbf{X} = \sum_{p=1}^r \mathbf{v}_{1,p} \otimes \dots \otimes \mathbf{v}_{d,p} = \sum_{p=1}^r \bigotimes_{i=1}^d \mathbf{v}_{i,p}$$

for $\mathbf{v}_{i,p} \in \mathbb{R}^{n_i}$. Define the matrices

$$\mathbf{V}_i = [\mathbf{v}_{i,1} | \dots | \mathbf{v}_{i,r}] \quad (1)$$

and show that

$$\mathbf{X}[i_1, \dots, i_d] = \sum_{p=1}^r \mathbf{V}_1[i_1, p] \mathbf{V}_2[i_2, p] \dots \mathbf{V}_d[i_d, p] \quad (2)$$

Exercise 2 Consider the sequence

$$\mathbf{X}_n = n \left(\mathbf{u} + \frac{1}{n} \mathbf{v} \right) \otimes \left(\mathbf{u} + \frac{1}{n} \mathbf{v} \right) \otimes \left(\mathbf{u} + \frac{1}{n} \mathbf{v} \right) \otimes -n \mathbf{u} \otimes \mathbf{u} \otimes \mathbf{u}$$

with $\mathbf{u}, \mathbf{v} \in \mathbb{R}^m$, $\|\mathbf{v}\| = \|\mathbf{u}\| = 1$ and $\langle \mathbf{v}, \mathbf{u} \rangle \neq 1$.

i) Show that $\mathbf{X}_n \in \mathcal{M}_{\leq r}$ for all $n \in \mathbb{N}$

ii) Show that

$$\lim_{n \rightarrow \infty} \mathbf{X}_n = \mathbf{v} \otimes \mathbf{u} \otimes \mathbf{u} + \mathbf{u} \otimes \mathbf{v} \otimes \mathbf{u} + \mathbf{u} \otimes \mathbf{u} \otimes \mathbf{v} \notin \mathcal{M}_{\leq r}$$

Exercise 3 In this exercise, you will implement and compare the different higher-order SVDs introduced in class.

i) Implement the Truncated Higher-Order SVD (THOSVD)

Input: Target tensor $\mathbf{A} \in \mathbb{R}^{n_1 \times \dots \times n_d}$, target rank $r = (r_1, \dots, r_d)$

Output: Core tensor $\mathbf{C} \in \mathbb{R}^{r_1 \times \dots \times r_d}$, basis matrices $\mathbf{U}_k \in \mathbb{R}^{r_k \times n_k}$ for $1 \leq k \leq d$

ii) Implement the Sequential Truncated Higher Order SVD (STHOSVD)

Input: Target tensor $\mathbf{A} \in \mathbb{R}^{n_1 \times \dots \times n_d}$, target rank $r = (r_1, \dots, r_d)$

Output: Core tensor $\mathbf{C} \in \mathbb{R}^{r_1 \times \dots \times r_d}$, basis matrices $\mathbf{U}_k \in \mathbb{R}^{r_k \times n_k}$ for $1 \leq k \leq d$

iii) Implement the Randomized Sequential Truncated Higher Order SVD (R-STHOSVD)

Input: Target tensor $\mathbf{A} \in \mathbb{R}^{n_1 \times \dots \times n_d}$, target rank $r = (r_1, \dots, r_d)$, oversampling parameter $p \in \mathbb{N}$

Output: Core tensor $\mathbf{C} \in \mathbb{R}^{r_1 \times \dots \times r_d}$, basis matrices $\mathbf{U}_k \in \mathbb{R}^{r_k \times n_k}$ for $1 \leq k \leq d$

iv) Implement the sketched Sequential Truncated Higher Order SVD (sketched-STHOSVD)

Input: Target tensor $\mathbf{A} \in \mathbb{R}^{n_1 \times \dots \times n_d}$, target rank $r = (r_1, \dots, r_d)$, oversampling parameter $\ell = (\ell_1, \dots, \ell_d)$

Output: Core tensor $\mathbf{C} \in \mathbb{R}^{r_1 \times \dots \times r_d}$, basis matrices $\mathbf{U}_k \in \mathbb{R}^{r_k \times n_k}$ for $1 \leq k \leq d$

- v) Implement the sub-sketch Sequential Truncated Higher Order SVD (sub-sketch-STHOSVD)
 Input: Target tensor $\mathbf{A} \in \mathbb{R}^{n_1 \times \dots \times n_d}$, target rank $r = (r_1, \dots, r_d)$, oversampling parameter $\ell = (\ell_1, \dots, \ell_d)$, power $q \in \mathbb{N}$
 Output: Core tensor $\mathbf{C} \in \mathbb{R}^{r_1 \times \dots \times r_d}$, basis matrices $\mathbf{U}_k \in \mathbb{R}^{r_k \times n_k}$ for $1 \leq k \leq d$

- vi) Run experiment 1 outlined in class:

Consider the Hilbert tensor $\mathbf{A} \in \mathbb{R}^{n_1 \times \dots \times n_d}$ defined as

$$\mathbf{A}[i_1, \dots, i_d] = \frac{1}{i_1 + \dots + i_d}, \quad 1 \leq i_j \leq n_j, \quad 1 \leq j \leq d$$

Set the tensor parameters:

- $d = 5$
- $n_j = 25, j = 1, 2, \dots, d$
- Target rank is (r, r, r, r, r) , where $r \in \llbracket 1, 25 \rrbracket$

Choose the computational parameters:

- Oversampling $p = 5$
- $l_i = r_i + 2, i = 1, 2, \dots, d$
- Power parameter $q = 1$

Reproduce the plots shown in classes (lecture 17)

- vii) Now consider a Hilbert tensor $\mathbf{A} \in \mathbb{R}^{500 \times 500 \times 500}$. Run experiments targeting the ranks

$$(i \cdot 10, i \cdot 10, i \cdot 10) \quad \text{for } i \in \llbracket 10 \rrbracket$$

Report the CPU time in seconds and the relative error. Reproduce the numbers shown in class (lecture 17).

- viii) Run experiment 2 outlined in class:

Consider the sparse tensor $\mathbf{A} \in \mathbb{R}^{200 \times 200 \times 200}$:

$$\mathbf{A} = \sum_{i=1}^{10} \frac{\gamma}{i^2} \mathbf{x}_i \otimes \mathbf{y}_i \otimes \mathbf{z}_i + \sum_{i=11}^{200} \frac{1}{i^2} \mathbf{x}_i \otimes \mathbf{y}_i \otimes \mathbf{z}_i,$$

where $\mathbf{x}_i, \mathbf{y}_i, \mathbf{z}_i \in \mathbb{R}^{200}$ are sparse vectors with 5% nonzeros each (generated using the **sprand** command in MATLAB)

Set the tensor parameters:

- $\gamma = 2, 10, 200$
- $n_j = 25, j = 1, 2, \dots, d$
- Target rank is (r, r, r) , where $r \in \llbracket 20, 100 \rrbracket$

Choose the computational parameters:

- Oversampling $p = 5$
- $l_i = r_i + 2, i = 1, 2, \dots, d$
- Power parameter $q = 1$

Reproduce the plots shown in class (lecture 17).