

Algorithms and Uniqueness of Tensor Decompositions

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What is a matrix?



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A **matrix** is an element of $\mathbb{F}^{d_x} \otimes \mathbb{F}^{d_y}$ $\begin{bmatrix} 2 & 4 \\ 6 & 12 \end{bmatrix} \in \mathbb{F}^2 \otimes \mathbb{F}^2$

A **rank-one** matrix is a matrix of the form $x \otimes y = xy^T = (x_i y_j)_{(i,j)}$

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What is a ~~matrix~~ tensor?



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A ~~matrix~~ **tensor** is an element of $\mathbb{F}^{d_x} \otimes \mathbb{F}^{d_y} \otimes \mathbb{F}^{d_z}$

$$\begin{bmatrix} 15 & 18 \\ 20 & 24 \end{bmatrix} \begin{bmatrix} 30 & 36 \\ 40 & 48 \end{bmatrix} \in \mathbb{F}^2 \otimes \mathbb{F}^2 \otimes \mathbb{F}^2$$

A **product tensor** is a tensor of the form $x \otimes y \otimes z = (x_i y_j z_k)_{(i,j,k)}$

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$$\begin{bmatrix} 15 & 18 \\ 20 & 24 \end{bmatrix} \begin{bmatrix} 30 & 36 \\ 40 & 48 \end{bmatrix} = \begin{pmatrix} 1 \\ 2 \end{pmatrix} \otimes \begin{pmatrix} 3 \\ 4 \end{pmatrix} \otimes \begin{pmatrix} 5 \\ 6 \end{pmatrix}$$

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Tensor decompositions

Definition: Let $n \in \mathbb{N}$ and $[n] := \{1, \dots, n\}$.



For $T \in \mathbb{F}^{d_x} \otimes \mathbb{F}^{d_y} \otimes \mathbb{F}^{d_z}$, an expression $T = \sum_{a \in [n]} x_a \otimes y_a \otimes z_a \in \mathbb{F}^{d_x} \otimes \mathbb{F}^{d_y} \otimes \mathbb{F}^{d_z}$ is called a **decomposition** of T into product tensors

$\text{rank}(T) := \text{smallest } n$

Uniqueness of tensor decompositions

Definition: Let $n \in \mathbb{N}$ and $[n] := \{1, \dots, n\}$.



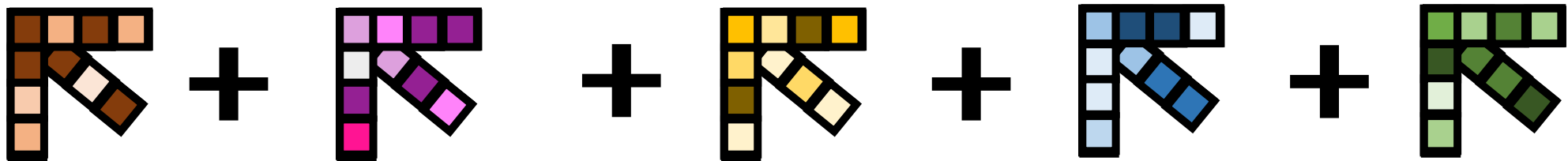
A rank decomposition

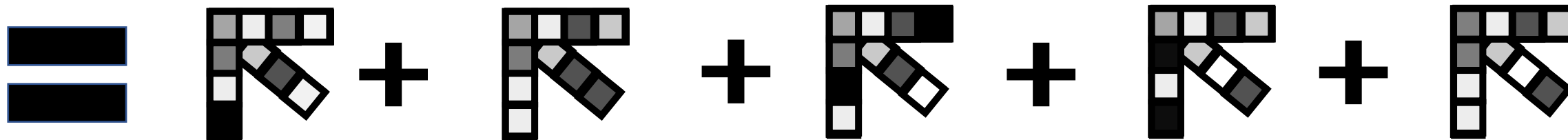
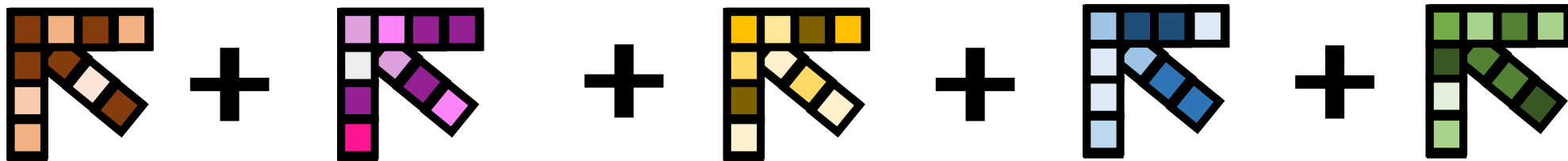
$$T = \sum_{a \in [n]} x_a \otimes y_a \otimes z_a \in \mathbb{F}^{d_x} \otimes \mathbb{F}^{d_y} \otimes \mathbb{F}^{d_z}$$

is called the **unique (rank) decomposition** of T if for any other decomposition

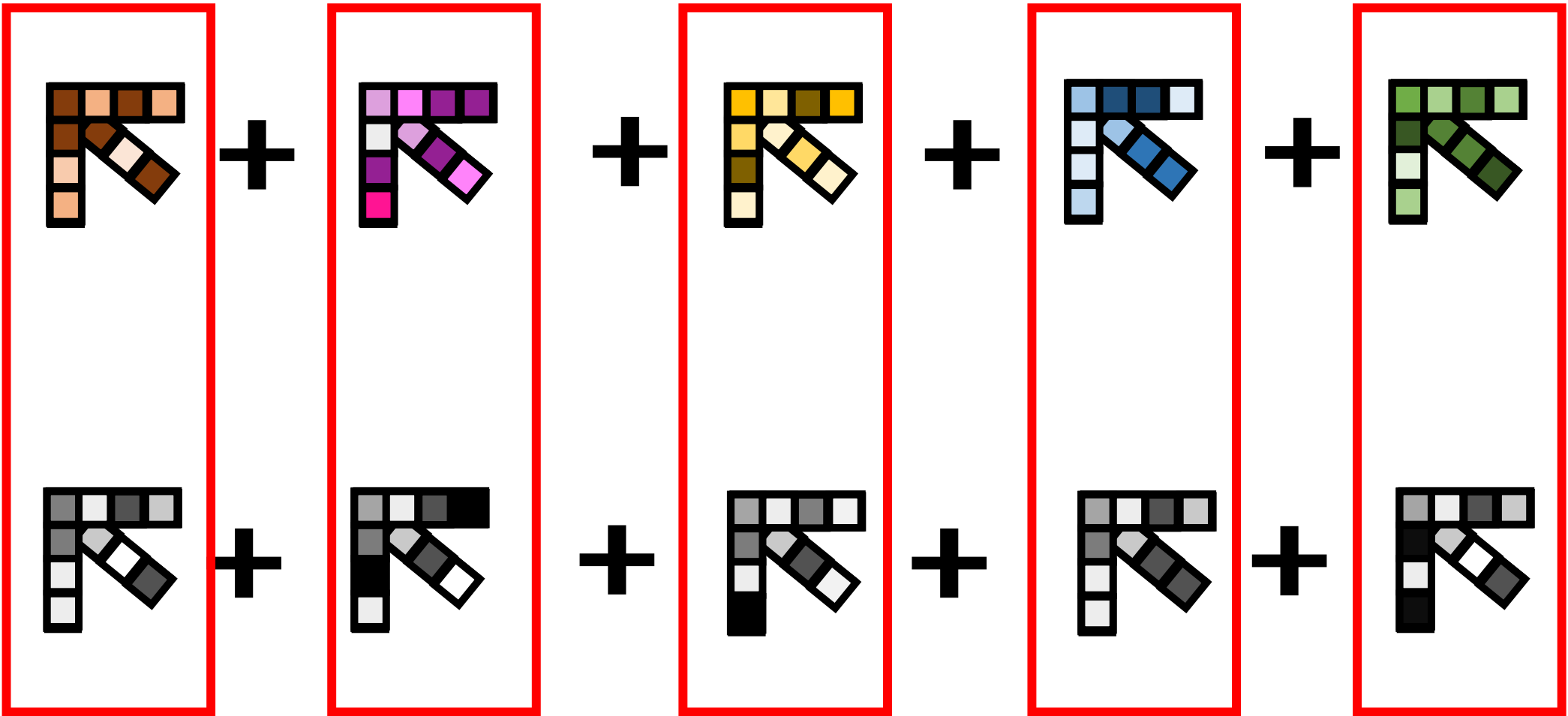
$$T = \sum_{a \in [n]} x'_a \otimes y'_a \otimes z'_a \in \mathbb{F}^{d_x} \otimes \mathbb{F}^{d_y} \otimes \mathbb{F}^{d_z}$$

there is a permutation $\sigma \in S_n$ such that $x_a \otimes y_a \otimes z_a = x'_{\sigma(a)} \otimes y'_{\sigma(a)} \otimes z'_{\sigma(a)}$ for all $a \in [n]$.

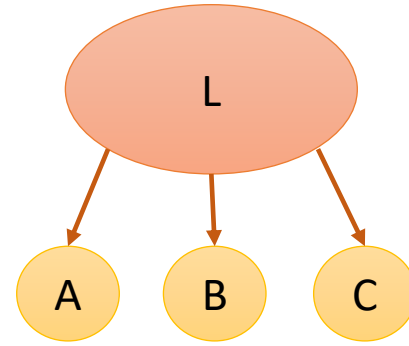




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Application: Latent parameter learning



- Let A, B, C, L be finite random variables such that A, B, C are conditionally independent, i.e.

$$\Pr(a, b, c|l) = \Pr(a|l) \Pr(b|l) \Pr(c|l) \quad \text{for all } a, b, c, l.$$

- Goal: Given the probability vector $\Pr(A, B, C)$, determine $\Pr(A, B, C, L)$.
- Method:

$$\Pr(A, B, C) = \sum_l \Pr(l) \Pr(A, B, C|l) = \sum_l \underbrace{\Pr(l) \Pr(A|l) \otimes \Pr(B|l) \otimes \Pr(C|l)}$$

... If $\Pr(A, B, C)$ has a unique decomposition, then we can recover $\Pr(A, B, C, l)$,

- Applications: Learning mixtures of spherical gaussians, phylogenetic tree reconstruction, hidden Markov models, orbit retrieval, blind signal separation, document topic models, ...

Two goals

$$T = \sum_{a \in [n]} x_a \otimes y_a \otimes z_a \in \mathbb{F}^{d_x} \otimes \mathbb{F}^{d_y} \otimes \mathbb{F}^{d_z} \quad (1)$$

1. Algorithms

[JLV 2023, published in FOCS]

Given a tensor $T \in \mathbb{F}^{d_x} \otimes \mathbb{F}^{d_y} \otimes \mathbb{F}^{d_z}$, find a rank decomposition (1).

2. Uniqueness

[LP 2023, published in FoM Sigma]

Given a rank decomposition (1), prove that it is the unique rank decomposition.

