

# On minimal ranks and the approximate block-term tensor decomposition

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# Summary

**Goal:** study **existence** of soln's to the **approximation** problem

$$\min_{\mathbf{H}_i, \mathbf{w}_i} \left\| \mathbf{y} - \sum_{i=1}^k \mathbf{H}_i \otimes \mathbf{w}_i \right\|, \quad \text{subj. to rank } \mathbf{H}_i \leq r_i$$

- ▶ Block-term decomposition, a.k.a. BTM (De Lathauwer, 2008)
  - ▶ Useful for source separation (uniqueness results)
  - ▶ One block per source

**Motivation:** In practice,  $\mathbf{y} = (\text{BTM}) + \mathcal{E}$

- ▶ Computation: typically, optimization with fixed  $k$  and  $r_i$
- ▶ But does a global minimum always exist?



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- ▶ But does a global minimum always exist?
  - ▶ **(Spoiler alert)** No



## Summary

Tensors and tensor decompositions

Minimal ranks of BTDs and of matrix polynomials

Non-existence of best approximate BTB

Final remarks



# Tensors

- ▶ Formally, elements of tensor products of abstract vector spaces

$$\mathcal{U} \otimes \mathcal{V} \otimes \mathcal{W}$$

- ▶ Can be viewed as multilinear maps



# Tensors

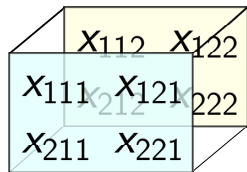
- Formally, elements of tensor products of abstract vector spaces

$$\mathcal{U} \otimes \mathcal{V} \otimes \mathcal{W}$$

- Can be viewed as multilinear maps
- Finite dimensions, fixed basis: tensors  $\leftrightarrow$  multi-way arrays

$$\mathbb{R}^m \otimes \mathbb{R}^n \otimes \mathbb{R}^d \simeq \mathbb{R}^{m \times n \times d}$$

Example:  $\mathcal{X} \in \mathbb{R}^{2 \times 2 \times 2}$  (third order: three-way array)



$$\left( \begin{array}{cc|cc} X_{111} & X_{121} & X_{112} & X_{122} \\ X_{211} & X_{221} & X_{212} & X_{222} \end{array} \right)$$



# Tensor product (Segre outer product)

## Definition

- ▶ Binary operator

$$\otimes : \mathbb{R}^{m_1 \times \dots \times m_p} \times \mathbb{R}^{n_1 \times \dots \times n_q} \rightarrow \mathbb{R}^{m_1 \times \dots \times m_p \times n_1 \times \dots \times n_q}$$

$$[\mathbf{u} \otimes \mathbf{v}]_{i_1, \dots, i_p, j_1, \dots, j_q} = u_{i_1, \dots, i_p} v_{j_1, \dots, j_q}$$

- ▶ Associative, non-commutative and distributive w.r.t. +

$$(\mathbf{u} \otimes \mathbf{v}) \otimes \mathbf{w} = \mathbf{u} \otimes (\mathbf{v} \otimes \mathbf{w}) = \mathbf{u} \otimes \mathbf{v} \otimes \mathbf{w} \neq \mathbf{v} \otimes \mathbf{u} \otimes \mathbf{w}$$

$$(\mathbf{u}_1 + \mathbf{u}_2) \otimes \mathbf{v} = \mathbf{u}_1 \otimes \mathbf{v} + \mathbf{u}_2 \otimes \mathbf{v} = \mathbf{u}_1 \mathbf{v}^T + \mathbf{u}_2 \mathbf{v}^T$$



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## Elementary tensors

- ▶ Tensor products of vectors:  $\mathbf{u} \otimes \mathbf{v} \otimes \mathbf{w}$
- ▶ Every tensor can be written as a sum of elementary ones

$$\mathbf{x} = \sum_i \sum_j \sum_l x_{ijl} \mathbf{e}_i \otimes \mathbf{e}_j \otimes \mathbf{e}_l \quad (\text{trivial})$$





# Canonical polyadic (CP, tensor rank) decomposition

## Definition

- ▶ Minimal decomposition as a sum of elementary tensors

$$\mathcal{X} = \sum_{i=1}^r \mathbf{a}_i \otimes \mathbf{b}_i \otimes \mathbf{c}_i$$

