MAT224 Notes

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1.1 Vector spaces

Definition A real 1 vector space is a set V together with two vector operations vector addition and scalar multiplication such that

- 1. **AC** Additive Closure: $\forall \vec{x}, \vec{y} \in V, \vec{x} + \vec{y} \in V$
- 2. C Commutative: $\forall \vec{v}, \vec{y} \in V, \vec{x} + \vec{y} = \vec{y} + \vec{x}$
- 3. **AA** Additive Associative: $\forall \vec{x}, \vec{y}, \vec{z} \in V, (\vec{x} + \vec{y}) + \vec{z} = \vec{x} + (\vec{y} + \vec{z})$
- 4. **Z** Zero Vector: $\exists \vec{0} \in Vs.t. \forall \vec{x} \in V, \vec{x} + \vec{0} = \vec{x}$
- 5. **AI** Additive Inverse: $\forall \vec{x} \in V, \exists -\vec{x} \in V s.t.\vec{x} + (-\vec{x}) = \vec{0}$
- 6. **SC** Scalar Closure: $\forall \vec{x}, c \in \mathbb{R}, c\vec{x} \in V$
- 7. **DVA** Distributive Vector Additions: $\forall \vec{x}, \vec{y} \in V, c \in \mathbb{R}, c(\vec{x} + \vec{y}) = c\vec{x} + c\vec{y}$
- 8. **DSA** Distributive Scalar Additions: $\forall \vec{x} \in V, c, d \in \mathbb{R}, (c+d)\vec{x} = c\vec{x} + d\vec{x}$
- 9. **SMA** Scalar Multiplication Associative: $\forall \vec{x} \in V, c, d \in \mathbb{R}, (cd)\vec{x} = c(d\vec{x})$
- 10. **O** One: $\forall \vec{x} \in V, 1\vec{x} = \vec{x}$

Note For V to be a vector space, need to know or be given operations of vector additions multiplication and check all 10 properties hold.

¹A vector space is real if scalar which defines scalar multiplication is real.

1.2 Examples of vector spaces

Example 1 \mathbb{R}^n w.r.t.² usual component-wise addition and scalar multiplication.

Example 2 $\mathbb{M}_{m \times n}(\mathbb{R})$ set of all $m \times n$ matrices with real entry. w.r.t. usual entry-wise addition and scalar multiplication.

Example 3 $\mathbb{P}_n(\mathbb{R})$ set of polynomials with real coefficients, of degree less or equal to n, w.r.t. usual degree-wise polynomial addition and scalar multiplication.

Note If define $\mathbb{P}_n^{\star}(\mathbb{R})$ as set of all polynomials of degree <u>exactly equal</u> to n w.r.t. normal degree-wise multiplication and addition.

Then it is **NOT** a vector space.

Explanation: $(1+x^n), (1-x^n) \in \mathbb{P}_n^{\star}(\mathbb{R})$ but $(1+x^n) + (1-x^n) = 2 \notin \mathbb{P}_n^{\star}(\mathbb{R})$

Example 4 Something unusual, define V as

$$V = \{(x_1, x_2) | x_1, x_2 \in \mathbb{R}\}\$$

with vector addition

$$(x_1, x_2) + (y_1, y_2) = (x_1 + y_1 + 1, x_2 + y_2 + 1)$$

and scalar multiplication

$$c(x_1, x_2) = (cx_1 + c - 1, cx_2 + c - 1)$$

This is a vector space.

1.3 Some properties of vector spaces

Suppose V is a vector space, then it has the following properties.

²w.r.t. is the abbreviation of "with respect to".

Property 1 The zero vector is unique. *proof.*

Assume
$$\vec{0}, \vec{0^{\star}}$$
 are two zero vectors in V WTS: $\vec{0} = \vec{0^{\star}}$ Since $\vec{0}$ is the zero vector, by Z $\vec{0^{\star}} + \vec{0} = \vec{0^{\star}}$ Similarly, $\vec{0} + \vec{0^{\star}} = \vec{0}$ Also, $\vec{0} + \vec{0^{\star}} = \vec{0^{\star}} + \vec{0}$ by commutative vector addition. So, $\vec{0^{\star}} = \vec{0}$

Property 2 $\forall \vec{x} \in V$, the additive inverse $-\vec{x}$ is unique. *proof.*

Exercise. (By Cancellation Law)

Property 3 $\forall \vec{x} \in V, 0\vec{x} = \vec{0}.$ proof.

By property of number 0:
$$0\vec{x} = (0+0)\vec{x}$$

By DSA: $0\vec{x} = 0\vec{x} + 0\vec{x}$
By AI, $\exists (-0\vec{x})s.t.$
 $0\vec{x} + (-0\vec{x}) = 0\vec{x} + 0\vec{x} + (-0\vec{x})$
By AA
 $\implies 0\vec{x} = \vec{0}$

Property 4
$$\forall c \in \mathbb{R}, c\vec{0} = \vec{0}$$
 proof.
$$c\vec{0} = c(\vec{0} + \vec{0}) = c\vec{0} + c\vec{0}$$

2 Lecture Jan. 10 2018

2.1 Some properties of vector spaces-Cont'd

Property 5 For a vector space V, $\forall \vec{x} \in V$, $(-1)\vec{x} = (-\vec{x})$. (we could use this property to find the <u>additive inverse</u> with scalar multiplication with (-1))³. proof.

$$(-\vec{x})=(-\vec{x})+\vec{0}$$
 By property of zero vector
$$=(-\vec{x})+0\vec{x}$$
 By property3
$$=(-\vec{x})+(1+(-1))\vec{x}$$
 By property of zero as real number
$$=(-\vec{x})+1\vec{x}+(-1)\vec{x}$$

$$=\vec{0}+(-1)\vec{x}$$

$$=(-1)\vec{x}$$

Property 6 For a vector space V, let $\vec{x} \in V$ and $c \in \mathbb{R}$, then,

$$c\vec{x} = \vec{0} \implies c = 0 \lor \vec{x} = \vec{0}$$

proof.

if
$$c = 0 \implies True$$

else $c^{-1}c\vec{x} = c^{-1} = \vec{0}$
 $\implies (c^{-1}c)\vec{x} = \vec{0}$
 $\implies 1\vec{x} = \vec{0}$
 $\implies \vec{x} = \vec{0}$
 $\implies True$

2.2 Subspaces

Loosely A subspace is a space contained within a vector space.

 $^{^3}$ The scalar multiplication here is the one defined in vector space V.

Definition Let V be a vector space and $W \subseteq V$, W is a **subspace** of V if W is itself a vector space w.r.t. operations of vector addition and scalar multiplication from V.

Theorem Let V be a vector space, and $W \subseteq V$, W has the <u>same</u>⁴ operations of vector addition and scalar multiplication as in V. Then, W is a subspace of V <u>iff</u>:

- 1. W is non-empty. $W \neq \emptyset$.
- 2. W is closed under addition. $\forall \vec{x}, \vec{y} \in W, \ \vec{x} + \vec{y} \in W$.
- 3. W us closed under scalar multiplication. $\forall \vec{x} \in W, c \in \mathbb{R}, c\vec{x} \in W$.

Proof.

Forward:

If W is a subspace

$$\implies \vec{0} \in W$$

$$\implies W \neq \emptyset$$

Also, additive and scalar multiplication closures \implies (ii), (iii)

Backward:

Let $W \neq \emptyset \land (ii) \land (iii)$

WTS. 10 axioms in definition of vector space hold

- $(ii) \implies \text{Additive Closure}$
- $(iii) \implies \text{Scalar Multiplication Clousure}$

Because $W \subseteq V$, and V is a vector space, so properties hold $\forall \vec{w} \in W$.

Additive inverse: by property 5 and scalar multiplication closure,

$$\forall \vec{x} \in W, -\vec{x} = (-1)\vec{x} \in W.$$

Also, existence of additive identity: $(-\vec{x}) + \vec{x} = \vec{0} \in W$.

2.3 Examples of subspaces

Example 1 Let $V = \mathbb{M}_{n \times n}(\mathbb{R})$, V is a subspace.

⁴Other properties of vector spaces related to vector addition and scalar multiplication are immediately inherited from the parent vector space.

