

# Decomposing Tensors into Frames

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# Symmetric Tensors

$T$  is an  $\underbrace{n \times \dots \times n}_{d \text{ times}}$  symmetric tensor with elements in a field  $\mathbb{K}(= \mathbb{R}, \mathbb{C})$  if

$$T_{i_1 i_2 \dots i_d} = T_{i_{\sigma_1} i_{\sigma_2} \dots i_{\sigma_d}}$$

for all permutations  $\sigma$  of  $\{1, 2, \dots, d\}$ . Notation:  $T \in S^d(\mathbb{K}^n)$ .

Example (symmetric matrices ( $d = 2$ ))

$$T = \begin{pmatrix} T_{11} & T_{12} & \cdots & T_{1n} \\ T_{12} & T_{22} & \cdots & T_{2n} \\ & & \vdots & \\ T_{1n} & T_{2n} & \cdots & T_{nn} \end{pmatrix}$$

# Symmetric Tensor Decomposition

For a tensor  $T \in S^d(\mathbb{K}^n)$ , a symmetric (or Waring) decomposition has the form

$$T = \sum_{i=1}^r \lambda_i v_i^{\otimes d}.$$

The smallest  $r$  for which such a decomposition exists is the *Waring rank* of  $T$ .

# Orthogonal Tensor Decomposition

An *orthogonal decomposition* of a symmetric tensor  $T \in S^d(\mathbb{R}^n)$  is a decomposition

$$T = \sum_{i=1}^r \lambda_i v_i^{\otimes d}$$

such that the vectors  $v_1, \dots, v_r$  are orthonormal. In particular,  $r \leq n$ .

## Definition

A tensor  $T \in S^d(\mathbb{R}^n)$  is *orthogonally decomposable*, for short *odeco*, if it has an orthogonal decomposition.

## Examples

1. All symmetric matrices are odeco: by the spectral theorem

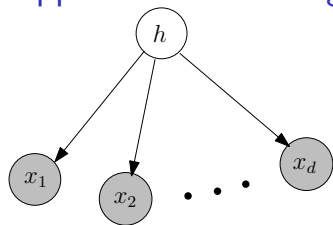
$$\begin{aligned} T = V^T \Lambda V &= \begin{bmatrix} | & \cdots & | \\ v_1 & \cdots & v_n \\ | & \cdots & | \end{bmatrix} \begin{bmatrix} \lambda_1 & & \\ & \ddots & \\ & & \lambda_n \end{bmatrix} \begin{bmatrix} - & v_1 & - \\ & \vdots & \\ - & v_n & - \end{bmatrix} \\ &= \sum_{i=1}^n \lambda_i v_i v_i^T = \sum_{i=1}^n \lambda_i v_i^{\otimes 2}, \end{aligned}$$

where  $v_1, \dots, v_n$  is an orthonormal basis of eigenvectors.

2. The Fermat polynomial: If  $v_i = e_i$ , for  $i = 1, \dots, n$ , then

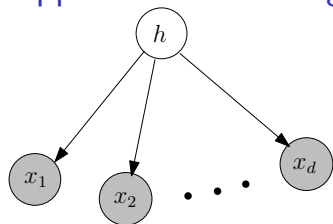
$$T = e_1^{\otimes d} + e_2^{\otimes d} + \cdots + e_n^{\otimes d}.$$

## An Application: Exchangeable Single Topic Models



Pick a topic  $h \in \{1, 2, \dots, k\}$  with distribution  $(w_1, \dots, w_k) \in \Delta_{k-1}$ . Given  $h = j$ ,  $x_1, \dots, x_d$  are *i.i.d.* random variables taking values in  $\{1, 2, \dots, n\}$  with distribution  $\mu_j = (\mu_{j1}, \dots, \mu_{jn}) \in \Delta_{n-1}$ .

