## Week 1

# **Tensors**

#### 1.1 Course Motivation

3/28: • Motivation for the course and an overview of Guillemin and Haine (2018).

### 1.2 Defining Tensors and Their Operations

3/30: • Plan:

- More (multi)linear algebra.

• Dual spaces.

 $\bullet$  Let V be an n-dimensional real vector space.

• Hom  $(V,\mathbb{R})$ : The set of all homomorphisms (i.e., linear maps) from V to  $\mathbb{R}$ . Also known as  $V^*$ .

• Dual basis (for  $V^*$ ): The set of linear transformations from V to  $\mathbb{R}$  defined by

$$e_j \mapsto \begin{cases} 1 & j = i \\ 0 & j \neq i \end{cases}$$

where  $e_1, \ldots, e_n$  is a basis of V. Denoted by  $e_1^*, \ldots, e_n^*$ .

• Check:  $e_1^*, \ldots, e_n^*$  are a basis for  $V^*$ .

– Are they linearly independent? Let  $c_1e_1^* + \cdots + c_ne_n^* = 0 \in \text{Hom}(V, \mathbb{R})$ . Then

$$c_i = (c_1 e_1^* + \dots + c_n e_n^*)(e_i) = 0 \in \mathbb{R}$$

as desired.

- Span? Let  $\varphi \in \text{Hom}(V, \mathbb{R})$ . Then we can verify that

$$\varphi(e_1)e_1^* + \dots + \varphi(e_n)e_n^* = \varphi$$

- $\blacksquare$  We prove this by verifying the previous statement on the basis of V (if two linear transformations have the same action on the basis of a vector space, they are equal).
- With a choice of basis for V, we obtain an isomorphism  $\varepsilon: V \to V^*$  with the mapping  $e_i \mapsto e_i^*$  for all i.
- The dual space is known as such because  $(V^*)^* \cong V$ , where  $\cong$  is **canonical** (no choice of basis is needed).

- One more property of dual spaces: **functoriality**.
  - Given a linear transformation  $A: V \to W$ , we know that  $A^*: W^* \to V^*$  where  $A^*$  is the transpose of A. In particular, if  $\varphi \in W^*$ , then  $\varphi \circ A: V \to \mathbb{R}$ .
  - Claim:  $A^*$  is linear.
- Functoriality: If  $A: V \to W$  and  $B: W \to U$ , then  $B^*: U^* \to W^*$  and  $A^*: W^* \to V^*$ . The functoriality statement is that  $(B \circ A)^* = A^* \circ B^*$ .
- $A^*$  is the **pullback** (or transpose) of A.
- Let  $v_1, \ldots, v_n$  be a basis for V and  $w_1, \ldots, w_m$  be a basis for W. Then  $[A]_{v_1, \ldots, v_n}^{w_1, \ldots, w_m} = A$  is the matrix of the linear transformation A with respect to these bases. Then if  $v_1^*, \ldots, v_n^*$  and  $w_1^*, \ldots, w_m^*$  are the corresponding dual bases, then  $[A^*]_{v_1^*, \ldots, v_n^*}^{w_1^*, \ldots, w_m^*} = A^T$ . We can and should verify this for ourselves.
- This is over the real numbers, so  $A^*$  is just the transpose because there are no complex numbers of which to take the conjugate!
- A generalization: Tensors.
- k-tensor: A multilinear map

$$T: \underbrace{V \times \cdots \times V}_{k \text{ times}} \to \mathbb{R}$$

• Multilinear (map T): A function T such that

$$T(v_1, \dots, v_i^1 + v_i^2, \dots, v_k) = T(v_1, \dots, v_i^1, \dots, v_k) + T(v_1, \dots, v_i^2, \dots, v_k)$$
$$T(v_1, \dots, \lambda v_i, \dots, v_k) = \lambda T(v_1, \dots, v_i, \dots, v_k)$$

for all  $(v_1, \ldots, v_k) \in V^k$ .

- The determinant is an n-tensor!
- 1-tensors are just covectors.
- $\mathcal{L}^k(V)$ : The vector space of all k-tensors on V.
- Calculating dim  $\mathcal{L}^k(V)$ . (Answer not given in this class.)
- Let  $A: V \to W$ . Then  $A^*: \mathcal{L}^k(W) \to \mathcal{L}^k(V)$ .
  - Check  $(A \circ B)^* = B^* \circ A^*$ .
- Multi-index of n of length k: A k-tuple  $(i_1, \ldots, i_k)$  where each  $i_j \in \mathbb{N}$  satisfies  $1 \leq i_j \leq n$   $(j = 1, \ldots, k)$ . Denoted by I.
- Let  $e_1, \ldots, e_n$  be a basis for V.
- **Tensor product** (of  $T_1 \in \mathcal{L}^k(V)$ ,  $T_2 \in L^l(V)$ ): The function from  $V^{k+l}$  to  $\mathbb{R}$  defined by

$$(v_1, \ldots, v_{k+l}) \mapsto T_1(v_1, \ldots, v_k) T_2(v_{k+1}, \ldots, v_{k+l})$$

Denoted by  $T_1 \otimes T_2$ .

- Claims:
  - 1.  $T_1 \otimes T_2 \in L^{k+l}(V)$ .
  - 2.  $A^*(T_1 \otimes T_2) = A^*(T_1) \otimes A^*(T_2)$ .
- $e_I^*$ : The function  $e_{i_1}^* \otimes \cdots \otimes e_{i_k}^*$ , where  $I = (i_1, \dots, i_k)$  is a multi-index of n of length k.

