Orthogonally Decomposable Tensors

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Symmetric Tensors

T is an $\underbrace{n \times \ldots \times n}_{d \text{ times}}$ symmetric tensor with elements in $\mathbb R$ if

$$T_{i_1i_2\ldots i_d} = T_{i_{\sigma_1}i_{\sigma_2}\ldots i_{\sigma_d}}$$

for all permutations σ of $\{1, 2, ..., d\}$. Notation: $T \in S^d(\mathbb{R}^n)$. Example (d = 2)

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

Example (n = 3, d = 3)

$$T = \underbrace{\begin{pmatrix} T_{111} & T_{112} & T_{113} \\ T_{112} & T_{122} & T_{123} \\ T_{113} & T_{123} & T_{133} \end{pmatrix}}_{T_{1..}}, \underbrace{\begin{pmatrix} T_{112} & T_{122} & T_{123} \\ T_{122} & T_{222} & T_{223} \\ T_{123} & T_{223} & T_{233} \end{pmatrix}}_{T_{2..}}, \underbrace{\begin{pmatrix} T_{113} & T_{123} & T_{133} \\ T_{123} & T_{223} & T_{233} \\ T_{133} & T_{233} & T_{333} \end{pmatrix}}_{T_{3..}}.$$

Symmetric Tensors and Polynomials

An equivalent way of representing a symmetric tensor $T \in S^d(\mathbb{R}^n)$ is by a homogeneous polynomial $f_T \in \mathbb{R}[x_1, ..., x_n]$ of degree d.

Example (d = 2)

In the case of matrices,

$$f_{T}(x_{1},...,x_{n}) = x^{T}Tx$$

$$= \begin{pmatrix} x_{1} & x_{2} & \cdots & x_{n} \end{pmatrix} \begin{pmatrix} T_{11} & T_{12} & \cdots & T_{1n} \\ T_{12} & T_{13} & \cdots & T_{2n} \\ & & \vdots \\ T_{1n} & T_{2n} & \cdots & T_{nn} \end{pmatrix} \begin{pmatrix} x_{1} \\ x_{2} \\ \vdots \\ x_{n} \end{pmatrix}$$

$$= \sum_{i,j} T_{ij}x_{i}x_{j}.$$

Symmetric Tensors and Polynomials

For general $T \in S^d(\mathbb{R}^n)$,

$$f_T(x_1,...,x_n) = T \cdot x^d := \sum_{i_1,...,i_d=1}^n T_{i_1...i_d} x_{i_1}...x_{i_d}$$

Symmetric Tensors and Polynomials

For general $T \in S^d(\mathbb{R}^n)$,

$$f_{T}(x_{1},...,x_{n}) = T \cdot x^{d} := \sum_{i_{1},...,i_{d}=1}^{n} T_{i_{1}...i_{d}} x_{i_{1}}...x_{i_{d}}$$
$$= \sum_{j_{1}+\cdots+j_{n}=d} {\binom{d}{j_{1},...,j_{n}}} T_{\underbrace{1...1}_{j_{1}}\cdots\underbrace{n...n}_{j_{n}}} x_{1}^{j_{1}}\dots x_{n}^{j_{n}}$$
$$= \sum_{j_{1}+\cdots+j_{n}=d} u_{j_{1},...,j_{n}} x_{1}^{j_{1}}\dots x_{n}^{j_{n}}.$$

Example (n = 3, d = 2)

For 3×3 matrices,

$$f_{T}(x_{1}, x_{2}, x_{3}) = \sum_{i_{1}, i_{2}=1}^{3} T_{i_{1}i_{2}}x_{i_{1}}x_{i_{2}}$$

= $\underbrace{T_{11}}_{u_{2,0,0}} x_{1}^{2} + \underbrace{2T_{12}}_{u_{1,1,0}}x_{1}x_{2} + \underbrace{2T_{13}}_{u_{1,0,1}}x_{1}x_{3} + \underbrace{T_{22}}_{u_{0,2,0}}x_{2}^{2} + \underbrace{2T_{23}}_{u_{0,1,1}}x_{2}x_{3} + \underbrace{T_{33}}_{u_{0,0,2}}x_{3}^{2}.$