Example 2 Define W as

$$W = \{A \in \mathbb{M}_{n \times n}(\mathbb{R}) | A \text{ is } \underline{\text{not}} \text{ symmetric} \}$$

Explanation: Let
$$A_1=\begin{bmatrix}0&-2\\-1&0\end{bmatrix}$$
 and $A_2=\begin{bmatrix}0&2\\1&0\end{bmatrix}$ $A_1,A_2\in W$ but

$$A_1 + A_2 = \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix} \notin W.$$

Since there's no additive identity in set W, so W failed to be a vector space, therefore W is not a subspace.

Example 3 Let $V = \mathbb{P}_2(\mathbb{R})$, is W defined as following,

$$W = \{ p(x) \in V | p(1) = 0 \}$$

a subspace of V?

proof.

WTS: (i)

Let
$$z(x) = 0$$
 or $z(x) = x^2 - 1, \forall x \in \mathbb{R}$

$$\implies W \neq \emptyset$$

WTS: (ii)

Let $p_1, p_2 \in W$, which means $p_1(1) = p_2(1) = 0$

$$(p_1 + p_2)(1) = p_1(1) + p_2(1) = 0 + 0 = 0$$

$$\implies p_1 + p_2 \in W$$

 $\implies W$ is closed under addition.

WTS: (iii) Let $p \in W$ and $c \in \mathbb{R}$

$$\implies p(1) = 0$$

Since
$$(c * p)(x) = c * p(x)$$
, we have $(c * p)(1) = c * p(1) = c * 0 = 0$

$$\implies cp \in W$$
.

So W is a subspace of V.

4 Recall from MAT223

Let $A \in \mathbb{M}_{m \times n}(\mathbb{R})$, then Nul(A) is a subspace of \mathbb{R}^n and Col(A) is a subspace of \mathbb{R}^m .

3 Lecture 3Jan. 16 2018

3.1 Linear Combination

Definition Let V be a vector space, $\vec{v_1}, \ldots, \vec{v_n} \in V$, $a_1, \ldots, a_n \in \mathbb{R}$ the expression

$$c_1\vec{v_1} + \cdots + c_n\vec{v_n}$$

is called a linear combination of $\vec{v_1}, \ldots, \vec{v_n}$.

Theorem Let V be a vector space, W is a subspace of V, $\forall \vec{w_1}, \dots \vec{w_k} \in W, c_1, \dots, c_k \in \mathbb{R}$, we have

$$c_1\vec{w_1} + \dots + c_k\vec{w_k} \in W$$

Subspaces are <u>closed under linear combinations</u>, since subspaces are closed under scalar multiplication and vector addition.

Theorem Let V be a vector space, let $\vec{v_1}, \ldots, \vec{v_k} \in V$ then the set of all linear combination of $\vec{v_1}, \ldots, \vec{v_k}$

$$W = \{ \sum_{i=1}^{k} c_i \vec{v_i} | c_i \in \mathbb{R} \forall i \}$$

is a subspace of V. *proof.*

Consider
$$\vec{0} \in W$$

So, $W \neq \emptyset$

Let $c \in \mathbb{R}$, Let $\vec{x} \in W \land \vec{y} \in W$

By definition of span, we have,

$$\vec{x} = \sum_{i=1}^{k} a_i \vec{v_i}, \quad \vec{y} = \sum_{i=1}^{k} b_i \vec{v_i}$$

Consider, $\vec{x} + c\vec{y}$

$$\vec{x} + c\vec{y} = \sum_{i=1}^{k} a_i \vec{v_i} + c \sum_{i=1}^{k} b_i \vec{v_i} = \sum_{i=1}^{k} (a_i + cb_i) \vec{v_i} \in W$$

Definition Let V be a vector space, $\vec{v_1}, \ldots, \vec{v_k} \in V$, **span** of the set of vectors $\{\vec{v_i}\}_{i=1}^k$ is defined as the collection of all possible linear combinations of $\{\vec{v_i}\}_{i=1}^k$. By pervious theorem, span is a subspace.

3.2 Combination of subspaces

Definition Let W_1, W_2 be two sets, then the **union** of W_1, W_2 is defined as:

$$W_1 \cup W_2 = \{ \vec{w} \mid \vec{w} \in W_1 \lor \vec{w} \in W_2 \}$$

the **intersection** of W_1, W_2 is defined as:

$$W_1 \cap W_2 = \{ \vec{w} \mid \vec{w} \in W_1 \land \vec{w} \in W_2 \}$$

Now consider W_1, W_2 to be two subspaces of vector space V, then we have,

- 1. $W_1 \cup W_2$ is **not** a subspace.
- 2. $W_1 \cap W_2$ is a subspace.

proof.

Falsify the statement by providing counter-example:

$$W_{1} = \{(x_{1}, x_{2}) \mid x_{1} \in \mathbb{R}, x_{2} = 0\}$$

$$W_{2} = \{(x_{1}, x_{2}) \mid x_{2} \in \mathbb{R}, x_{1} = 0\}$$

$$\binom{0}{1} \in W_{1} \cup W_{2} \quad \binom{1}{0} \in W_{1} \cup W_{2}$$

$$\text{But}, \quad \binom{0}{1} + \binom{1}{0} = \binom{1}{1} \notin W_{1} \cup W_{2}$$

proof.

Because W_1 and W_2 are both subspaces, so $\vec{0} \in W_1 \cap W_2 \implies W_1 \cap W_2 \neq \emptyset$ Let $\vec{x}, \vec{y} \in W_1 \cap W_2, c \in \mathbb{R}$ Consider, $\vec{x} + c\vec{y}$ Sine W_1, W_2 are subspaces, $\vec{x} + c\vec{y} \in W_1 \wedge \vec{x} + c\vec{y} \in W_2$ $\implies \vec{x} + c\vec{y} \in W_1 \cap W_2$ So, $W_1 \cap W_2$ is a subspace.

Definition Let W_1, W_2 be subspaces of vector space V, define the **sum** of two subspaces as:

$$W_1 + W_2 = \{ \vec{x} + \vec{y} \mid \vec{x} \in W_1 \land \vec{y} \in W_2 \}$$

Note Let $\vec{x} = \vec{0} \in W_1$, $\forall \vec{y} \in W_2$, $\vec{y} \in W_1 + W_2$ so that, $W_2 \subseteq W_1 + W_2$. Similarly, let $\vec{y} = 0 \in W_2$, $\forall \vec{x} \in W_1$, $\vec{x} \in W_1 + W_2$. so that, $W_1 \subseteq W_1 + W_2$. So we have $\forall \vec{v} \in W_1 \cap W_2$, $\vec{v} \in W_1 + W_2$. So that,

$$W_1 \cap W_2 \subseteq W_1 + W_2$$

Note $W_1 + W_2$ is a subspace of V. proof.

Let
$$\vec{x_1}, \vec{x_2} \in W_1, \vec{y_1}, \vec{y_2} \in W_2$$

By properties of subspaces,
 $\forall c \in \mathbb{R}, \vec{x_1} + c\vec{x_1} \in W_1 \land \vec{y_2} + c\vec{y_2} \in W_2$
Consider, $\vec{x_1} + \vec{y_1} \in W_1 + W_2, \vec{x_2} + \vec{y_2} \in W_1 + W_2$
 $(\vec{x_1} + \vec{y_1}) + c(\vec{x_2} + \vec{y_2})$
 $= (\vec{x_1} + c\vec{x_2}) + (\vec{y_1} + c\vec{y_2}) \in W_1 + W_2$

Definition(Unique Representation) Let W_1, W_2 be subspaces of vector space V, say V is **direct sum** of W_1 and W_2 , written as $V = W_1 \bigoplus W_2$, if every $\vec{x} \in V$ can be written <u>uniquely</u> as $\vec{x} = \vec{w_1} + \vec{w_2}$ where $\vec{w_1} \in W_1$ and $\vec{w_2} \in W_2$.

Equivalently Let W_1 and W_2 be subspaces of V, $V = W_1 \bigoplus W_2 \iff V = W_1 + W_2 \wedge W_1 \cap W_2 = \{\vec{0}\}.$

4 Lecture 4 Jan. 17 2018

4.1 Cont'd

Cont'd Proof of Theorem proof.

