

Tensor Kalman Filter

For Large-Scale Systems

Delft University of Technology

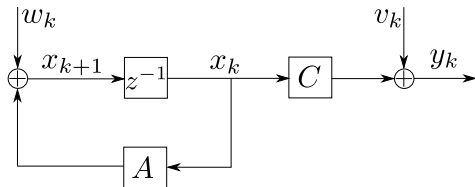
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Outline

- 1 Introduction / Problem Description
- 2 Tensors
- 3 Tensor Kalman Filter
- 4 Application
- 5 Conclusion / Future Work

Introduction / Problem Description

Filtering of a system in state-space description



Problem of Kalman filter for large-scale systems

States n and outputs p large

- Computational complexity of order $\mathcal{O}(n^3)$
for covariance update $P_{k|k}$ and $P_{k+1|k}$, Kalman gain K_k
- Storage of system matrices with $A \in \mathbb{R}^{n \times n}$, $C \in \mathbb{R}^{p \times n}$

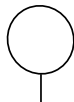
\Rightarrow Design a Kalman filter for large scale real-time problems

Introduction to Tensors

- Tensor as multidimensional array $\mathcal{A} \in \mathbb{R}^{I_1 \times I_2 \times \dots \times I_d}$, $I_k = 1, \dots, n_k$
- Each entry $a_{i_1 i_2 \dots i_d}$ determined by d indices \rightarrow Order- d tensor
- Visualization in Tensor Network (TN) Diagrams:



Scalar
 $a \in \mathbb{R}$



Vector
 $\mathbf{a} \in \mathbb{R}^{I_1}$



Matrix
 $\mathbf{A} \in \mathbb{R}^{I_1 \times I_2}$



Tensor Order-3
 $\mathcal{A} \in \mathbb{R}^{I_1 \times I_2 \times I_3}$

Basic Operations (1)

- Define $\mathcal{A} \in \mathbb{R}^{I_1}$, $\mathcal{B} \in \mathbb{R}^{I_1}$

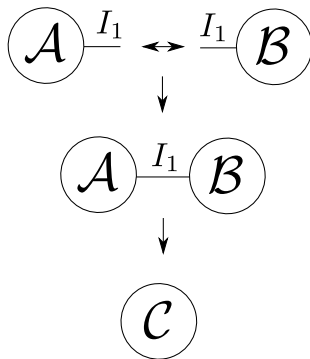
- Elementwise notation:

$$\begin{aligned} c &= \sum_{i_1=1}^{n_1} a_{i_1} b_{i_1} \\ &= \mathbf{a} \cdot \mathbf{b} \end{aligned}$$

- Tensor notation:

$$\mathcal{C} = \mathcal{B} \times_1^1 \mathcal{A}$$

\Rightarrow Inner Product



Basic Operations (2)

- Define $\mathcal{A} \in \mathbb{R}^{I_1}$, $\mathcal{B} \in \mathbb{R}^{I_2}$
 $\rightarrow \mathcal{A} \in \mathbb{R}^{I_1 \times 1}$, $\mathcal{B} \in \mathbb{R}^{I_2 \times 1}$

- Elementwise notation:

$$c_{i_1 i_2} = \sum_{i=1}^1 a_{i_1} b_{i_2}$$

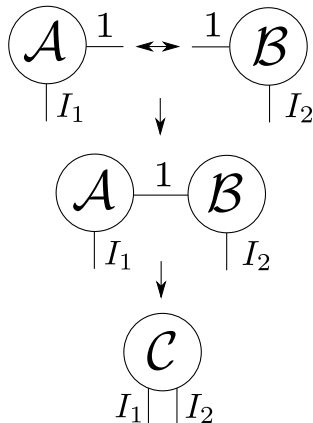
- Vector notation:

$$\mathbf{c} = \mathbf{a}\mathbf{b}^T = \mathbf{a} \circ \mathbf{b}$$

- Tensor notation:

$$\mathcal{C} = \mathcal{B} \times_2^2 \mathcal{A}$$

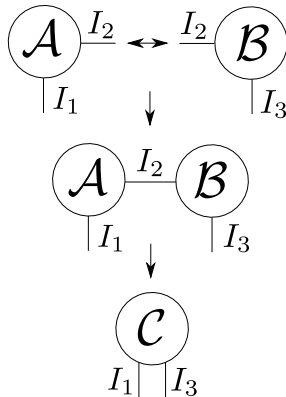
\Rightarrow Outer Product



Basic Operations (3)

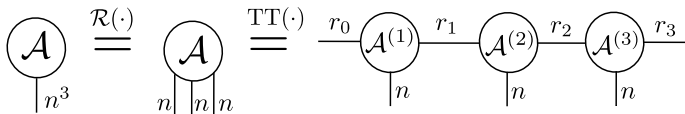
- Define $\mathcal{A} \in \mathbb{R}^{I_1 \times I_2}$, $\mathcal{B} \in \mathbb{R}^{I_2 \times I_3}$
- Elementwise Notation:
$$c_{i_1 i_3} = \sum_{i_2=1}^{n_2} a_{i_1 i_2} b_{i_2 i_3}$$
- Matrix Notation:
 $\mathbf{C} = \mathbf{AB} = \mathbf{A} \circ \mathbf{B}^T$
- Tensor Notation:
 $\mathcal{C} = \mathcal{B} \times_1^2 \mathcal{A}$