Symmetric Tensor Decomposition

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

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

If $f_T \in \mathbb{R}[x_1, ..., x_n]$ is the corresponding polynomial, then

$$f_T(x_1,...,x_n) = \sum_{i=1}^r \lambda_i (v_i \cdot x)^d = \sum_{i=1}^r \lambda_i (v_{i1}x_1 + v_{i2}x_2 + \cdots + v_{in}x_n)^d.$$

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

Orthogonal Tensor Decomposition

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

$$\mathcal{T} = \sum_{i=1}^r \lambda_i v_i^{\otimes d}$$
 with corresponding $f_{\mathcal{T}} = \sum_{i=1}^r \lambda_i (v_i \cdot x)^d$

such that the vectors $v_1, ..., v_r \in \mathbb{R}^n$ are orthonormal. In particular, $r \leq n$.

Definition

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

Examples

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

$$T = V^{T} \Lambda V = \begin{bmatrix} | & \cdots & | \\ v_{1} & \cdots & v_{n} \\ | & \cdots & | \end{bmatrix} \begin{bmatrix} \lambda_{1} & & \\ & \ddots & \\ & & \lambda_{n} \end{bmatrix} \begin{bmatrix} - & v_{1} & - \\ & \vdots \\ - & v_{n} & - \end{bmatrix}$$

$$=\sum_{i=1}^n \lambda_i v_i v_i^T = \sum_{i=1}^n \lambda_i v_i^{\otimes 2},$$

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

2. <u>The Fermat polynomial</u>: If $v_i = e_i$, for i = 1, ..., n, then $f_T(x_1, ..., x_n) = x_1^d + x_2^d + \cdots + x_n^d,$

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

An Application: Exchangeable Single Topic Models



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

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

$$T = \sum_{j=1}^{k} \mathbb{P}(h=j)\mathbb{P}(x_1|h=j) \otimes \cdots \otimes \mathbb{P}(x_d|h=j) = \sum_{j=1}^{k} w_j \mu_j^{\otimes d}$$

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Given T, to recover the parameters w, μ , use a transformation $T \mapsto T_{od}$ and decompose T_{od} [Anandkumar et al.].

Eigenvectors of Tensors

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

Example (d = 2)

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

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

Example (d = 3)

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

$$Tw^{2} := \begin{pmatrix} \vdots \\ \sum_{j,k=1}^{n} T_{i,j,k} w_{j} w_{k} \\ \vdots \end{pmatrix} = \lambda w.$$

Definition

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

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

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

▶ For the corresponding $f_T \in \mathbb{R}[x_1, ..., x_n]$, $w \in \mathbb{C}^n$ is an eigenvector with eigenvalue λ if

$$\nabla f_T(w) = d\lambda w.$$

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▶ For the corresponding $f_T \in \mathbb{R}[x_1, ..., x_n]$, $w \in \mathbb{C}^n$ is an *eigenvector* with *eigenvalue* λ if

$$\nabla f_T(w) = d\lambda w.$$

Therefore, the eigenvectors of f are given by the vanishing of the

 2×2 minors of the matrix $[\nabla f_T(x)|x]$.

Example

Let d = 2 and T be an $n \times n$ symmetric matrix. Then

$$f_T(x_1,\ldots,x_n)=\sum_{i,j}T_{ij}x_ix_j.$$

Thus,

$$\nabla f_T(x_1,\ldots,x_n) = \begin{pmatrix} 2(\sum_{i=1}^n T_{1i}x_i)\\ 2(\sum_{i=1}^n T_{2i}x_i)\\ \vdots\\ 2(\sum_{i=1}^n T_{ni}x_i) \end{pmatrix} = 2Tx.$$

So, x is an eigenvector with eigenvalue λ if and only if $\nabla f_T(x) = 2\lambda x$.