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Then, the joint distribution of  $x_1, \dots, x_d$  is an  $\underbrace{n \times n \times \dots \times n}_{d \text{ times}}$  symmetric tensor  $T \in S^d(\mathbb{R}^n)$  whose entries sum to 1. Moreover,

$$T_{i_1, \dots, i_d} = \mathbb{P}(x_1 = i_1, \dots, x_d = i_d) = \sum_{j=1}^k \mathbb{P}(h = j) \prod_{k=1}^d \mathbb{P}(x_k = i_k | h = j).$$

$$\text{So, } T = \sum_{j=1}^k w_j \mu_j^{\otimes d}.$$

# Eigenvectors of Symmetric Tensors

Consider a symmetric tensor  $T \in S^d(\mathbb{K}^n)$ .

Example ( $d = 2$ )

$T$  is an  $n \times n$  matrix and  $w \in \mathbb{K}^n$  is an eigenvector if

$$Tw = \begin{pmatrix} \vdots \\ \sum_{j=1}^n T_{ij} w_j \\ \vdots \end{pmatrix} = \lambda w.$$



# Eigenvectors of Symmetric Tensors

## Definition

Given a symmetric tensor  $T \in S^d(\mathbb{K}^n)$ , an *eigenvector* of  $T$  with *eigenvalue*  $\lambda$  is a vector  $w \in \mathbb{K}^n$  such that

$$T w^{d-1} := \begin{pmatrix} \vdots \\ \sum_{i_2, \dots, i_d=1}^n T_{i, i_2, \dots, i_d} w_{i_2} \dots w_{i_d} \\ \vdots \end{pmatrix} = \lambda w.$$

Two eigenvector-eigenvalue pairs  $(w, \lambda)$  and  $(w', \lambda')$  are equivalent if there exists  $t \in \mathbb{K} \setminus \{0\}$  such that  $t^{d-2}\lambda = \lambda'$  and  $tw = w'$ .

# Eigenvectors of Symmetric Tensors

## Example

Let

$$T = e_1^{\otimes 3} + e_2^{\otimes 3} + e_3^{\otimes 3}.$$

Then,  $(x, y, z)^T$  is an eigenvector of  $T$  with eigenvalue  $\lambda$  if

$$T \cdot \begin{pmatrix} x \\ y \\ z \end{pmatrix}^2 = \begin{pmatrix} 3x^2 \\ 3y^2 \\ 3z^2 \end{pmatrix} = \lambda \begin{pmatrix} x \\ y \\ z \end{pmatrix}.$$

Equivalently, the  $2 \times 2$  minors of the matrix  $\begin{bmatrix} 3x^2 & x \\ 3y^2 & y \\ 3z^2 & z \end{bmatrix}$  vanish. Therefore,

$$x^2y - xy^2 = x^2z - xz^2 = y^2z - yz^2 = 0.$$

The solutions are (up to scaling):

$$\{(1, 0, 0), (0, 1, 0), (0, 0, 1), (1, 1, 0), (1, 0, 1), (0, 1, 1), (1, 1, 1)\}.$$

## Eigenvectors of Odeco Tensors

If  $T = \sum_{i=1}^n \lambda_i v_i^{\otimes d}$  is an odeco tensor, i.e.  $v_1, \dots, v_n$  are orthonormal, then  $v_1, \dots, v_n$  are eigenvectors of  $T$  with eigenvalues  $\lambda_1, \dots, \lambda_n$  respectively.

- ▶ Is there an easy way of finding these vectors, i.e. finding the orthogonal decomposition of an odeco tensor?
- ▶ Are these all of the eigenvectors of an odeco tensor?

# Robust Eigenvectors and the Tensor Power Method

## Definition

A unit vector  $u \in \mathbb{R}^n$  is a *robust eigenvector* of a tensor  $T \in S^d(\mathbb{R}^n)$  if there exists  $\epsilon > 0$  such that for all  $\theta \in \mathcal{B}_\epsilon(u) = \{u' : \|u - u'\| < \epsilon\}$ , repeated iteration of the map

$$\theta \mapsto \frac{T\theta^{d-1}}{\|T\theta^{d-1}\|}, \quad (1)$$

starting from  $\theta$  converges to  $u$ .

## Theorem (Anandkumar et al.)

Let  $T$  have an orthogonal decomposition  $T = \sum_{i=1}^k \lambda_i v_i^{\otimes d}$  with  $\lambda_1, \dots, \lambda_k > 0$ .