Two goals

$$T = \sum_{a \in [n]} x_a \otimes y_a \otimes z_a \in \mathbb{F}^{d_x} \otimes \mathbb{F}^{d_y} \otimes \mathbb{F}^{d_z} \quad (1)$$

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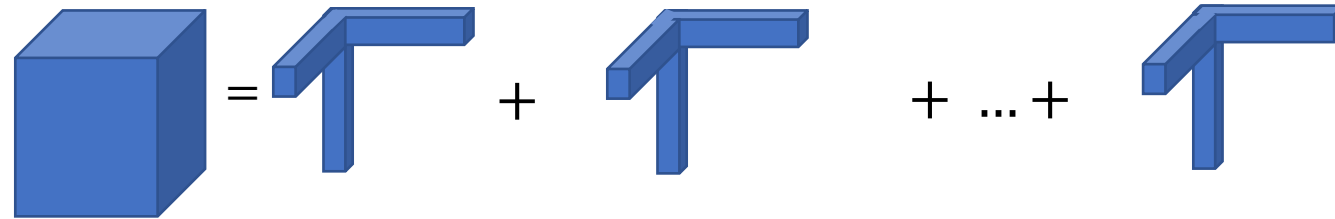
Given a rank decomposition (1), prove that it is the unique rank decomposition.

Algorithm idea

Tensor: $T \in \mathbb{F}^d \otimes \mathbb{F}^d \otimes \mathbb{F}^k$

Decomposition: Sum of R product tensors

$$T = x_1 \otimes y_1 \otimes z_1 + x_2 \otimes y_2 \otimes z_2 + \cdots + x_n \otimes y_n \otimes z_n$$



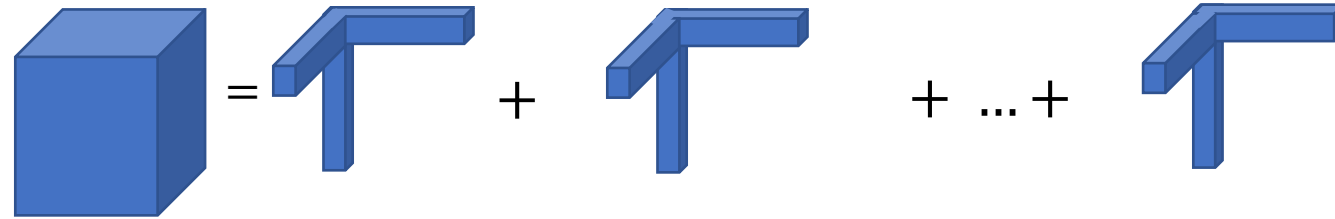
Idea: If we view T as an $d^2 \times k$ matrix, then the image is in the span of the $x_i \otimes y_i$.

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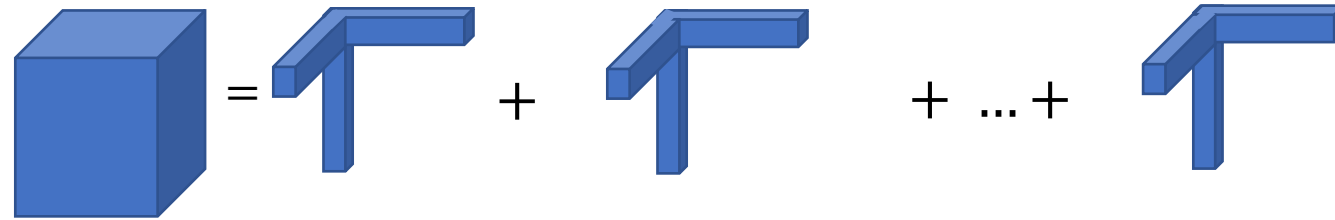
- Finding rank-one matrices in $\text{im}(T)$ \leftrightarrow Finding tensor decompositions of T

Algorithm idea

Tensor: $T \in \mathbb{F}^d \otimes \mathbb{F}^d \otimes \mathbb{F}^k$

Decomposition: Sum of R product tensors

$$T = x_1 \otimes y_1 \otimes z_1 + x_2 \otimes y_2 \otimes z_2 + \cdots + x_n \otimes y_n \otimes z_n$$



Idea: If we view T as an $d^2 \times k$ matrix, then the image is in the span of the $x_i \otimes y_i$.

- Finding rank-one matrices in $\text{im}(T)$ \leftrightarrow Finding tensor decompositions of T
- Finding other types of matrices in $\text{im}(T)$ \leftrightarrow Finding other types of decompositions of T

(X, k) -decompositions

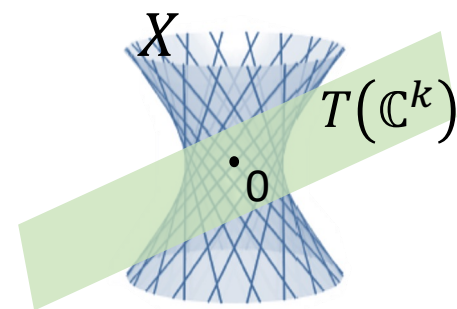
For $T \in \mathbb{F}^D \otimes \mathbb{F}^k$, $X \subseteq \mathbb{C}^D$,

an (X, k) -decomposition is an expression
$$T = \sum_{i=1}^n v_i \otimes z_i \in \mathbb{F}^D \otimes \mathbb{F}^k$$

where $v_1, \dots, v_R \in X$

Example: When $X = X_1 = \{\text{rank 1 matrices}\} \subseteq \mathbb{F}^d \otimes \mathbb{F}^d$, an (X, k) -decomposition is just a tensor decomposition.

Viewing T as a map $\mathbb{F}^k \rightarrow \mathbb{F}^D$, each $v_i \in T(\mathbb{F}^k) \cap X$,
so computing $T(\mathbb{F}^k) \cap X \leftrightarrow (X, k)$ -decomposing T



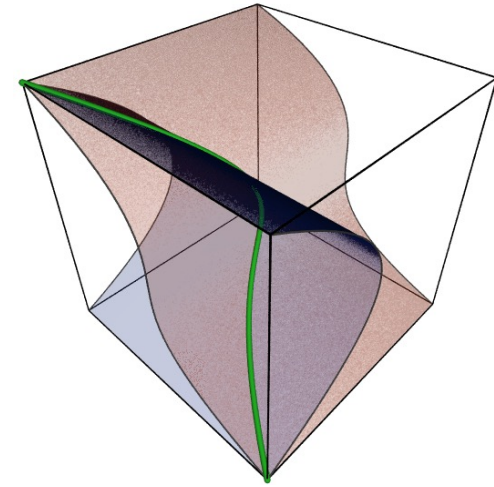
Theorem (informal) [JLV 2023]: For many algebraic varieties X ,
we can recover low-rank (X, k) -decompositions efficiently.

Algebraic Varieties

Variety: common zeroes of a set of polynomials

$$X = \{x \in \mathbb{F}^D : f_1(x) = \cdots = f_p(x) = 0\}$$

- f_1, f_2, \dots, f_p cut out the variety X



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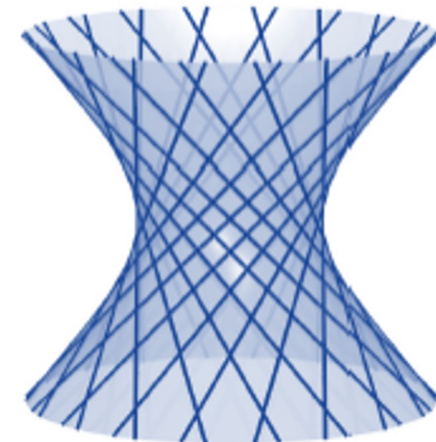
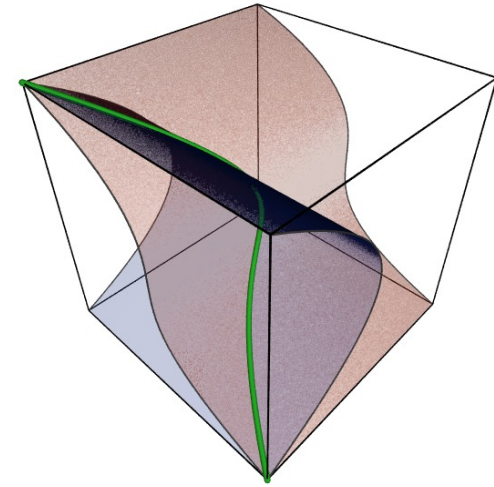
$$X = \{x \in \mathbb{F}^D : f_1(x) = \cdots = f_p(x) = 0\}$$

- f_1, f_2, \dots, f_p cut out the variety X

$X \subseteq \mathbb{F}^D$ is a (conic) variety iff

$$v \in X \implies \forall \lambda \in \mathbb{F}, \lambda v \in X$$

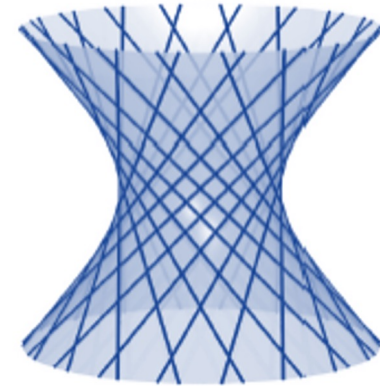
- Conic variety: f_1, f_2, \dots, f_p can be homogenous of same degree ℓ



Running example: rank-1 matrices

$$X_1 = \{u_1 \otimes u_2 \mid u_1 \in \mathbb{F}^{d_1}, u_2 \in \mathbb{F}^{d_2}\}$$

$u_1 \otimes u_2 = u_1 u_2^T$ is vector outer product.

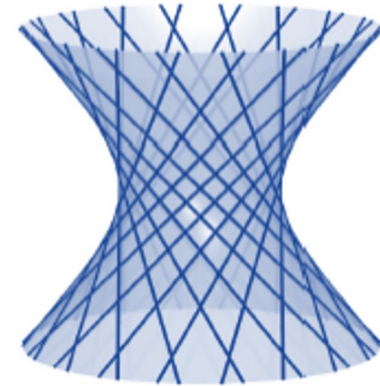


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- $X_1 \subset \mathbb{F}^{d_1 \times d_2}$ is a conic variety cut out by degree-2 polynomials

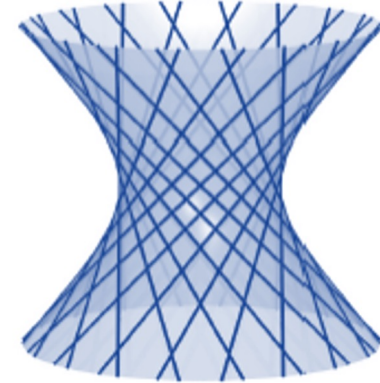


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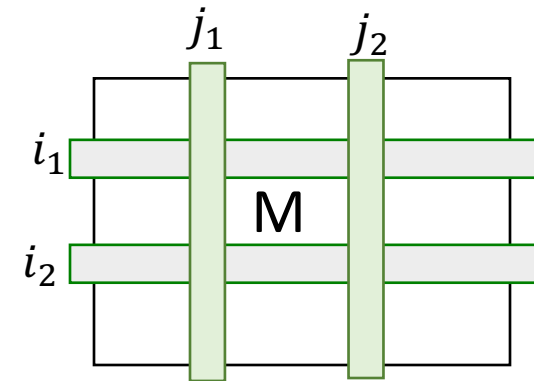
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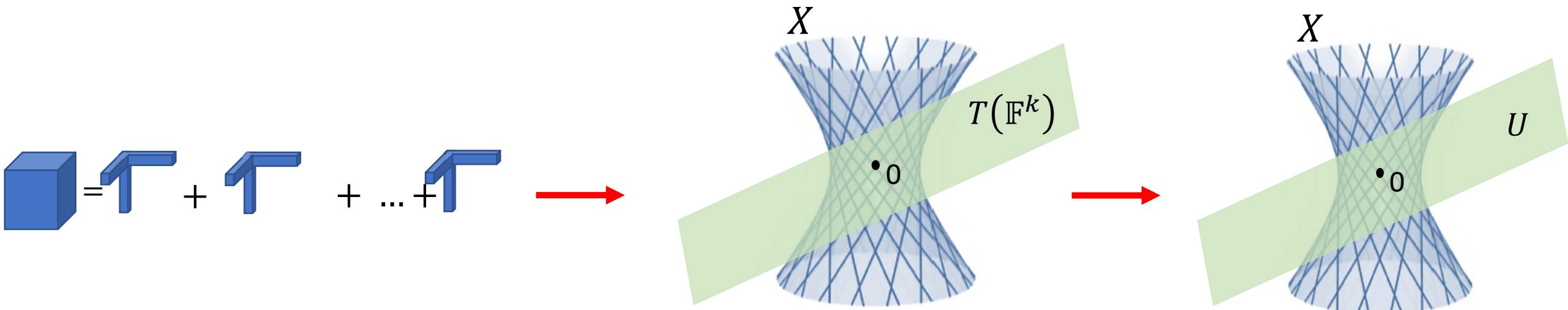
$$M \in X_1 \text{ iff}$$
$$\begin{aligned} \forall 1 \leq i_1 < j_1 \leq n_1, \\ \forall 1 \leq i_2 < j_2 \leq n_2, \end{aligned} \quad M_{i_1 j_1} M_{i_2 j_2} - M_{i_1 j_2} M_{i_2 j_1} = 0$$



$$\det(2 \times 2 \text{ submatrix}) = 0$$

- X_1 cut out by $\binom{n_1}{2} \binom{n_2}{2}$ homogenous degree 2 polynomials

Reduction: computing (X, k) -decompositions \longrightarrow computing linear sections