Example

Let

$$T = e_1^{\otimes 3} + e_2^{\otimes 3} + e_3^{\otimes 3}$$
 and $f_T(x, y, z) = x^3 + y^3 + z^3$.

Then, $(x, y, z)^T$ is an eigenvector of f_T if and only if the 2 × 2 minors of the matrix $\begin{bmatrix} x \\ \nabla f & y \\ z \end{bmatrix} = \begin{bmatrix} 3x^2 & x \\ 3y^2 & y \\ 3z^2 & z \end{bmatrix}$ vanish. Therefore,

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

This is equivalent to

$$xy(x-y) = xz(x-z) = yz(y-z) = 0.$$

The solutions are (up to scaling):

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

If $T = \sum_{i=1}^{n} \lambda_i v_i^{\otimes d}$ is an odeco tensor, i.e. $v_1, ..., v_n$ are orthonormal, then the vectors v_k , k = 1, ..., n are eigenvectors of T with corresponding eigenvalues λ_k , k = 1, ..., n:

$$Tv_k^{d-1} = \sum_{i=1}^n \lambda_i v_i (v_k \cdot v_i)^{d-1} = \lambda_k v_k.$$

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- 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

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

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

starting from θ 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 v_1, \ldots, v_k orthonormal, and assume that $\lambda_1, \ldots, \lambda_k > 0$.

- 1. The set of $\theta \in \mathbb{R}^n$ which do not converge to some v_i under repeated iteration of (1) has measure 0.
- 2. The set of robust eigenvectors of T is equal to $\{v_1, v_2, ..., v_k\}$.

The Tensor Power Method

Algorithm

 $\frac{\text{Input: An odeco tensor } \mathcal{T}.}{\text{Output: An orthogonal representation of } \mathcal{T}.}$

Repeat

Find $v_i \leftarrow$ power method output starting from a random $u \in \mathbb{R}^n$. Recover $\lambda_i = T \cdot v_i^d$. $T \leftarrow T - \lambda_i v_i^{\otimes d}$. Return $v_1, ..., v_k$ and $\lambda_1, ..., \lambda_k$.

The tensor power method consists of repeated iteration of the map

1

$$u\mapsto \frac{Tu^{d-1}}{||Tu^{d-1}||},$$

or equivalently,

$$u\mapsto \frac{\nabla f(u)}{||\nabla f(u)||}.$$

The Number of Eigenvectors of a Tensor

<u>Recall</u>: Given a tensor $T \in S^d(\mathbb{C}^n)$ with corresponding polynomial f_T , the eigenvectors $x \in \mathbb{C}^n$ are the solutions to the equations given by the 2×2 minors of the matrix

 $\left[\nabla f_T(x) | x \right].$

Theorem (Sturmfels and Cartwright)

If a tensor $T \in S^d(\mathbb{C}^n)$ has finitely many eigenvectors, then their number is $\frac{(d-1)^n-1}{d-2}$.

Odeco tensors are nice because we can characterize all of their eigenvectors.

Theorem Let $T \in S^d(\mathbb{C}^n)$ be an odeco tensor with $d \ge 3$ and $T = \sum_{i=1}^n \lambda_i v_i^{\otimes d}$. Then, T has $\frac{(d-1)^n - 1}{d-2}$ eigenvectors. Explicitly, they are

$$\begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix} = \lambda_{\sigma(1)}^{-\frac{1}{d-2}} v_{\sigma(1)} + \eta_2 \lambda_{\sigma(2)}^{-\frac{1}{d-2}} v_{\sigma(2)} + \dots + \eta_k \lambda_{\sigma(k)}^{-\frac{1}{d-2}} v_{\sigma(k)},$$

where k = 1, ..., n, $\eta_2, ..., \eta_k$ are (d - 2)-nd roots of unity and σ is a permutation on $\{1, ..., n\}$.