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Example: diagonal tensor ( $m = n = d$ )

$$\mathcal{X} = \sum_{i=1}^n x_{iii} \mathbf{e}_i \otimes \mathbf{e}_i \otimes \mathbf{e}_i \quad \implies \quad \text{rank } \mathcal{X} = \# \text{ of nonzero } x_{iii}$$



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Example: rank-one tensor

$$\mathcal{X} = \mathbf{a} \otimes \mathbf{b} \otimes \mathbf{c}_1 + \mathbf{a} \otimes \mathbf{b} \otimes \mathbf{c}_2 = \mathbf{a} \otimes \mathbf{b} \otimes (\mathbf{c}_1 + \mathbf{c}_2) \quad \implies \quad \text{rank } \mathcal{X} = 1$$



# Multilinear transformation

## Definition

Given

$$\mathbf{P} \in \mathbb{R}^{m' \times m}, \quad \mathbf{U} \in \mathbb{R}^{n' \times n}, \quad \mathbf{T} \in \mathbb{R}^{d' \times d}$$
$$\mathcal{X} = \sum_i \mathbf{a}_i \otimes \mathbf{b}_i \otimes \mathbf{c}_i \quad \in \mathbb{R}^{m \times n \times d}$$

we define:

$$(\mathbf{P}, \mathbf{U}, \mathbf{T}) \cdot \mathcal{X} = \sum_i (\mathbf{P}\mathbf{a}_i) \otimes (\mathbf{U}\mathbf{b}_i) \otimes (\mathbf{T}\mathbf{c}_i) \in \mathbb{R}^{m' \times n' \times d'}$$



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- ▶ If  $(\mathbf{P}, \mathbf{U}, \mathbf{T}) \in \text{GL}_m \times \text{GL}_n \times \text{GL}_d$ : basis change (**preserves rank**)



# Block-term decomposition

- Generally, a sum of tensors (blocks) having low multilinear ranks

We focus on the particular form:

$$\mathbf{x} = \sum_{i=1}^k \mathbf{H}_i \otimes \mathbf{w}_i \quad \text{where} \quad \text{rank } \mathbf{H}_i \leq r_i$$



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- ▶ For  $\mathbf{w} \in \mathbb{R}^d$ :

$$\begin{aligned} \mathbf{H} \otimes \mathbf{w} &= \mathbf{H} \otimes \left( \sum_i w_i \mathbf{e}_i \right) = \sum_i (w_i \mathbf{H}) \otimes \mathbf{e}_i \\ &= \underbrace{\begin{pmatrix} w_1 h_{11} & \dots & w_1 h_{1n} & \dots & w_d h_{11} & \dots & w_d h_{1n} \\ \vdots & & \vdots & & \vdots & & \vdots \\ w_1 h_{m1} & \dots & w_1 h_{mn} & \dots & w_d h_{m1} & \dots & w_d h_{mn} \end{pmatrix}}_{d \text{ "frontal slices" }} \end{aligned}$$



# Approximate canonical polyadic decomposition

$\arg \min_{\mathbf{a}_i, \mathbf{b}_i, \mathbf{c}_i} \left\| \mathcal{Y} - \sum_{i=1}^k \mathbf{a}_i \otimes \mathbf{b}_i \otimes \mathbf{c}_i \right\|$  may be **empty** for a given rank  $k > 1$





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Famous example: sequence of rank-2 tensors  $\rightarrow$  rank-3 tensor

► If  $\{\mathbf{a}_1, \mathbf{a}_2\}, \{\mathbf{b}_1, \mathbf{b}_2\}, \{\mathbf{c}_1, \mathbf{c}_2\}$  are l.i. :

$$\text{rank}(\mathbf{a}_2 \otimes \mathbf{b}_1 \otimes \mathbf{c}_1 + \mathbf{a}_1 \otimes \mathbf{b}_2 \otimes \mathbf{c}_1 + \mathbf{a}_1 \otimes \mathbf{b}_1 \otimes \mathbf{c}_2) = 3$$



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► But:

$$\begin{aligned} \mathbf{x}_p &\triangleq p \left( \mathbf{a}_1 + \frac{1}{p} \mathbf{a}_2 \right) \otimes \left( \mathbf{b}_1 + \frac{1}{p} \mathbf{b}_2 \right) \otimes \left( \mathbf{c}_1 + \frac{1}{p} \mathbf{c}_2 \right) - p \mathbf{a}_1 \otimes \mathbf{b}_1 \otimes \mathbf{c}_1 \\ &= \mathbf{a}_2 \otimes \mathbf{b}_1 \otimes \mathbf{c}_1 + \mathbf{a}_1 \otimes \mathbf{b}_2 \otimes \mathbf{c}_1 + \mathbf{a}_1 \otimes \mathbf{b}_1 \otimes \mathbf{c}_2 + \frac{1}{p} \mathcal{R}_1 + \frac{1}{p^2} \mathcal{R}_2 \end{aligned}$$



