SCE 594: Special Topics in Intelligent Automation & Robotics

Lecture 4: Vector Spaces II



- Recap: Last Lectures
- Dual spaces
- Tensor spaces
- Bases and components



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Recap: Structure hierarchy

- A recurrent theme in mathematics is the classification of spaces by means of structure-preserving maps between them.
- Space = set + some structure

	Set	Group	Vector space over a Field
Mathematical object	S	(<i>G</i> ,⊕)	(V, \bigoplus, \bigcirc) over $(K, +, \cdot)$
Structure-preserving map	$ Map \\ f\colon \mathbb{S} \to \mathbb{T}$	Group homomorphism $\rho: G \to H$	Linear map $A: V \stackrel{\sim}{\rightarrow} W$
Isomorphic spaces	$\begin{array}{c} Bijection \\ \mathbb{S} \cong_{set} \mathbb{T} \end{array}$	Group isomorphism $G \cong_{\operatorname{grp}} H$	Linear isomorphism $V \cong_{\text{vec}} W$



Recap: Vector space

- A vector space (V, \bigoplus, \odot) over a field $(K, +, \cdot)$ is the set V equipped with two operations:
 - \bigoplus : $V \times V \to V$ called vector addition
 - $\bigcirc: K \times V \to V$ called scalar multiplication

that should satisfy the rules:

- (V,⊕) is an Abelian group
- The map \odot is an action of K on (V, \oplus)

• An element of $v \in V$ is called a **vector**.



Recap: Vector space

• Example is $(\mathbb{R}^n, \oplus, \odot)$:

•
$$\begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix} \oplus \begin{pmatrix} y_1 \\ \vdots \\ y_n \end{pmatrix} := \begin{pmatrix} x_1 + y_1 \\ \vdots \\ x_n + y_n \end{pmatrix}$$

•
$$\lambda \odot \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix} := \begin{pmatrix} \lambda \cdot x_1 \\ \vdots \\ \lambda \cdot x_n \end{pmatrix}$$

• Identity element of (\mathbb{R}^n, \oplus) is the zero **vector** $\begin{pmatrix} 0 \\ \vdots \\ 0 \end{pmatrix}$

• Example is $(\mathbb{R}^{m \times n}, \oplus, \odot)$:

$$\begin{pmatrix} a_{11} & \cdots & a_{1n} \\ \vdots & \ddots & \vdots \\ a_{m1} & \cdots & a_{mn} \end{pmatrix} \oplus \begin{pmatrix} b_{11} & \cdots & b_{1n} \\ \vdots & \ddots & \vdots \\ b_{m1} & \cdots & b_{mn} \end{pmatrix} := \begin{pmatrix} a_{11} + b_{11} & \cdots & a_{1n} + b_{1n} \\ \vdots & \ddots & \vdots \\ a_{m1} + b_{m1} & \cdots & a_{mn} + b_{mn} \end{pmatrix}$$

• Identity element of $(\mathbb{R}^{m \times n}, \oplus)$ is the zero **matrix** $\begin{pmatrix} 0 & \cdots & 0 \\ \vdots & \ddots & \vdots \\ 0 & \cdots & 0 \end{pmatrix}$



Recap: Linear Maps

• A linear map $A: V \to W$ between the vector spaces (V, \bigoplus, \odot) and (W, \boxplus, \boxdot) over the same field K is defined such that:

- $A(v_1 \oplus v_2) = A(v_1) \boxplus A(v_2), \quad \forall v_1, v_2 \in V$
- $A(\lambda \odot v) = \lambda \odot A(v)$, $\forall v \in V, \lambda \in K$
- If we drop the special notation for operators, $A(\lambda v_1 + v_2) = \lambda A(v_1) + A(v_2)$

 The set of all linear maps from a vector space V to W is itself a vector-space, denoted by L(V; W).