Example
$$(d = 3, n = 3)$$

Let

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

Then, V = I, the identity matrix and the eigenvectors of T are:

$$k = 1 \quad (1:0:0)^{T}, (0:1:0)^{T}, (0:0:1)^{T}$$

$$k = 2 \quad (1:1:0)^{T}, (1:0:1)^{T}, (0:1:1)^{T}$$

$$k = 3 \quad (1:1:1)^{T}.$$

The Set of Odeco Tensors

Parametric representation:

The set of orthogonally decomposable tensors can be parametrized by $\mathbb{R}^n \times O_n(\mathbb{R})$:

$$\lambda, V \mapsto \sum_{i=1}^n \lambda_i v_i^{\otimes d}.$$

Implicit representation:

The set of orthogonally decomposable tensors can also be represented as the solutions to a set of equations.

Definition

The *odeco variety* is the set of all odeco tensors in $S^d(\mathbb{R}^n)$.

Goal: find equations defining this variety.

The Odeco Variety

Let $T \in S^d(\mathbb{R}^n)$. For $q = 1, \ldots, d$, consider the tensor $T *_q T \in S^2(S^{d-1}(\mathbb{R}^n))$.

The Odeco Variety

Let $T \in S^d(\mathbb{R}^n)$. For q = 1, ..., d, consider the tensor $T *_q T \in S^2(S^{d-1}(\mathbb{R}^n))$. Let \mathcal{F} be the set of equations defining by the condition

$$T*_q T \in S^{2(d-1)}(\mathbb{R}^n).$$

Theorem (Boralevi, Draisma, Horobeț, R.)

The odeco variety is equal to zero set of \mathcal{F} for every n.

Conjecture

The ideal defined by \mathcal{F} is prime.

Odeco Tensors as Algebras

Let $T \in S^3(\mathbb{R}^n)$ be a symmetric tensor of order 3. Let $V_T = \mathbb{R}^n$ be equipped with a positive definite inner product $(\cdot|\cdot)$. Then, V has the structure of an algebra via the product

$$V_T \times V_T \to V_T$$

 $(x, y) \mapsto x \star y := T(x, y, \cdot).$

The inner product is compatible with the product in the sense that

$$(x \star y|z) = T(x, y, z) = (z \star x|y).$$

And the product is commutative

$$x \star y = y \star x.$$

Theorem (Boralevi, Draisma, Horobeț, R.) The tensor T is odeco if and only if (V, \star) is associative, i.e.

$$x \star (y \star z) = (x \star y) \star z.$$

Nonsymmetric Tensor Decomposition

Let $T \in \mathbb{R}^n \otimes \mathbb{R}^n \otimes \cdots \otimes \mathbb{R}^n = (\mathbb{R}^n)^{\otimes d}$. A *decomposition* of T is an expression of the form

$$T = \sum_{i=1}^r \lambda_i a_i \otimes b_i \otimes c_i \otimes \cdots$$

A tensor $T \in \mathbb{R}^n \otimes \cdots \otimes \mathbb{R}^n$ is orthogonally decomposable, or odeco, if we can decompose it as

$$\mathcal{T} = \sum_{i=1}^n \lambda_i \mathbf{a}_i \otimes \mathbf{b}_i \otimes \mathbf{c}_i \otimes \cdots,$$

so that $a_1, ..., a_n \in \mathbb{R}^n$ are orthonormal, $b_1, ..., b_n \in \mathbb{R}^n$ are orthonormal, $c_1, ..., c_n \in \mathbb{R}^n$ are orthonormal, etc.

Nonsymmetric Odeco Tensors

Example

1. If $T \in \mathbb{R}^n \otimes \mathbb{R}^n$ is a matrix, then T has singular value decomposition

$$T = U\Sigma V^{T} = \sum_{i=1}^{k} \sigma_{i} u_{i} v_{i}^{T} = \sum_{i=1}^{k} \sigma_{i} u_{i} \otimes v_{i}$$

where $u_1, ..., u_k$ are orthonormal and $v_1, ..., v_k$ are orthonormal.