# Approximate CPD: main ingredients

- ▶ Euclidean tensor norm:

- ▶  $\langle \mathbf{a}_1 \otimes \mathbf{b}_1 \otimes \mathbf{c}_1, \mathbf{a}_2 \otimes \mathbf{b}_2 \otimes \mathbf{c}_2 \rangle \triangleq \langle \mathbf{a}_1, \mathbf{a}_2 \rangle \langle \mathbf{b}_1, \mathbf{b}_2 \rangle \langle \mathbf{c}_1, \mathbf{c}_2 \rangle$

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- ▶ We need similar concepts in the BTD case





Summary

Tensors and tensor decompositions

Minimal ranks of BTDs and of matrix polynomials

Non-existence of best approximate BTB

Final remarks



# Sets of BTDs

- ▶ For  $r_1 \geq r_2 \geq \dots \geq r_d \geq 0$ , we define:

$$\mathcal{B}_{r_1, \dots, r_d} \triangleq \left\{ \sum_{i=1}^d \mathbf{H}_i \otimes \mathbf{w}_i \mid \text{rank}(\mathbf{H}_i) \leq r_i \text{ and } \mathbf{w}_1, \dots, \mathbf{w}_d \text{ are l.i.} \right\}$$



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Example:

$$\left( \sum_{i=1}^3 \mathbf{u}_i \otimes \mathbf{v}_i \right) \otimes \mathbf{d}_1 + \mathbf{x} \otimes \mathbf{y} \otimes \mathbf{d}_2 \in \mathcal{B}_{3,1} \subset \mathcal{B}_{3,2} \subset \mathcal{B}_{4,4}$$



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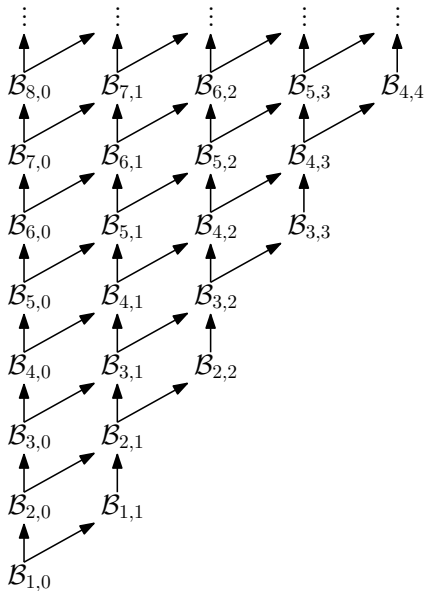
Richer than the hierarchy induced by tensor rank



# Hierarchy of BTD sets

Notation:

$$\mathcal{B} \rightarrow \mathcal{C} \equiv \mathcal{B} \subset \mathcal{C}$$



# Minimal ranks

- ▶ Need parallel to: “ $\mathcal{X}$  has rank  $r$ ”  $\Leftrightarrow \mathcal{X} \in \mathcal{S}_r \setminus \mathcal{S}_{r-1}$

Definition: Minimal ranks of  $m \times n \times d$  tensor  $\mathcal{X}$

$$\rho(\mathcal{X}) = (r_1, \dots, r_d) \quad \text{when} \quad \mathcal{X} \in \mathcal{B}_{s_1, \dots, s_d} \Leftrightarrow \forall i, s_i \geq r_i$$



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- ▶ Must be true for some  $(r_1, \dots, r_d)$



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$\rho(\mathcal{X})$  is  $GL_m \times GL_n \times GL_d$ -invariant

$$\mathcal{X} \in \mathcal{B}_{r_1, \dots, r_d} \Leftrightarrow (\mathbf{P}, \mathbf{U}, \mathbf{T}) \cdot \mathcal{X} \in \mathcal{B}_{r_1, \dots, r_d}$$