Recap: Linear Maps

- Example: $V = \mathbb{R}^n$, $W = \mathbb{R}^m$
- You can prove that any linear map $A: \mathbb{R}^n \to \mathbb{R}^m$ must have the form:

$$A(v) = \left(\sum a_i^1 v^i, \cdots, \sum a_i^m v^i\right) ,$$

with $a_i^j \in \mathbb{R}$ for $i \in \{1, \dots, n\}, j \in \{1, \dots, m\}$.

• $L(\mathbb{R}^n; \mathbb{R}^m) = \mathbb{R}^{m \times n}$ in this case

 $A(v) = \begin{pmatrix} a_1^1 & \cdots & a_n^1 \\ \vdots & \ddots & \vdots \\ a_1^m & \cdots & a_n^m \end{pmatrix} \cdot \begin{pmatrix} v^1 \\ \vdots \\ v^n \end{pmatrix}$

m – rows, n - columns



Example: Map that is not linear

- Example: $V = \mathbb{R}^n$, $W = \mathbb{R}^m$
- Consider the map $\Psi: \mathbb{R}^n \to \mathbb{R}^m$ between vector spaces defined by:

$$\Psi(\mathbf{v}) = \left(\sum a_i^1 v^i + \mathbf{b^1}, \cdots, \sum a_i^m v^i + \mathbf{b^m}\right),$$

with $a_i^j, b^j \in \mathbb{R}$ for $i \in \{1, \dots, n\}, j \in \{1, \dots, m\}$.

In components

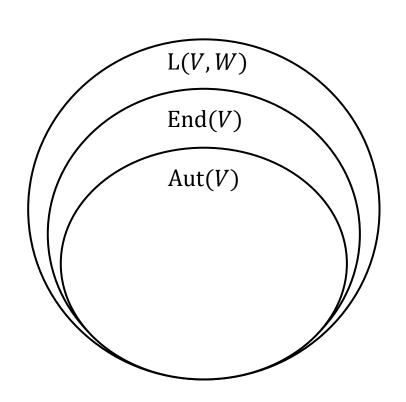
$$\Psi(v) = \begin{pmatrix} a_1^1 & \cdots & a_n^1 \\ \vdots & \ddots & \vdots \\ a_1^m & \cdots & a_n^m \end{pmatrix} \cdot \begin{pmatrix} v^1 \\ \vdots \\ v^n \end{pmatrix} + \begin{pmatrix} b^1 \\ \vdots \\ b^m \end{pmatrix} = A(v) + b$$

 A map Ψ of the form above is called an affine map and is not a linear map.



Recap: Subsets of L(V, W)

- $L(V; W) := \{A \mid A: V \stackrel{\sim}{\to} W\}$
- End $(V) := L(V, V) := \{A \in L(V, W) \mid W = V\}$
- $Aut(V) := \{A \in End(V) | A \text{ is an isomorphism} \}$
- Example $V = \mathbb{R}^n$:
 - $\operatorname{End}(\mathbb{R}^n) = \mathbb{R}^{n \times n}$
 - $\operatorname{Aut}(\mathbb{R}^n) = \operatorname{GL}(n, \mathbb{R})$





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Dual space

- The notion of a dual (vector) space V^* to a vector space V is extremely important in mechanics.
- However, it is usually overlooked !!
- While an element of V is called a vector, an element of V* is called a covector*.
 - Velocity-like variables are vectors.
 - Force-like variables are covectors.



Dual space

- Let *V* be a vector space over the field *K*
- The dual space to V is:

$$V^* \coloneqq L(V;K),$$

where $(K, +, \cdot)$ is considered a vector space over itself.

