

Tensor Algebra

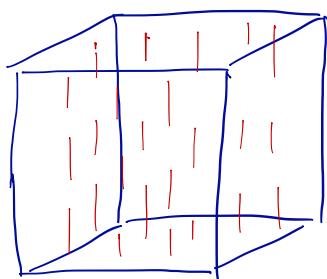
Indexing

Tensor $T \in \mathbb{R}^{l_1 \times l_2 \times \dots \times l_n}$ has elements

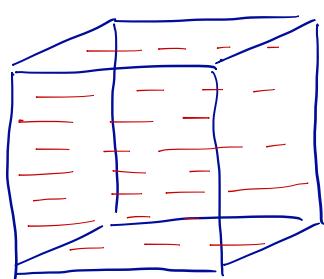
$$T = (t_{i_1 i_2 \dots i_n})$$

We can write $\vec{i} = i_1 i_2 \dots i_n$ to denote an index vector $\vec{i} \in [l_1] \times [l_2] \times \dots \times [l_n]$, where $[n] = \{1, 2, 3, \dots, n\}$.

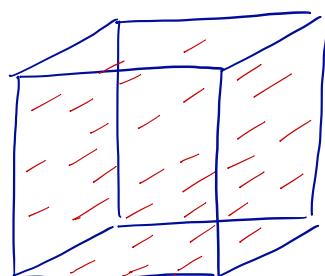
The fibres are denoted $\vec{t}_{i_1 i_2 \dots i_{n-1} : i_n \dots i_n}$



$$t_{: i_2 i_3}$$



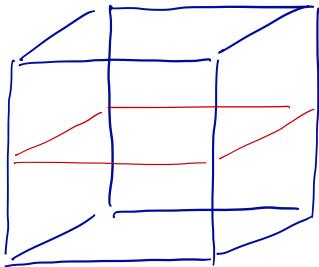
$$t_{i_1 : i_3}$$



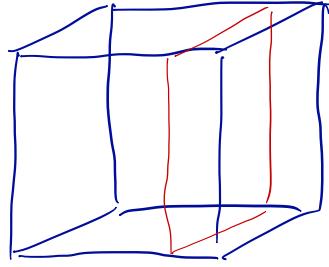
$$t_{i_1 i_2 :}$$

The slices are denoted

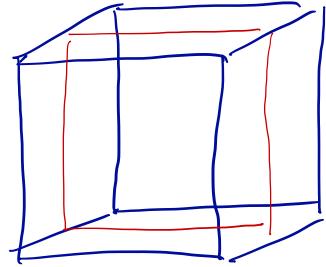
$$T_{i_1 \dots i_j \dots i_l \dots i_k \dots i_n}$$



$$T_{i_1 \dots}$$



$$T_{\dots i_2 \dots}$$



$$T_{\dots \dots i_3}$$

For 3-way tensors, the frontal slice $T_{\dots \dots i_3}$ has a special short-hand notation

$$T_{\dots \dots i_3} = T_{i_3}$$

Matricization

A tensor T can be matricized or unfolded or flattened in a matrix by reordering its elements. Mode-n matricization takes the mode-n fibres and stacks them as the columns of

the new matrix. If $T \in \mathbb{R}^{l_1 \times l_2 \times \dots \times l_n}$, its mode-1 matricization is a matrix

$$T_{(1)} \in \mathbb{R}^{l_1 \times (l_2 l_3 \dots l_n)}$$

and in general, the mode-n matricization is a matrix

$$T_{(n)} \in \mathbb{R}^{l_n \times (l_1 l_2 \dots l_{n-1} l_{n+1} \dots l_n)}$$

In mode-n matricization, element at (i_1, i_2, \dots, i_n) maps to (i_n, j) , where

$$j = 1 + \sum_{\substack{k=1 \\ k \neq n}}^N (i_k - 1) J_k, \text{ where}$$

$$J_k = \prod_{\substack{m=1 \\ m \neq n}}^{k-1} l_m.$$

An example (next page) will clarify the process.