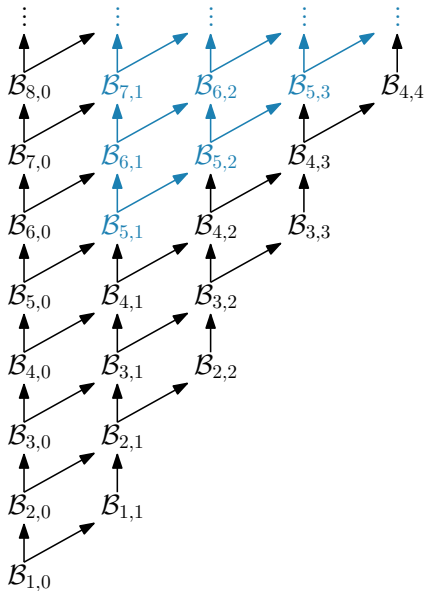


# Example of minimal ranks

$$\rho(\mathcal{X}) = (5, 1)$$

Notation:

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## Third-order tensors as matrix polynomials

- ▶ Consider the isomorphism  $\Phi : \mathbb{R}^{m \times n \times d} \rightarrow \mathcal{P}_{d-1}[\mathbb{R}^{m \times n}]$

$$\mathbf{X} = \sum_{i=1}^d \mathbf{X}_i \otimes \mathbf{e}_i \quad \mapsto \quad X(\lambda) = \sum_{i=1}^d \lambda^{i-1} \mathbf{X}_i$$

- ▶ Tensor slices are mapped to matrix coefficients
- ▶ In particular:  $\mathbf{A} \otimes \mathbf{e}_1 + \mathbf{B} \otimes \mathbf{e}_2 \quad \mapsto \quad \mathbf{A} + \lambda \mathbf{B}$



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### Minimal ranks of matrix polynomial

$$\rho(X) \triangleq \rho(\Phi^{-1}(X))$$



## Minimal ranks of matrix pencil ( $\simeq$ 2-block BTB)

- Basic idea: given  $\mathbf{A} + \lambda\mathbf{B}$ , find  $\mathbf{T} = \begin{bmatrix} t_{11} & t_{12} \\ t_{21} & t_{22} \end{bmatrix} \in \text{GL}_2$  that yields

$$\mathbf{A}' + \lambda\mathbf{B}' = (t_{11}\mathbf{A} + t_{12}\mathbf{B}) + \lambda(t_{21}\mathbf{A} + t_{22}\mathbf{B})$$

such that the ranks of  $\mathbf{A}'$  and  $\mathbf{B}'$  are minimal:

$$(t_{21}, t_{22}) = \arg \min_{(y,z)} \text{rank}(y\mathbf{A} + z\mathbf{B})$$

$$(t_{11}, t_{12}) = \arg \min_{(y,z) \neq \alpha(t_{21}, t_{22})} \text{rank}(y\mathbf{A} + z\mathbf{B})$$



# Kronecker canonical form of matrix pencil

- ▶  $P(\lambda) = \mathbf{A} + \lambda\mathbf{B} \sim P'(\lambda) = \mathbf{A}' + \lambda\mathbf{B}' \implies \rho(P) = \rho(P')$
- ▶ Easier to determine  $\rho(P')$

$$P'(\lambda) = \underbrace{S(\lambda)}_{\text{singular part}} \oplus \underbrace{R(\lambda)}_{\text{regular part}}$$

$$R(\lambda) = \underbrace{N(\lambda)}_{\text{infinite elementary divisors}} \oplus \underbrace{J(\lambda)}_{\text{finite elementary divisors}} \oplus \underbrace{Q(\lambda)}_{\text{finite elementary divisors } \neq \text{ powers of linear forms}}$$

- ▶ Legend: Can reduce rank / Cannot reduce rank



# Minimal ranks of pencils

Theorem (G. & C., 2017)

$$\rho(P) = \rho(\mathbf{A}, \mathbf{B}) = (r_0 + r', r_0 + s')$$

where:

- ▶  $r_0$  = sum of minimal indices = rank of singular part
- ▶  $s'$  = size( $R$ ) – largest number of elementary divisors of  $N \oplus J$  having a common factor
- ▶  $r'$  = size( $R$ ) – second largest number of elementary divisors of  $N \oplus J$  having a common factor