• An element of $\alpha \in V^*$ is a linear map from V to K called a **covector** or **one-form**.

$$\alpha: V \stackrel{\sim}{\to} K$$
$$v \mapsto \alpha(v)$$



Example: Dual space

- Example: $V = \mathbb{R}^n$, $K = \mathbb{R}$
- The dual space $(\mathbb{R}^n)^*$ consists of all linear maps:

$$\alpha: \mathbb{R}^n \stackrel{\sim}{\to} \mathbb{R}$$

$$v \mapsto \alpha(v)$$

that have the form

$$\alpha(v) = \sum_{i=1}^{n} \alpha_i v^i$$

In components

$$\alpha(v) = (\alpha_1 \quad \cdots \quad \alpha_n) \cdot \begin{pmatrix} v^1 \\ \vdots \\ v^n \end{pmatrix}$$



Dual of a linear map

 It is conventional to introduce a duality pairing between vectors and covectors denoted by:

$$\langle \cdot | \cdot \rangle : V^* \times V \xrightarrow{\sim} K$$

 $(\alpha, v) \mapsto \langle \alpha | v \rangle \coloneqq \alpha(v)$

• Let $A: U \to V$ be a linear map between the K-vector spaces U and V.

Then the linear map $A^*: V^* \rightarrow U^*$ defined (implicitly) by:

$$\langle A^*(\alpha)|u\rangle = \langle \alpha|A(u)\rangle, \qquad \forall u \in U, \alpha \in V^*$$

is called the dual of the map A.



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Tensors

• Let V be a vector space over K. A (p,q) tensor T on V is a multilinear map

$$T: \underbrace{V^* \times \cdots \times V^*}_{p \text{ times}} \times \underbrace{V \times \cdots \times V}_{q \text{ times}} \xrightarrow{\sim} K$$

i.e., T is a map that eats p-covectors and q-vectors.

- The term multi-linear means T is a linear map in each of its entries.
- The rank of a tensor T is the sum:

$$rank(T) = p + q$$



Tensors

Cases of interest:

 (0,1) tensor is a covector 	$\alpha:V \stackrel{\sim}{\to} K$
 (1,0) tensor is a vector 	$v:V^* \overset{\sim}{\to} K$
• (1,1) tensor	$A: V^* \times V \stackrel{\sim}{\to} K$
• (0,2) tensor	$B \colon V \times V \overset{\sim}{\to} K$
• (2,0) tensor	$C: V^* \times V^* \stackrel{\sim}{\to} K$

• The set of all (p,q) tensor T on V is denoted by $T_q^p V$ and we can make it into a vector-space structure.

$$T_q^p V \coloneqq \{T \mid T \text{ is a } (p,q) \text{ tensor on V}\}$$



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Components of tensors

- So far, the mathematical objects on vector spaces we introduced are abstractly defined.
 - Vectors, covectors, linear maps, dual of linear maps, (p,q) tensors.
- All these objects can be written in components once a basis has been chosen.
- However, the geometric nature of these objects should be respected independent of the basis we choose.



Basis for a vector space

- Let V be a vector space. A basis S for a vector space V is a collection of vectors in V that are:
 - Linearly independent from each other
 - Generate V



Basis for a vector space

- Let V be a vector space over \mathbb{R} . A basis S for a vector space V is a collection of vectors in V that are:
 - Linearly independent from each other
 - A set $S \subset V$ of vectors is linearly independent if, for every finite subset $\{e_1, \dots, e_k\} \subset S$, the equality

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\sum_{i=i}^k c^i e_k = c^1 e_1 + \dots + c^k e_k = 0, for some constants c^i \in \mathbb{R}, implies that these constants should be zero, i.e., c^i = 0 \ \forall i \in \{1, \dots, k\}.
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Generate V



Basis for a vector space

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 $\sum_{i=i}^k c^i e_k = c^1 e_1 + \dots + c^k e_k = 0, \text{ for some constants } c^i \in \mathbb{R},$ implies that these constants should be zero, i.e., $c^i = 0 \ \forall i \in \{1, \dots, k\}.$

Generate V

• A set $S \subset V$ of vectors generates a vector space V, if every vector $v \in V$ can be written as the linear combination

$$v = c^1 e_1 + \dots + c^k e_k$$
 , for some constants $c^i \in \mathbb{R}$

• We usually write that $V = \operatorname{span}_{\mathbb{R}}(S)$