- Claim: Letting I range over all  $n^k$  multi-indices of n of length k, the  $e_I^*$  are a basis for  $\mathcal{L}^k(V)$ .
- If  $V = \mathbb{R}$ , then  $V = \mathbb{R}e_1$ . If  $V = \mathbb{R}^2$ , then  $V = \mathbb{R}e_1 \oplus \mathbb{R}e_2$ .
- We know that  $L^1(V) = V^* = Re_1^*$ . Thus,  $e_1^* \otimes e_2^* : V \times V \to \mathbb{R}$ . Thus, for example,

$$(e_1^* \otimes e_2^*)((1,2),(3,4)) = e_1^*(1,2) \cdot e_2^*(3,4) = 1 \cdot 4 = 4$$

#### 1.3 The Tensor Product and Permutations

- 4/1: Plan: More multilinear algebra.
  - Properties of the tensor product.
  - Sign of a permutation.
  - Alternating tensors (lead into differential forms down the road).
  - Recall: V is an n-dimensional vector space over  $\mathbb{R}$  with basis  $e_1, \ldots, e_n$ .  $\mathcal{L}^k(V)$  is the vector space of k-tensors on V.  $\{e_I^* \mid I \text{ a multiindex of } n \text{ of length } k\}$  is a basis for  $\mathcal{L}^k(V)$ .
  - For example, if  $V = \mathbb{R}^2$  and  $T \in \mathcal{L}^2(V)$ , then

$$T(a_1e_1 + a_2e_2, b_1e_1 + b_2e_2) = a_1b_1T(e_1, e_1) + a_1b_2T(e_1, e_2) + a_2b_1T(e_2, e_1) + a_2b_2T(e_2, e_2)$$

- A basis of  $\mathcal{L}^2(V)$  is

$$\{e_1^* \otimes e_1^*, e_1^* \otimes e_2^*, e_2^* \otimes e_1^*, e_2^* \otimes e_2^*\}$$

- Recall that some basic properties are

$$e_1^* \otimes e_2^*((1,2),(3,4)) = 1 \cdot 4 = 4$$
  $e_2^* \otimes e_1^*((1,2),(3,4)) = 2 \cdot 3 = 6$ 

- It follows by the initial decomposition of T that

$$T = a_1b_1e_1^* \otimes e_1^* + a_1b_2e_1^* \otimes e_2^* + a_2b_1e_2^* \otimes e_1^* + a_2b_2e_2^* \otimes e_2^*$$

- Important consequence: To know the action of T on an arbitrary pair of vectors, you need only know its action on the basis; a higher-dimensional generalization of the earlier property.
- Note that

$$e_I^*(e_J) = \delta_{IJ} = \begin{cases} 1 & I = J \\ 0 & I \neq J \end{cases}$$

- Basic properties of the tensor product.
  - 1. Right-distributive: If  $T_1 \in \mathcal{L}^k(V)$  and  $T_2, T_3 \in \mathcal{L}^{\ell}(V)$ , then

$$T_1 \otimes (T_2 + T_3) = T_1 \otimes T_2 + T_1 \otimes T_3$$

2. Left-distributive: If  $T_1, T_2 \in \mathcal{L}^k(V)$  and  $T_3 \in \mathcal{L}^{\ell}(V)$ , then

$$(T_1 + T_2) \otimes T_3 = T_1 \otimes T_3 + T_2 \otimes T_3$$

3. Associative: If  $T_1 \in \mathcal{L}^k(V)$ ,  $T_2 \in \mathcal{L}^\ell(V)$ , and  $T_3 \in \mathcal{L}^m(V)$ , then

$$T_1 \otimes (T_2 \otimes T_3) = (T_1 \otimes T_2) \otimes T_2 = T_1 \otimes T_2 \otimes T_3$$

4. Scalar multiplication: If  $T_1 \in \mathcal{L}^k(V)$ ,  $T_2 \in \mathcal{L}^{\ell}(V)$ , and  $\lambda \in \mathbb{R}$ , then

$$(\lambda T_1) \otimes T_2 = \lambda (T_1 \otimes T_2) = T_1 \otimes (\lambda T_2)$$

- Note that the tensor product is not commutative.
- Aside: Defining the sign of a permutation.
- $S_A$ : The set of all automorphisms of A (bijections from A to A), where A is a set.
- $S_n$ : The set  $S_{[n]}$ .
- Given  $\sigma_1, \sigma_2 \in S_n, \sigma_1 \circ \sigma_2 \in S_n$ .
  - Thus,  $S_n$  is a group.
- Transposition: A function  $\tau \in S_n$  such that

$$\tau(k) = \begin{cases} j & k = i \\ i & k = j \\ k & k \neq i, j \end{cases}$$

for some  $i, j \in [n]$ . Denoted by  $\tau_{i,j}$ .