(Forward direction) Suppose
$$V = W_1 \bigoplus W_2$$

WTS. $V = W_1 + W_2 \wedge W_1 \cap W_2 = \{\vec{0}\}$

Let $V = W_1 \bigoplus W_2$
 $\Rightarrow \forall \vec{x} \in V$, can be written uniquely as $\vec{x} = \vec{w_1} + \vec{w_2}, \ \vec{w_1} \in W_1, \ \vec{w_2} \in W_2$
 $\Rightarrow V = W_1 + W_2$ by definition of sum .

Let $\vec{x} \in W_1 \cap W_2$

Decomposition, let $\vec{z} \in W_1, \vec{0} \in W_2$
 $\vec{z} = \vec{z} + \vec{0}, \ \vec{z} \in W_1, \vec{0} \in W_2$
 $\vec{z} = \vec{0} + \vec{z}, \ \vec{0} \in W_1, \vec{z} \in W_2$

Since decomposition is unique, $\vec{z} = \vec{0}$

So, $W_1 \cap W_2 = \{\vec{0}\}$

(Backward direction) Suppose $V = W_1 + W_2 \wedge W_1 \cap W_2 = \{\vec{0}\}$

WTS. $V = W_1 \bigoplus W_2$

Assume $\vec{x} = \vec{w_1} + \vec{w_2}, \ \vec{w_1} \in W_1, \vec{w_2} \in W_2$
 $\vec{x} = \vec{w_1}' + \vec{w_2}', \ \vec{w_1}' \in W_1, \vec{w_2}' \in W_2$
 $\Rightarrow \vec{w_1} + \vec{w_2} = \vec{w_1}' + \vec{w_2}'$
 $\Rightarrow \vec{w_1} - \vec{w_1}' = \vec{w_2}' - \vec{w_2}$

Where, by definition of subspace, $\vec{w_1} - \vec{w_1}' \in W_1 \wedge \vec{w_2}' - \vec{w_2} \in W_2$

So, $\vec{w_1} - \vec{w_1}' = \vec{w_2}' - \vec{w_2} \in W_1 \cap W_2$

Since $W_1 \cap W_2 = \{\vec{0}\}$
 $\Rightarrow \vec{w_1} = \vec{w_1}' \wedge \vec{w_2} = \vec{w_2}'$

So the decomposition is unique.

4.2 Linear Independence

Theorem (Redundancy theorem) Let V be a vector space, $\{\vec{x_1}, \dots \vec{x_n}\}$, let $\vec{x} \in \{\vec{x_1}, \dots \vec{x_n}\}$, then

$$span\{\vec{x_1}, \dots \vec{x_n}, \vec{x}\} = span\{\vec{x_1}, \dots \vec{x_n}\}$$

we say \vec{x} is the **redundant** vector that contributes nothing to the span. proof.

let
$$\vec{x} \in span\{\vec{x}, \dots, \vec{x_n}\}$$

$$\vec{x} = \sum_{i=1}^{n} c_i \vec{x_i} \text{ for } c_i \in \mathbb{R} \ \forall i$$
So, $span\{\vec{x_1}, \dots, \vec{x_n}, \vec{x}\} = \{\sum_{i=1}^{n} a_i \vec{x_i} + z \vec{x} \mid a_i, z \in \mathbb{R} \forall i\}$

$$= \{\sum_{i=1}^{n} a_i \vec{x_i} + z \sum_{i=1}^{n} c_i \vec{x_i} \mid a_i, c_i \in \mathbb{R} \forall i\}$$

$$= \{\sum_{i=1}^{n} (a_i + z c_i) \vec{x_i} \mid a_i, c_i \in \mathbb{R} \forall i\}$$

$$\text{Let } d_i = a_i + z c_i \in \mathbb{R}$$

$$= \{\sum_{i=1}^{n} d_i \vec{x_i} \mid d_i \in \mathbb{R} \forall i\}$$

$$= span\{\vec{x_1}, \dots, \vec{x_n}\}$$

Definition Let V be a vector space, let $\{\vec{x_1}, \dots, \vec{x_n}\} \in V$, we say $\{v_i\}_{i=1}^n$ is **linearly independent** if the only set of scalars $\{c_1, \dots, c_n\}$ that satisfies,

$$\sum_{i=1}^{n} c_i \vec{x_i} = 0$$

is $\{0, ..., 0\}$.

Definition In contrast, we say a set of vector, with size n, is **linearly** dependent if

$$\exists \vec{c} \neq \vec{0} \in \mathbb{R}^n, \ s.t. \ \sum_{i=1}^n c_i \vec{v_i} = 0$$

Theorem Let V be a vector space, $\{\vec{v_i}\}_{i=1}^n \in V$ is linearly dependent if and only if,

$$\exists \vec{x} \in \{\vec{v_i}\}_{i=1}^n \ s.t. \ \vec{x_j} \in span\{\{\vec{v_i}\}_{i=1}^n \setminus \{\vec{x}\}\}$$

Theorem Let V be a vector space, $\{\vec{v_i}\}_{i=1}^n \in V$ is linearly independent if and only if,

$$\forall \vec{x} \in \{\vec{v_i}\}_{i=1}^n, \ \vec{x_i} \notin span\{\{\vec{v_i}\}_{i=1}^n \setminus \{\vec{x}\}\}\$$

5 Lecture Jan. 23 2018

5.1 Linear independence, recall definitions

Acknowledgement: special thanks to Frank Zhao.

Definition Let $\{\vec{x_1}, \dots \vec{x_k}\}$ is **linearly independent** if only scalars $c_1 \dots c_k$ s.t.

$$\sum_{i=1}^{k} c_1 \vec{x_k} = 0(\star)$$

are
$$c_1 = \cdots = c_k = 0$$

linearly dependent means at least one $c_i \neq 0$, (\star) still holds.

5.1.1 Alternative definitions of linear independency

Definition(Alternative.1) $\{\vec{x_1} \dots \vec{x_k}\}$ is linearly independent iff none of them can be written as a linear combination of the remaining k-1 vectors.⁵

Definition(Alternative.2) $\{\vec{x_1} \dots \vec{x_k}\}$ is **linearly dependent** iff at least one of them can be written as a linear combination of the remaining k-1 vectors. ⁶

5.2 Basis

Definition Let V be a vector space, a non-empty⁷ set S of vectors from V is a **basis** for V if

1.
$$V = span\{S\}$$

 $^{^5\}mathrm{See}$ theorem from the pervious lecture.

 $^{^6\}mathrm{See}$ theorem from the pervious lecture.

⁷Specially, for an empty set, we define $span\{\emptyset\} = \{\vec{0}\}$

2. S is linearly independent.

Theorem (characterization of basis) A non-empty subset $S = \{\vec{x_i}\}_{i=1}^n$ of vector space V is basis for V iff every $\vec{x} \in V$ can be written <u>uniquely</u> as linear combination for vectors in S.

proof.

Forwards

Suppose S is a basis for V

So every $\vec{x} \in V$ can be written as a linear combination of vectors in S

To prove the uniqueness, assume two expressions of $\vec{x} \in V$

$$\vec{x} = \begin{cases} c_1 \vec{x_1} + \dots + c_k \vec{x_k} \\ b_1 \vec{x_1} + \dots + d_k \vec{x_k} \end{cases}$$

Consider.

$$c_1\vec{x_1} + \dots + c_k\vec{x_k} - (b_1\vec{x_1} + \dots + d_k\vec{x_k}) = \vec{0}$$

$$\iff \sum_{i=1}^{k} (c_i - b_i) \vec{x_1} = \vec{0}$$

Since vectors in basis S are linear independent,

$$c_i = b_i \forall i \in \mathbb{Z} \cap [1, k]$$

So the representation is unique.

Backwards

Suppose every $\vec{x} \in V$ can be written uniquely as linear combination of vectors in S.

WTS: $V = span\{S\} \wedge S$ is linearly independent

By the assumption, spanning set is shown.

All we need to show is linear independence.