Algorithm for (X, k) -decompositions \longrightarrow Algorithm to compute $U \cap X$ for linear subspace $U = T(\mathbb{F}^k)$

Suppose $U \subseteq \mathbb{F}^D$ has a basis $\{v_1, \dots, v_n\}$ such that each $v_i \in X$.

Problem: Given some other basis $\{u_1, \dots, u_n\}$ of U , recover $\{v_1, \dots, v_n\}$ (up to scale).

Example: Jennrich's Algorithm: If $U' = \text{span} \{v_1^{\otimes \ell}, \dots, v_n^{\otimes \ell}\}$ with $\{v_1, \dots, v_n\}$ linearly independent, then $\{v_1^{\otimes \ell}, \dots, v_n^{\otimes \ell}\}$ can be recovered from any basis of U' in $D^{O(\ell)}$ -time.

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Jennrich's Algorithm:

Pick $T_j \in U'$, $j = 1, 2$ at random, view these as maps $T_j: (\mathbb{F}^D)^{\otimes \ell-1} \rightarrow \mathbb{F}^D$

$$T_j = \sum_{i=1}^n \alpha_{j,i} v_i (v_i^t)^{\otimes \ell-1} \quad T_j^{-1} = \sum_i \frac{1}{\alpha_{j,i}} (w_i)^{\otimes \ell-1} w_i^t \quad \text{where } w_i^t v_{i'} = \delta_{i,i'}$$

So $T_1 T_2^{-1} = \sum_i \frac{\alpha_{1,i}}{\alpha_{2,i}} v_i w_i^t$. E-vectors / E-values of $T_1 T_2^{-1}$ are v_i , $\frac{\alpha_{1,i}}{\alpha_{2,i}}$  Distinct for different i

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Lifted Jennrich's Algorithm [JLV 23, DLCC 07]: Run Jennrich on $U' = U^{\otimes \ell} \cap X^\ell$, where $X^\ell = \text{span}\{v^{\otimes \ell} : v \in X\}$.

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Example [JLV 23]: If $U \subseteq \mathbb{F}^d \otimes \mathbb{F}^d$ is spanned by $n \leq \frac{1}{4}(d-1)^2$ generic product tensors, then these can be recovered from any basis of U in $\text{poly}(d)$ -time.

Corollary [JLV 23]: A generic tensor

$T \in \mathbb{F}^d \otimes \mathbb{F}^d \otimes \mathbb{F}^{d^2}$ with

$$\text{rank}(T) \leq \frac{1}{4}(d-1)^2$$

has a unique rank decomposition, that can be recovered in $\text{poly}(d)$ -time by applying lifted Jennrich to $\text{im}(T)$.

- Maximum possible rank up to constant
- Quadratic improvement over Jennrich's algorithm, which can handle rank $O(d)$.

Corollary [JLV 23]: A generic tensor

$T \in (\mathbb{F}^d)^{\otimes m}$ of tensor rank

$$\text{rank}(T) = O(d^{\lfloor m/2 \rfloor})$$

has a unique tensor rank decomposition, which is recovered in $n^{O(m)}$ -time by applying our algorithm to $T \left((\mathbb{F}^d)^{\otimes \lfloor m/2 \rfloor} \right)$.

- r -aided rank: $T = \sum_i v_i \otimes w_i$, where $v_i \in \text{rank} - r$ matrices
- Applications in signal processing and machine learning [Comon, Jutten 2010]

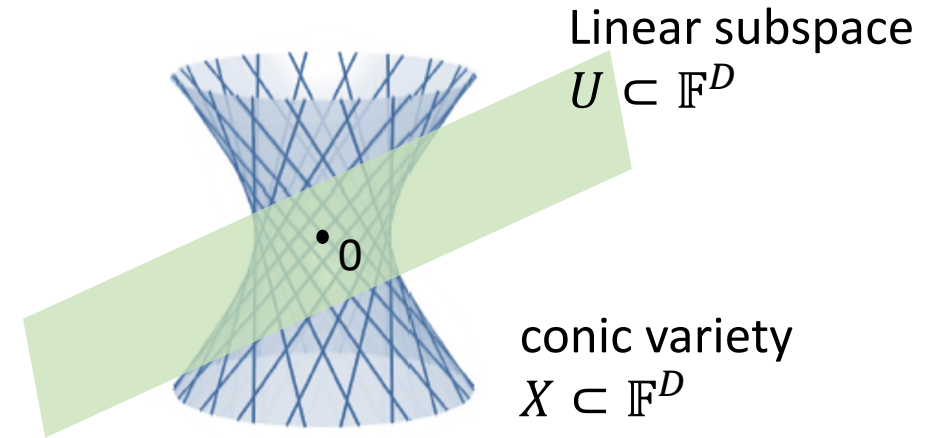
Corollary [JLV 23]: A generic tensor $T \in \mathbb{F}^d \otimes \mathbb{F}^d \otimes \mathbb{F}^k$
of r -aided rank

$$r\text{-aided rank}(T) \leq \min\{\Omega_r(d^2), k\}$$

has a unique r -aided rank decomposition, which is recovered in $d^{O(r)}$ -time by applying our algorithm to $T(\mathbb{F}^k)$.

Algorithm: Takeaways

- Natural algorithmic problem
- Captures wide array of decomposition problems
- NP-hard even for $X = \text{rank-one matrices}$

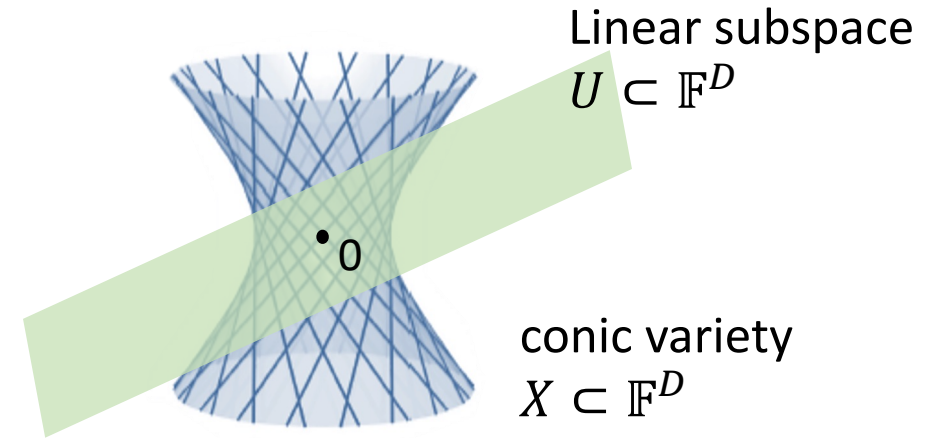


Aim: Compute intersection of variety X and linear subspace U

Main Result: Can design polynomial time algorithm if U is generic and $\dim(U)$ is not too large

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Aim: Compute intersection of variety X and linear subspace U

Main Result: Can design polynomial time algorithm if U is generic and $\dim(U)$ is not too large

Future Directions:

- New applications for different choices of varieties?
- Robust versions of the statement?
- Using algebraic geometry ideas for other algorithmic problems

Two goals

$$T = \sum_{a \in [n]} x_a \otimes y_a \otimes z_a \in \mathbb{F}^{d_x} \otimes \mathbb{F}^{d_y} \otimes \mathbb{F}^{d_z} \quad (1)$$

1. Algorithms

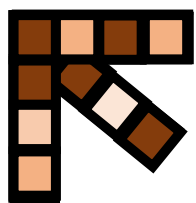
[JLV 2023, published in FOCS]

Given a tensor $T \in \mathbb{F}^{d_x} \otimes \mathbb{F}^{d_y} \otimes \mathbb{F}^{d_z}$, find a rank decomposition (1).

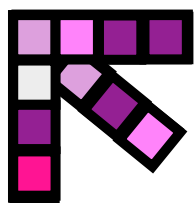
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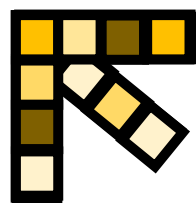
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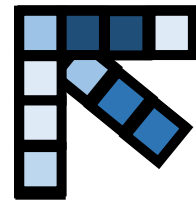
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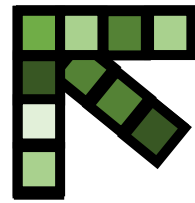
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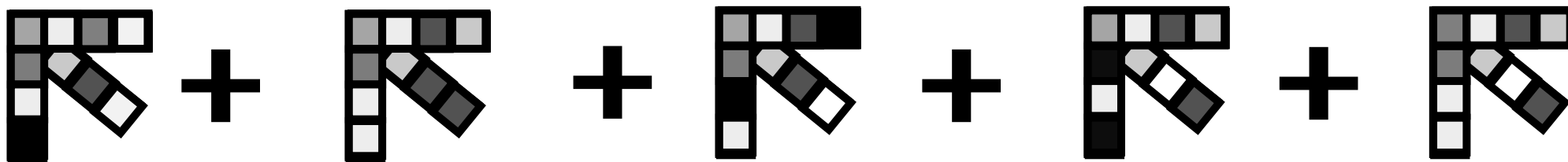
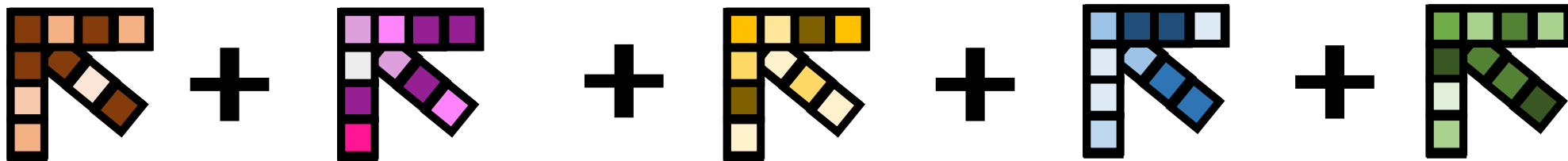