- Theorem: An element of  $S_n$  can be written as the product of transpositions (i.e., for all  $\sigma \in S_n$ , there exist  $\tau_1, \ldots, \tau_m \in S_n$  such that  $\sigma = \tau_1 \circ \cdots \circ \tau_m$ ).
- Sign (of  $\sigma \in S_n$ ): The number (mod 2) of transpositions whose product equals  $\sigma$ . Denoted by  $(-1)^{\sigma}$ , sign  $(\sigma)$ .
- Theorem: The sign of  $\sigma$  is well-defined. Additionally,

$$(-1)^{\sigma_1 \sigma_2} = (-1)^{\sigma_1} \cdot (-1)^{\sigma_2}$$

- Example: Consider the identity permutation.  $(-1)^{\sigma} = +1$ . We can think of this as the product of zero transpositions or, for instance, as the product of the two transpositions  $\tau_{1,2} \circ \tau_{1,2}$ . Another example would be  $\tau_{2,3} \circ \tau_{1,2} \circ \tau_{1,2} \circ \tau_{2,3}$ .
- Theorem: Let  $X_i$  be a rational or polynomial function for each  $i \in \mathbb{N}$ . Then

$$(-1)^{\sigma} = \prod_{i < j} \frac{X_{\sigma(i)} - X_{\sigma(j)}}{X_i - X_j}$$

• Example: For the permutation  $\sigma = (1, 2, 3)$ , we have

$$(-1)^{\sigma} = \frac{X_{\sigma(1)} - X_{\sigma(2)}}{X_1 - X_2} \cdot \frac{X_{\sigma(1)} - X_{\sigma(3)}}{X_1 - X_3} \cdot \frac{X_{\sigma(2)} - X_{\sigma(3)}}{X_2 - X_3}$$

$$= \frac{X_2 - X_3}{X_1 - X_2} \cdot \frac{X_2 - X_1}{X_1 - X_3} \cdot \frac{X_3 - X_1}{X_2 - X_3}$$

$$= \frac{-(X_1 - X_2)}{X_1 - X_2} \cdot \frac{-(X_1 - X_3)}{X_1 - X_3} \cdot \frac{X_2 - X_3}{X_2 - X_3}$$

$$= -1 \cdot -1 \cdot 1$$

$$= +1$$

which squares with the fact that  $\sigma = \tau_{1,2} \circ \tau_{2,3}$ .

- Claims to verify with the above formula:
  - 1.  $sign(\sigma) \in \{\pm 1\}.$
  - 2.  $sign(\tau_{i,i}) = -1$ .
  - 3.  $\operatorname{sign}(\sigma_1 \sigma_2) = \operatorname{sign}(\sigma_1) \operatorname{sign}(\sigma_2)$ .

### 1.4 Chapter 1: Multilinear Algebra

From Guillemin and Haine (2018).

3/31: • Guillemin and Haine (2018) defines real vector spaces, the operations on them, their basic properties,
 and the zero vector.

- Linearly independent (vectors  $v_1, \ldots, v_k$ ): A finite set of vectors  $v_1, \ldots, v_k \in V$  such that the map from  $\mathbb{R}^k$  to V defined by  $(c_1, \ldots, c_k) \mapsto c_1 v_1 + \cdots + c_k v_k$  is injective.
- Spanning (vectors  $v_1, \ldots, v_k$ ): We require that the above map is surjective.
- Guillemin and Haine (2018) defines basis, finite-dimensional vector space, dimension, subspace, linear map, and kernel.
- Image (of  $A: V \to W$ ): The range space of A, a subspace of W. Also known as im (A).
- Guillemin and Haine (2018) defines the matrix of a linear map.
- Inner product (on V): A map  $B: V \times V \to \mathbb{R}$  with the following three properties.
  - Bilinearity: For vectors  $v, v_1, v_2, w \in V$  and  $\lambda \in \mathbb{R}$ , we have

$$B(v_1 + v_2, w) = B(v_1, w) + B(v_2, w)$$

and

$$B(\lambda v, w) = \lambda B(v, w)$$

- Symmetry: For vectors  $v, w \in V$ , we have B(v, w) = B(w, v).
- Positivity: For every vector  $v \in V$ , we have  $B(v,v) \geq 0$ . Moreover, if  $v \neq 0$ , then B(v,v) > 0.
- **W-coset**: A set of the form  $\{v + w \mid w \in W\}$ , where W is a subspace V and  $v \in V$ . Denoted by v + W.
  - If  $v_1 v_2 \in W$ , then  $v_1 + W = v_2 + W$ .
  - It follows that the distinct W-cosets decompose V into a disjoint collection of subsets of V.
- Quotient space (of V by W): The set of distinct W-cosets in V, along with the following definitions of vector addition and scalar multiplication.

$$(v_1 + W) + (v_2 + W) = (v_1 + v_2) + W$$
  $\lambda(v + W) = (\lambda v) + W$ 

Denoted by V/W.

• Quotient map: The linear map  $\pi: V \to V/W$  defined by

$$\pi(v) = v + W$$

- $-\pi$  is surjective.
- Note that  $\ker(\pi) = W$  since for all  $w \in W$ ,  $\pi(w) = w + W = 0 + W$ , which is the zero vector in V/W.
- If V, W are finite dimensional, then

$$\dim(V/W) = \dim(V) - \dim(W)$$

- Proposition 1.2.9: Let  $A: V \to U$  be a linear map. If  $W \subset \ker(A)$ , then there exists a unique linear map  $A^{\sharp}: V/W \to U$  with the property that  $A = A^{\sharp} \circ \pi$ , where  $\pi: V \to V/W$  is the quotient map.
  - This proposition rephrases in terms of quotient spaces the fact that if  $w \in W$ , then A(v+w) = Av.

• Dual space (of V): The set of all linear functions  $\ell: V \to \mathbb{R}$ , along with the following definitions of vector addition and scalar multiplication.

$$(\ell_1 + \ell_2)(v) = \ell_1(v) + \ell_2(v) \qquad (\lambda \ell)(v) = \lambda \cdot \ell(v)$$

Denoted by  $V^*$ .