... and some examples

### Example 1

$$\mathbf{H} = \begin{bmatrix} a & 1 & 0 \\ 0 & a & 1 \\ 0 & 0 & a \end{bmatrix}, \quad \mathbf{H}' = \begin{bmatrix} a & 1 & 0 \\ 0 & a & 0 \\ 0 & 0 & a \end{bmatrix} \quad \text{and} \quad \mathbf{H}'' = \begin{bmatrix} a & 1 & 0 \\ 0 & a & 0 \\ 0 & 0 & b \end{bmatrix}$$

$$\rho(\mathbf{H}, \mathbf{E}) = (3, 2), \quad \rho(\mathbf{H}', \mathbf{E}) = (3, 1), \quad \rho(\mathbf{H}'', \mathbf{E}) = (2, 2)$$



## ... and some examples

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### Example 2

$$\lambda \mathbf{E} - \mathbf{Q} = \lambda \mathbf{E} - \begin{bmatrix} a & b \\ -b & a \end{bmatrix} \quad \text{has one elem. divisor: } (\lambda - a)^2 + b^2$$

$$\rho(-\mathbf{Q}, \mathbf{E}) = (2, 2) \text{ in } \mathbb{R} \quad \left( \text{but } \rho(-\mathbf{Q}, \mathbf{E}) = (1, 1) \text{ in } \mathbb{C}! \right)$$





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### Example 3

$$\rho \left( \begin{bmatrix} 1 & \lambda & 0 \\ 0 & 1 & 0 \\ 0 & 0 & \lambda + a \end{bmatrix} \right) = \rho \left( \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & a \end{bmatrix} + \lambda \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix} \right) = (2, 2)$$



# Spaces of real $n \times n$ pencils

## Corollary

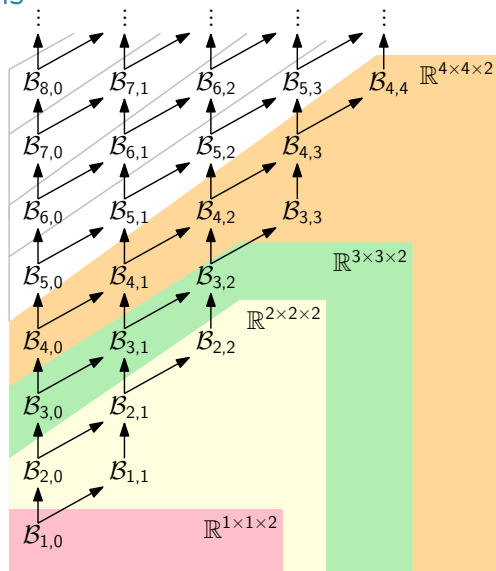
$X(\lambda)$  has full ranks

$\Leftrightarrow$

$X(\lambda) = Q(\lambda)$

$\Leftrightarrow$

$X(\lambda)$  is regular with  
all eigenvalues in  $\mathbb{C}$



Summary

Tensors and tensor decompositions

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## Generalization of celebrated CPD example

- ▶ Let  $\mathbf{A}, \mathbf{B} : m \times s$  and  $\mathbf{C}, \mathbf{D} : n \times s$  and consider

$$X(\lambda) = (\mathbf{AC}^T + \mathbf{BD}^T) + \lambda \mathbf{BC}^T$$



## Generalization of celebrated CPD example

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$$X(\lambda) = \left( \mathbf{AC}^T + \mathbf{BD}^T \right) + \lambda \mathbf{BC}^T$$

- ▶ Define the sequence of pencils  $X_p(\lambda) \in \mathcal{B}_{s,s}$  converging to  $X(\lambda)$ :

$$\begin{aligned} X_p(\lambda) &\triangleq p \left[ \left( \mathbf{B} + \frac{1}{p} \mathbf{A} \right) \left( \mathbf{C} + \frac{1}{p} \mathbf{D} \right)^T \right] \left( \mathbf{1} + \frac{1}{p} \lambda \right) - p \left( \mathbf{BC}^T \right) \\ &= X(\lambda) + \frac{1}{p} R_1(\lambda) + \frac{1}{p^2} R_2(\lambda) \end{aligned}$$



## Generalization of celebrated CPD example

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$$X(\lambda) = \left( \mathbf{A}\mathbf{C}^T + \mathbf{B}\mathbf{D}^T \right) + \lambda \mathbf{B}\mathbf{C}^T$$

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- ▶ If  $\min \{ \text{rank} [\mathbf{A} \ \mathbf{B}], \text{rank} [\mathbf{C} \ \mathbf{D}] \} > 3s/2$ , then  $X \notin \mathcal{B}_{s,s}$ .

$$\text{Therefore: } \arg \min_{Z(\lambda) \in \mathcal{B}_{s,s}} \|X(\lambda) - Z(\lambda)\| = \emptyset$$