Matricization example

Let $\mathcal{T} \in \mathbb{R}^{4 \times 3 \times 2}$ with frontal slices

$$T_1 = \begin{pmatrix} 1 & 5 & 9 \\ 2 & 6 & 10 \\ 3 & 7 & 11 \\ 4 & 8 & 12 \end{pmatrix} \text{ and } T_2 = \begin{pmatrix} 13 & 17 & 21 \\ 14 & 18 & 22 \\ 15 & 19 & 23 \\ 16 & 20 & 24 \end{pmatrix}.$$

Now we have:

$$T_{(1)} = \begin{pmatrix} 1 & 5 & 9 & 13 & 17 & 21 \\ 2 & 6 & 10 & 14 & 18 & 22 \\ 3 & 7 & 11 & 15 & 19 & 23 \\ 4 & 8 & 12 & 16 & 20 & 24 \end{pmatrix}$$

$$T_{(2)} = \begin{pmatrix} 1 & 2 & 3 & 4 & 13 & 14 & 15 & 16 \\ 5 & 6 & 7 & 8 & 17 & 18 & 19 & 20 \\ 9 & 10 & 11 & 12 & 21 & 22 & 23 & 24 \end{pmatrix}$$

$$T_{(3)} = \begin{pmatrix} 1 & 2 & 3 & 4 & 5 & \dots & 9 & 10 & 11 & 12 \\ 13 & 17 & 15 & 16 & 18 & \dots & 21 & 22 & 23 & 24 \end{pmatrix}$$

Vectorizations

A vectorization of a tensor stacks the columns of its mode-1 matricization in a column vector. \mathcal{T} as above, we have $\text{vec}(\mathcal{T}) = (1, 2, 3, \dots, 24)^T$.

Tensor times {scalar, vector, matrix, tensor}

Tensor times a scalar, $\alpha \underline{J}$, scales every element of J by $\alpha \in \mathbb{R}$,

$$\alpha \underline{J} = (\alpha t_{ij}).$$

The n-mode vector-tensor product of a tensor $\underline{J} \in \mathbb{R}^{l_1 \times l_2 \times \dots \times l_n}$ with vector $\vec{v} \in \mathbb{R}^{l_n}$ is

$$\begin{aligned} \underline{J} \vec{x}_n \vec{v} &\in \mathbb{R}^{l_1 \times l_2 \times \dots \times l_{n-1} \times l_{n+1} \times \dots \times l_N} \\ (\underline{J} \vec{x}_n \vec{v})_{i_1 \dots i_{n-1} i_n i_{n+1} \dots i_N} &= \sum_{i_n=1}^{l_n} t_{i_1 i_2 \dots i_n i_{n+1} \dots i_N} v_{i_n} \\ &= [\langle \vec{t}_{i_1 \dots i_{n-1} i_n i_{n+1} \dots i_N}, \vec{v} \rangle]_{i_1 \dots i_{n-1} i_n i_{n+1} \dots i_N} \end{aligned}$$

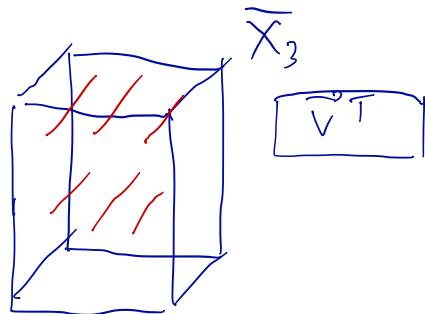
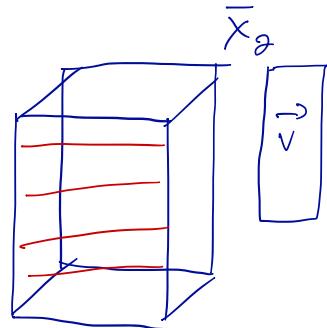
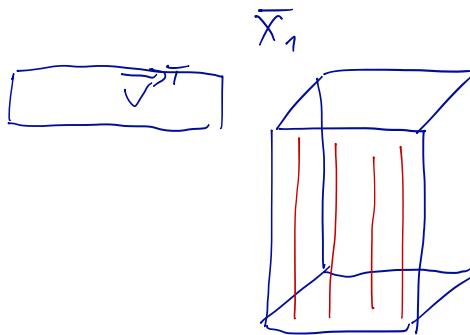
where $\langle \cdot, \cdot \rangle$ is the inner product of two vectors. The idea is to take the inner products between all mode-n fibres and \vec{v} .