• **Dual basis** (of  $e_1, \ldots, e_n$  a basis of V): The basis of  $V^*$  consisting of the n functions that take every  $v = c_1 e_1 + \cdots + c_n e_n$  to one of the  $c_i$ . Denoted by  $e_1^*, \ldots, e_n^*$ . Given by

$$e_i^*(v) = c_i$$

for all  $v \in V$ .

• Claim 1.2.12: If V is an n-dimensional vector space with basis  $e_1, \ldots, e_n$ , then  $e_1^*, \ldots, e_n^*$  is a basis of  $V^*$ .

*Proof.* We will first prove that  $e_1^*, \ldots, e_n^*$  spans  $V^*$ . Let  $\ell \in V^*$  be arbitrary. Set  $\lambda_i = \ell(e_i)$  for all  $i \in [n]$ . Define  $\ell' = \sum_{i=1}^n \lambda_i e_i^*$ . Then

$$\ell'(e_j) = \sum_{i=1}^n \lambda_i e_i^*(e_j) = \lambda_j \cdot 1 = \ell(e_j)$$

for all  $j \in [n]$ . Therefore, since  $\ell, \ell'$  take identical values on the basis of  $V, \ell = \ell'$ , as desired. We now prove that  $e_1^*, \ldots, e_n^*$  is linearly independent. Let  $\sum_{i=1}^n \lambda_i e_i^* = 0$ . Then for all  $j \in [n]$ ,

$$\lambda_j = \left(\sum_{i=1}^n \lambda_i e_i^*\right)(e_j) = 0$$

as desired.  $\Box$ 

- Transpose (of A): The map from  $W^*$  to  $V^*$  defined by  $\ell \mapsto \ell \circ A$  for all  $\ell \in W^*$ . Denoted by  $A^*$ .
- Claim 1.2.15: If  $e_1, \ldots, e_n$  is a basis of  $V, f_1, \ldots, f_m$  is a basis of  $W, e_1^*, \ldots, e_n^*$  and  $f_1^*, \ldots, f_m^*$  are the corresponding dual bases, and  $[a_{i,j}]$  is the  $m \times n$  matrix of A with respect to  $\{e_j\}, \{f_i\}$ , then the linear map  $A^*$  is defined in terms of  $\{f_i^*\}, \{e_j^*\}$  by the transpose matrix  $(a_{j,i})$ .

*Proof.* Let  $[c_{j,i}]$  be the  $n \times m$  matrix of  $A^*$  with respect to  $\{f_i^*\}, \{e_j^*\}$ . We seek to prove that  $a_{i,j} = c_{j,i}$   $(1 \le i \le m, 1 \le j \le n)$ .

By the definition of  $[a_{i,j}]$  and  $[c_{j,i}]$ , we have that

$$A^* f_i^* = \sum_{k=1}^n c_{k,i} e_k^*$$
 
$$Ae_j = \sum_{k=1}^m a_{k,j} f_k$$

It follows that

$$[A^*f_i^*](e_j) = \left[\sum_{k=1}^n c_{k,i} e_k^*\right](e_j) = c_{j,i}$$

and

$$[A^*f_i^*](e_j) = f_i^*(Ae_j) = f_i^*\left(\sum_{k=1}^m a_{k,j}f_k\right) = a_{i,j}$$

so transitivity implies the desired result.

4/4: •  $V^k$ : The set of all k-tuples  $(v_1, \ldots, v_k)$  where  $v_1, \ldots, v_k \in V$  a vector space.

- Note that

$$V^k = \underbrace{V \times \dots \times V}_{k \text{ times}}$$

where "x" denotes the Cartesian product.

- **Linear** (function in its  $i^{\text{th}}$  variable): A function  $T: V^k \to \mathbb{R}$  such that the map from V to  $\mathbb{R}$  defined by  $v \mapsto T(v_1, \ldots, v_{i-1}, v, v_{i+1}, \ldots, v_k)$  is linear, where all  $v_i$  save  $v_i$  are fixed.
- **k-linear** (function T): A function  $T: V^k \to \mathbb{R}$  that is linear in its  $i^{\text{th}}$  variable for i = 1, ..., k. Also known as **k-tensor**.
- $\mathcal{L}^{k}(V)$ : The set of all k-tensors in V.
  - Since the sum  $T_1 + T_2$  of two k-linear functions  $T_1, T_2 : V^k \to \mathbb{R}$  is just another k-linear function, and  $\lambda T_1$  is k-linear for all  $\lambda \in \mathbb{R}$ , we have that  $\mathcal{L}^k(V)$  is a vector space.
- Convention: 0-tensors are just the real numbers. Mathematically, we define

$$\mathcal{L}^0(V) = \mathbb{R}$$

- Note that  $\mathcal{L}^1(V) = V^*$ .
- Defines multi-indices of n of length k.
- Lemma 1.3.5: If  $n, k \in \mathbb{N}$ , then there are exactly  $n^k$  multi-indices of n of length k.
- $T_I$ : The real number  $T(e_{i_1}, \ldots, e_{i_k})$ , where  $T \in \mathcal{L}^k(V)$ ,  $e_1, \ldots, e_n$  is a basis of V, and I is a multi-index of n of length k.
- Proposition 1.3.7: The real numbers  $T_I$  determine T, i.e., if T, T' are k-tensors and  $T_I = T'_I$  for all I, then T = T'.

