

*Khatri-Rao Products and Conditions for the
Uniqueness of PARAFAC Solutions for $I \times J \times K$
Arrays*

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Outline

- Column and Orthogonal Complement (OC) Spaces
- Uniqueness questions and OC Spaces
- Results from OC Spaces

Uniqueness

- \mathbf{X} is a 3-way array of order $I \times J \times K$
- PARAFAC decomposes the slices of the array as
$$\mathbf{X}_k = \mathbf{A} \mathbf{C}_k \mathbf{B}^t + \mathbf{E}_k$$
 - $\mathbf{A}_{I \times R}$; $\mathbf{B}_{J \times R}$; $\mathbf{C}_{K \times R}$
- Suppose there exists another decomposition
$$\mathbf{X}_k = \mathbf{G} \mathbf{D}_k \mathbf{H}^t + \mathbf{E}_k$$
 - $\mathbf{G}_{I \times R}$; $\mathbf{H}_{J \times R}$; $\mathbf{D}_{K \times R}$
- The decomposition is unique if every alternative satisfies: $\mathbf{G} = \mathbf{A} \mathbf{\Pi} \mathbf{\Lambda}_1$; $\mathbf{H} = \mathbf{B} \mathbf{\Pi} \mathbf{\Lambda}_2$; $\mathbf{D} = \mathbf{C} \mathbf{\Pi} \mathbf{\Lambda}_3$
 - $\mathbf{\Pi}$ is a permutation matrix and $\mathbf{\Lambda}_i$ is diagonal and $\mathbf{\Lambda}_1 \mathbf{\Lambda}_2 \mathbf{\Lambda}_3 = \mathbf{I}_R$

KR Products

$$\mathbf{A} = \begin{pmatrix} 1 & 0 & 0 & a_1 \\ 0 & 1 & 0 & a_2 \\ 0 & 0 & 1 & a_3 \end{pmatrix}$$

$$\mathbf{B} = \begin{pmatrix} 1 & 0 & 0 & b_1 \\ 0 & 1 & 0 & b_2 \\ 0 & 0 & 1 & b_3 \end{pmatrix}$$

$\mathbf{A} \circ \mathbf{B}$ 

$$\begin{pmatrix} 1 & 0 & 0 & a_1 b_1 \\ 0 & 0 & 0 & a_1 b_2 \\ 0 & 0 & 0 & a_1 b_3 \\ 0 & 0 & 0 & a_2 b_1 \\ 0 & 1 & 0 & a_2 b_2 \\ 0 & 0 & 0 & a_2 b_3 \\ 0 & 0 & 0 & a_3 b_1 \\ 0 & 0 & 0 & a_3 b_2 \\ 0 & 0 & 1 & a_3 b_3 \end{pmatrix}$$

ten Berge, J.M. and Sidiropoulos, N.D. 2002. On uniqueness in CANDECOMP/PARAFAC. *Psychometrika* 67: 399-409.

Summary

- The decomposition is considered without error
$$\mathbf{X} = (\mathbf{A} \circ \mathbf{B}) \mathbf{C}^t$$
- KR products are full column rank (Liu and Sidiropoulos, 2001)
- The KR product is a basis for the column space
- Is the KR product $(\mathbf{A} \circ \mathbf{B})$ the only KR product that generates the columns of \mathbf{X} ?
- The alternative is assumed to be $\mathbf{X} = (\mathbf{G} \circ \mathbf{H}) \mathbf{D}^t$
- The component matrices are investigated in reduced form (characterizing the k-rank and rank)

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Column Spaces

$$\mathbf{X} = (\mathbf{A} \circ \mathbf{B})\mathbf{C}'$$

- The KR product representation of the PARAFAC suggests that the columns of \mathbf{X} are linear combinations of the columns of the KR product

$$\mathbf{X} = \left[(\mathbf{A} \circ \mathbf{B})\mathbf{c}_1^t \quad (\mathbf{A} \circ \mathbf{B})\mathbf{c}_2^t \quad \cdots \quad (\mathbf{A} \circ \mathbf{B})\mathbf{c}_k^t \right]_{IJ \times K}$$

- So...the columns of \mathbf{X} are generated from the columns of the KR product $(\mathbf{A} \circ \mathbf{B})$

ten Berge, J.M. and Sidiropoulos, N.D. 2002. On uniqueness in CANDECOMP/PARAFAC. *Psychometrika* 67: 399-409.