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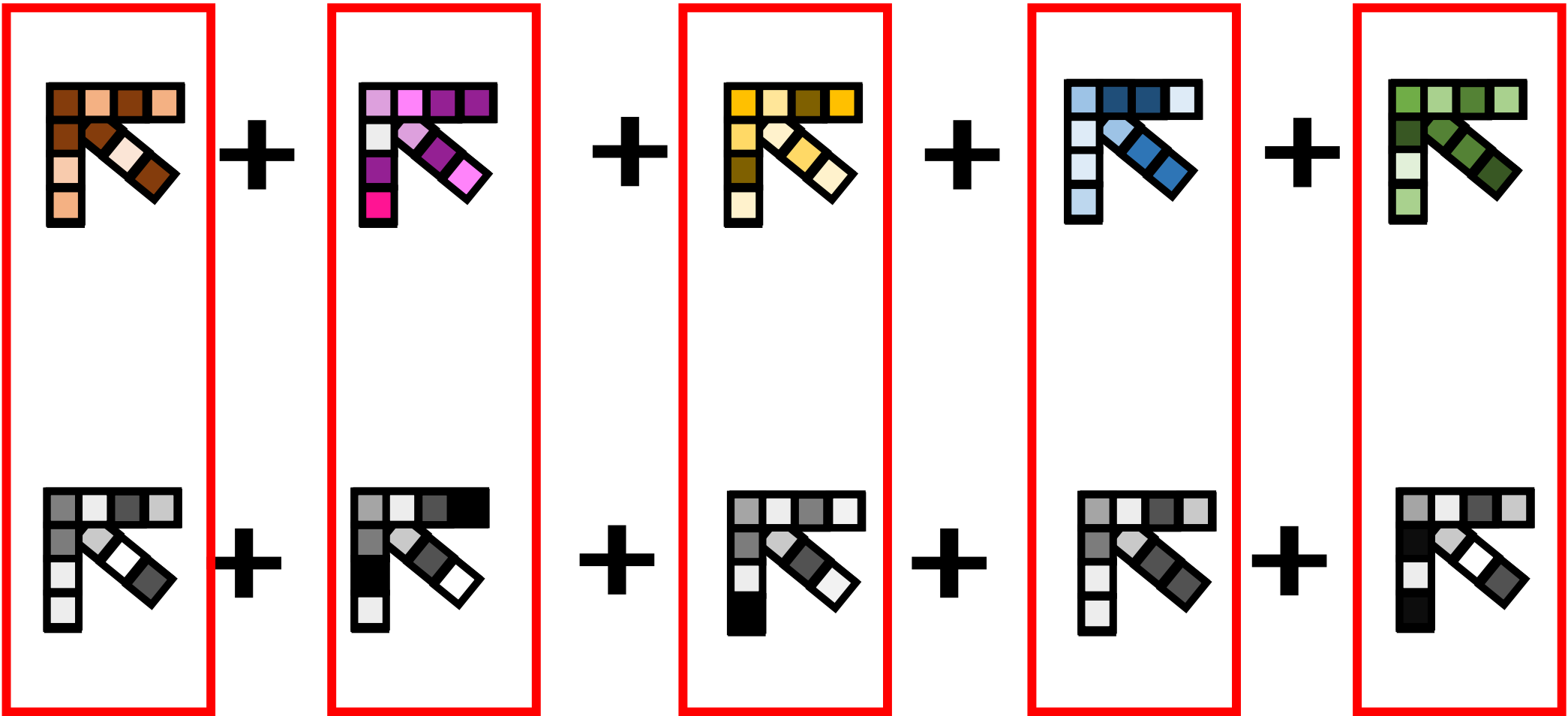


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Uniqueness

Jennrich's Uniqueness Theorem: Given a rank decomposition

$$T = \sum_{a \in [d]} x_a \otimes y_a \otimes z_a \in \mathbb{F}^d \otimes \mathbb{F}^d \otimes \mathbb{F}^2 \quad (1)$$

If it holds that

1. $\{x_1, \dots, x_d\} \subseteq \mathbb{F}^d$ is linearly independent,
2. $\{y_1, \dots, y_d\} \subseteq \mathbb{F}^d$ is linearly independent,
3. and $\{z_1, \dots, z_d\} \subseteq \mathbb{F}^2$ are non-parallel

then (1) is the unique rank decomposition of T .

Uniqueness

Jennrich's Uniqueness Theorem: Given a rank decomposition

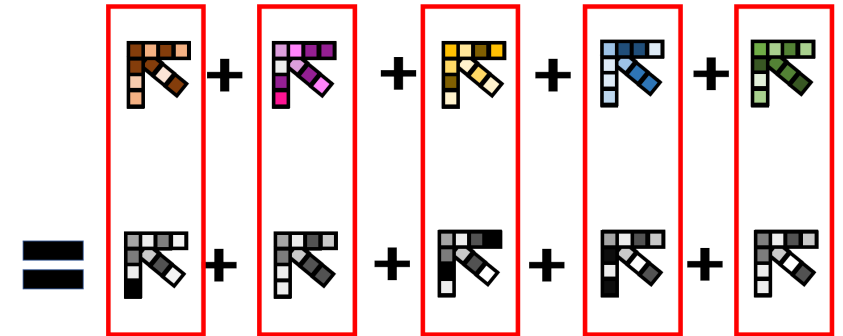
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1. $\{x_1, \dots, x_d\} \subseteq \mathbb{F}^d$ is linearly independent,
2. $\{y_1, \dots, y_d\} \subseteq \mathbb{F}^d$ is linearly independent,
3. and $\{z_1, \dots, z_d\} \subseteq \mathbb{F}^2$ are non-parallel

then (1) is the unique rank decomposition of T .

Jennrich's Algorithm: Finds the decomposition (1) efficiently!



Kruskal's theorem

$$T = \sum_{a \in [n]} x_a \otimes y_a \otimes z_a \in \mathbb{F}^{d_x} \otimes \mathbb{F}^{d_y} \otimes \mathbb{F}^{d_z} \quad (1)$$

Definition: The **Kruskal rank** of $\{x_1, \dots, x_n\} \in \mathbb{F}^{d_x}$ is the largest integer k_x such that every subset $S \subseteq \{x_1, \dots, x_n\}$ of size $|S| = k_x$ is linearly independent.

Kruskal rank is NP-Hard!

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$$T = e_1^{\otimes 3} + e_2^{\otimes 3} + e_3^{\otimes 3} + e_4^{\otimes 3} + (e_1 + e_2) \otimes (e_2 + e_3) \otimes (e_1 + e_4)$$

$$\begin{array}{ccccccc} x_1 \otimes y_1 \otimes z_1 & & \dots & & x_5 & \otimes & y_5 & \otimes & z_5 \end{array}$$

$$\{x_1, \dots, x_5\} = \{e_1, e_2, e_3, e_4, e_1 + e_2\}, \quad k_x = 2.$$

Kruskal's theorem

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Matroid theory for product tensors

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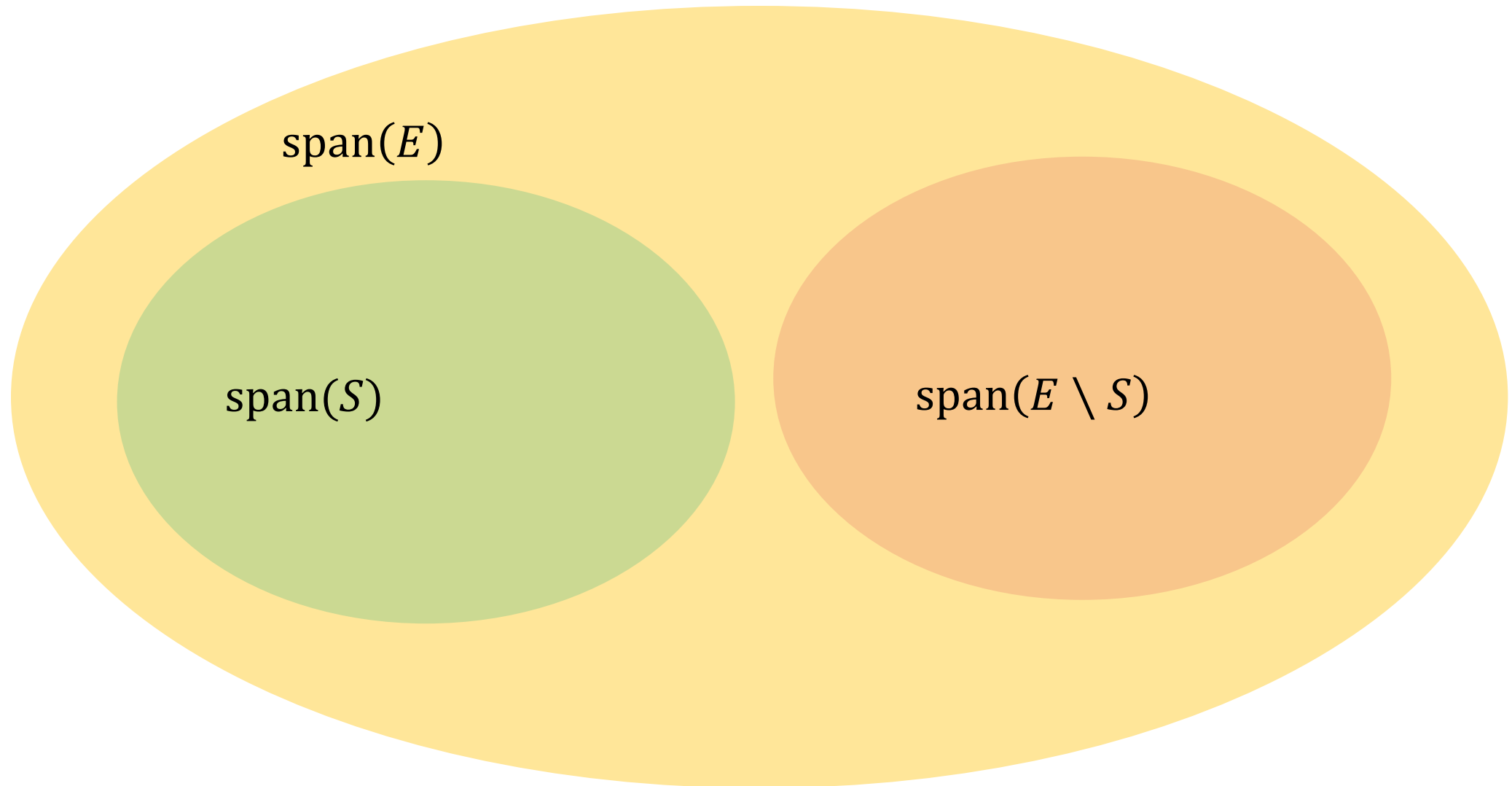
Rest of talk: A splitting theorem for product tensors

Splitting

Definition: A set of vectors $E = \{v_1, \dots, v_n\}$ **splits** if there exists a non-trivial subset $S \subseteq E$ such that