*Proof.* We induct on n. For the base case n = 1,  $T \in (\mathbb{R}^k)^*$  and we have already proven this result. Now suppose inductively that the assertion is true for n - 1. For each  $e_i$ , let  $T_i$  be the (k - 1)-tensor defined by

$$(v_1, \ldots, v_{n-1}) \mapsto T(v_1, \ldots, v_{n-1}, e_i)$$

Then for an arbitrary  $v = c_1 e_1 + \cdots + c_n e_n$ ,

$$T(v_1, \dots, v_{n-1}, v) = \sum_{i=1}^{n} c_i T_i(v_1, \dots, v_{n-1})$$

so the  $T_i$ 's determine T. Applying the inductive hypothesis completes the proof.

• **Tensor product**: The function  $\otimes : \mathcal{L}^k(V) \times \mathcal{L}^{\ell}(V) \to \mathcal{L}^{k+\ell}(V)$  defined by

$$(T_1 \otimes T_2)(v_1, \dots, v_{k+\ell}) = T_1(v_1, \dots, v_k)T_2(v_{k+1}, \dots, v_{k+\ell})$$

for all  $T_1 \in \mathcal{L}^k(V)$  and  $T_2 \in \mathcal{L}^{\ell}(V)$ .

• Note that by the definition of 0-tensors as real numbers, if  $a \in \mathbb{R}$  and  $T \in \mathcal{L}^k(V)$ , then

$$a \otimes T = T \otimes a = aT$$

- Proposition 1.3.9: Associativity, distributivity of scalar multiplication, and left and right distributive laws for the tensor product.
- Decomposable (k-tensor): A k-tensor T for which there exist  $\ell_1, \ldots, \ell_k \in V^*$  such that

$$T = \ell_1 \otimes \cdots \otimes \ell_k$$

- Defines  $e_I^*$ .
- Theorem 1.3.13: V a vector space with basis  $e_1, \ldots, e_n$  and  $0 \le k \le n$  implies the k-tensors  $e_I^*$  form a basis of  $\mathcal{L}^k(V)$ .

*Proof.* Spanning: Let  $T \in \mathcal{L}^k(V)$  be arbitrary. Define

$$T' = \sum_{I} T_{I} e_{I}^{*}$$

Since

$$T'_J = T'(e_{j_1}, \dots, e_{j_k}) = \sum_I T_I e_I^*(e_{j_1}, \dots, e_{j_k}) = T_J e_J^*(e_{j_1}, \dots, e_{j_k}) = T_J$$

for all J, Proposition 1.3.7 asserts that T = T'. Therefore, since every  $T_I \in \mathbb{R}$ ,  $T = T' \in \text{span}(e_I^*)$ .

Linear independence: Suppose

$$T = \sum_{I} c_I e_I^* = 0$$

for some set of constants  $c_I \in \mathbb{R}$ . Then

$$0 = T(e_{j_1}, \dots, e_{j_k}) = \sum_{I} c_I e_I^*(e_{j_1}, \dots, e_{j_k}) = c_J$$

for all J, as desired.

• Corollary 1.3.15: If dim V = n, then dim $(\mathcal{L}^k(V)) = n^k$ .

*Proof.* Follows immediately from Lemma 1.3.5.

• **Pullback** (of T by the map A): The k-tensor  $A^*T: V^k \to \mathbb{R}$  defined by

$$(A^*T)(v_1,\ldots,v_k) = T(Av_1,\ldots,Av_k)$$

where V, W are finite-dimensional vector spaces,  $A: V \to W$  is linear, and  $T \in \mathcal{L}^k(W)$ .

- Proposition 1.3.18: The map  $A^*: \mathcal{L}^k(W) \to \mathcal{L}^k(V)$  defined by  $T \mapsto A^*T$  is linear.
- Identities:

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- If  $T_1 \in \mathcal{L}^k(W)$  and  $T_2 \in \mathcal{L}^m(W)$ , then

$$A^*(T_1 \otimes T_2) = A^*(T_1) \otimes A^*(T_2)$$

- If U is a vector space,  $B: U \to V$  is linear, and  $T \in \mathcal{L}^k(W)$ , then  $(AB)^*T = B^*(A^*T)$ . Hence,

$$(AB)^* = B^*A^*$$

•  $\Sigma_k$ : The set containing the natural numbers 1 through k. Given by

$$\Sigma_k = \{1, 2, \dots, k\}$$

- Permutation of order k: A bijection on  $\Sigma_k$ . Denoted by  $\sigma$ .
- **Product** (of  $\sigma_1, \sigma_2$ ): The composition  $\sigma_1 \circ \sigma_2$ , i.e., the map

$$i \mapsto \sigma_1(\sigma_2(i))$$

Denoted by  $\sigma_1 \sigma_2$ .

- Inverse (of  $\sigma$ ): The permutation of order k which is the inverse bijection of  $\sigma$ . Denoted by  $\sigma^{-1}$ .
- Permutation group (of  $\Sigma_k$ ): The set of all permutations of order k. Also known as symmetric group on k letters. Denoted by  $S_k$ .
- Lemma 1.4.2: The group  $S_k$  has k! elements.
- Transposition: A permutation of order k defined by

$$\ell \mapsto \begin{cases} j & \ell = i \\ i & \ell = j \\ \ell & \ell \neq i, j \end{cases}$$

for all  $\ell \in \Sigma_k$ , where  $i, j \in \Sigma_k$ . Denoted by  $\tau_{i,j}$ .

- Elementary transposition: A transposition of the form  $\pi_{i,i+1}$ .
- Theorem 1.4.4: Every  $\sigma \in S_k$  can be written as a product of (a finite number of) transpositions.

*Proof.* We induct on k.

