

Multilinear Algebra

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Overview

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 - Multilinearity
 - Dual Space
- 2 Tensors
 - Tensor Product
 - Basis of $\mathcal{T}_q^p(V)$
- 3 Component Representation
 - Kronecker Product
 - Components
 - Comparison

Multilinear Functions

Definition

A function $f : V \mapsto W$, where V and W are vector spaces over a field F , is linear if for all x, y in V and all α, β in F

$$f(\alpha x + \beta y) = \alpha f(x) + \beta f(y).$$

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A function $f : V \times U \mapsto W$, where V , U , and W are vector spaces over a field F , is **bilinear** if for all x, y in V and all α, β in F

$$f(\alpha x_1 + \beta x_2, y) = \alpha f(x_1, y) + \beta f(x_2, y), \text{ and}$$
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A function $f : V_1 \times \cdots \times V_s \mapsto W$, where $\{V_i\}_{i=1}^s$ and W are vector spaces over a field F , is **s-linear** if for all x_i, y_i in V_i and all α, β in F

$$f(v_1, \dots, \alpha x_i + \beta y_i, \dots, v_s) = \\ \alpha f(v_1, \dots, x_i, \dots, v_s) + \beta f(v_1, \dots, y_i, \dots, v_s),$$

for all indices i in $\{1, \dots, s\}$.

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For any collection of vector spaces $\{V_i\}_{i=1}^s$, and any collection of **linear functions** $f_i : V_i \mapsto \mathbb{R}$, the function

$$f(v_1, \dots, v_s) = \prod_{i=1}^s (f_i(v_i))$$

is s -linear.

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- Elements of V are vectors while elements of V^* are **covectors**
- $(V^*)^*$ is identical to V .
- Notation: $\langle v^*, v \rangle$ denotes the value of v^* evaluated at v . For our purposes, consider it the inner product of v and $(v^*)^T$.

Tensors

Definition

A **tensor of order (p, q)** is a $(p + q)$ -linear map

$$T : \underbrace{V^* \times \cdots \times V^*}_{p \text{ times}} \times \underbrace{V \times \cdots \times V}_{q \text{ times}} \mapsto \mathbb{R}.$$

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- $\mathcal{T}_1^0(V) = V^*$, $\mathcal{T}_0^1(V) = V$, and $\mathcal{T}_1^1(V) \cong L(V : V)$.
- That is, lower order tensors are the 1 and 2 dimensional arrays we usually work with.

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Example

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- Recall that this is a special case of our earlier example, as $(v \otimes v^*)$ is the product of two linear functions.

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$$v_1 \otimes \cdots \otimes v_p \otimes v^1 \otimes \cdots \otimes v^q : \underbrace{V^* \times \cdots \times V^*}_{p \text{ times}} \times \underbrace{V \times \cdots \times V}_{q \text{ times}} \mapsto \mathbb{R},$$

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$$\begin{aligned} & (v_1 \otimes \cdots \otimes v_p \otimes v^1 \otimes \cdots \otimes v^q)(u^1, \dots, u^p, u_1, \dots, u_q) \\ &= \langle u^1, v_1 \rangle \cdots \langle u^p, v_p \rangle \langle v^1, u_1 \rangle \cdots \langle v^q, u_q \rangle \end{aligned}$$

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- However, we can use simple tensors to build a basis of $\mathcal{T}_q^p(V)$.

Theorem

For any basis of V , $B = \{e_i\}_{i=1}^N$, there exists a unique **dual basis** of V^* relative to B , denoted $\{e^j\}_{j=1}^N$ and defined as

$$\langle e^j, e_i \rangle = \delta_i^j = \begin{cases} 1, & \text{if } i = j \\ 0, & \text{if } i \neq j \end{cases}.$$

Theorem

For any basis $\{e_i\}_{i=1}^N$ of V , and the corresponding dual basis $\{e^j\}_{j=1}^N$ of V^* , the set of simple tensors

$$\{e_{i_1} \otimes \cdots \otimes e_{i_p} \otimes e^{j_1} \otimes \cdots \otimes e^{j_q}\}$$

for all combinations of $\{i_k\}_{k=1}^p \in \{1, \dots, N\}$ and $\{j_z\}_{z=1}^q \in \{1, \dots, N\}$, forms a basis of $\mathcal{T}_q^p(V)$.

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- The size of this basis is $N^{(p+q)}$.
- Simplified proof in my paper, but our relation of linear dependence is nasty ($p + q$ nested sums).

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For two matrices $A_{m \times n}$ and $B_{p \times q}$, the **Kronecker product** of A and B is defined as

$$A \otimes B = \begin{pmatrix} [A]_{1,1}B & [A]_{1,2}B & \cdots & [A]_{1,n}B \\ [A]_{2,1}B & [A]_{2,2}B & \cdots & [A]_{2,n}B \\ \vdots & \vdots & \ddots & \vdots \\ [A]_{m,1}B & [A]_{m,2}B & \cdots & [A]_{m,n}B \end{pmatrix}$$

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- Can be represented by 2-dimensional array, but we consider this product to be a list of lists, table of lists, list of tables, table of tables, ect.

