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# High Order Singular Value Decomposition (HOSVD)

## Problem 1

For a third-order tensor  $\mathcal{X} \in \mathbb{C}^{I \times J \times K}$  implement the truncated high-order singular value decomposition (HOSVD), using the following prototype function:

$$\left[\mathcal{S}, \mathbf{U}^{(1)}, \mathbf{U}^{(2)}, \mathbf{U}^{(3)}\right] = \operatorname{hosvd}(\mathcal{X})$$
 (1)

Hint: Use the file "hosvd\_test.mat" to validate your results.

### Results

#### Simulation setup

- The algorithm that uses the SVD was applied to the given initial tensor  $\mathcal{X} \in \mathbb{C}^{I \times J \times K}$ ;
- I, J, K = 3, 4, 5.

#### Discussion

To compare the given data with the estimated factors, we may use two main experiment results: the orthogonality between the subtensors (slices) of  $\mathcal{S}$  and the NMSE between the given data and obtained as output to HOSVD.

The orthogonality assessment consists in compute the function  $f_{ort}(\mathcal{S})$ , that acumulates the scalar product bewteen the slices, following the equation below.

$$f_{ort}(\mathcal{S}) = \sum_{k_M=1}^K \sum_{k_N=1}^K ext{vec}(\mathcal{S}_{:,:,k_M})^ op ext{vec}(\mathcal{S}_{:,:,k_N}) \quad for \,\, k_M 
eq k_N$$

We obtain  $f_{ort}(\mathcal{S})=0$ , as expected for successful HOSVD.

The values obtained for NMSE present low SNR, with an emphasis to  $\mathrm{NMSE}(\mathcal{X},\hat{\mathcal{X}})$  value, with a very low SNR.

```
NMSE: -618.62 dB
NMSE: 6.51 dB
NMSE between the factor matrices U and their estimation ----

NMSE between U1 and its estimation: 8.52 dB
NMSE between U2 and its estimation: 6.02 dB
NMSE between U3 and its estimation: 4.12 dB
```

We can see that both results, Orthogonality and NMSE, support the proper algorithm estimation hypothesis.

Problem 1 script.

## Problem 2

Consider the two third-order tensors  $\mathcal{X} \in \mathbb{C}^{8 \times 4 \times 10}$  and  $\mathcal{Y} \in \mathbb{C}^{5 \times 5 \times 5}$  provided in the data file "hosvd\_denoising.mat". By using your HOSVD prototype function, find a low multilinear rank approximation for these tensors, defined as

 $\tilde{\mathcal{X}} \in \mathbb{C}^{R1 \times R2 \times R3}$  and  $\tilde{\mathcal{Y}} \in \mathbb{C}^{P1 \times P2 \times P3}$ . Then, calculate the normalized mean square error (NMSE) between the original tensor and its approximation, i.e.,:

$$\mathrm{NMSE}(\tilde{\mathcal{X}}) = \frac{||\tilde{\mathcal{X}} - \mathcal{X}||_F^2}{||\mathcal{X}||_F^2} \quad , \quad \mathrm{NMSE}(\tilde{\mathcal{Y}}) = \frac{||\tilde{\mathcal{Y}} - \mathcal{Y}||_F^2}{||\mathcal{Y}||_F^2}$$

Hint: The multilinear ranks of X and Y can be found by analysing the profile of the 1-mode, 2-mode and 3-mode singular values of these tensors.

Results

## Simulation setup

- The algorithm that uses the SVD was applied to the given initial tensor  $\mathcal{X} \in \mathbb{C}^{R1 \times R2 \times R3}$  and  $\mathcal{Y} \in \mathbb{C}^{P1 \times P2 \times P3}$ ;
- R1, R2, R3 = 8, 4, 10;
- P1, P2, P3 = 5, 5, 5.

## Discussion

To compare the both random tensor estimation with given multilinear ranks, we may use NMSE results between the given data and obtained as output to HOSVD. We may assess also by comparing the multilinear rank obtained in the tensor core  $S_X$  and  $S_Y$  estimated with the given ones.

```
------ NMSE between a given tensor X and its estimation ----

NMSE: -600.49 dB
------ NMSE between a given tensor Y and its estimation ----

NMSE: -610.64 dB
```

The values obtained for NMSE present very low SNR, less than  $-600\,\mathrm{dB}.$ 

As defined in the proposed problem, the given ranks of  $\mathcal{X}$   $\mathcal{Y}$  are R1,R2,R3=8,4,10, and P1,P2,P3=5,5,5, respectively.

```
Tensor X multilinear rank: [8 4 10]
Tensor Y multilinear rank: [5 5 5]
```

We can see that that the algorithm provide the expected result, with the given ranks equal to the estimated. In conclusion, both results, NMSE and ranks estimation using the tensor core, support the proper algorithm estimation hypothesis.

Problem 2 script.