

Copositive optimization and completely positive matrix factorization

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Joint work with Patrick Groetzner

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First of all ...

Congratulations to
Manuel
and many happy returns!

Standard Quadratic Problem

We have:

$$x^T Ax = 0$$

$$\Leftrightarrow$$

$$\begin{aligned} \langle A, X \rangle &= 0 \\ X &= xx^T \end{aligned}$$

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$$\mathcal{CP} := \text{conv}\{x x^T : x \geq 0\}.$$

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Theorem (Bomze, D., et al. 2000)

StQP is equivalent to the following linear problem over CP:

$$\min \langle A, X \rangle \quad \text{s. t.} \quad \langle E, X \rangle = 1, \quad X \in \mathcal{CP}.$$

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More precisely:

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If X^* is optimal for cpStQP $\implies X^* = \sum_{i=1}^k x_i x_i^T$ with $x_i \in \mathbb{R}_+^n$,
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So: given $X \in \mathcal{CP}$, how to **find factorization** $X = BB^T$?



Factorizations are not unique

Example (Dickinson 2010)

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Note: in $A = BB^T$, the **number of columns of B** can vary!

The minimal possible number of columns is called the **cp-rank of A** :

$$\text{cprank}(A) = \inf\{r \in \mathbb{N} \mid \exists B \in \mathbb{R}^{n \times r}, B \geq 0, A = BB^T\}$$

Orthogonal matrices

Good news:

∃ upper bound cp_n such that

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Lemma (Bomze, Dickinson, Still 2015): For all $A \in \mathcal{CP}_n$ we have:

$$cprank(A) \leq cp_n = \begin{cases} n+1 & n \in \{2, 3, 4\} \\ \frac{1}{2}n(n+1) - 3 & n \geq 5 \end{cases}$$

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Fundamental Lemma for our approach:

Let $B, C \in \mathbb{R}^{n \times r}$. Then $BB^T = CC^T$ if and only if there exists an orthogonal matrix $Q \in \mathbb{R}^{r \times r}$ with $BQ = C$.



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- **Output:** $A = (BQ)(BQ)^T$ is a \mathcal{CP} factorization

Finding Q

Feasibility Problem:

find Q

s.t. $BQ \geq 0$

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\implies use **alternating projection method** between \mathcal{P} and \mathcal{O}_r .

Example

Given A , the eigen-decomposition gives an initial factorization

$$\underbrace{\begin{pmatrix} 4 & 1 & 1 & 1 & 1 \\ 1 & 2 & 1 & 1 & 1 \\ 1 & 1 & 2 & 1 & 1 \\ 1 & 1 & 1 & 2 & 1 \\ 1 & 1 & 1 & 1 & 2 \end{pmatrix}}_A = \underbrace{\begin{pmatrix} 0.0000 & 0.0000 & 0.0000 & -1.2308 & 1.5764 \\ 0.5413 & -0.5152 & 0.4377 & 0.4805 & 1.0950 \\ 0.0533 & 0.8062 & 0.3119 & 0.4805 & 1.0950 \\ -0.8108 & -0.2910 & -0.0890 & 0.4805 & 1.0950 \\ 0.2163 & 0.0000 & -0.8386 & 0.4805 & 1.0950 \end{pmatrix}}_{\tilde{B}} \cdot \underbrace{\begin{pmatrix} 0.0000 & 0.0000 & 0.0000 & -1.2308 & 1.5764 \\ 0.5413 & -0.5152 & 0.4377 & 0.4805 & 1.0950 \\ 0.0533 & 0.8062 & 0.3119 & 0.4805 & 1.0950 \\ -0.8108 & -0.2910 & -0.0890 & 0.4805 & 1.0950 \\ 0.2163 & 0.0000 & -0.8386 & 0.4805 & 1.0950 \end{pmatrix}^T}_{\tilde{B}^T}$$

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Column replication $A = BB^T$:

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Alternative: Add zero-columns \rightarrow numerically not so promising

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$$Q_0 = I \quad \longrightarrow \quad \text{our algorithm} \quad \longrightarrow \quad BQ_k$$



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$$Q_0 = I \quad \longrightarrow \quad \text{our algorithm} \quad \longrightarrow \quad BQ_k$$