Investigating Column Spaces

- Replacement
 - Used when the elements can be switched without changing the column space
- Transformation
 - Used when one of the loading matrices has full-column rank
- Finding a non-trivial alternative basis for the column space implies a non-unique decomposition
 - Once one is found, all decompositions with that KR product can also be considered non-unique

ten Berge, J.M. and Sidiropoulos, N.D. 2002. On uniqueness in CANDECOMP/PARAFAC. *Psychometrika* 67: 399-409.

Reduced Forms

Instead of considering loading matrices of the form

$$\begin{pmatrix} \alpha_1 & \beta_1 & \gamma_1 & a\alpha_1 + b\beta_1 \\ \alpha_2 & \beta_2 & \gamma_2 & a\alpha_2 + b\beta_2 \\ \alpha_3 & \beta_3 & \gamma_3 & a\alpha_3 + b\beta_3 \end{pmatrix}$$

loading matrices in reduced form were considered

$$\begin{pmatrix} 1 & 0 & 0 & a \\ 0 & 1 & 0 & b \\ 0 & 0 & 1 & 0 \end{pmatrix}$$

ten Berge, J.M. and Sidiropoulos, N.D. 2002. On uniqueness in CANDECOMP/PARAFAC. *Psychometrika* 67: 399-409.

Key Elements Learned

- The KR product allows us to talk about decompositions in column space language
- The properties of symmetry that applied for the “slab” notation also apply
- Matrices and the resulting KR products can be considered in “reduced form”.
- More to uniqueness than k-rank?

When k-rank didn't give the full story

- $R = 4$
- \mathbf{M}_1 : rank = 3 and k-rank = 2
- \mathbf{M}_2 : rank = 3 and k-rank = 2
- \mathbf{M}_3 : full column rank (rank = k-rank = 4)

$$\begin{pmatrix} 1 & 0 & 0 & x \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & z \end{pmatrix}$$

$$\begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & y \\ 0 & 0 & 1 & z \end{pmatrix}$$

$$\begin{pmatrix} 1 & 0 & 0 & x \\ 0 & 1 & 0 & y \\ 0 & 0 & 1 & 0 \end{pmatrix}$$

When k-rank didn't give the full story

$$\begin{pmatrix} 1 & 0 & 0 & X \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & X \\ 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & X \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & X \end{pmatrix}$$

Same position



$$\begin{pmatrix} 1 & 0 & 0 & X \\ 0 & 0 & 0 & X \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & X \\ 0 & 0 & 0 & X \\ 0 & 0 & 1 & 0 \end{pmatrix}$$

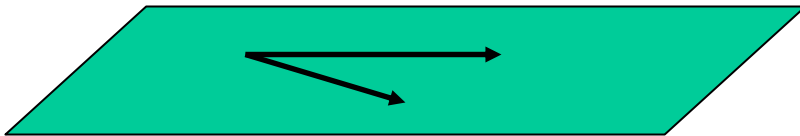
Different position

Orthogonal Complement Spaces (OCS)

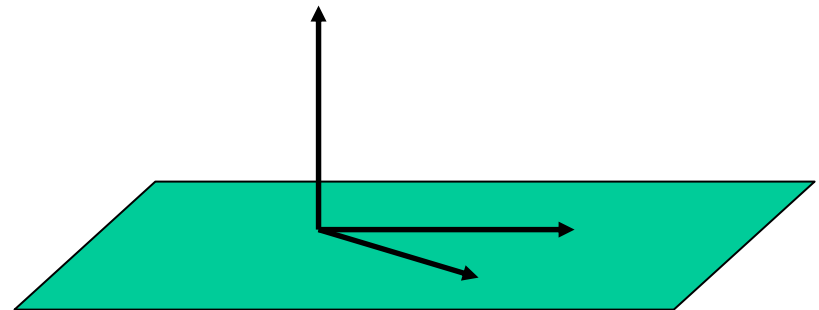
What's different about an OCS approach?