Consider,

$$\sum_{i=1}^{n} c_i \vec{x}_i = \vec{0}$$

Also, we know

$$\sum_{i=1}^{n} 0\vec{x_i} = \vec{0}$$

By the uniqueness of representation

We have identical expression
$$\sum_{i=1}^{n} c_i \vec{x}_i = \sum_{i=1}^{n} 0 \vec{x}_i$$

$$\therefore c_i = 0 \ \forall i \in \mathbb{Z} \cap [1, n]$$

Example

$$V = \{(x_1, x_2) \mid x_1, x_2 \in \mathbb{R}\}$$
$$(x_1, x_2) + (y_1, y_2) = (x_1 + y_1 + 1, x_2 + y_2 + 1)$$
$$c(x_1, x_2) = (cx_1 + c - 1, cx_2 + c - 1)$$

Show that $\{(1,0),(6,3)\}$ is a basis of V.

By theorem, $\{(1,0),(6,3)\}$ is basis if every $(a,b) \in V$ can be written uniquely as linear combination of $\{(1,0),(6,3)\}$.

 \exists unique scalars $c_1, c_2 \in \mathbb{R}$ s.t. $c_1(1,0) + c_2(6,3) = (a,b)$

proof.

By definition of scalar multiplication and vector addition in this space,

Consider
$$(a, b) = c_1(1, 0) + c_2(6, 3) = (2c_1 - 1, c_1 - 1) + (7c_2 - 1, 4c_2 - 1)$$

= $(2c_1 + 7c_2 - 1, c_1 + 4c_2 - 1)$

Consider the coefficients of variables

$$\begin{cases} 2c_1 + 7c_2 - 1 = a \\ c_1 + 4c_2 - 1 = b \end{cases}$$

WTS, the above system of linear equations has unique solution for all a, b

The system has a unique solution $\forall a, b \in \mathbb{R}$

Since the coefficient matrix has rank 2

$$rank(\begin{pmatrix} 2 & 7 \\ 1 & 4 \end{pmatrix}) = 2$$

Since obviously the columns are linearly independent.

5.3 Dimensions

Definition For a vector space V, the **dimension** of V is the minimum number of vectors required to span V.

Fundamental Theorem if V vector space is spanned by m vectors, then any set of more than m vectors from V must be <u>linearly dependent</u>.

Fundamental Theorem (Alternative) If V is vector space spanned by m vectors, then any <u>linearly independent</u> set in V must contain less or equal to m vectors.

5.3.1 Consequences of fundamental theorem

Theorem if $S = \{\vec{v}_i\}_{i=1}^k$ and $T = \{\vec{w}_i\}_{i=1}^l$ are two bases of vector space V then l = k. Bases have the same size.

proof.

Since S spans V and T is linearly independent

$$\therefore l \leq k$$

(flip) Since T spans V and S is linearly independent

Definition So we can define the **dimension** of V, as dim(V) as the number vectors in <u>any</u> basis for V. For special case $V = \{\vec{0}\}$, dim(V) = 0.

Example

- $dim(\mathbb{R}^n) = n$
- $dim(\mathbb{P}_n(\mathbb{R})) = n+1$
- $dim(\mathbb{M}_{m \times n}(\mathbb{R})) = m \times n$

5.3.2 Use dimension to prove facts about linearly (in)dependent sets and subspaces

Theorem If V is a vector space, dim(V) = n, $S = \{\vec{x_k}\}_{i=1}^k$ is subset of V, if k > n then S is <u>linearly dependent</u>.

Note $k \leq n \Rightarrow S$ is linear dependent.

Theorem If W is subspace of vector space V, then

- 1. $dim(W) \leq dim(V)$
- 2. $dim(W) = dim(V) \iff W = V$

proof.

(1) Suppose
$$dim(V) = n, dim(W) = k$$

WTS, $k \le n$

Any basis for W is a linearly independent set of k vectors from V.

Since V is spanned by n vectors, since dim(V) = n

By fundamental theorem, $k \leq n$

$$\iff dim(W) \le dim(V)$$

(2) By contradiction, assume dim(V) = dim(W) = n but $V \neq W$ Then $\exists \vec{x} \in V \land \vec{x} \notin W$

Take S as a basis of W, then $\vec{x} \notin span\{S\}$

Then $S \cup \vec{x}$ is linearly independent

 $\implies S \cup \{\vec{x}\}\$ is linearly independent in V containing n+1 vectors

This contradicts the assumption by fundamental theorem since dim(V) = n so it could not contain more than n linearly independent vectors

6 Lecture Jan. 24 2018

6.1 Basis and Dimension

Theorem Let V be a vector space , S is a spanning set of V, and I is a linearly independent subset of V, s.t. $I \subseteq S$, then \exists basis B for V s.t. $I \subseteq B \subseteq S$.

Explaining

- 1. Any spanning set for V cab be **reduced** to basis for V by removing the linearly dependent(redundant) vector in the spanning set, using <u>redundancy theorem</u> to get a linearly independent spanning set.
- 2. Linear independent set can be **enlarged** to a basis for V.

proof.

omitted.

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Corollary Let V be a vector space and dim(V) = n, any set of n linearly independent vectors from V is a basis for V.

proof. If n linearly independent vectors did not span V, then could be enlarged to a basis of V by pervious theorem, but then have a basis containing more than n vectors from V, which is impossible by the fundamental theorem since we given the dim(V) = n, proven by contradiction.

Example Let
$$V = P_2(\mathbb{R})$$
, $p_1(x) = 2 - 5x$, $p_2(x) = 2 - 5x + 4x^2$, find $p_3 \in P_2(\mathbb{R})$ s.t. $\{p_1(x), p_2(x), p_3(x)\}$ is basis for $P_2(\mathbb{R})$

Note Since $dim(P_2(\mathbb{R})) = 3$ so any 3 linearly independent vectors from $P_2(\mathbb{R})$ will be a basis for $P_2(\mathbb{R})$.

Solutions e.g. constant function $p_3(x) = 1$, since $1 \notin span\{p_1(x), p_2(x)\}$, so $\{p_1(x), p_2(x), p_3(x)\}$ is a basis of $P_2(\mathbb{R})$. e.g. $p_3(x) = x$, since $x \notin span\{p_1(x), p_2(x)\}$

Theorem Let U and W be subspaces of vector space V, then we have

$$dim(U+W) = dim(U) + dim(W) - dim(U \cap W)$$

proof.

Let
$$\{\vec{v_i}\}_1^k$$
 be basis for $U \cap W$
 $\implies dim(U \cap W) = k$

Since $\{\vec{v_i}\}_1^k$ is basis for $U \cap W$ then it's a linearly independent subset of U So it could be enlarged to basis for U, $\{\vec{v_1}, \dots, \vec{v_k}, \vec{y_1}, \dots, \vec{y_r}\}$

So
$$dim(U) = k + r$$

We also could enlarge a basis for W $\{\vec{v_1}, \dots, \vec{v_k}, \vec{z_1}, \dots, \vec{z_s}\}$

$$\implies dim(V) = k + s$$

WTS. $\{\vec{v_1}, \dots, \vec{v_k}, \dots, \vec{y_1}, \dots, \vec{y_r}, \vec{z_1}, \dots, \vec{z_s}\}$ is a basis for U + W

(If we could show this) dim(U+W)=k+r+s=(k+r)+(k+s)-k

$$= dim(U) + dim(W) - dim(U \cap W)$$

Obviously, the above set spans U+W

WTS. $\{\vec{v_1},\ldots,\vec{v_k},\ldots,\vec{y_1},\ldots,\vec{y_r},\vec{z_1},\ldots,\vec{z_s}\}$ is linearly independent

Consider $a_1 \vec{v_1} + \dots + a_k \vec{v_k} + b_1 \vec{y_1} + \dots + b_r \vec{y_r} + c_1 \vec{z_1} + \dots + c_s \vec{z_s} = \vec{0} (\star)$

From
$$(\star) \implies \sum (c_i \vec{z_i}) = -\sum (a_i \vec{v_i}) - \sum b_i \vec{y_i}$$

 $\implies \sum (c_i \vec{z_i}) \in U \land \sum (c_i \vec{z_i}) \in W$
 $\iff \sum (c_i \vec{z_i}) \in U \cap W$

Since $\{\vec{v_i}\}$ is a basis for $U \cap W$

$$\implies \sum (c_i \vec{z_i}) = \sum (d_i \vec{v_i})$$

$$\iff \sum (c_i \vec{z_i}) - \sum (d_i \vec{v_i}) = \vec{0} \in W$$

 $\implies c_i = d_i = 0$ since $\{\vec{z_i}, \vec{v_i}\}$ is a basis

Rewrite (\star)

$$\sum (a_i \vec{v_i}) + \sum b_i \vec{y_i} = 0 \in U$$

 $\implies a_i = b_i = 0 \text{ since } \{\vec{v_i}, \vec{y_i}\} \text{ is a basis for } U$

Corollary For direct sum, since the intersection is $\{\vec{0}\}$

$$dim(U \bigoplus W) = dim(U) + dim(W)$$

Example Let U, W are subspaces of \mathbb{R}^3 such shat dim(U) = dim(W) = 2, why is $U \cap W \neq \{\vec{0}\}$

Solutions Geometrically, U and W are planes through origin then the intersection would be a line through $\operatorname{origin}(U \neq W)$ or a plane through $\operatorname{origin}(U = W)$, so shown.