Then, the set of robust eigenvectors of  $T$  is equal to  $\{v_1, v_2, \dots, v_k\}$ .

# Decomposing Tensors into Frames

Question: How to enlarge the set of odedco tensors to contain tensors of higher ranks?

Idea:

- ▶ Odedco tensors come from an orthonormal basis  $V := (\mathbf{v}_1, \dots, \mathbf{v}_n) \in (\mathbb{R}^n)^n$ , i.e.

$$VV^T = I_n \quad \text{and} \quad \|\mathbf{v}_j\|^2 = 1, j = 1, \dots, n.$$

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- ▶ Instead let  $V := (\mathbf{v}_1, \dots, \mathbf{v}_r) \in (\mathbb{R}^n)^r$  be a *finite unit norm tight frame* (or *funtf*), i.e.

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A tensor  $T \in S^d(\mathbb{R}^n)$  is *frame decomposable* (or *fradeco*) if it can be written as

$$T = \sum_{i=1}^r \lambda_i \mathbf{v}_i^{\otimes d},$$

where  $(\mathbf{v}_1, \dots, \mathbf{v}_r)$  form a finite unit norm tight frame.

# Finite Unit Norm Tight Frames

## Examples

▶ The Mercedes Benz Frame  $V = \begin{pmatrix} 0 & \frac{\sqrt{3}}{2} & -\frac{\sqrt{3}}{2} \\ 1 & -\frac{1}{2} & -\frac{1}{2} \end{pmatrix}$ .



▶  $V = \frac{1}{3\sqrt{3}} \begin{pmatrix} -5 & 1 & 1 & 3 \\ 1 & -5 & 1 & 3 \\ 1 & 1 & -5 & 3 \end{pmatrix}$ .

▶  $V = \begin{pmatrix} 1 & 0 & \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \\ 0 & 1 & \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} \end{pmatrix}$ .

The *funtf variety*, denoted by  $\mathcal{F}_{r,n}$  is the sub variety of  $\mathbb{C}^{nr}$  defined by the equations

$$VV^T = \frac{r}{n}I_n \quad \text{and} \quad \|\mathbf{v}_j\|^2 = 1, j = 1, \dots, r.$$



# Finite Unit Norm Tight Frames

## Theorem (Cahill, Mixon, Strawn, Dykema)

The funtf variety  $\mathcal{F}_{r,n}$  is irreducible when  $r \geq n + 2 > 4$ . We have

$$\dim(\mathcal{F}_{r,n}) = (n-1) \cdot \left(r - \frac{n}{2} - 1\right) \quad \text{when } r > n \geq 2.$$

$r$	$n$	$\dim \mathcal{F}_{r,n}$	$\deg \mathcal{F}_{r,n}$	# components & degrees
3	2	1	$8 \cdot 2$	8 components, each degree 2
4	2	2	$12 \cdot 4$	12 components, each degree 4
5	2	3	112	irreducible
6	2	4	240	irreducible
7	2	5	496	irreducible
4	3	3	$16 \cdot 8$	16 components, each degree 8
5	3	5	1024	irreducible
6	3	7	2048	irreducible
7	3	9	4096	irreducible
5	4	6	$32 \cdot 40$	32 components, each degree 40
6	4	9	20800	irreducible
7	4	12	65536	irreducible

Table: Dimension and degree of the funtf variety in some small cases

# The Fradeco Variety

## Questions:

- ▶ Does the tensor power method recover the decomposition of a fradeco tensor?
- ▶ Is there another way to recover the decomposition?

The *variety of frame decomposable tensors* or the *fradeco variety*, denoted by  $\mathcal{T}_{r,n,d}$  is the Zariski closure in  $S^d(\mathbb{C}^n)$  of all tensors that can be written as

$$T = \sum_{j=1}^r \lambda_j \mathbf{v}_j^{\otimes d},$$

where  $(\mathbf{v}_1, \dots, \mathbf{v}_r) \in \mathcal{F}_{r,n}$ .