For the base case k = 2, the identity permutation of  $S_2$  is the "product" of zero transpositions, and the only other permutation is a transposition (the "product" of one transposition, namely itself).

Now suppose inductively that we have proven the claim for k-1. Let  $\sigma \in S_k$  be arbitrary. Suppose  $\sigma(k) = i$ . Then  $\tau_{i,k}\sigma(k) = k$ . Since  $(\tau_{i,k}\sigma)|_{\Sigma_{k-1}} \in S_{k-1}$ , we have by the inductive hypothesis that  $(\tau_{i,k}\sigma)|_{\Sigma_{k-1}} = \tau_1 \cdots \tau_m$  for some set of permutations  $\tau_1, \ldots, \tau_m \in S_{k-1}$ . For each  $\tau_j$   $(1 \leq j \leq m)$ , define  $\tau'_i \in S_k$ 

$$\tau'_{j}(\ell) = \begin{cases} \tau_{j}(\ell) & \ell < k \\ \ell & \ell = k \end{cases}$$

It follows that

$$\tau_{i,k}\sigma = \tau_1' \cdots \tau_m'$$
$$\sigma = \tau_{i,k}\tau_1' \cdots \tau_m'$$

as desired.

• Theorem 1.4.5: Every transposition can be written as a product of elementary transpositions.

*Proof.* Let  $\tau_{i,j} \in S_k$ , and let i < j WLOG. Then we have that

$$\tau_{i,j} = \prod_{\ell=i}^{i-1} \tau_{\ell,\ell+1}$$

as desired.

- Corollary 1.4.6: Every permutation can be written as a product of elementary transpositions.
- Sign (of  $\sigma$ ): The number  $\pm 1$  assigned to  $\sigma$  by the expression

$$\prod_{i < j} \frac{x_{\sigma(i)} - x_{\sigma(j)}}{x_i - x_j}$$

where  $x_1, \ldots, x_k$  are coordinate functions on  $\mathbb{R}^k$ . Denoted by  $(-1)^{\sigma}$ .

• Claim 1.4.9: The sign defines a group homomorphism  $S_k \to \{\pm 1\}$ . That is, for  $\sigma_1, \sigma_2 \in S_k$ , we have

$$(-1)^{\sigma_1 \sigma_2} = (-1)^{\sigma_1} (-1)^{\sigma_2}$$

*Proof.* For all i < j, define p, q such that p is the lesser of  $\sigma_2(i), \sigma_2(j)$  and q is the greater of  $\sigma_2(i), \sigma_2(j)$ . Formally,

$$p = \begin{cases} \sigma_2(i) & \sigma_2(i) < \sigma_2(j) \\ \sigma_2(j) & \sigma_2(j) < \sigma_2(i) \end{cases} \qquad q = \begin{cases} \sigma_2(j) & \sigma_2(i) < \sigma_2(j) \\ \sigma_2(i) & \sigma_2(j) < \sigma_2(i) \end{cases}$$

It follows that if  $\sigma_2(i) < \sigma_2(j)$ , then

$$\frac{x_{\sigma_{1}\sigma_{2}(i)}-x_{\sigma_{1}\sigma_{2}(j)}}{x_{\sigma_{2}(i)}-x_{\sigma_{2}(j)}} = \frac{x_{\sigma_{1}(p)}-x_{\sigma_{1}(q)}}{x_{p}-x_{q}}$$

and if  $\sigma_2(j) < \sigma_2(i)$ , then

$$\frac{x_{\sigma_1\sigma_2(i)} - x_{\sigma_1\sigma_2(j)}}{x_{\sigma_2(i)} - x_{\sigma_2(j)}} = \frac{x_{\sigma_1(q)} - x_{\sigma_1(p)}}{x_q - x_p} = \frac{x_{\sigma_1(p)} - x_{\sigma_1(q)}}{x_p - x_q}$$

Therefore,

$$\begin{split} (-1)^{\sigma_1\sigma_2} &= \prod_{i < j} \frac{x_{\sigma_1\sigma_2(i)} - x_{\sigma_1\sigma_2(j)}}{x_i - x_j} \\ &= \prod_{i < j} \frac{x_{\sigma_1\sigma_2(i)} - x_{\sigma_1\sigma_2(j)}}{x_{\sigma_2(i)} - x_{\sigma_2(j)}} \cdot \frac{x_{\sigma_2(i)} - x_{\sigma_2(j)}}{x_i - x_j} \\ &= \prod_{i < j} \frac{x_{\sigma_1(p)} - x_{\sigma_1(q)}}{x_p - x_q} \cdot \prod_{i < j} \frac{x_{\sigma_2(i)} - x_{\sigma_2(j)}}{x_i - x_j} \\ &= (-1)^{\sigma_1} (-1)^{\sigma_2} \end{split}$$

as desired.  $\Box$ 

• Proposition 1.4.11: If  $\sigma$  is the product of an odd number of transpositions, then  $(-1)^{\sigma} = -1$ , and if  $\sigma$  is the product of an even number of transpositions, then  $(-1)^{\sigma} = +1$ .

*Proof.* Follows from the fact that  $(-1)^{\sigma} = -1$  (see Exercise 1.4.ii).

•  $T^{\sigma}$ : The k-tensor defined by

$$T^{\sigma}(v_1,\ldots,v_k) = T(v_{\sigma^{-1}(1)},\ldots,v_{\sigma^{-1}(k)})$$

where  $T \in \mathcal{L}^k(V)$ , V is an n-dimensional vector space, and  $\sigma \in S_k$ .