Question V is a vector space, dim(V) = n, $U \neq W$ are subspaces of V but dim(U) = dim(V) = (n-1), proof:

- 1. V = U + W
- 2. $dim(U \cap W) = (n-z)$

7 Lecture 7 Jan. 30, 2018

7.1 Linear Transformations

Definition Let V, W be vector spaces, a function $T: V \to W$ is a **linear transformation**⁸ if

1.
$$T(\vec{x} + \vec{y}) = T(\vec{x}) + T(\vec{y}) \ \forall \vec{x}, \vec{y} \in V^9$$

2.
$$T(c\vec{x}) = cT(\vec{x}) \ \forall \vec{x} \in V, \ c \in \mathbb{R}^{10}$$

Linear transformation preserves <u>vector additions ans saclar multiplications</u> on vector spaces.

Theorem(Alternative definition) Transformation $T: V \to W$ is linear if and only if

$$T(c\vec{x} + d\vec{y}) = cT(\vec{x}) + dT(\vec{y}), \ \forall \vec{x}, \vec{y} \in V, c, d \in \mathbb{R}$$

Linear transformations preserves <u>linear combinations</u>.

Example (form 223) Rotation through angle θ about the origin in \mathbb{R}^2 .

⁸In some textbooks, this is annotated as **linear mapping**.

 $^{^{9}}$ Notice that the vector additions on the left and right sides of the equation are defined in different vector spaces, in V and W respectively.

 $^{^{10}}$ Notice that the scalar multiplication on the left and right sides of the equation are defined in different vector spaces, in V and W respectively.

Example (from 223) <u>Matrix transformation</u>, let $A \in M_{m \times n}(\mathbb{R})$, transformation $T : \mathbb{R}^n \to \mathbb{R}^m$ defined as

$$T(\vec{x}) = A\vec{x}$$

is linear.

Example Derivative $T: P_n(\mathbb{R}) \to P_{n-1}(\mathbb{R})$ defined by

$$T(\vec{p}(x)) = \vec{p}'(x)$$

Example Matrix transpose $T: M_{m \times n}(\mathbb{R}) \to M_{n \times m}(\mathbb{R})$ defined by

$$T(A) = A^T$$

7.2 Properties of linear transformations

Property(i) Linear transformation $T: V \to W$ are <u>uniquely</u> defined by their values on <u>any</u> basis for V.

proof.

Let
$$\{\vec{v_1}, \dots, \vec{v_k}\}$$
 be any basis for V

Every vector $\vec{x} \in V$ can be uniquely written as some linear combination of the $\{\vec{v}_i\}_{i=1}^k$

$$\vec{x} = \sum_{i=1}^{k} c_i \vec{v_i}, \ c_i \in \mathbb{R}, \text{ and } c_i \text{ are uniquely determined } \forall \vec{x} \in V$$

$$\implies T(\vec{x}) = T(\sum_{i=1}^{k} c_i \vec{v_i})$$

 $= \sum_{i=1}^{k} c_i T(\vec{v_i}) \text{ since the transformation } T \text{ is linear.}$

Since c_i s are uniquely determined by $\{\vec{v_i}\}_{i=1}^k$

so the value of $T(\vec{x})$ is uniquely determined by its value on basis vectors $\{\vec{v_i}\}_{i=1}^k$.

Property(ii) Let $T: V \to W$ be a linear transformation, let A be a subspace of vector space V, then the **image** T(A) defined as

$$T(A) = \{T(\vec{x}) \mid \vec{x} \in A\}$$

called the image of A under linear transformation T is a subspace of W. Linear transformation maps subspaces of V to subspaces of W.

proof.

Since A is a subspace so it's non-empty, therefore $\exists T(\vec{x}), \ \vec{x} \in A$

So
$$T(A) \neq \emptyset$$

Let $\vec{w_1}, \vec{w_2} \in T(A)$
 $\implies \vec{w_1} = T(\vec{x_1}), \vec{w_2} = T(\vec{x_2}), \ \vec{x_1}, \vec{x_2} \in A$
 $\implies \vec{w_1} + \vec{w_2} = T(\vec{x_1}) + T(\vec{x_2}) = T(\vec{x_1} + \vec{x_2}) \text{ since } T \text{ is linear.}$
Since $\vec{x_1} + \vec{x_2} \in A$ by the definition of subspaces.

$$\implies \vec{w_1} + \vec{w_2} \in T(A)$$

So T(A) is closed under vector addition.

Let
$$\vec{w} \in T(A)$$

$$\implies \vec{w} = T(\vec{x}), \vec{x} \in A$$
Let $c \in \mathbb{R}$
Consider $c\vec{w} = cT(\vec{x}) = T(c\vec{x})$
Since $c\vec{x} \in A$
So $c\vec{w} \in T(A)$

So T(A) is closed under scalar multiplication.

Property(derived from the definition) For all linear transformation $T: V \to W$, we have ¹¹

$$T(\vec{0}) = \vec{0}$$

Property(iii) Let transformation $T: V \to W$ be linear, let B be a subspace of W, then its **pre-image** defined as

$$T^{-1}(B) = \{ \vec{x} \in V \mid T(x) \in B \}$$

is a subspace of V. ¹²

¹¹In the equation, clearly, the zero vector on the left side of the equation is in space V and the zero vector on the right side is in space W.

 $^{^{12}}$ The pre-image and inverse share the same notation, but in this case, transformation T is not necessarily invertible.

proof.

Let
$$\vec{w_1}, \vec{w_2} \in T^{-1}(B)$$

$$\implies T(\vec{w_1}), T(\vec{w_2}) \in B$$

$$\implies aT(\vec{w_1}) + b(\vec{w_2}) \in B, \ \forall a, b \in \mathbb{R} \text{ since } B \text{ is a subspace.}$$

$$\implies T(a\vec{w_1} + b\vec{w_2}) \in B$$

$$\implies a\vec{w_1} + b\vec{w_2} \in T^{-1}(B)$$

So $T^{-1}(B)$ is closed under both vector addition and scalar multiplication, So $T^{-1}(B)$ is a subspace.

7.3 Definitions

Let $T: V \to W$ to be a linear transformation,

Definition the **Image** of transformation T is defined as

$$Im(T) = T(V) = \{T(\vec{x}) \mid \vec{x} \in V\}$$

Definition the **Rank** of transformation T is defined as

$$Rank(T) = dim(Im(T))$$

Definition the **Kernel** of transformation T is defined as

$$Ker(T) = T^{-1}(\{\vec{0}\}) = \{\vec{x} \in V \mid T(\vec{x}) = \vec{0}\}\$$

Definition the **Nullity** of transformation T is defined as

$$Nullity(T) = dim(ker(T))$$

Example $T: P_3(\mathbb{R}) \to P_3(\mathbb{R})$ is <u>linear</u> defined by

$$T(\vec{p}(x)) = \vec{p}(2x+1) - 8\vec{p}(x)$$

find Ker(T).

Theorem Let $T: V \to W$ be a linear transformation, let $\{\vec{v_1}, \dots, \vec{v_k}\}$ be the spanning set of V^{13} , then $\{T(\vec{v_1}), \dots, T(\vec{v_k})\}$ spans Im(T)

proof.