Components as Basis Images

Definition

In general, we define the **components** of $T \in \mathcal{T}_q^p(V)$ to be the $(p + q)$ -indexed scalars

$$A_{j_1, \dots, j_q}^{i_1, \dots, i_p} = A(e^{i_1}, \dots, e^{i_p}, e_{j_1}, \dots, e_{j_q}).$$

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$$A_{j_1, \dots, j_q}^{i_1, \dots, i_p} = A(e^{i_1}, \dots, e^{i_p}, e_{j_1}, \dots, e_{j_q}).$$

- For vectors, this is exactly how we define components ($\langle v, e_i \rangle = [v]_i$).

Components as Basis Images

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- For vectors, this is exactly how we define components ($\langle v, e_i \rangle = [v]_i$).
- If T is a simple tensor, then the $(p + q)$ -dimensional array formed by $A_{j_1, \dots, j_q}^{i_1, \dots, i_p}$ is equal to the Kronecker product of the vectors and covectors which make up T .

Example

For $V = \mathbb{R}^2$, consider the vectors $u = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$, $v^* = [2 \ 1]$, and $w^* = [1 \ 3]$. Let $A = u \otimes v^* \otimes w^*$ and consider

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$$\begin{aligned} A_1^{1,1} &= A(e^1, e_1, e_1) \\ &= \langle [1 \ 0], \begin{bmatrix} 1 \\ 1 \end{bmatrix} \rangle \langle [2 \ 1], \begin{bmatrix} 1 \\ 0 \end{bmatrix} \rangle \langle [1 \ 3], \begin{bmatrix} 1 \\ 0 \end{bmatrix} \rangle \\ &= 2 \end{aligned}$$

$$A_1^{2,1} = A(e^1, e_2, e_1) = 1$$

$$A_1^{1,2} = A(e^1, e_1, e_2) = 6$$

$$\vdots$$

Example

Or, we can take the Kronecker product $u \otimes v^* \otimes w^*$ to get

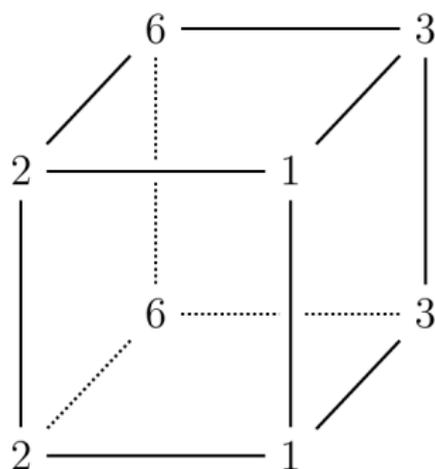
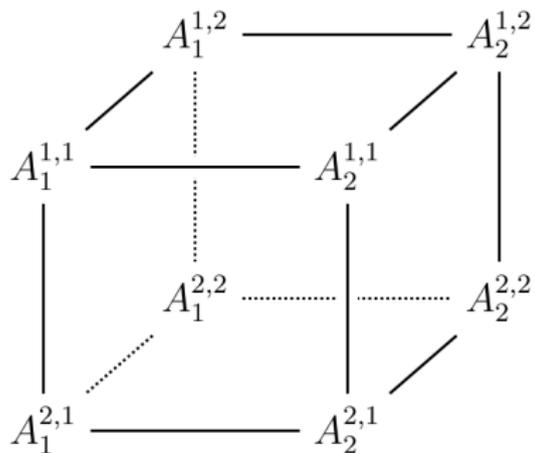
Example

Or, we can take the Kronecker product $u \otimes v^* \otimes w^*$ to get

$$\begin{aligned}
 u \otimes v^* \otimes w^* &= \begin{bmatrix} 1 \\ 1 \end{bmatrix} \otimes [2 \ 1] \otimes [1 \ 3] \\
 &= \begin{bmatrix} 2 & 1 \\ 2 & 1 \end{bmatrix} \otimes [1 \ 3] \\
 &= \begin{bmatrix} 2 \begin{bmatrix} 1 & 3 \end{bmatrix} & 1 \begin{bmatrix} 1 & 3 \end{bmatrix} \\ 2 \begin{bmatrix} 1 & 3 \end{bmatrix} & 1 \begin{bmatrix} 1 & 3 \end{bmatrix} \end{bmatrix} \\
 &= \begin{bmatrix} \begin{bmatrix} 2 & 6 \end{bmatrix} & \begin{bmatrix} 1 & 3 \end{bmatrix} \\ \begin{bmatrix} 2 & 6 \end{bmatrix} & \begin{bmatrix} 1 & 3 \end{bmatrix} \end{bmatrix},
 \end{aligned}$$

Either way, we get

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The End