Column Spaces and Orthogonal Complement Spaces

Column Space Approach



Orthogonal Complement Space Approach



Finding OC Constraints: The Steps


1. Suppose you have $\mathbf{A}^\circ\mathbf{B}$
2. Assume that an alternative $\mathbf{A}^\circ\mathbf{B}$ exists, represented by $\mathbf{G}^\circ\mathbf{H}$
3. Find basis vectors for the null space of $(\mathbf{A}^\circ\mathbf{B})^t$
4. Take the inner product of the columns of the $\mathbf{G}^\circ\mathbf{H}$ and the null space basis vectors
5. Set equal to 0 (orthogonal) and solve

The constraints that result will determine if non-trivial alternatives are possible

OCSA: Finding Constraints

$$M_1 = \begin{pmatrix} 1 & 0 & 0 & a_1 \\ 0 & 1 & 0 & a_2 \\ 0 & 0 & 1 & 0 \end{pmatrix}$$

$$M_2 = \begin{pmatrix} 1 & 0 & 0 & b_1 \\ 0 & 1 & 0 & b_2 \\ 0 & 0 & 1 & 0 \end{pmatrix}$$

$M_1 \circ M_2$ 

$$\begin{pmatrix} 1 & 0 & 0 & a_1 b_1 \\ 0 & 0 & 0 & a_1 b_2 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & a_2 b_1 \\ 0 & 1 & 0 & a_2 b_2 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{pmatrix}$$


OCSA: Finding Constraints

$$\begin{bmatrix} 0 \\ -\frac{b_1 a_2}{b_2 a_1} \\ 0 \\ 1 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 1 \\ 0 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ 1 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 1 \\ 0 \end{bmatrix} \longrightarrow \frac{h_2 g_1}{h_1 g_2} = \frac{b_2 a_1}{b_1 a_2}$$

OCSA: Finding Constraints

$$M_1 = \begin{pmatrix} 1 & 0 & 0 & a_1 \\ 0 & 1 & 0 & a_2 \\ 0 & 0 & 1 & 0 \end{pmatrix}$$

$$M_2 = \begin{pmatrix} 1 & 0 & 0 & b_1 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & b_3 \end{pmatrix}$$

$M_1 \circ M_2$ 

$$\begin{pmatrix} 1 & 0 & 0 & a_1 b_1 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & a_1 b_3 \\ 0 & 0 & 0 & a_2 b_1 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & a_2 b_3 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{pmatrix}$$

OCSA: Finding Constraints

$$\left[\begin{array}{c} 0 \\ 0 \\ \frac{b_1 a_2}{b_3 a_1} \\ 1 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \end{array} \right], \left[\begin{array}{c} 0 \\ 0 \\ \frac{a_2}{a_1} \\ 0 \\ 0 \\ 1 \\ 0 \\ 0 \\ 0 \end{array} \right], \left[\begin{array}{c} 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 1 \\ 0 \\ 0 \end{array} \right], \left[\begin{array}{c} 0 \\ 0 \\ 0 \\ 1 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \end{array} \right], \left[\begin{array}{c} 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 1 \\ 0 \\ 0 \\ 0 \end{array} \right] \rightarrow \begin{array}{l} \underline{g_1} = \underline{a_1} \\ \underline{g_2} = \underline{a_2} \\ \underline{h_1} = \underline{b_1} \\ \underline{h_3} = \underline{b_3} \end{array}$$

OCSA: OC Constraints

Non-Separable Constraints

$$\frac{h_2 g_1}{h_1 g_2} = \frac{b_2 a_1}{b_1 a_2}$$

Non-trivial transformations

Separable Constraints

$$\frac{g_1}{g_2} = \frac{a_1}{a_2},$$
$$\frac{h_1}{h_3} = \frac{b_1}{b_3}$$

Trivial transformations

OCSA: Utilizing Symmetry

- If an alternative KR product can be found that is non-trivial transformation, then all PARAFAC decompositions with that KR product are non-unique
- A PARAFAC decomposition is unique if every alternative KR product can only be a trivial transformation.