Let
$$\vec{w} \in Im(T)$$

Since
$$V = span\{\vec{v_1}, \dots, \vec{v_k}\}$$

For any $\vec{x} \in V$ can be written as

$$\vec{x} = \sum_{i=1}^{k} c_i \vec{v_i}, \ c_i \in \mathbb{R}$$

$$\implies \vec{w} = T(\vec{x}) = T(\sum_{i=1}^{k} c_i \vec{v_i})$$

$$= \sum_{i=1}^{k} c_i T(\vec{v_i})$$

as a linear combination of $\{T(\vec{v_1}), \dots, T(\vec{v_k})\}$

So
$$Im(T) = span\{T(\vec{v_1}), \dots, T(\vec{v_k})\}$$

8 Lecture 8 Jan. 31 2018

8.1 Linear Transformations

Example $T: P_3(\mathbb{R}) \to P_3(\mathbb{R})$

$$T(p(x)) = p(2x+1) - 8p(x)$$

Find the image of T.

We know $B = \{1, x, x^2, x^3\}$ is the standard basis for $P_3(\mathbb{R})$, consider the set P(B)

$$P(B) = \{-7, 1 - 6x, 1 + 4x - 4x^2, 1 + 6x + 12x^2\}$$

spans Im(T). Notice the first three vectors in the set is linearly independent, the last vector is clearly dependent to the pervious three.¹⁴. So by the <u>redundancy theorem</u> we could remove the last vector. There we have

$$Im(T) = span\{-7, 1 - 6x, 1 + 4x - 4x^2\}$$

¹³The set is only the spanning set of V, it's not necessarily to be a basis of V.

¹⁴Notice that the first three vectors is a basis of $P_2(\mathbb{R})$.

as basis.

In this example, the dimension of Ker(T) is 1 and the dimension of Im(T) is 3, and dimension of $P_3(\mathbb{R})$ is 4. We have, $dim(P_3(\mathbb{R})) = Nullity(T) + Rank(T)$

Theorem(Dimension Theorem) Let $T: V \to W$ be a linear transformation,

$$dim(V) = Nullity(T) + Rank(T)$$

Proof.

Say
$$dim(V) = n$$

Let $\{\vec{v_1}, \dots, \vec{v_k}\}$ be a basis for Ker(T)

Since Ker(T) is a subspace of V, the set $\{\vec{v_i}\}_1^k$ is a subset of V,

It can be extended to a basis $\{\vec{v_i}\}_1^k \cup \{\vec{v_i}\}_{k+1}^n$ for V.

Claim:
$$\{T(\vec{v_{k+1}}), \dots, T(\vec{v_n})\}\$$
 is basis for $Im(T)$

If the claim is true, this prove the theorem since

$$\dim(Ker(T))+\dim(Im(T))=k+n-k=n=\dim(V)$$

$$T(\vec{v_i}) = \vec{0}, \ \forall i \in \mathbb{Z}_1^k$$

and by the definition of kernel of linear transformation,

$$T: \{T(\vec{v_i})\}_{k=1}^n \text{ spans } Im(T)$$

Show if
$$\sum_{i=k+1}^{n} c_i T(\vec{v_i}) = \vec{0} \implies c_i = 0$$

$$\implies T(\sum_{i=k+1}^{n} c_i \vec{v_i}) = \vec{0}$$

$$\implies \sum_{i=k+1}^n c_i \vec{v_i} \in Ker(T)$$

$$\implies \sum_{i=k+1}^{n} c_i \vec{v_i} = \sum_{i=1}^{k} c_i \vec{v_i}$$

$$\implies c_1 \vec{v_1} + \dots + c_k \vec{v_k} - c_{k+1} \vec{v_{k+1}} - \dots - c_n \vec{v_n} = \vec{0}$$

Since $\{\vec{v_i}\}_i^n$ is a basis for V.

$$\implies c_i = 0 \ \forall i$$

8.2 Applications of dimension theorem

Definition A linear transformation $T: V \to W$ is called **injective**(one-to-one) if and only if

$$T(\vec{v_1}) = T(\vec{v_2}) \implies \vec{v_1} = \vec{v_2}$$

Definition A linear transformation $T: V \to W$ is called **surjective**(onto) if and only if

$$Im(T) = W$$

Every vector in W has a pre-image in V.

Definition A linear transformation $T: V \to W$ is called **bijective** if it's both injective and surjective.

Theorem Let transformation $T: V \to W$ is linear, T is injective if and only if dim(Ker(T)) = 0.

Proof.

Exercise

Theorem T is surjective if and only if dim(Im(T)) = dim(W).

Example $T: P_2(\mathbb{R}) \to \mathbb{R}^2$ defined by

$$T(p(x)) = \begin{pmatrix} p(1) \\ p(2) \end{pmatrix}$$

is T injective? surjective?

Not injective but surjective.

Solution

$$Ker(T) = span\{(x-1)(x-2)\}$$

So T has nullity of 1 and since $dim(P_2(\mathbb{R})) = 3$, by the <u>dimension theorem</u> we have Rank(T) = 2 and since Im(T) is a subspace of \mathbb{R}^2 which has dimension of 2, we could conclude that $Im(T) = \mathbb{R}^2$.

9 Lecture 9 Feb. 6 2018

9.1 Applications of dimension theorem

Recall Dimension Theorem $T: V \to W$ is linear transformation,

$$dim(V) = dim(Ker(T)) + dim(Im(T))$$

Recall T is **injective** if and only if dim(Ker(T)) = 0.

Recall T is surjective if and only if dim(Im(T)) = dim(W).

Example $T: P_2(\mathbb{R}) \to \mathbb{R}^3$ defined by

$$T(p(x)) = (p(1), p(2), p(3))$$

Take $p(x) = a + bx + cx^2 \in P_2(\mathbb{R}), p(x) \in Ker(T) \text{ iff } T(p(x)) \in \vec{0}.$ Let $p(x) \in Ker(T),$

Obviously the only solution for the system

$$\begin{cases} a+b+c = 0 \\ a+2b+4c = 0 \\ a+3b+9c = 0 \end{cases}$$

is a = b = c = 0, So dim(Ker(T)) = 0. Therefore, T is **injective**. By dimension theorem,

$$dim(V) = 3 = 0 + dim(Im(T)) \implies dim(Im(T)) = 3 = dim(\mathbb{R}^3)$$

therefore T is surjective. Therefore, T is called **bijective**.

Question $T: P_n(\mathbb{R}) \to P_n(\mathbb{R})$

$$T(p(x)) = xp'(x)$$

Solution Not injective because any constant function in $P_n(\mathbb{R})$ is mapped onto $\vec{0} \in P_n(\mathbb{R})$. Also not surjective by the dimension theorem.

Theorem Let $T: V \to W$ be an <u>injective</u> linear transformation, if $\{\vec{v_i}\}_{i=1}^k$ is linearly independent in V, then the set $\{T(\vec{v_i})\}_{i=1}^k$ is linearly independent in W.

 $\label{linear linear linear$

Proof.

If $\sum c_i T(\vec{v_i}) = \vec{0}$, then we have $T(\sum c_i \vec{v_i}) = \vec{0}$, which means $\sum c_i v_i \in Ker(T)$. By definition of injective transformation, $\sum c_i v_i = \vec{0}$. Since $\{\vec{v_i}\}_{i=1}^k$ is linearly independent, so $c_i = 0$, $\forall i$.

Theorem $T: V \to W$ is a linearly transformation, $\{\vec{v_i}\}_{i=1}^n$ is a basis for V then, if $\{T(\vec{v_i})\}_{i=1}^n$ is linear independent, then T is <u>injective</u>. A criteria for T to be injective based on image of a basis.

Proof.

Let
$$\{\vec{v_i}\}_{i=1}^n$$
 be a basis of V
Consider $T(\vec{x}) = \vec{0}$
Since $\{\vec{v_i}\}_{i=1}^n$ is a basis
Let $x = \sum c_i \vec{v_i}$
 $T(\vec{x}) = \vec{0} \iff T(\sum c_i \vec{v_i}) = \vec{0}$
 $\implies \sum c_i T(\vec{v_i}) = \vec{0} \implies c_i = 0$
 $\therefore \vec{x} = \sum 0 \vec{v_i} = \vec{0}$
Therefore $Ker(T) = \{\vec{0}\}$
Therefore $dim(Ker(T)) = 0$
 \implies injective

Theorem Let $T: V \to W$ be a linear transformation,

- 1. If dim(V) > dim(W), then T cannot be injective.
- 2. If dim(V) < dim(W), then T cannot be surjective.

