

ENCODING ORTHONORMAL TENSOR DECOMPOSITION AS SVD

Theorem-1: $u_1, u_2 \in \mathbb{R}^m$ $v_1, v_2 \in \mathbb{R}^n$ and $u_1 \perp u_2$, then

i) $\langle A, B \rangle = \text{Tr}(A^T B) = 0$ where $A = u_1 v_1^T$, $B = u_2 v_2^T$

ii) $\langle C, D \rangle = \text{Tr}(C^T D) = 0$ where $C = v_1 u_1^T$, $D = v_2 u_2^T$

Proof: $u_1 \perp u_2 \Leftrightarrow u_1^T u_2 = 0 = u_2^T u_1$

i) $\langle A, B \rangle = \langle u_1 v_1^T, u_2 v_2^T \rangle = \text{Tr}((u_1 v_1^T)^T u_2 v_2^T)$
 $= \text{Tr}(v_1 \underbrace{u_1^T u_2}_{=0} v_2^T) = \text{Tr}(v_1 \cdot 0 \cdot v_2^T) = \text{Tr}(0) = 0$

ii) $\langle C, D \rangle = \langle A^T, B^T \rangle = \text{Tr}(A^{TT} B^T) = \text{Tr}(A B^T)$
 $= \text{Tr}(u_1 v_1^T (u_2 v_2^T)^T) = \text{Tr}(u_1 v_1^T v_2 u_2^T)$
 $= \text{Tr}(\underbrace{u_2^T u_1}_{=0} v_1^T v_2) = \text{Tr}(0) = 0$

Alternatively, note $\langle A^T, B^T \rangle = \langle A, B \rangle$.

Orthonormal CP-decomposition

Let $T \in \mathbb{R}^{m \times n \times p}$, then $T = \sum_i \alpha_i (\hat{a}_i \circ \hat{b}_i \circ \hat{c}_i)$ is called the rank- α orthonormal CP-decomposition of T iff $a_i \perp a_j \forall i \neq j$, $b_i \perp b_j \forall i \neq j$, and $\|a_i\|_2 = 1$, $\|b_i\|_2 = 1$ and $\|c_i\|_2 = 1$.

This can be said more compactly as $\hat{c}_i^T \hat{c}_j = \delta_{ij} = \hat{a}_i^T \hat{a}_j$

where $\delta_{ij} = \begin{cases} 0, & i \neq j \\ 1, & \text{otherwise} \end{cases}$

Theorem-2: Let \vec{u} & \vec{v} be in \mathbb{R}^m & \mathbb{R}^n respectively.

Let $u^T \cdot u = 1 = v^T \cdot v$

Then $\|A\|_F = 1 = \|\text{vec}(A)\|_2$



Proof:- $A = uv^T$, $\|A\|_F^2 = \langle A, A \rangle = \text{Tr}(A^T A) = \text{Tr}((uv^T)^T uv^T)$

$$1 = \text{Tr}(\underbrace{v^T v}_1) = \text{Tr}(vv^T) = \text{Tr}(\underbrace{v u^T u}_1 v^T)$$

Rank r - Orthogonal CP decomposition \equiv rank r SVD

Let $T = \sum_{i=1}^r \alpha_i (\vec{a}_i \circ \vec{b}_i \circ \vec{c}_i)$ be the orthogonal CP decomposition of $T \in \mathbb{R}^{m \times n \times p}$. Let $A_i = a_i \circ b_i = a_i b_i^T$. Let $\vec{z}_i = \text{vec}(A_i)$

$$\vec{z}_i^T \vec{z}_j = \langle A_i, A_j \rangle = 0 \text{ as } a_i \perp a_j \text{ (theorem-1)}$$

$$\text{and } \|A_i\|_F^2 = \|\vec{z}_i\|_2^2 = \langle A_i, A_i \rangle = 1$$

$$\text{Hence } \vec{z}_i^T \vec{z}_j = \delta_{ij} = \begin{cases} 1, & \text{if } i=j \\ 0, & \text{otherwise} \end{cases}$$

Let $\mathbb{R}^{mn} \ni \underline{\text{dflate}}(T)_{:,i} := \text{vec}(T_{:,:,i})$. Corollary
 $\underline{\text{dflate}}(a \circ b \circ c) = \text{vec}(a \circ b) \circ c$

such that $M = \underline{\text{dflate}}(T) \in \mathbb{R}^{mn \times p}$

Linearity of $\text{vec}(\cdot)$ & $\underline{\text{dflate}}(\cdot)$:

It isn't hard to see $\text{vec}(\alpha A + \beta B) = \alpha \text{vec}(A) + \beta \text{vec}(B)$
 $\& \underline{\text{dflate}}(\alpha_1 T_1 + \alpha_2 T_2) = \alpha_1 \underline{\text{dflate}}(T_1) + \alpha_2 \underline{\text{dflate}}(T_2)$
 $\forall \alpha, \beta \in \mathbb{R}$

$$M = \underline{\text{dflate}}(T) = \underline{\text{dflate}}\left(\sum_{i=1}^r \alpha_i a_i \circ b_i \circ c_i\right) = \sum_{i=1}^r \alpha_i \underline{\text{dflate}}(a_i \circ b_i \circ c_i)$$

$$= \sum_{i=1}^r \alpha_i \text{vec}(a_i \circ b_i) \circ c_i = \sum_{i=1}^r \alpha_i \text{vec}(A_i) \circ c_i = \sum_{i=1}^r \alpha_i \vec{z}_i \circ c_i$$

where $Z = \begin{bmatrix} \vec{z}_1 \\ \vdots \\ \vec{z}_r \end{bmatrix}$ $\alpha = \text{diag}([\alpha_i])$ $C = \begin{bmatrix} c_1 \\ \vdots \\ c_r \end{bmatrix}$

$$\left\{ \begin{array}{l} Z \propto C^T \\ \text{(SVD)} \end{array} \right. = \sum_{i=1}^r \alpha_i \vec{z}_i c_i^T$$