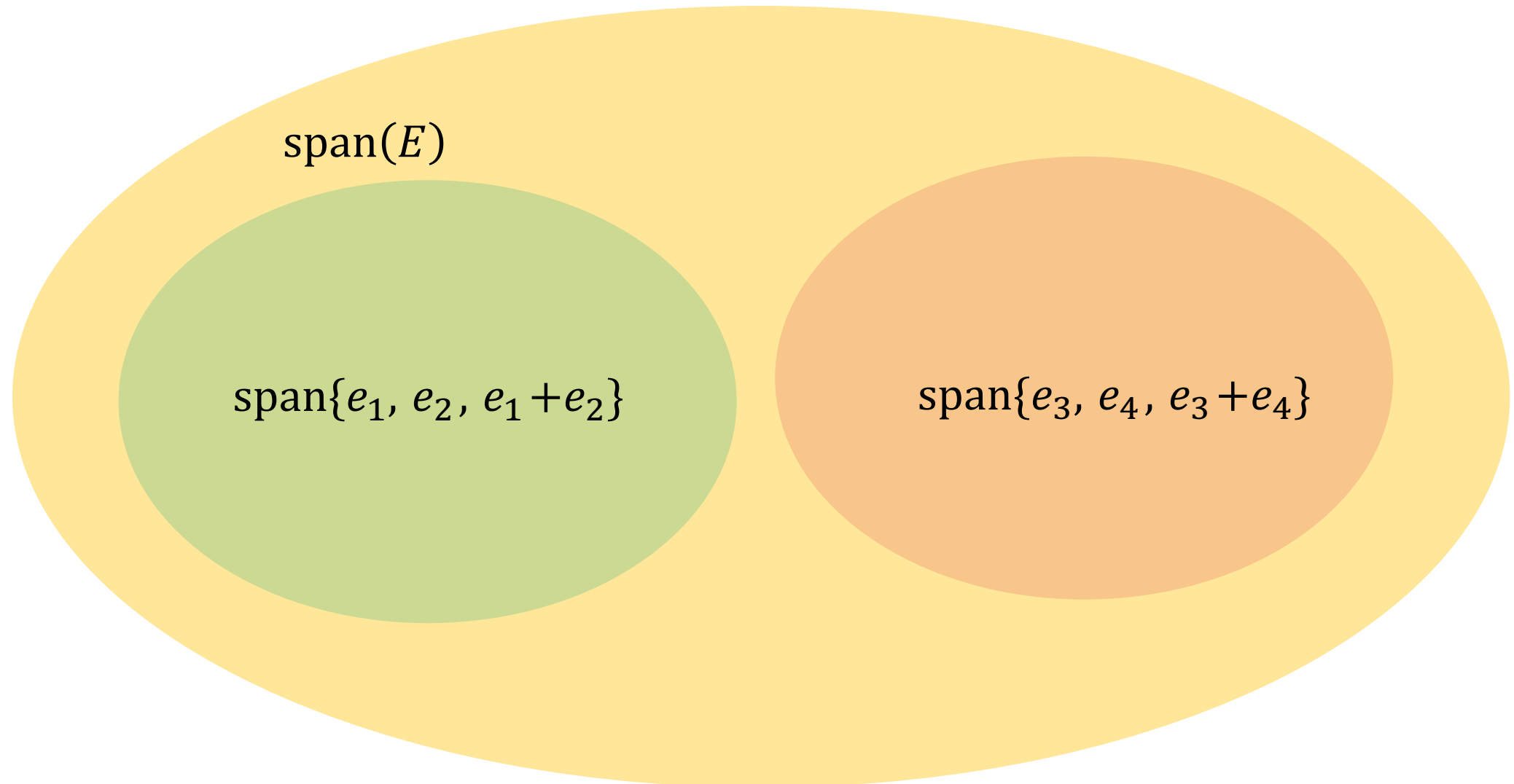
$$\text{span}(S) \cap \text{span}(E \setminus S) = \{0\} \quad (2)$$

E **splits** if there exists $S \subseteq E$ such that
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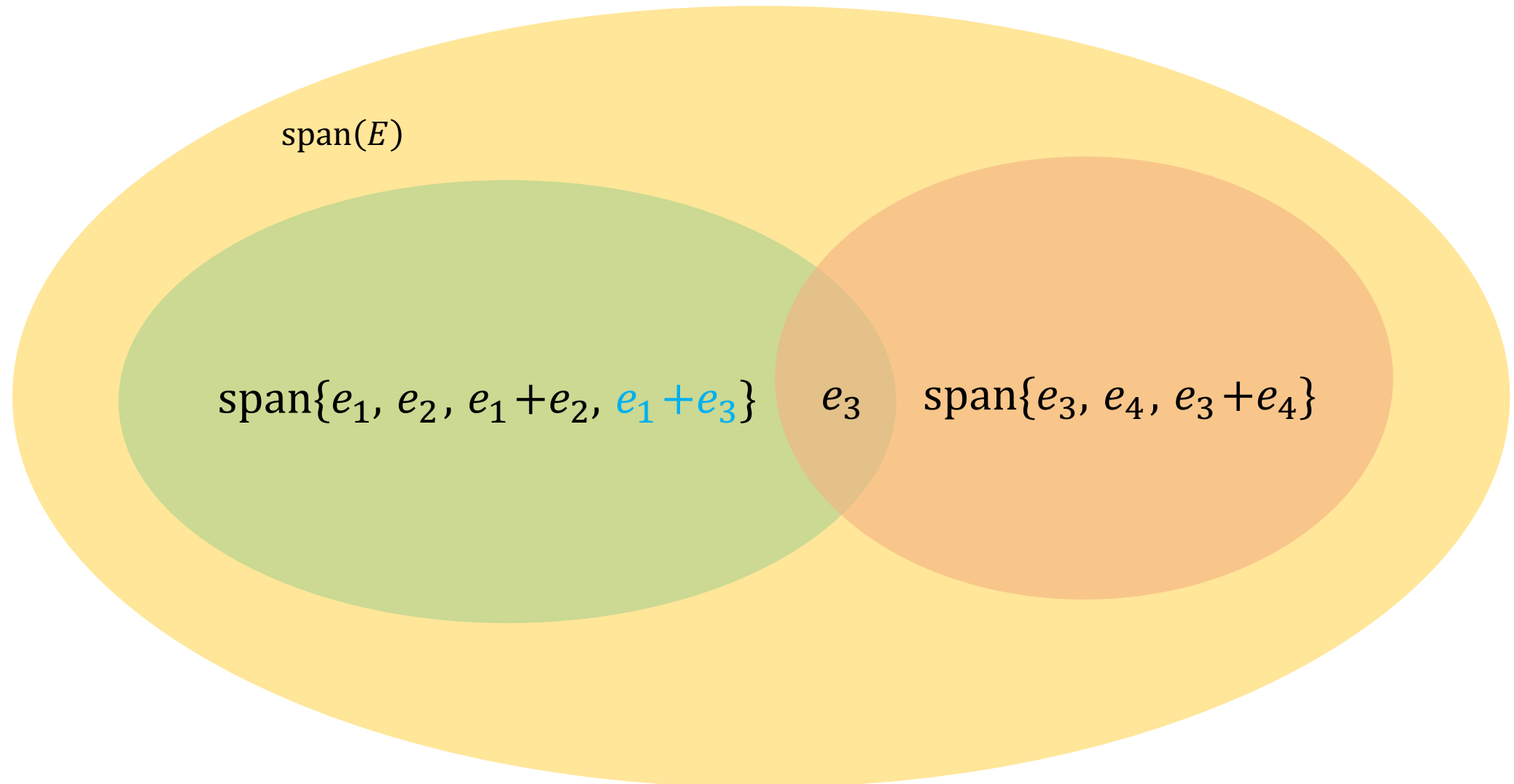
$$E = \{e_1, e_2, e_1 + e_2, e_3, e_4, e_3 + e_4\}$$

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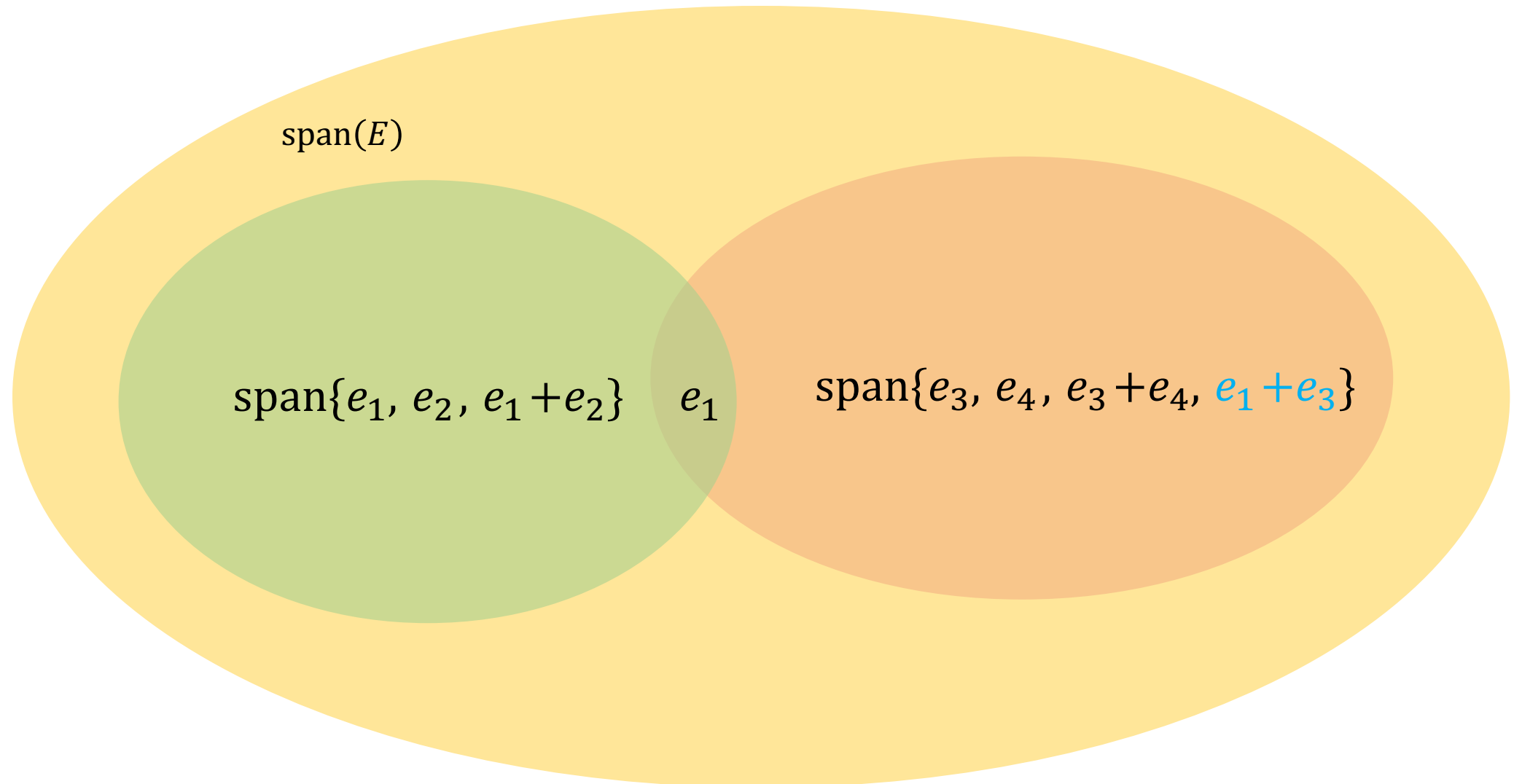
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Fact: If (2) holds, and $\sum(E) = 0$, then $\sum(S) = \sum(E \setminus S) = 0$

Proof: $\sum(E) = 0$

$$\Rightarrow \sum(S) = -\sum(E \setminus S) \in \text{span}(S) \cap \text{span}(E \setminus S) = \{0\}$$



Splitting theorem

E **splits** if there exists $S \subseteq E$ such that $\text{span}(S) \cap \text{span}(E \setminus S) = \{0\}$


Splitting theorem [LP 2023]: Let $E = \{x_a \otimes y_a : a \in [n]\}$.