2. The tensor $T \in \mathbb{R}^n \otimes \cdots \otimes \mathbb{R}^n$ given by

$$T = \sum_{i=1}^n \lambda_i e_i \otimes e_i \otimes \cdots \otimes e_i$$

is odeco.

Singular Vector Tuples

Example

Given a matrix $T \in \mathbb{R}^n \otimes \mathbb{R}^n$, (u, v) is a singular vector tuple of T if there exists λ such that

$$Tu = \begin{bmatrix} \vdots \\ \sum_{j} T_{ij} u_{j} \\ \vdots \end{bmatrix} = \lambda v \quad \text{and} \quad T^{T} v = \begin{bmatrix} \vdots \\ \sum_{i} T_{ij} v_{i} \\ \vdots \end{bmatrix} = \lambda u.$$

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Definition

Given a tensor $T \in \mathbb{R}^n \otimes \cdots \otimes \mathbb{R}^n$ a singular vector tuple is a *d*-tuple $(x_1, \cdots, x_d) \in \mathbb{C}^n \times \cdots \times \mathbb{C}^n$ such that for every $1 \le k \le d$,

$$T(x_1,\ldots,x_{k-1},\cdot,x_{k+1},\ldots,x_d)=\lambda x_k,$$

for some $\lambda \in \mathbb{C}$.

Example

1. If $T \in \mathbb{R}^n \otimes \mathbb{R}^n$ is a generic matrix with singular value decomposition

$$T = U\Sigma V^T = \sum_{i=1}^n \sigma_i u_i v_i^T,$$

 $(u_1, v_1), ..., (u_n, v_n)$ are all of the singular vector pairs of T. 2. Let $T \in \mathbb{R}^n \otimes \mathbb{R}^n \otimes \mathbb{R}^n$ be odeco with

$$T = \sum_{i=1}^n \lambda_i a_i \otimes b_i \otimes c_i.$$

Then, $(a_1, b_1, c_1), \ldots, (a_n, b_n, c_n)$ are singular vector triples, but there are many more additional ones.

Tensor Power Method for Nonsymmetric Odeco Tensors

Start with odeco $T = \sum \lambda_i a_i \otimes b_i \otimes c_i$. While $T \neq 0$ repeat Choose $x^{(0)}, y^{(0)}, z^{(0)} \in \mathbb{R}^n$. For *i* from 1 to *N* repeat

$$\begin{split} x^{(i+1)} &= T(\cdot, y^{(i)}, z^{(i)}) \\ y^{(i+1)} &= T(x^{(i)}, \cdot, z^{(i)}) \\ z^{(i+1)} &= T(x^{(i)}, y^{(i)}, \cdot). \end{split}$$

End for

Find

$$\lambda = T(x^{(N)}, y^{(N)}, z^{(N)}).$$

Set

$$T = T - \lambda x^{(N)} \otimes y^{(N)} \otimes z^{(N)}.$$

End while

Lemma

With probability 1 the tensor power method converges to one of the (a_i, b_i, c_i) . And each of them has a positive probability of happening.

Singular Vectors of Odeco Tensors

Theorem

Let $T \in \mathbb{R}^n \otimes \cdots \otimes \mathbb{R}^n$ be odeco with decomposition $T = \sum_{i=1}^n \lambda_i a_i \otimes b_i \otimes c_i \otimes \cdots$. Let $A = \begin{pmatrix} a_1 & a_2 & \cdots & a_n \end{pmatrix}$, $B = \begin{pmatrix} b_1 & b_2 & \cdots & b_n \end{pmatrix}$, etc., so that A, B, C, \ldots are orthogonal matrices. Then, the singular vector tuples of T are given as follows: Type I