OCSA: When $R = 4$

k-rank (M_1)	k-rank (M_2)	rank(M_1)	rank(M_2)	P/S Only?
2	2	2	2	No
2	2	2	3	No
2	2	3	3	Yes/No
2	3	2	3	No
2	3	3	3	Yes
2	4	2	4	Yes
2	4	3	4	Yes
3	3	3	3	Yes*
3	4	3	4	Yes

k-rank (M_1)	rank(M_1)	k-rank (M_2)	rank(M_2)	k-rank (M_3)	rank(M_3)	Uniqueness?
2	2	2	2	2	2	No
2	2	4	4	2	2	No
2	3	4	4	2	2	No
3	3	3	3	2	2	No
3	3	4	4	2	2	No
2	2	2	3	2	2	No
2	2	3	3	2	2	No
2	2	3	3	4	4	No
2	3	2	3	2	2	No
2	3	3	3	2	2	No
2	3	2	3	2	3	No/Yes
2	3	3	3	2	3	No/Yes
2	3	4	4	2	3	No/Yes
3	3	3	3	2	3	Yes
3	3	3	3	3	3	Yes*
3	3	4	4	2	3	Yes

OCSA: Conclusions for $R = 4$

- Necessary and Sufficient Conditions for uniqueness (when $k\text{-rank} = \text{rank}$)
 - A PARAFAC decomposition is unique if and only if $r(\mathbf{M}_i) + r(\mathbf{M}_j) \geq R + 2$, for all $i \neq j$
- Conditions for uniqueness
 - If two of the matrices have $k\text{-rank} < \text{rank}$, you will need to look at the number of OC constraints
 - Otherwise, A PARAFAC decomposition is unique if and only if $r(\mathbf{M}_i) + r(\mathbf{M}_j) \geq R + 2$, for all $i \neq j$

OCSA: Conclusions

- The OCSA provided a method for determining if alternative KR products could have non-trivial transformations
- The OC constraints offered an explanation of uniqueness when k-rank couldn't.
- Based on the OC constraints, it was possible to determine if PARAFAC decompositions were unique
- Being able to “look” at decomposition uniqueness provided necessary and sufficient conditions for uniqueness for $R = 4, 5, \text{ and } 6$ (k-rank = rank)

Discussion

- The tools we have...
 - KR products allow the use of linear algebra
 - Simplification allows for a better “view” of the decomposition
 - The OCSA provides a straightforward approach for determining if decompositions are unique for any R
- We need...
 - A better “tool” for determining which alternatives are truly alternatives
 - With more information on the decompositions that are unique, it will be possible to provide further empirical evidence for what causes uniqueness
- It looks like the key to defining necessary and sufficient conditions could rest with orthogonal complement spaces

Selected References

- Harshman, R.A. 1972. Determination and proof of minimum uniqueness conditions for PARAFAC1. *UCLA Working Papers in Phonetics* 22: 111-117.
- Kruskal, J.B. 1977. Three-way arrays: Rank and uniqueness of trilinear decompositions with application to arithmetic complexity and statistics. *Linear Algebra Appl.* 18: 95-138.
- Bro, R. 1998. Multi-way Analysis in the Food Industry. Models, algorithms and Applications, Ph.D. Thesis, University of Amsterdam, The Netherlands.
- ten Berge, J.M. and Sidiropoulos, N.D. 2002. On uniqueness in CANDECOP/PARAFAC. *Psychometrika* 67: 399-409.

Extra Slides

OC Spaces and Column Spaces

- Column Space of : $\mathcal{C}(\mathbf{A}^\circ\mathbf{B})$
- Null Space of $(\mathbf{A}^\circ\mathbf{B})^t$: $\mathcal{N}((\mathbf{A}^\circ\mathbf{B})^t)$
- ◊ $\mathcal{C}(\mathbf{A}^\circ\mathbf{B}) = [\mathcal{N}((\mathbf{A}^\circ\mathbf{B})^t)]^\perp$
- All vectors orthogonal to $[\mathcal{N}((\mathbf{A}^\circ\mathbf{B})^t)]$ are elements of $\mathcal{C}(\mathbf{A}^\circ\mathbf{B})$
- So...we need to find an alternative KR product that has columns that are orthogonal to $[\mathcal{N}((\mathbf{A}^\circ\mathbf{B})^t)]$