## Example

When  $r = n$ ,  $\mathcal{T}_{n,n,d}$  is the odeco variety.

- ▶ Can we give equations that define the variety  $\mathcal{T}_{r,n,d}$ .

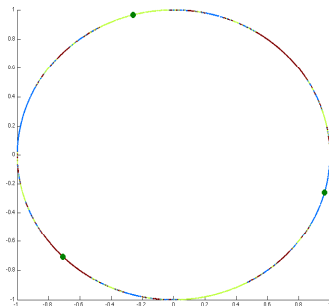
# The tensor power method

## Conjecture

Let  $r = n + 1 < d$  and  $T = \sum_{j=1}^{n+1} \lambda_j \mathbf{v}_j^{\otimes d} \in \mathcal{T}_{n+1, n, d}$  with  $\lambda_1, \dots, \lambda_{n+1} > 0$ . Then,  $\mathbf{v}_1, \dots, \mathbf{v}_{n+1}$  are the robust eigenvectors of  $T$ , so they are found by the tensor power method.

## Example (The Mercedes Benz Frame)

Let  $T = \begin{pmatrix} 0 \\ 1 \end{pmatrix}^{\otimes 5} + \begin{pmatrix} \frac{\sqrt{3}}{2} \\ -\frac{1}{2} \end{pmatrix}^{\otimes 5} + \begin{pmatrix} -\frac{\sqrt{3}}{2} \\ -\frac{1}{2} \end{pmatrix}^{\otimes 5}$ . The dynamics of the power method looks like this



# The tensor power method

## Example

Let  $n = 2, r = 4, d = 5$  and consider the tensor

$$T = \alpha \begin{pmatrix} 1 \\ 0 \end{pmatrix}^{\otimes 4} + \begin{pmatrix} 0 \\ 1 \end{pmatrix}^{\otimes 4} + \begin{pmatrix} 1 \\ 1 \end{pmatrix}^{\otimes 4} + \begin{pmatrix} 1 \\ -1 \end{pmatrix}^{\otimes 4} \in \mathcal{T}_{4,2,5},$$

where  $\alpha > 6$ . The vector  $\begin{pmatrix} 1 \\ 0 \end{pmatrix}$  is an eigenvector, but none of the other eigenvectors are real. Therefore, the frame decomposition of  $T$  cannot be recovered from its eigenvectors.

# Symmetric $2 \times 2 \times \cdots \times 2$ fradeco tensors

Given a symmetric tensor  $T \in S^d(\mathbb{C}^2)$ , denote

$$t_i = T_{i_1, \dots, i_d}$$

whenever  $i$  of  $i_1, \dots, i_d$  are equal to 1 and the rest are equal to 0.

## Theorem

Fix  $r \in \{3, \dots, 9\}$  and  $d \geq 2r - 2$ . There exists a matrix  $\mathcal{M}_r$  with

- (a)  $\dots$   $r - 1$  rows and  $d - r + 1$  columns, whose entries are linear forms in the coordinates  $t_0, t_1, \dots, t_d$  on  $\mathbb{P}^d$ .
- (b) The columns of  $\mathcal{M}_r$  involve  $r + 1$  of the unknowns  $t_i$ , and they are identical up to index shifts.
- (c) The maximal minors of  $\mathcal{M}_r$  form a **Gröbner basis** for the homogeneous prime ideal of  $\mathcal{T}_{r,2,d}$ .

## Lemma

The fradeco variety  $\mathcal{T}_{r,2,d}$  has dimension  $2r - 3$  in  $\mathbb{P}^d$ .