In tensor-vector products, precedence matters:

$$\underline{J} \vec{x}_m \vec{a} \vec{x}_n \vec{b} = (\underline{J} \vec{x}_m \vec{a}) \vec{x}_{n-1} \vec{b} = (\underline{J} \vec{x}_n \vec{b}) \vec{x}_m \vec{a}$$

If \underline{J} is 2-way (a matrix), we have

$$\underline{J} \bar{x}_1 \vec{v} = \vec{v}^T \underline{J} \quad \text{and} \quad \underline{J} \bar{x}_2 \vec{v} = \underline{J} \vec{v}.$$



Let $\underline{J} \in \mathbb{R}^{4 \times 3 \times 2}$ with frontal slices

$$T_1 = \begin{pmatrix} 1 & 5 & 9 \\ 2 & 6 & 10 \\ 3 & 7 & 11 \\ 4 & 8 & 12 \end{pmatrix} \quad \text{and} \quad T_2 = \begin{pmatrix} 13 & 17 & 21 \\ 14 & 18 & 22 \\ 15 & 19 & 23 \\ 16 & 20 & 24 \end{pmatrix},$$

and let $\vec{v} = (5, 10)^T$. Then

$$\underline{J} \bar{x}_3 \vec{v} = \begin{pmatrix} 5+130 & 25+170 & 45+210 \\ 10+140 & 30+180 & 50+220 \\ 15+150 & 35+190 & 55+230 \\ 20+160 & 40+200 & 60+240 \end{pmatrix} = \begin{pmatrix} 135 & 195 & 255 \\ 150 & 210 & 270 \\ 165 & 225 & 285 \\ 180 & 240 & 300 \end{pmatrix}.$$

The n -mode matrix-tensor product of tensor $\underline{T} \in \mathbb{R}^{l_1 \times l_2 \times \dots \times l_n}$ and matrix $M \in \mathbb{R}^{j \times l_n}$ is

$$\begin{aligned} \underline{T} \times_n M &\in \mathbb{R}^{l_1 \times \dots \times l_{n-1} \times j \times l_{n+1} \times \dots \times l_n} \\ (\underline{T} \times_n M)_{i_1 \dots i_{n-1} j i_{n+1} \dots i_n} &= \sum_{i_n=1}^{l_n} t_{i_1 i_2 \dots i_n} m_{j i_n} \\ &= [M \vec{x}_{i_1 \dots i_{n-1} i_{n+1} \dots i_n}]_{i_1 \dots i_n} \end{aligned}$$

That is, we multiply each mode- n fibre with M . Equivalently, we can use unfolding:

$$\underline{S} = \underline{T} \times_n M \iff S_{(n)} = M T_{(n)}.$$

If \underline{T} is 2-way (i.e. a matrix), $T_{(1)} = \underline{T}$ and $T_{(2)} = \underline{T}^T$. So

$$\underline{T} \times_1 M = M \underline{T} \quad \text{and} \quad \underline{T} \times_2 M = M \underline{T}^T$$

The order of tensor-matrix multiplications over different modes doesn't matter: if $m \neq n$, then

$$\underline{\mathcal{I}} \times_m A \times_n B = \underline{\mathcal{I}} \times_n B \times_m A.$$