If

$$\text{dimspan}(E) \leq d_x^{[n]} + d_y^{[n]} - 2$$

then E splits.


$d_x^{[n]} = \text{dimspan}\{x_1, \dots, x_n\}$



A generalization of Kruskal's theorem on tensor decomposition

Part of: Basic linear algebra Designs and configurations

Published online by Cambridge University Press: 05 April 2023

Benjamin Lovitz  and Fedor Petrov 

Forum of Mathematics, Sigma

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More matroid theory for product tensors?

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Splitting theorem \Rightarrow Corollary: If E is linearly independent, then it splits. Otherwise,

$$\text{dimspan}(E) \leq n - 1 \leq d_x^{[n]} + d_y^{[n]} + d_z^{[n]} - 3, \quad \text{so } E \text{ splits by splitting theorem.}$$



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Corollary: If

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Replaces Kruskal ranks with standard ranks

Corollary: If $2n \leq d_x^{[n]} + d_y^{[n]} + d_z^{[n]} - 2$, then for any other set of product tensors $E' = \{x'_a \otimes y'_a \otimes z'_a : a \in [n]\}$, $E \cup E'$ splits.

Recall the tensor decomposition setup...

We are handed a rank decomposition

$$T = \sum_{a \in [n]} x_a \otimes y_a \otimes z_a$$

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Poly time to check!

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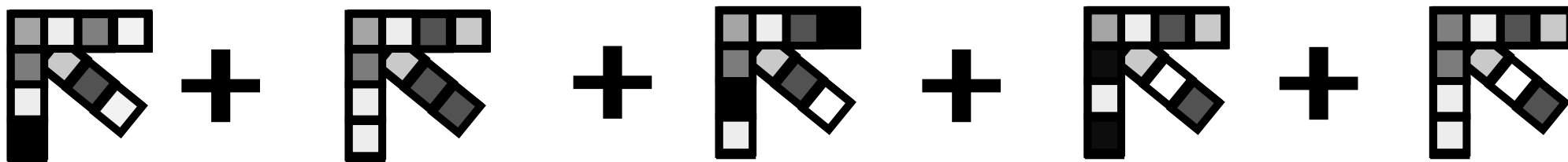
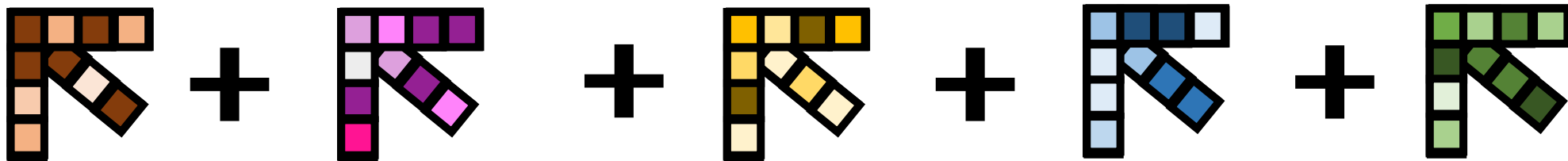


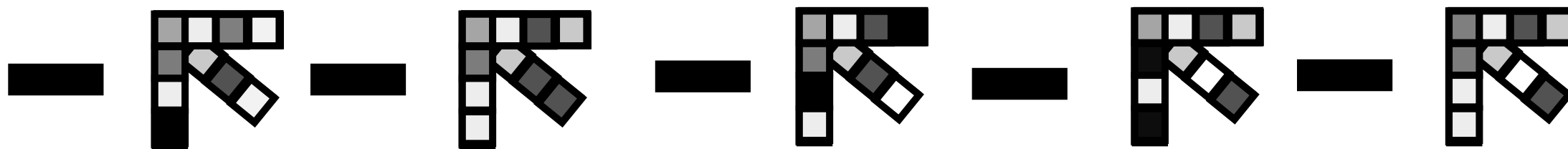
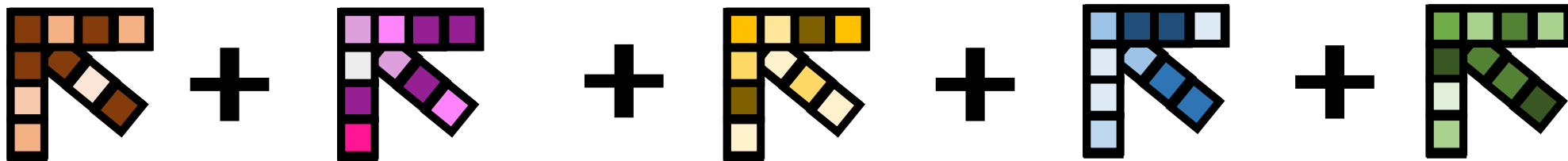
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Proof:

By previous corollary, $\{x_a \otimes y_a \otimes z_a, x'_a \otimes y'_a \otimes z'_a : a \in [n]\}$ splits