• Proposition 1.4.14:

1. If 
$$T = \ell_1 \otimes \cdots \otimes \ell_k$$
 ( $\ell_i \in V^*$ ), then  $T^{\sigma} = \ell_{\sigma(1)} \otimes \cdots \otimes \ell_{\sigma(k)}$ .

*Proof.* If  $v_1, \ldots, v_k \in V$ , then

$$T^{\sigma}(v_{1},...,v_{k}) = T(v_{\sigma^{-1}(1)},...,v_{\sigma^{-1}(k)})$$

$$= [\ell_{1} \otimes \cdots \otimes \ell_{k}](v_{\sigma^{-1}(1)},...,v_{\sigma^{-1}(k)})$$

$$= \ell_{1}(v_{\sigma^{-1}(1)}) \cdots \ell_{k}(v_{\sigma^{-1}(k)})$$

$$= \ell_{\sigma(1)}(v_{1}) \cdots \ell_{\sigma(k)}(v_{2})$$

$$= [\ell_{\sigma(1)} \otimes \cdots \otimes \ell_{\sigma(k)}](v_{1},...,v_{k})$$

as desired. Note that we can justify the fourth equality by nothing that if  $\sigma^{-1}(i) = q$ , then the  $i^{\text{th}}$  term in the product is  $\ell_{\sigma(q)}(v_q)$ , so since  $\sigma$  is a bijection, the product can be arranged to the form on the right-hand side of equality four.

2. The assignment  $T \mapsto T^{\sigma}$  is a linear map from  $\mathcal{L}^k(V) \to \mathcal{L}^k(V)$ .

*Proof.* See Exercise 1.4.iii. 
$$\Box$$

3. If  $\sigma_1, \sigma_2 \in S_k$ , we have  $T^{\sigma_1 \sigma_2} = (T^{\sigma_1})^{\sigma_2}$ .

*Proof.* Let  $T = \ell_1 \otimes \cdots \otimes \ell_k^{[1]}$ . Then

$$T^{\sigma_1} = \ell_{\sigma_1(1)} \otimes \cdots \otimes \ell_{\sigma_1(k)} = \ell'_1 \otimes \cdots \otimes \ell'_k$$

and thus

$$(T^{\sigma_1})^{\sigma_2} = \ell'_{\sigma_2(1)} \otimes \cdots \otimes \ell'_{\sigma_2(k)}$$

Let  $\sigma_2(i) = j$ . Then since  $\ell_p' = \ell_{\sigma_1(p)}$  by definition, we have that  $\ell_{\sigma_2(j)}' = \ell_{\sigma_1(\sigma_2(j))}$ . Therefore,

$$(T^{\sigma_1})^{\sigma_2} = \ell'_{\sigma_2(1)} \otimes \cdots \otimes \ell'_{\sigma_2(k)}$$

$$= \ell_{\sigma_1(\sigma_2(1))} \otimes \cdots \otimes \ell_{\sigma_1(\sigma_2(k))}$$

$$= \ell_{\sigma_1\sigma_2(1)} \otimes \cdots \otimes \ell_{\sigma_1\sigma_2(k)}$$

$$= T^{\sigma_1\sigma_2}$$

as desired.  $\Box$ 

- Alternating (k-tensor): A k-tensor  $T \in \mathcal{L}^k(V)$  such that  $T^{\sigma} = (-1)^{\sigma}T$  for all  $\sigma \in S_k$ .
- $\mathcal{A}^k(V)$ : The set of all alternating k-tensors in  $\mathcal{L}^k(V)$ .
  - Proposition 1.4.14(2) implies that  $(T_1 + T_2)^{\sigma} = T_1^{\sigma} + T_2^{\sigma}$  and  $(\lambda T)^{\sigma} = \lambda T^{\sigma}$ ; it follows that  $\mathcal{A}^k(V)$  is a vector space.
- Alternation operation: The function from  $\mathcal{L}^k(V) \to \mathcal{L}^k(V)$  defined by

$$T \mapsto \sum_{\tau \in S_k} (-1)^{\tau} T^{\tau}$$

Denoted by Alt.

- Proposition 1.4.17: For  $T \in \mathcal{L}^k(V)$  and  $\sigma \in S_k$ , we have that
  - 1. Alt $(T)^{\sigma} = (-1)^{\sigma}$  Alt T.

*Proof.* We have that

$$\operatorname{Alt}(T)^{\sigma} = \left(\sum_{\tau \in S_k} (-1)^{\tau} T^{\tau}\right)^{\sigma}$$

$$= \sum_{\tau \in S_k} (-1)^{\tau} (T^{\tau})^{\sigma} \qquad \text{Proposition 1.4.14(2)}$$

$$= \sum_{\tau \in S_k} (-1)^{\tau} T^{\tau\sigma} \qquad \text{Proposition 1.4.14(3)}$$

$$= (-1)^{\sigma} \sum_{\tau \in S_k} (-1)^{\tau\sigma} T^{\tau\sigma}$$

$$= (-1)^{\sigma} \sum_{\tau \sigma \in S_k} (-1)^{\tau\sigma} T^{\tau\sigma}$$

$$= (-1)^{\sigma} \operatorname{Alt} T$$

as desired.

 $<sup>^{1}</sup>$ What gives us the right to assume T is decomposable?