$$\begin{pmatrix} A \begin{bmatrix} \lambda_1^{-\frac{1}{d-2}} \\ \chi_{12}\eta_{2}\lambda_{2}^{-\frac{1}{d-2}} \\ \vdots \\ \chi_{1k}\eta_{k}\lambda_{k}^{-\frac{1}{d-2}} \\ 0 \\ \vdots \\ 0 \end{bmatrix}, B \begin{bmatrix} \lambda_1^{-\frac{1}{d-2}} \\ \chi_{22}\eta_{2}\lambda_{2}^{-\frac{1}{d-2}} \\ \vdots \\ \chi_{2k}\eta_{k}\lambda_{k}^{-\frac{1}{d-2}} \\ 0 \\ \vdots \\ 0 \end{bmatrix}, C \begin{bmatrix} \lambda_1^{-\frac{1}{d-2}} \\ \chi_{32}\eta_{2}\lambda_{2}^{-\frac{1}{d-2}} \\ \vdots \\ \chi_{3k}\eta_{k}\lambda_{k}^{-\frac{1}{d-2}} \\ 0 \\ \vdots \\ 0 \end{bmatrix}, \cdots \end{pmatrix},$$

where $1 \le k \le n$, χ_{ij} is a 2-nd root of unity, η_i is a (d-2)-nd root of unity, up to permutation.

Type II

$$(Ax_1, Bx_2, \ldots, Cx_3, \ldots),$$

where the matrix $X = (x_{ij})_{ij}$ has at least two zeros in each column and no row is identical to 0.

The Set of Odeco Tensors

Definition

The *odeco variety* is the Zariski closure of the set of all odeco tensors in $\mathbb{R}^n \otimes \cdots \otimes \mathbb{R}^n$.

For $q = 1, \ldots, d$ consider $T *_q T \in S^2(\mathbb{R}^{n_1} \otimes \cdots \otimes \mathbb{R}^{n_{q-1}} \otimes \mathbb{R}^{n_{q+1}} \otimes \cdots \otimes \mathbb{R}^{n_d})$.

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Let ${\mathcal F}$ be the deal defined by the condition that for every $q=1,\ldots,d$

$$T *_q T \in S^2(\mathbb{R}^{n_1}) \otimes \cdots \otimes S^2(\mathbb{R}^{n_{q-1}}) \otimes S^2(\mathbb{R}^{n_{q+1}}) \otimes \cdots \otimes S^2(\mathbb{R}^{n_d})$$

Theorem (Boralevi, Draisma, Horobeț, R.)

The set of orthogonally decomposable tensors equals $\mathcal{V}(\mathcal{F})$.

Conjecture

The ideal \mathcal{F} is prime.

Decomposing Tensors into Frames

A general tensor $T \in S^d(\mathbb{R}^n)$ has rank $\lfloor \frac{1}{n} \binom{n+d-1}{d} \rfloor$. An odeco tensor $T = \sum_{i=1}^n \lambda_i v_i^{\otimes d}$ has rank n.

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

<u>Idea:</u> Let $V := (\mathbf{v}_1, \cdots, \mathbf{v}_r) \in (\mathbb{R}^n)^r$ be a finite unit norm tight frame , i.e.

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

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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 $(\mathbf{v}_1, ..., \mathbf{v}_r)$ 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 tensor power method

Conjecture

Let r = n + 1 and $T = \sum_{j=1}^{n+1} \lambda_j \mathbf{v}_j^{\otimes d}$ with $\lambda_1, ..., \lambda_{n+1} > 0$. Then, $\mathbf{v}_1, ..., \mathbf{v}_{n+1}$ 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

$$T = \alpha \begin{pmatrix} 1 \\ 0 \end{pmatrix}^{\otimes 4} + \begin{pmatrix} 0 \\ 1 \end{pmatrix}^{\otimes 4} + \begin{pmatrix} 1 \\ 1 \end{pmatrix}^{\otimes 4} + \begin{pmatrix} 1 \\ -1 \end{pmatrix}^{\otimes 4},$$

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.

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