Sample MAPLE Code

with(LinearAlgebra):

```
A:=Matrix(3,4,[[1,0,0,a[1]],[0,1,0,0],[0,0,1,a[3]]]);
```

```
B:=Matrix(3,4,[[1,0,0,b[1]],[0,1,0,0],[0,0,1,b[3]]]);
```

```
AB:=Matrix(9,4,[convert((OuterProductMatrix(B[1..3,1],A[1..3,1])),Vector),convert((OuterProductMatrix(B[1..3,2],A[1..3,2])),Vector),convert((OuterProductMatrix(B[1..3,3],A[1..3,3])),Vector),convert((OuterProductMatrix(B[1..3,4],A[1..3,4])),Vector)]);
```

```
G:=Matrix(3,4,[[1,0,0,g[1]],[0,1,0,g[2]],[0,0,1,g[3]]]);
```

```
H:=Matrix(3,4,[[1,0,0,h[1]],[0,1,0,h[2]],[0,0,1,h[3]]]);
```

```
GH:=Matrix(9,4,[convert((OuterProductMatrix(H[1..3,1],G[1..3,1])),Vector),convert((OuterProductMatrix(H[1..3,2],G[1..3,2])),Vector),convert((OuterProductMatrix(H[1..3,3],G[1..3,3])),Vector),convert((OuterProductMatrix(H[1..3,4],G[1..3,4])),Vector)]);
```

```
NAB:=NullSpace(Transpose(AB));
```

```
eqns:={Multiply(Transpose(NAB[1]),GH[1..9,4])=0,Multiply(Transpose(NAB[2]),GH[1..9,4])=0,Multiply(Transpose(NAB[3]),GH[1..9,4])=0,Multiply(Transpose(NAB[4]),GH[1..9,4])=0,Multiply(Transpose(NAB[5]),GH[1..9,4])=0,a[1]<>0,a[2]<>0,a[3]<>0,b[1]<>0,b[2]<>0,b[3]<>0};
```

```
solve(eqns);
```

OCSA: When $R = 5$ (k -rank = rank)

k-rank (M1)	k-rank (M2)	rank(M1)	rank(M2)	P/S Only?
2	2	2	2	No
2	3	2	3	No
2	4	2	4	No
2	5	2	5	Yes
3	3	3	3	No
3	4	3	4	Yes
3	5	3	5	Yes
4	4	4	4	Yes
4	5	4	5	Yes

k-rank (M1)	rank(M1)	k-rank (M2)	rank(M2)	k-rank (M3)	rank(M3)	Uniqueness?
2	2	2	2	2	2	No
2	2	3	3	2	2	No
2	2	4	4	2	2	No
2	2	5	5	2	2	No
3	3	3	3	2	2	No
3	3	4	4	2	2	No
3	3	5	5	2	2	No
4	4	4	4	2	2	No
4	4	5	5	2	2	No
2	2	5	5	3	3	No
3	3	3	3	3	3	No
3	3	4	4	3	3	No
3	3	5	5	3	3	No
4	4	4	4	3	3	Yes

OCSA: When $R = 6$ ($k\text{-rank} = \text{rank}$)

k-rank (M1)	k-rank (M2)	rank(M1)	rank(M2)	P/S Only?
2	2	2	2	No
2	3	2	3	No
2	4	2	4	No
2	5	2	5	No
2	6	2	6	Yes
3	3	3	3	No
3	4	3	4	No
3	5	3	5	Yes
3	6	3	6	Yes
4	4	4	4	Yes
4	5	4	5	Yes
4	6	4	6	Yes
5	5	5	5	Yes
5	6	5	6	Yes

k-rank (M1)	rank(M1)	k-rank (M2)	rank(M2)	k-rank (M3)	rank(M3)	Uniqueness?
2	2	2	2	2	2	N
2	2	3	3	2	2	N
2	2	4	4	2	2	N
2	2	5	5	2	2	N
2	2	6	6	2	2	N
3	3	3	3	2	2	N
3	3	4	4	2	2	N
3	3	5	5	2	2	N
3	3	6	6	2	2	N
4	4	4	4	2	2	N
4	4	5	5	2	2	N
4	4	6	6	2	2	N
5	5	5	5	2	2	N
5	5	6	6	2	2	N
3	3	3	3	3	3	N
3	3	4	4	3	3	N
3	3	5	5	3	3	N
3	3	6	6	3	3	N
4	4	4	4	3	3	N
4	4	5	5	3	3	N
4	4	6	6	3	3	N
5	5	5	5	3	3	Y
4	4	4	4	4	4	Y
4	4	5	5	4	4	Y

Kruskal and Pairwise Combinations

- $k_A + k_B + k_C \geq 2R + 2$
- $k_C \leq R$

$$\begin{aligned}k_A + k_B &\geq 2R + 2 - k_C \\ &\geq 2R + 2 - R \\ &= R + 2\end{aligned}$$

$$r_A + r_B \geq 2R + 2$$