 \Rightarrow Matrix Product
- Complexity: $\mathcal{O}(n_1 n_2 n_3)$ Flops

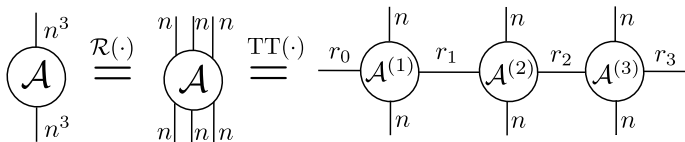


Tensor Train (TT) Decomposition

- One order- d tensor $\Rightarrow d$ order-3 TN cores.
- Vector, example:



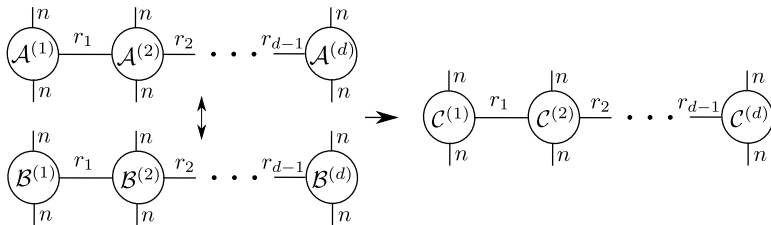
- Matrix, example:



- TT-rank boundary conditions: $r_0 = r_d = 1$

Tensor Train (TT) Operations

- Matrix-Format:
 - Define $\mathbf{A} \in \mathbb{R}^{n^d \times n^d}$, $\mathbf{B} \in \mathbb{R}^{n^d \times n^d}$
 - $\mathbf{C} = \mathbf{AB}$
 - Complexity: $\mathcal{O}(n^{3d})$
- TT-Format:
 - Define $\mathcal{A}^{(i)}, \mathcal{B}^{(i)} \in \mathbb{R}^{r_{i-1} \times n \times n \times r_i}$
 - $\mathcal{C} = \mathcal{B} \times_2^3 \mathcal{A}$

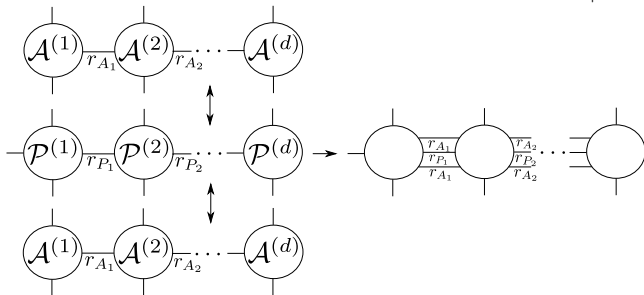


- Complexity: $\mathcal{O}(dn^3r^2)$

Tensor Kalman Filter

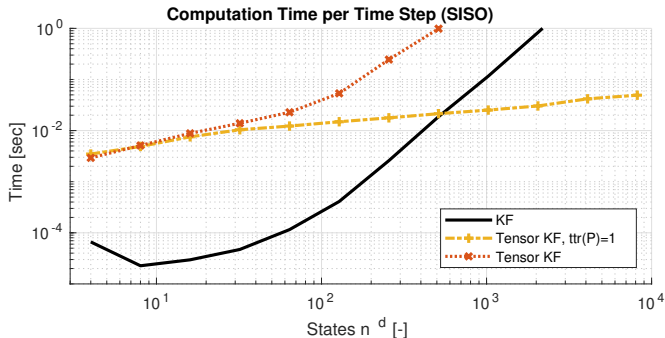
- 1 Transform system matrices \mathbf{A} , \mathbf{C} , (\mathbf{Q}, \mathbf{R}) in TT-format.
 $\mathbf{A} \in \mathbb{R}^{n^d \times n^d}$, $\mathbf{C} \in \mathbb{R}^{p \times n^d}$
- 2 Rewrite all variables in the KF in TT-format
 Covariance, Kalman gain, ...
- 3 Rewrite KF equations with multilinear operations

Example: Time update of covariance (Matrix form: $\mathbf{A}\mathbf{P}_{k|k}\mathbf{A}^T + \mathbf{Q}$)