Solution after 83 iterations and 25 milliseconds:

$$BQ_k = \begin{pmatrix} 0.0265 & 0.0000 & 0.0621 & 0.0000 & 0.0000 & 0.1058 & 0.0011 & 0.6029 & 0.0000 & 0.8386 & 1.7081 & 0.0000 \\ 0.0239 & 0.0000 & 0.1370 & 0.0000 & 0.0000 & 0.1175 & 0.0001 & 0.0000 & 0.0067 & 1.0477 & 0.0584 & 0.9304 \\ 0.0000 & 0.0719 & 0.3288 & 0.0000 & 0.0010 & 0.1472 & 0.0016 & 0.0000 & 1.1392 & 0.4069 & 0.3646 & 0.5184 \\ 0.0194 & 0.0000 & 0.9413 & 0.0015 & 0.1146 & 0.2921 & 0.0000 & 0.0000 & 0.0062 & 0.0073 & 0.5292 & 0.8573 \\ 0.1175 & 0.0000 & 0.6165 & 0.0000 & 0.0000 & 0.9972 & 0.1690 & 0.0274 & 0.2500 & 0.7074 & 0.1437 & 0.0506 \end{pmatrix} \geq 0$$

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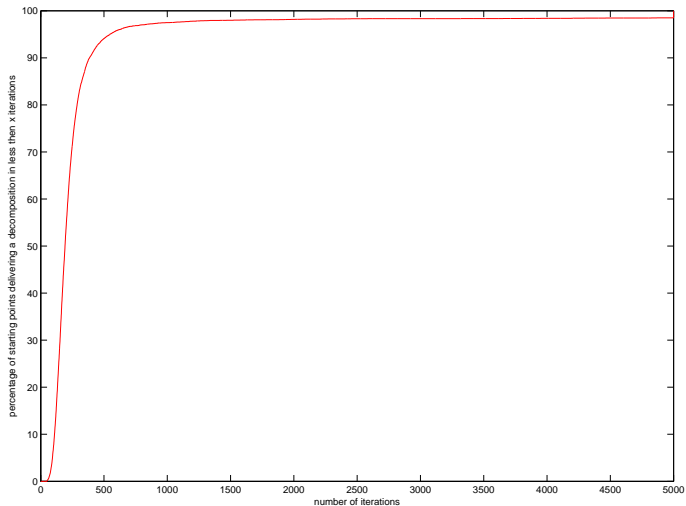


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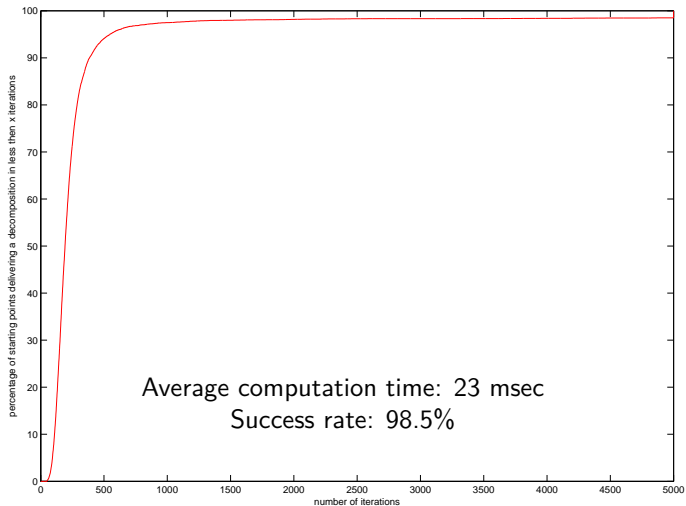
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Hence: $A = (BQ_k)(BQ_k)^T \in \mathcal{CP}_5$!

Performance with 10,000 different Q_0

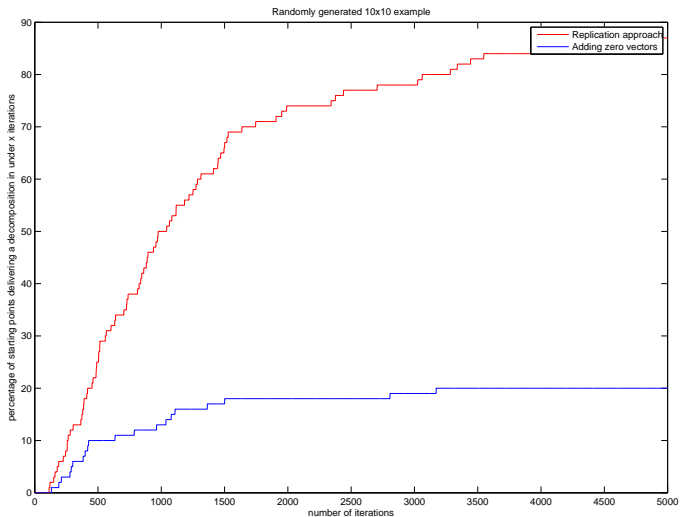


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Why not add zero-columns?



Results on randomly generated instances

- construct $R \in \mathbb{R}^{n \times r}$, $R \geq 0$
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See preprint:

P.Groetzner and M.Dür: A factorization method for completely positive matrices. *Preprint* 2018, available on OptimizationOnline

Thank you for your attention!

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