MATRIX FACTORIZATION AS TENSOR DECOMPOSITION

Notations: Let $\vec{v} \in \mathbb{R}^{q \times p}$.

define $\text{mat}(\vec{v}) \in \mathbb{R}^{q \times p}$ the matrix such that

$$\vec{v} = \text{vec}(\text{mat}(\vec{v})) \text{ and } A = \text{mat}(\text{vec}(A)) \text{ for } A \in \mathbb{R}^{q \times p}$$

$$\text{i.e. } \text{mat} = \text{vec}^{-1}$$

\Rightarrow Let $M \in \mathbb{R}^{m \times n \times p}$ be any matrix.

\Rightarrow Define $T = \text{nflate}(M) \in \mathbb{R}^{m \times n \times p}$ such that

$$\text{nflate}(M)_{:, :, i} = \text{mat}(M_{:, i})$$

\Rightarrow Clearly nflate is linear, i.e. $\text{nflate}(\gamma_1 M_1 + \gamma_2 M_2) = \gamma_1 \text{nflate}(M_1) + \gamma_2 \text{nflate}(M_2)$

$$\text{or } \text{nflate}\left(\sum_i \gamma_i M_i\right) = \sum_i \gamma_i \text{nflate}(M_i) \text{ for } \gamma_i \in \mathbb{R}.$$

Remark: $\text{dflate}(\text{nflate}(M))_{:, :, i} = \text{vec}(\text{nflate}(M)_{:, :, i})$
 $= \text{vec}(\text{mat}(M_{:, i})) = M_{:, i}$

$$\text{Hence, } \text{dflate}(\text{nflate}(M)) = M$$

$$\begin{aligned} \text{Similarly } \text{nflate}(\text{dflate}(T))_{:, :, i} &= \text{mat}(\text{dflate}(T)_{:, :, i}) \\ &= \text{mat}(\text{vec}(T_{:, :, i})) \\ &= T_{:, :, i} \end{aligned}$$

$$\text{Hence, } \text{nflate}(\text{dflate}(T)) = T$$

So, we have $\text{nflate} = \text{dflate}^{-1}$

Corollary: If $\vec{w} = \text{vec}(\vec{x} \circ \vec{y})$, then $\text{nflate}(\vec{w} \circ \vec{z}) = \vec{x} \circ \vec{y} \circ \vec{z}$
 $\text{nflate}(\text{vec}(\vec{x} \circ \vec{y}) \circ \vec{z}) = \vec{x} \circ \vec{y} \circ \vec{z}$



Matrix Factorization :- Let $M \in \mathbb{R}^{m \times n \times p}$ be a matrix such that

$$M = \sum_{i=1}^r d_i e_i^T = \sum_{i=1}^r \vec{d}_i \circ \vec{e}_i \text{ any rank-} r \text{ factorization of } M \text{ such that } \vec{d}_i \in \mathbb{R}^{mn} \text{ \& } \vec{e}_i \in \mathbb{R}^p$$

$$\text{Let } D_i = \text{mat}(\vec{d}_i) \in \mathbb{R}^{m \times n} \Rightarrow \text{vec}(D_i) = \vec{d}_i$$

$$\text{Let } D_i = \sum_{j=1}^k a_j b_j^T = \sum_{j=1}^k \vec{a}_j \circ \vec{b}_j \text{ be any rank-} k \text{ factorization of } D_i$$

$$\text{Hence } M = \sum_{i=1}^r \vec{d}_i \circ \vec{e}_i = \sum_{i=1}^r \text{vec}(D_i) \circ \vec{e}_i$$

$$= \sum_{i=1}^r \text{vec}\left(\sum_{j=1}^k \vec{a}_j \circ \vec{b}_j\right) \circ \vec{e}_i$$

$$= \sum_{i=1}^r \sum_{j=1}^k \text{vec}(\vec{a}_j \circ \vec{b}_j) \circ \vec{e}_i$$

$$\Rightarrow \text{nflate}(M) = \text{nflate}\left(\sum_{i=1}^r \sum_{j=1}^k \text{vec}(\vec{a}_j \circ \vec{b}_j) \circ \vec{e}_i\right)$$

$$T = \sum_{i=1}^r \sum_{j=1}^k \text{nflate}[\text{vec}(\vec{a}_j \circ \vec{b}_j) \circ \vec{e}_i]$$

$$T = \sum_{i=1}^r \sum_{j=1}^k \vec{a}_j \circ \vec{b}_j \circ \vec{e}_i \quad \because \text{Corollary on prev. page.}$$

$$\text{Let } s = rk, \text{ then } T = \sum_{s=1}^{rk} \vec{a}_{\lceil \frac{s}{r} \rceil} \circ \vec{b}_{\lceil \frac{s}{r} \rceil} \circ \vec{e}_{\lceil \frac{s}{k} \rceil}$$

where $\lceil \cdot \rceil$ is the ceil function or the "smallest integer value" function.

$$\text{nflate}(M) = T = \sum_{s=1}^{rk} \vec{a}_{\lceil \frac{s}{r} \rceil} \circ \vec{b}_{\lceil \frac{s}{r} \rceil} \circ \vec{e}_{\lceil \frac{s}{k} \rceil} \text{ is a rank 'rk' tensor}$$

decomposition/factorization of T where all of the tensor atoms are unique but the factors \vec{a}_i, \vec{b}_i & \vec{e}_i are sometimes repeated and chosen out of k of \vec{a}_s / \vec{b}_s and r of \vec{e}_s .