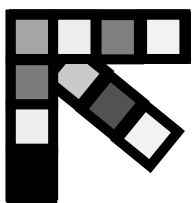


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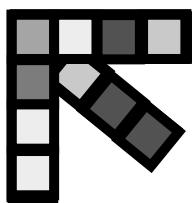
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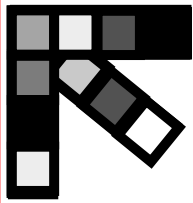
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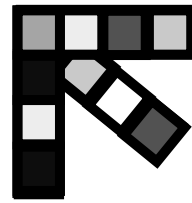
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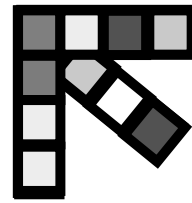
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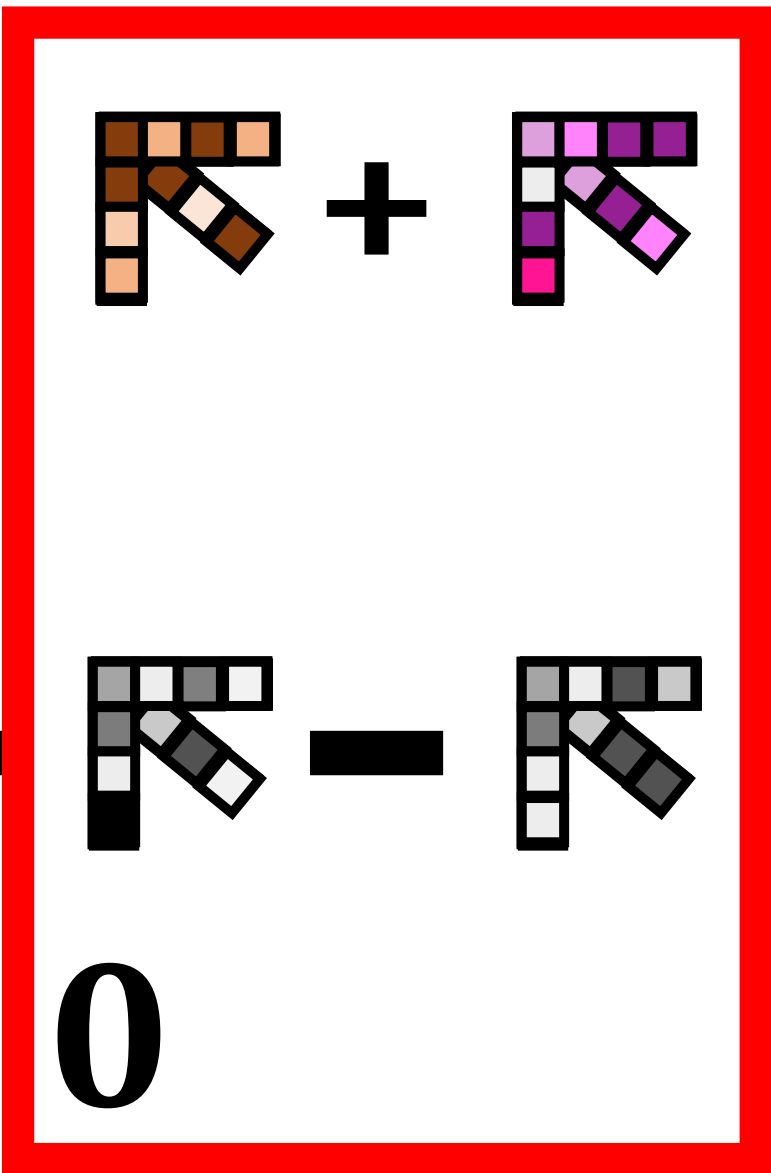
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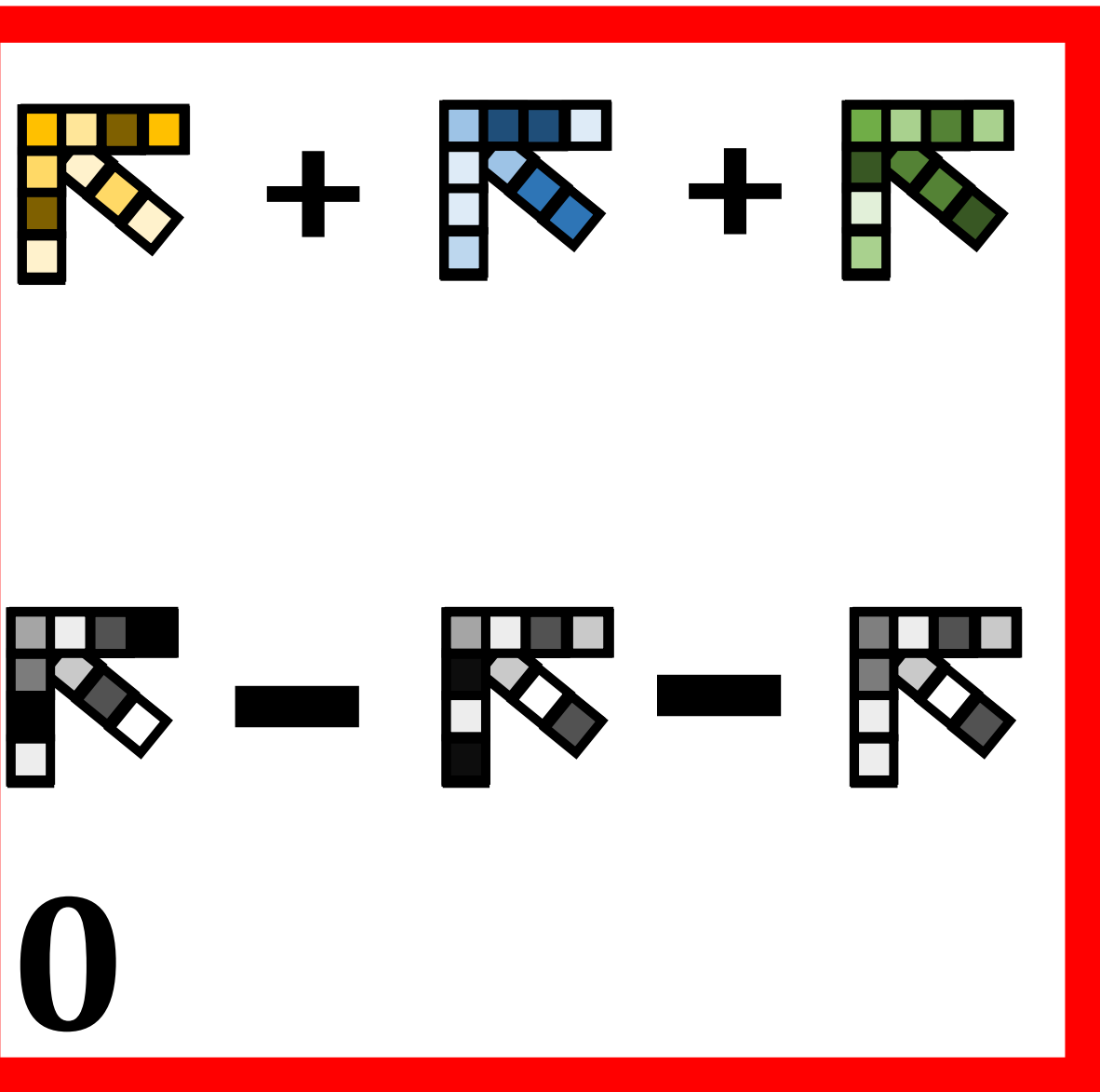
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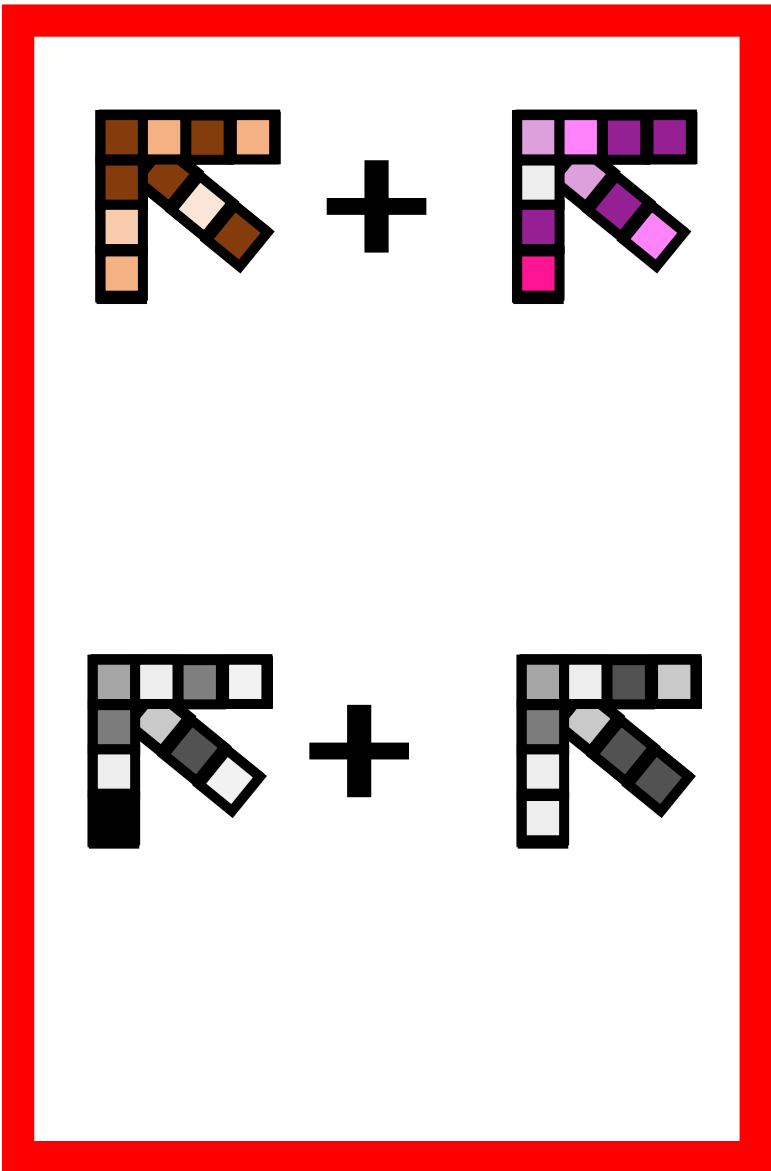
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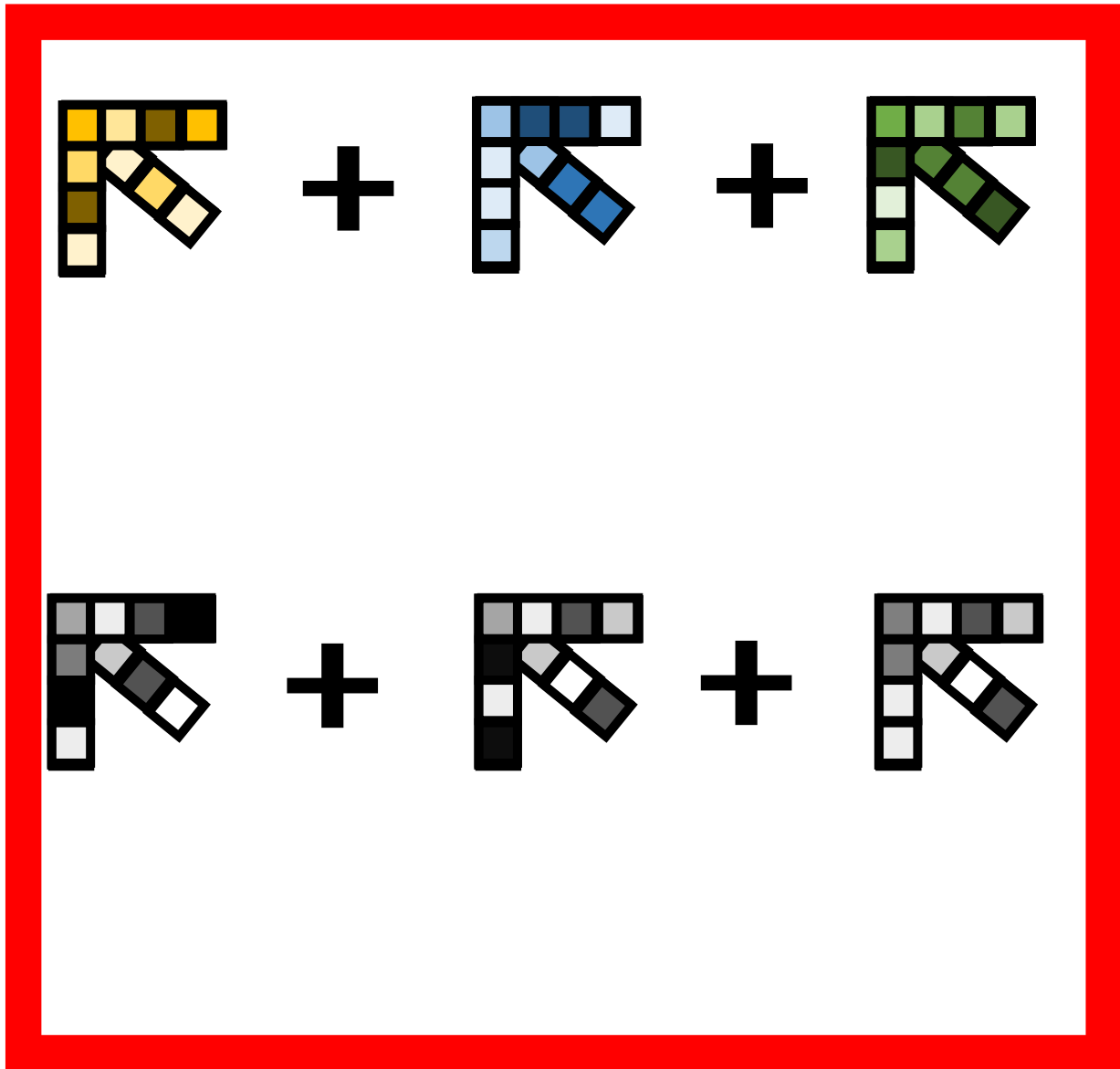
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Uniqueness

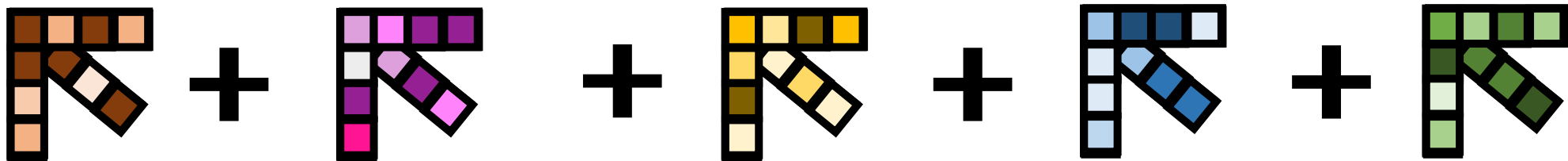
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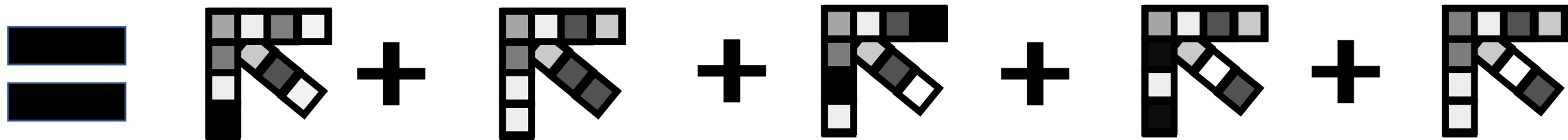
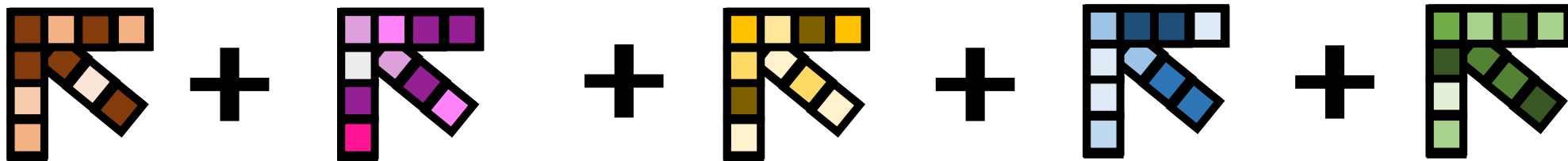
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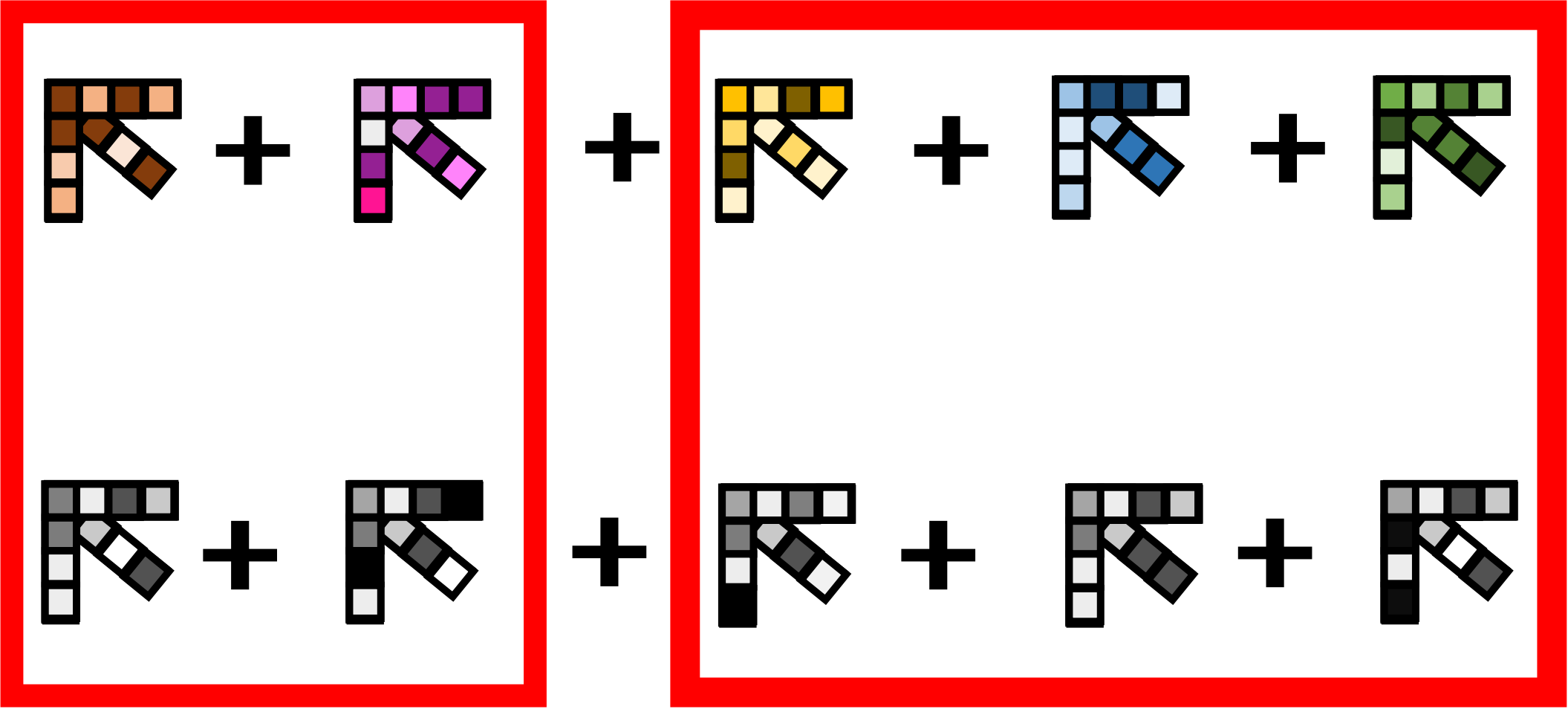