# Matrices for Binary Fradeco Tensors

$$\mathcal{M}_3 = \begin{pmatrix} t_0 - 3t_2 & t_1 - 3t_3 & t_2 - 3t_4 & t_3 - 3t_5 & \cdots & t_{d-3} - 3t_{d-1} \\ 3t_1 - t_3 & 3t_2 - t_4 & 3t_3 - t_5 & 3t_4 - t_6 & \cdots & 3t_{d-2} - t_d \end{pmatrix}$$

$$\mathcal{M}_4 = \begin{pmatrix} t_0 + t_4 & t_1 + t_5 & t_2 + t_6 & t_3 + t_7 & \cdots & t_{d-4} + t_d \\ t_1 - t_3 & t_2 - t_4 & t_3 - t_5 & t_4 - t_6 & \cdots & t_{d-3} + t_{d-1} \\ t_2 & t_3 & t_4 & t_5 & \cdots & t_{d-2} \end{pmatrix}$$

$$\mathcal{M}_5 = \begin{pmatrix} t_0 + 5t_2 & t_1 + 5t_3 & t_2 + 5t_4 & t_3 + 5t_5 & \cdots & t_{d-5} + 5t_{d-3} \\ t_1 - 3t_3 & t_2 - 3t_4 & t_3 - 3t_5 & t_4 - 3t_6 & \cdots & t_{d-4} - 3t_{d-2} \\ 3t_2 - t_4 & 3t_3 - t_5 & 3t_4 - t_6 & 3t_5 - t_7 & \cdots & 3t_{d-3} - t_{d-1} \\ 5t_3 + t_5 & 5t_4 + t_6 & 5t_5 + t_7 & 5t_6 + t_8 & \cdots & 5t_{d-2} + t_d \end{pmatrix}$$

$$\mathcal{M}_6 = \begin{pmatrix} t_0 + 3t_2 & t_1 + 3t_3 & t_2 + 3t_4 & t_3 + 3t_5 & \cdots & t_{d-6} + 3t_{d-4} \\ t_1 + t_5 & t_2 + t_6 & t_3 + t_7 & t_4 + t_8 & \cdots & t_{d-5} + t_{d-1} \\ t_2 - t_4 & t_3 - t_5 & t_4 - t_6 & t_5 - t_7 & \cdots & t_{d-4} - t_{d-2} \\ t_3 & t_4 & t_5 & t_6 & \cdots & t_{d-3} \\ 3t_4 + t_6 & 3t_5 + t_7 & 3t_6 + t_8 & 3t_7 + t_9 & \cdots & 3t_{d-2} + t_d \end{pmatrix}$$

... ..

We conjecture that this works for all  $r$ .

# Higher Dimensions

## Theorem

The following table describes the fradeco varieties  $\mathcal{T}_{r,n,d}$  in the cases when  $n \geq 3$  and  $1 \leq \dim(\mathcal{T}_{r,n,d}) \cdot \text{codim}(\mathcal{T}_{r,n,d}) \leq 100$ .

variety	dim	codim	degree	known equations
$\mathcal{T}_{4,3,3}$	6	3	17	3 cubics, 6 quartics
$\mathcal{T}_{4,3,4}$	6	8	74	6 quadrics, 37 cubics
$\mathcal{T}_{4,3,5}$	6	14	191	27 quadrics, 104 cubics
$\mathcal{T}_{5,3,4}$	9	5	210	1 cubic, 6 quartics
$\mathcal{T}_{5,3,5}$	9	11	1479	20 cubics, 213 quartics
$\mathcal{T}_{6,3,4}$	12	2	99	none in degree $\leq 5$
$\mathcal{T}_{6,3,5}$	12	8	4269	one quartic
$\mathcal{T}_{7,3,5}$	15	5	$\geq 38541$	none in degree $\leq 4$
$\mathcal{T}_{8,3,5}$	18	2	690	none in degree $\leq 5$
$\mathcal{T}_{10,3,6}$	24	3	$\geq 16252$	none in degree $\leq 7$
$\mathcal{T}_{5,4,3}$	10	9	830	none in degree $\leq 4$
$\mathcal{T}_{6,4,3}$	14	5	1860	none in degree $\leq 3$
$\mathcal{T}_{7,4,3}$	18	1	194	one in degree 194

## Proposition

For all  $r > n, d \geq 3$ , the dimension of  $\mathcal{T}_{r,n,d}$  is bounded above by

$$\min \left\{ (r-n)(n-1) + \binom{n-1}{2} + r - 1, \binom{n+d-1}{d} - 1 \right\}.$$

Thank you!



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