If the modes are the same and $A \in \mathbb{R}^{J \times I_n}$ and $B \in \mathbb{R}^{K \times J}$, then

$$\underline{\mathcal{I}} \times_n A \times_n B = \underline{\mathcal{I}} \times_n (BA).$$

Let $\underline{\mathcal{I}} \in \mathbb{R}^{4 \times 3 \times 2}$ with frontal slices

$$T_1 = \begin{pmatrix} 1 & 5 & 9 \\ 2 & 6 & 10 \\ 3 & 7 & 11 \\ 4 & 8 & 12 \end{pmatrix} \quad \text{and} \quad T_2 = \begin{pmatrix} 13 & 17 & 21 \\ 14 & 18 & 22 \\ 15 & 19 & 23 \\ 16 & 20 & 24 \end{pmatrix},$$

and let

$$M = \begin{pmatrix} 1 & 2 & 4 & 6 \\ 10 & 20 & 40 & 60 \end{pmatrix}$$

Then, if $\underline{S} = \underline{\mathcal{I}} \times_1 M$, we have

$$S_{111} = 1 \cdot 1 + 2 \cdot 2 + 4 \cdot 3 + 6 \cdot 4 = 41 \quad \text{and}$$

$$S_1 = \begin{pmatrix} 41 & 93 & 145 \\ 410 & 930 & 1450 \end{pmatrix} \quad \text{and} \quad S_2 = \begin{pmatrix} 192 & 299 & 301 \\ 1970 & 2490 & 3010 \end{pmatrix}$$

The tensor inner product of two tensors $\underline{S}, \underline{T} \in \mathbb{R}^{l_1 \times l_2 \times \dots \times l_N}$ is the sum of their element-wise products,

$$\langle \underline{S}, \underline{T} \rangle = \sum_{i_1=1}^{l_1} \sum_{i_2=1}^{l_2} \dots \sum_{i_N=1}^{l_N} S_{i_1 i_2 \dots i_N} T_{i_1 i_2 \dots i_N}$$

Tensor norm

The norm of a tensor $\underline{T} \in \mathbb{R}^{l_1 \times l_2 \times \dots \times l_N}$ is the square-root of the sum of the squares of its elements:

$$\|\underline{T}\| = \left(\sum_{i_1=1}^{l_1} \sum_{i_2=1}^{l_2} \dots \sum_{i_N=1}^{l_N} t_{i_1 i_2 \dots i_N}^2 \right)^{1/2}$$

This can be alternatively be defined as $\sqrt{\langle \underline{T}, \underline{T} \rangle}$, $\|\underline{T}_{(1)}\|_F$, $\|\underline{T}_{(i)}\|_F$, $\|\text{vec}(\underline{T})\|_2$, or $\sqrt{\text{trace}(\underline{T}_{(1)} \underline{T}_{(1)}^T)}$, or using any other way to define the Euclidean/Frobenius norm of a vector/matrix.

Symmetry

Tensor \underline{I} is cubical if all of its modes have the same dimensionality:

$$\underline{I} \in \mathbb{R}^{l \times l \times l \times \dots \times l}$$

Cubical tensor is (super-)symmetric, if its elements remain constant under any permutation of the indices. If $\underline{I} \in \mathbb{R}^{l \times l \times l}$, it is symmetric if and only if

$$t_{ijk} = t_{ikj} = t_{jik} = t_{jki} = t_{kij} = t_{ksi} \text{ for all } i, j, k \in [l]$$

Tensor $\underline{I} \in \mathbb{R}^{l_1 \times l_2 \times \dots \times l_N}$ is (hyper-)diagonal if $t_{i_1 i_2 \dots i_N} \neq 0$ only if $i_1 = i_2 = \dots = i_N$. If $l_1 = l_2 = \dots = l_N$, \underline{I} is also symmetric.

If the diagonal entries are all 1s, tensor behaves similarly to the identity matrix.