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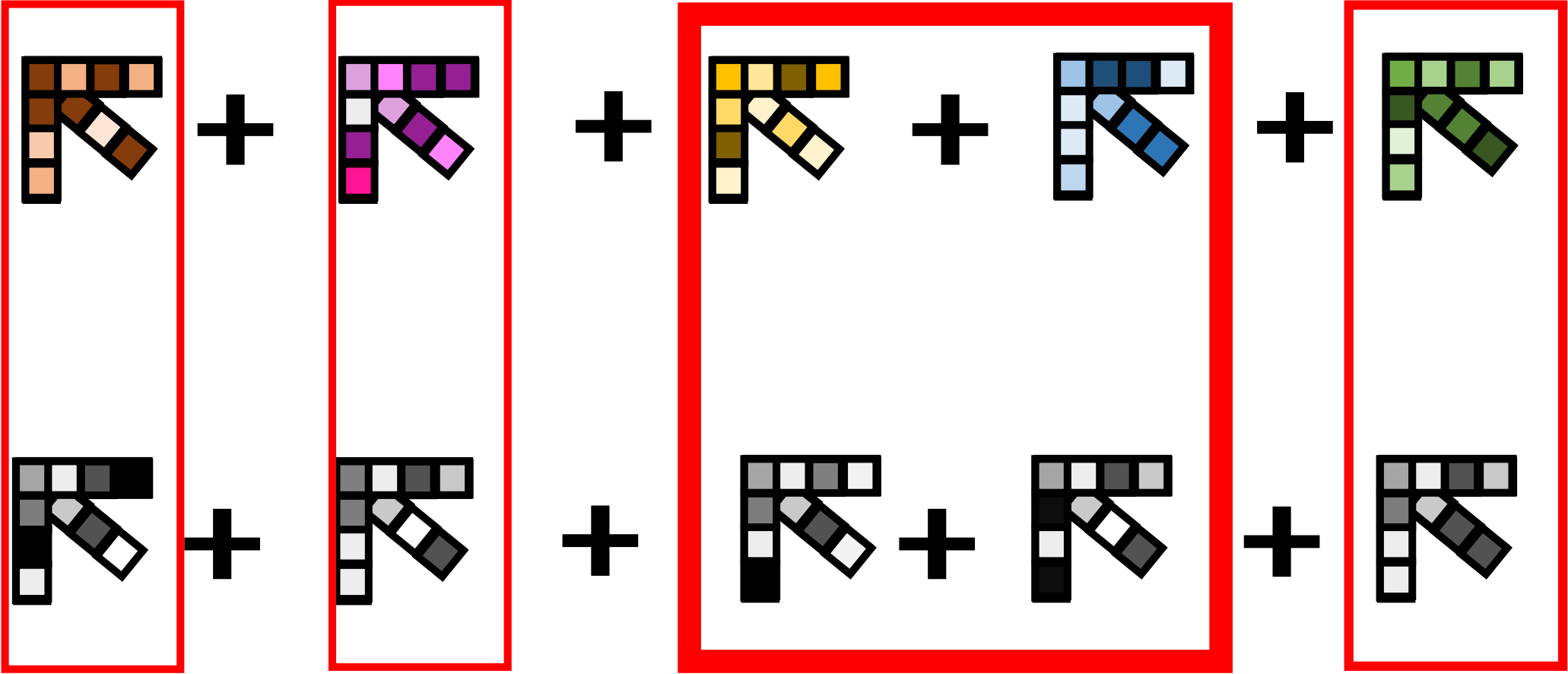




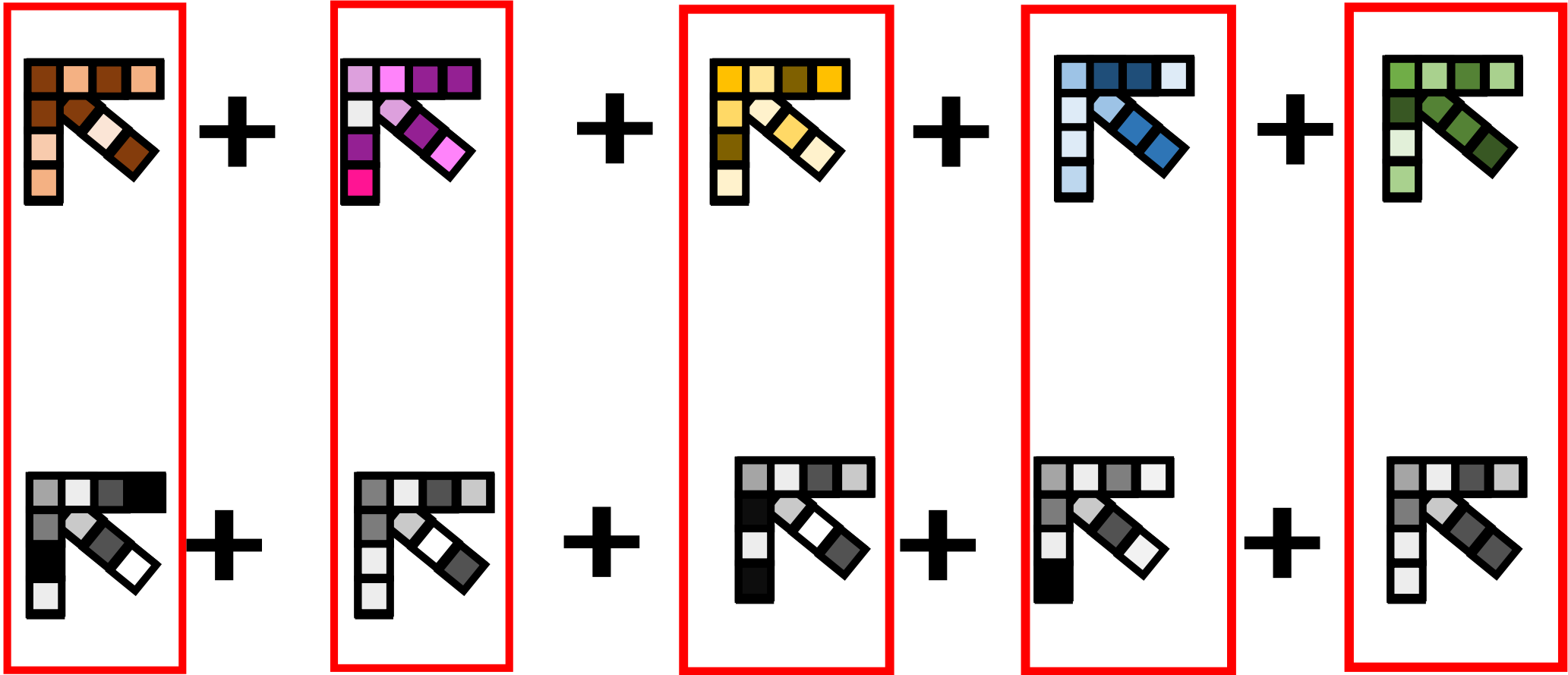
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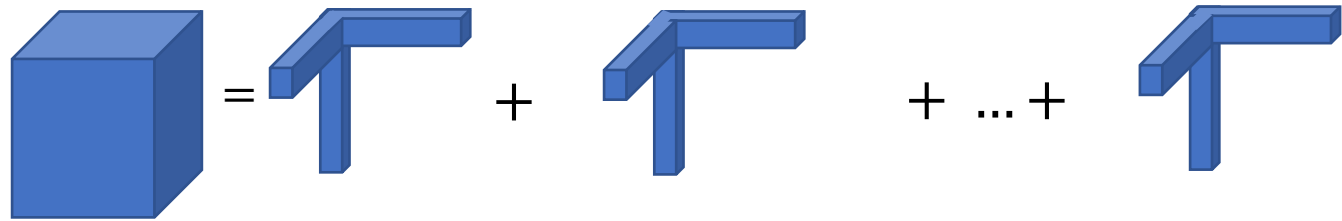

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Conclusion

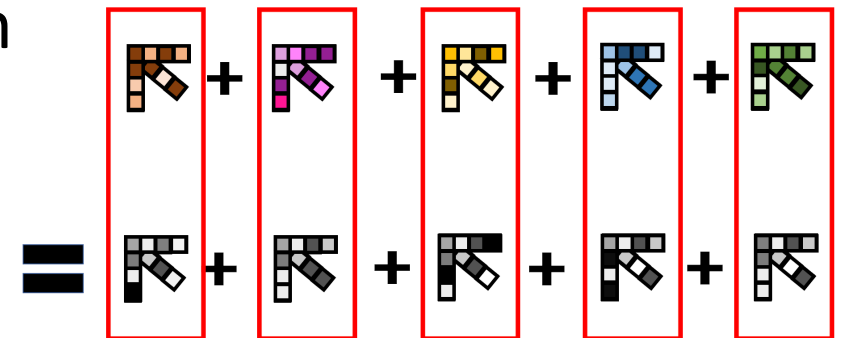
Algorithms:

- Intersecting variety X with subspace $\leftrightarrow (X, k)$ -decompositions
- Broad applications for different choices of X
- In particular, can decompose tensors of quadratically higher rank than Jennrich



Uniqueness:

- Splitting theorem “demystifies” Kruskal’s theorem
- More matroid theory for product tensors?



Algorithms and Uniqueness of Tensor Decompositions

Benjamin Lovitz
Northeastern University

UMass Boston Department of Mathematics

November 14, 2023



**Northeastern
University**