2. If  $T \in \mathcal{A}^k(V)$ , then Alt T = k!T.

*Proof.* Since  $T \in \mathcal{A}^k(V)$ , we know that  $T^{\sigma} = (-1)^{\sigma}T$ . Therefore,

Alt 
$$T = \sum_{\tau \in S_k} (-1)^{\tau} T^{\tau} = \sum_{\tau \in S_k} (-1)^{\tau} (-1)^{\tau} T = \sum_{\tau \in S_k} T = k! T$$

where the last equality holds because the cardinality of  $S_k$  is k!.

3.  $Alt(T^{\sigma}) = Alt(T)^{\sigma}$ .

*Proof.* We have that

$$\operatorname{Alt}(T^{\sigma}) = \sum_{\tau \in S_k} (-1)^{\tau} T^{\tau \sigma} = (-1)^{\sigma} \sum_{\tau \in S_k} (-1)^{\tau \sigma} T^{\tau \sigma} = (-1)^{\sigma} \operatorname{Alt}(T) = \operatorname{Alt}(T)^{\sigma}$$

as desired.  $\Box$ 

4. The alternation operation is linear.

*Proof.* Follows by Proposition 1.4.14.  $\Box$ 

- Repeating (multi-index I): A multi-index I of length k such that  $i_r = i_s$  for some  $r \neq s$ .
- Strictly increasing (multi-index I): A multi-index I of length k such that  $i_1 < i_2 < \cdots < i_r$ .
- $I^{\sigma}$ : The multi-index of length k defined by

$$I^{\sigma} = (i_{\sigma(1)}, \dots, i_{\sigma(k)})$$

- If I is non-repeating, there is a unique  $\sigma \in S_k$  such that  $I^{\sigma}$  is strictly increasing.
- $\psi_I$ : The following k-tensor. Given by

$$\psi_I = \text{Alt}(e_I^*)$$

- Proposition 1.4.20:
  - 1.  $\psi_{I^{\sigma}} = (-1)^{\sigma} \psi_{I}$ .

Proof. We have that

$$\psi_{I^{\sigma}} = \operatorname{Alt}(e_{I^{\sigma}}^*) = \operatorname{Alt}[(e_I^*)^{\sigma}] = \operatorname{Alt}(e_I^*)^{\sigma} = (-1)^{\sigma} \operatorname{Alt}(e_I^*) = (-1)^{\sigma} \psi_I$$

as desired.  $\Box$ 

2. If I is repeating, then  $\psi_I = 0$ .

*Proof.* Suppose  $I=(i_1,\ldots,i_k)$  is such that  $i_r=i_s$  for some distinct  $r,s\in\Sigma_k$ . Then  $e_I^*=e_{I^{\tau_{i_r,i_s}}}^*$ , so

$$\psi_I = \psi_{I^{\tau_{i_r,i_s}}} = (-1)^{\tau_{i_r,i_s}} \psi_I = -\psi_I$$

Therefore, we must have  $\psi_I = 0$ , as desired.

3. If I and J are strictly increasing, then

$$\psi_I(e_{j_1}, \dots, e_{j_k}) = \begin{cases} 1 & I = J \\ 0 & I \neq J \end{cases}$$

*Proof.* We have by definition that

$$\psi_I(e_{j_1},\ldots,e_{j_k}) = \sum_{\tau} (-1)^{\tau} e_{I^{\tau}}^*(e_{j_1},\ldots,e_{j_k})$$

This combined with the facts that

$$e_{I^{\tau}}^*(e_{j_1},\dots,e_{j_k}) = \begin{cases} 1 & I^{\tau} = J \\ 0 & I^{\tau} \neq J \end{cases}$$

 $I^{\tau}$  is strictly increasing iff  $I^{\tau} = I$ , and the above equation is nonzero iff  $I^{\tau} = I = J$  implies the desired result.

• Conclusion 1.4.22: If  $T \in \mathcal{A}^k(V)$ , then we can write T as a sum

$$T = \sum_{I} c_{I} \psi_{I}$$

with I's strictly increasing.

*Proof.* Let  $T \in \mathcal{A}^k(V)$  be arbitrary. By Theorem 1.3.13,

$$T = \sum_{I} a_{J} e_{J}^{*}$$

for some set of  $a_J \in \mathbb{R}$ . It follows since Alt(T) = k!T that

$$T = \frac{1}{k!} \sum a_J \operatorname{Alt}(e_J^*) = \sum b_J \psi_J$$

We can disregard all repeating terms in the sum since they are zero by Proposition 1.4.20(2); for every non-repeating term J, we can write  $J = I^{\sigma}$ , where I is strictly increasing and hence  $\psi_J = (-1)^{\sigma} \psi_I$ .  $\square$ 

• Claim 1.4.24: The  $c_I$ 's of Conclusion 1.4.22 are unique.

*Proof.* For J strictly increasing, we have

$$T_J = T(e_{j_1}, \dots, e_{j_k}) = \sum_I c_I \psi_I(e_{j_1}, \dots, e_{j_k}) = c_J$$

• Proposition 1.4.26: The alternating tensors  $\psi_I$  with I strictly increasing are a basis for  $\mathcal{A}^k(V)$ .

*Proof.* Spanning: See Conclusion 1.4.22.

Linear independence: See Claim 1.4.24.

• We have that

$$\dim \mathcal{A}^k(V) = \binom{n}{k} = \frac{n!}{(n-k)!k!}$$

- Hint in proving this claim: "Show that every strictly increasing multi-index of length k determines a k-element subset of  $\{1, \ldots, n\}$  and vice versa." (Guillemin & Haine, 2018, p. 16).
- Note also that if k > n, every multi-index has a repeat somewhere, meaning that dim  $\mathcal{A}^k(V) = \binom{n}{k} = 0$ .