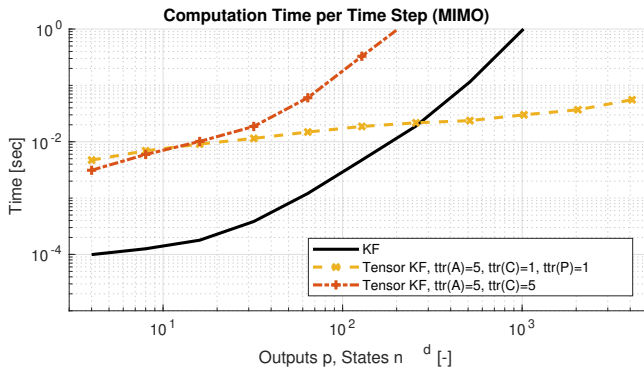
Tensor Kalman Filter - Improvement: TT-ranks

- Problem:
 - TT-ranks dominate the algorithmic complexity
 - No influence on TT-ranks $\text{ttr}(\mathcal{A})$ and $\text{ttr}(\mathcal{C})$
- Analysis SISO:
 - Effective only for systems with $\text{ttr}(\mathcal{A}) = \text{ttr}(\mathcal{C}) = 1$
 - Solution: Truncation of $\text{ttr}(\mathcal{P}) \rightarrow$ Approximation



Tensor Kalman Filter - Improvement: MIMO

- Analysis MIMO ($p = n^d$) including TT-rank solution:
 - Inversion $\mathbf{S} = (\mathbf{C}\mathbf{P}_{k|k-1}\mathbf{C}^T + \mathbf{R})$ with $\mathcal{O}(p^3)$ same as matrix filter
 - Special case for $\text{ttr}(\mathcal{S}) = 1$
 - all cores will be matrices ($1 \times n \times n \times 1$)
 - obtained by $\text{ttr}(\mathcal{C}) = 1$ and low measurement noise



Application - General

In General: Systems with low TT-ranks in \mathcal{A}, \mathcal{C}

- Recall: TT is outer product

$$\mathcal{A} = \mathcal{A}^{(d)} \times_1^3 \dots \times_1^3 \mathcal{A}^{(1)}$$

- If all TT-ranks are equal to one, then

$$\mathcal{A} = \mathcal{R} \left(\mathbf{A}^{(1)} \otimes \dots \otimes \mathbf{A}^{(d)} \right)$$

- Reason: Outer product is reshaped and permuted version of Kronecker product
 \Rightarrow every Kronecker Model has TT-rank of one in TT-format

Application - Specific: MIMO

For MIMO special case system:

$$\mathbf{x}_{k+1} = \mathbf{A}\mathbf{x}_k + \mathbf{w}_k$$
$$\mathbf{y}_k = \begin{bmatrix} \mathbf{C}_1 & 0 & \dots & 0 \\ 0 & \mathbf{C}_1 & & \vdots \\ \vdots & & \ddots & \\ 0 & \dots & & \mathbf{C}_1 \end{bmatrix} \mathbf{x}_k + \mathbf{v}_k$$

- Interconnected systems with distributed sensing
- Applications:
 - (Power) Networks
 - Partial Differential Equations
 - ...

Conclusion / Future Work

Conclusion:

- Solves the problem of exponentially large system matrices \mathbf{A} , \mathbf{C}
- Fast SISO filter
- Fast MIMO filter in special case

Limitations:

- General: Necessary small TT-ranks of system matrices
- MIMO filter: $\text{ttr}(\mathcal{C}) = 1$ and low measurement noise
- No square root implementation available

Future Work:

- 1 Improve filter algorithm
- 2 Find more applications



Thank you for your attention

Questions?

Appendix - Applications

- Recall: TT is outer product

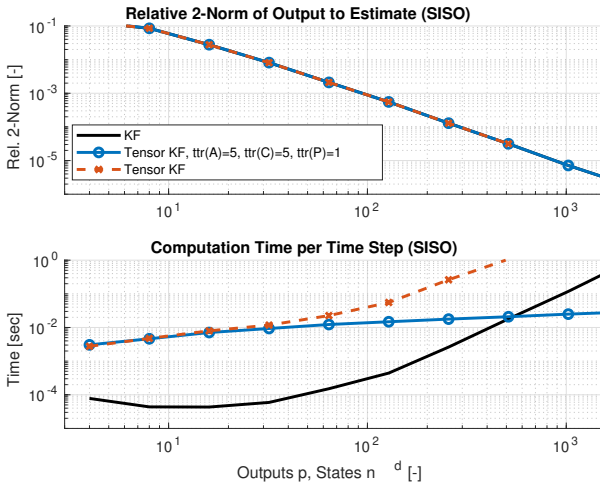
$$\mathcal{A} = \mathcal{A}^{(d)} \times_1^3 \dots \times_1^3 \mathcal{A}^{(1)}$$

- If all TT-ranks are equal to r , then

$$\mathcal{A} = \mathcal{R} \left(\sum_{k=1}^r \mathbf{A}_k^{(1)} \otimes \dots \otimes \mathbf{A}_k^{(d)} \right)$$

- Reason: Outer product is reshaped and permuted version of Kronecker product
 \Rightarrow every Kronecker Model with low r has low TT-ranks in TT-format

Appendix - SISO Filter



Appendix - MIMO Filter