For a linear transformation between spaces with different dimension, it could not be bijective.

Proof.

$$\begin{aligned} \dim(V) &= \dim(Ker(T)) + \dim(Im(T)) \\ & \because \dim(Im(T)) \leq \dim(W) \\ & \therefore \dim(V) \leq \dim(Ker(T) + \dim(W)) \\ & \Longrightarrow \dim(Ker(T)) \geq \dim(V) - \dim(W) \\ & \Longrightarrow \dim(Ker(T)) > 0 \\ & \text{So } T \text{ could not be injective} \\ \dim(V) &= \dim(Ker(T)) + \dim(Im(T)) \\ & \because \dim(Ker(T)) \geq 0 \\ & \therefore \dim(V) \geq \dim(Im(T)) \\ & \Longrightarrow \dim(Im(T)) < \dim(W) \\ & \text{So } T \text{ could not be surjective} \end{aligned}$$

Theorem Half is good enough Let $T: V \to W$ is linear, and dim(V) = dim(W). T is injective if and only if surjective.

Proof.

By dimension theorem
$$dim(V) = dim(Ker(T)) + dim(Im(T)) = dim(W)$$
 If injective
$$dim(Ker(T)) = 0$$

$$\implies dim(Im(T)) = dim(W)$$
 So surjective
$$\text{If surjective } dim(Im(T)) = dim(W) = dim(V)$$

$$\implies dim(Ker(T)) = 0$$
 So injective

9.2 Isomorphisms

Recall If $T: V \to W$ is both injective and surjective, say T is bijective.

Definition If $T:V\to W$ is bijective, we call T an **isomorphism**. If there exists an isomorphism $T:V\to W$ say V and W are **isomorphic** vector spaces.

Theorem V, W are isomorphic <u>iff</u> dim(V) = dim(W).

Proof.

$$\rightarrow V, W \text{ isomorphic } \implies dim(V) = dim(W)$$

Isomorphic means there exists a bijective transformation T

By dimension theorem
$$dim(V) = dim(Ker(T)) + dim(Im(T))$$

= $0 + dim(W)$

$$\leftarrow dim(V) = dim(W) \implies V, W \text{ isomorphic}$$

Equivalently, find a bijective transformation

Let
$$\{\vec{v_i}\}_{i=1}^n$$
 be basis for V

Let
$$\{\vec{w_i}\}_{i=1}^n$$
 be basis for W

Claim $T: V \to W$ is linear and s.t.

 $T(\vec{v_i}) = \vec{w_i}$ is an isomorphism.

If
$$\vec{x} \in Ker(T) \subseteq V$$

$$x = \sum c_i \vec{v_i}$$

$$\vec{0} = T(\vec{x})$$

$$= \sum c_i T(\vec{v_i})$$

$$= \sum (c_i \vec{w_i})$$

 $\implies c_i = 0$ since $\vec{w_i}$ are basis.

$$\implies \vec{x} = \vec{0}$$

$$\implies dim(Ker(T)) = 0$$

 \implies injective \iff surjective

Note if $T: V \to W$ is an isomorphism, then T maps a basis for V to a basis for W.

Example $T: P_2(\mathbb{R}) \to \mathbb{R}^3$,

$$T(p(x)) = (p(1), p(2), p(3))$$

is an isomorphism. And $P_2(\mathbb{R})$ and \mathbb{R}^3 are isomorphic.

Example $T: P_2(\mathbb{R}) \to \mathbb{R}^3$,

$$T(1) = \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}, \ T(x) = \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix}, \ T(x^2) = \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}$$

is an isomorphism.

Example $M_{2\times 2}(\mathbb{R}), P_3(\mathbb{R})$ and \mathbb{R}^4 are isomorphic.

Theorem Any n-dim vector space V is isomorphic to \mathbb{R}^n . What is an isomorphism $T:V\to\mathbb{R}^n$

Procedure:

Let
$$\{\vec{v_i}\}_{i=1}^n$$
 be any basis for V
We know that $\forall \vec{x} \in V$,
By property of basis,

$$\vec{x} = \sum c_i \vec{v_i}$$
 Then $T(\vec{x}) = \begin{pmatrix} c_1 \\ \vdots \\ c_n \end{pmatrix} \in \mathbb{R}^n$ is an isomorphism.

9.3 Coordinates

Definition Let V be a vector space, $\alpha = \{\vec{v_1}, \dots, \vec{v_n}\}$ be nay basis for V, $\forall \vec{x} \in V$ can be written uniquely as

$$\vec{x} = c_1 \vec{v_1} + \dots + c_n \vec{v_n}$$

then $c1, \ldots, c_n$ is called the **coordinates** for \vec{x} relative to α , with notation

$$[\vec{x}]_{\alpha} = \begin{pmatrix} c_1 \\ \vdots \\ c_n \end{pmatrix} \iff \vec{x} = \sum c_i \vec{v_i}$$

Claim $[\vec{x} + c\vec{y}]_{\alpha} = [\vec{x}]_{\alpha} + c[\vec{y}]_{\alpha} \quad \forall \vec{x}, \vec{y} \in V, \ c \in \mathbb{R}.$

Note if α, α' are any two bases for V then generally $[\vec{x}]_{\alpha} \neq [\vec{x}]_{\alpha'}$ (except $\vec{0}$).

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10.1 Matrix of linear transformation

Recall Let V be a vector space, let α be any basis for V.

$$\forall \vec{x} \in V, x = \sum c_i \vec{v_i}$$

$$[\vec{x}]_{\alpha} = \begin{pmatrix} c_1 \\ \vdots \\ c_n \end{pmatrix}$$

So transformation $\vec{x} \to [\vec{x}]_{\alpha}$ is an isomorphism that $V \to \mathbb{R}^n$.

Say W is a vector space and let $\beta = \{\vec{w_i}\}_1^m$ be any basis of W, say $T: V \to W$ is linear.

$$T(\vec{x}) = \sum c_i T(\vec{v_i})$$

So that

$$[T(\vec{x})]_{\beta} = [\sum c_i T(\vec{v_i})]_{\beta} = \sum c_i [T(\vec{v_i})]_{\beta}$$

$$= \begin{bmatrix} [T(\vec{v_1})]_{\beta} & \dots & [T(\vec{v_n})]_{\beta} \end{bmatrix} \begin{pmatrix} c_1 \\ \vdots \\ c_n \end{pmatrix}$$

 $[[T(\vec{v_1})]_{\beta}$... $[T(\vec{v_n})]_{\beta}]$ is called the <u>the matrix of T w.r.t.</u> α, β . Denoted as $[T]_{\alpha}^{\beta}$

$$[T(\vec{x})]_{\beta} = [T]_{\alpha}^{\beta} [\vec{x}]_{\alpha}$$

Example $T: P_2(\mathbb{R}) \to P_3(\mathbb{R})$

$$T(p(x)) = xp(x)$$

$$\alpha = \{1-x, 1-x^2, x\}, \ \beta = \{1, 1+x, 1+x+x^2, 1-x^3\}$$

Find $[T]^{\beta}_{\alpha}$.

$$T(1-x) = x(1-x) = x - x^{2}$$

$$x - x^{2} = (-1)(1) + 2(1+x) + (-1)(1+x+x^{2}) + 0(1-x^{3})$$

$$[T(1-x)]_{\beta} = (-1,2,-1,0)$$

$$T(1-x^{2}) = x - x^{3}$$

$$[T(1-x^{2})]_{\beta} = (-2,1,0,1)$$

$$[T(x)] = x^{2}$$

$$[T(x)]_{\beta} = (0,-1,1,0)$$

$$[T]_{\alpha}^{\beta} = \begin{bmatrix} -1 & -2 & 0\\ 2 & 1 & -1\\ -1 & 0 & 1\\ 0 & 1 & 0 \end{bmatrix}$$

Picture V, W are vectors spaces, $\alpha = \{\vec{v_1}, \dots, vecv_n\}$ is a basis for V and $\beta = \{\vec{w_1}, \dots, vecw_m\}$ is a basis for W.