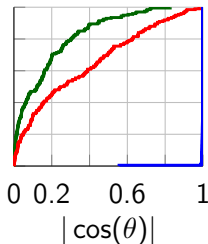
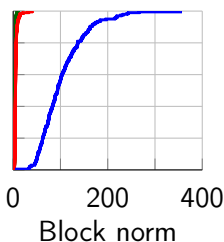
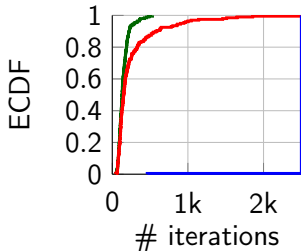
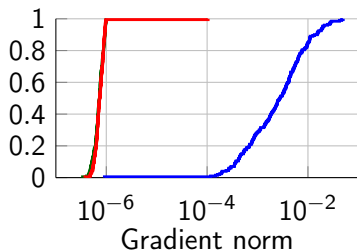
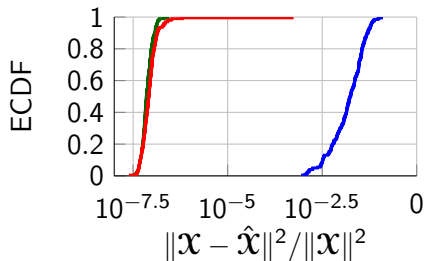
# Much trickier: does it happen for positive-volume sets?

Seems true: one can generate random examples where

- ▶ Optimization algorithm never seems to converge
- ▶ Keeps improving error but ultimately gives ill-conditioned estimate (nearly collinear blocks)
- ▶ Norms of these blocks “blow up,” but overall error stays bounded



Approx.  $\in \mathcal{B}_{3,3}^* \subset \mathbb{R}^{4 \times 4 \times 2}$  (50 init.)





# Main result

## Theorem (G. & C., 2017)

If  $\mathcal{X}$  is a real  $2k \times 2k \times 2$  tensor satisfying  $\rho(\mathcal{X}) = (2k, 2k)$ , then  $\mathcal{X}$  has no best approximation in  $\mathcal{B}_{2k-1, 2k-1}$ .



# Main result

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If  $\mathcal{X}$  is a real  $2k \times 2k \times 2$  tensor satisfying  $\rho(\mathcal{X}) = (2k, 2k)$ , then  $\mathcal{X}$  has no best approximation in  $\mathcal{B}_{2k-1, 2k-1}$ .

- ▶ Full minimal ranks  $\Leftrightarrow$  all eigenvalues are complex-valued
  - ▶ Open nonempty set  $\implies$  has positive volume
- ▶ Still true over  $\{\mathbf{H}_1 \otimes \mathbf{w}_1 + \mathbf{H}_2 \otimes \mathbf{w}_2 : \text{rank } \mathbf{H}_i \leq 2k - 1\}$



# Proof reasoning

- ▶  $\mathcal{C}_k = \mathcal{B}_{2k,2k} / \mathcal{B}_{2k,2k-1}$  (set of full-minimal rank tensors) is open
- ▶  $\implies \mathcal{B}_{2k,2k-1} = \mathbb{R}^{m \times n \times d} \setminus \mathcal{B}_{2k,2k}$  is closed
- ▶ Best approximation of  $\mathcal{X} \in \mathcal{C}_k$  in  $\mathcal{B}_{2k,2k-1}$  exists and has minimal ranks  $(2k, 2k - 1)$
- ▶ But tensors having minimal ranks  $(2k, 2k - 1)$  can be arbitrarily well approximated by tensors in  $\mathcal{B}_{2k-1,2k-1}$
- ▶  $\implies \mathcal{X}$  has no best approximation in  $\mathcal{B}_{2k-1,2k-1}$



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# Contributions and possible extensions

## Minimal ranks

- ▶ Study of BTD properties
- ▶ Coordinate-free: invariant under action of GL
- ▶ Connection with matrix pencils/matrix polynomials
- ▶ Computation of  $\rho$  for matrix pencils based on KCF
- ▶ What about matrix polynomials?

## Ill-posedness of best approximate BTD

- ▶ Extension of CPD degeneracy example
- ▶ Over  $\mathbb{R}^{2k \times 2k \times 2}$ , it happens for a positive-volume set



For more info, see [arXiv:1712.05742](https://arxiv.org/abs/1712.05742)

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