$$V \longrightarrow^{T} W$$

$$\downarrow^{[\]_{\alpha}} \qquad \downarrow^{[\]_{\beta}}$$

$$\mathbb{R}^{n} \rightarrowtail^{[T]_{\alpha}^{\beta}} \mathbb{R}^{m}$$

Note

1.
$$\vec{x} \in Ker(T) \iff T(\vec{x}) = \vec{0} \iff [T(x)]_{\beta} = [\vec{0}]_{\beta} \in \mathbb{R}^m \iff [T]_{\alpha}^{\beta}[\vec{x}]_{\alpha} = 0 \iff [\vec{x}]_{\alpha} \in Ker([T]_{\alpha}^{\beta})$$

2.
$$\vec{w} \in Im(T) \iff [\vec{w}]_{\beta} \in Col([T]_{\alpha}^{\beta})$$

Theorem(Rank nullity for transformation matrix)

$$\dim(Ker([T]_{\alpha}^{\beta}))+\dim(Col([T]_{\alpha}^{\beta}))=n$$

Example $T: P_2(\mathbb{R}) \to M_{2\times 2}(\mathbb{R})$

$$T(a+bx+c^{2}) = \begin{bmatrix} c & -c \\ a-c & a+c \end{bmatrix}$$

And given bases $\alpha = \{x^2 - x, x - 1, x^2 + 1\}$ and $\beta = \{\begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}, \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 0 \\ 1 & 1 \end{bmatrix}, \begin{bmatrix} -1 & 1 \\ 0 & 0 \end{bmatrix}\}$

Answer

$$\begin{split} [T]_{\alpha}^{\beta} &= \begin{bmatrix} 1 & 0 & 1 \\ -1 & 0 & -1 \\ 0 & -1 & 1 \\ 0 & 0 & 0 \end{bmatrix} \\ Nul([T]_{\alpha}^{\beta}) &= span\{ \begin{pmatrix} -1 \\ 1 \\ 1 \end{pmatrix} \} \\ Nul(T) &= span\{2x\} \\ Col([T]_{\alpha}^{\beta}) &= span\{ \begin{pmatrix} -1 \\ 1 \\ 0 \\ 0 \end{pmatrix}, \begin{pmatrix} 0 \\ 0 \\ -1 \\ 0 \end{pmatrix} \} \\ Col(T) &= span\{ \begin{bmatrix} -1 & 1 \\ 1 & -1 \end{bmatrix}, \begin{bmatrix} 0 & 0 \\ -1 & -1 \end{bmatrix} \} \end{split}$$

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11.1 Algebra of Transformation

Recall Let $T: V \to W$ be a linear transformation, where $\alpha = \{\vec{v_1}, \dots, \vec{v_n}\}$ and $\beta = \{\vec{w_1}, \dots, \vec{w_m}\}$ are bases for V, W respectively.

$$\vec{x} \in Ker(T) \iff [\vec{x}]_{\alpha} \in Ker([T]_{\alpha}^{\beta})$$

 $\vec{x} \in Im(T) \iff [\vec{x}]_{\beta} \in Col([T]_{\alpha}^{\beta})$

Definition $T_1, T_2: V \to W$ are linear transformations, define

$$(T_1 + T_2)(\vec{x}) = T_1(\vec{x}) + T_2(\vec{x}) \forall \vec{x} \in V$$
$$(cT_1)(\vec{x}) = c(T_1(\vec{x})) \forall \vec{x} \in V, \ c \in \mathbb{R}$$

And, let α and β be bases for V, W respectively, then,

$$[T_1]^{\beta}_{\alpha} + [T_2]^{\beta}_{\alpha} = [T_1 + T_2]^{\beta}_{\alpha}$$

 $c[T_1]^{\beta}_{\alpha} = [cT_1]^{\beta}_{\alpha}$

Definition $T: V \to W$ and $S: W \to U$ are linear transformations, then the **composition** $ST: V \to U$ is defined as

$$(ST)(\vec{x}) = S(T(\vec{x})) \quad \forall \vec{x} \in V$$

Note If S, T are linear then the composition ST is also linear. Check

Let
$$a, b \in \mathbb{R}, \ \vec{x}, \vec{y} \in V$$

$$ST(a\vec{x} + b\vec{y})$$

$$= S(T(a\vec{x} + b\vec{y}))$$

$$= S(aT(\vec{x}) + bT(\vec{y}))$$

$$= a(ST(\vec{x})) + b(ST(\vec{y}))$$

Example

omitted

11.2 Matrix of composition

Consider $T:V\to W$ and $S:W\to U$ as linear transformations, let α , β , γ be bases of V, W, U respectively. We know how to compute $[T]^{\beta}_{\alpha}$ and $[S]^{\gamma}_{\beta}$. Now want to find $[ST]^{\gamma}_{\alpha}$.

$$\begin{aligned} \forall \vec{x} \in V, [ST]_{\alpha}^{\gamma}[\vec{x}]_{\alpha} \\ &= [(ST)(\vec{x})]_{\gamma} \\ &= [S(T(\vec{x}))]_{\gamma} \\ &= [S]_{\beta}^{\gamma}[T(\vec{x})]_{\beta} \\ &= [S]_{\beta}^{\gamma}[T]_{\alpha}^{\beta}[\vec{x}]_{\alpha} \end{aligned}$$
 This holds true for all $\vec{x} \in V$

$$\therefore [ST]_{\alpha}^{\gamma} = [S]_{\beta}^{\gamma}[T]_{\alpha}^{\beta}$$

Conclusion the matrix of $ST = \text{matrix of } S \times \text{matrix of } T$.

11.3 Inverse transformations

Definition $T: V \to W$ is $isomorphism^{15}$ if and only if there exists function $S: W \to V$ such that

$$(ST)(\vec{v}) = \vec{v} \ \forall \vec{v} \in V \land (TS)(\vec{w}) = \vec{w} \ \forall \vec{w} \in W$$

And S is called the **inverse** of T, written as T^{-1} .

 $proof. \to T$ is an isomorphism means every vector in W has an unique preimage in V the function $S: W \to V$ maps every vector in W to its unique pre-image in V, so S is the inverse of T.

proof. ← Assume $S: W \to V$ is the inverse of $T: V \to W$ then $T(S(\vec{y})) = \vec{y} \, \forall \vec{y} \in V$, this means T is surjective since every $\vec{y} \in W$ has pre-image under T, that's $S(\vec{y}) \in V$. Now suppose $T(\vec{x_1}) = T(\vec{x_2})$, apply transformation S on both sides of the equation, $S(T(\vec{x_1})) = S(T(\vec{x_2}))$ we have $\vec{x_1} = \vec{x_2}$. This implies the transformation is injective. Therefore, transformation T is bijective, that's isomorphism.

Note $T^{-1}(\vec{y})$ is the <u>unique</u> vector \vec{x} , s.t. $T(\vec{x}) = \vec{y}$. That's

$$T(\vec{x}) = \vec{y} \iff T^{-1}(\vec{y}) = \vec{x}$$

¹⁵Recall that isomorphism is equivalent to bijective.

Theorem If $T: V \to W$ is an isomorphism then the inverse of T, T^{-1} , then $T^{-1}: W \to V$ is linear.¹⁶

Proof.

WTS
$$T^{-1}(a\vec{w_1} + b\vec{w_2}) = aT^{-1}(\vec{w_1}) + bT^{-1}(\vec{w_2}) \forall a, b \in \mathbb{R}, \forall \vec{w_1}, \vec{w_2} \in W$$

$$T^{-1}(\vec{w_1}) \text{ is the unique } \vec{x_1} \text{ s.t. } T(\vec{x_1}) = \vec{w_1}$$

$$T^{-1}(\vec{w_2}) \text{ is the unique } \vec{x_2} \text{ s.t. } T(\vec{x_2}) = \vec{w_2}$$

$$T^{-1}(a\vec{w_1} + b\vec{w_2}) \text{ is the unique } \vec{x} \text{ s.t. } T(\vec{x}) = a\vec{w_1} + b\vec{w_2}$$

$$\therefore T(\vec{x}) = a\vec{w_1} + b\vec{w_2}$$

$$= aT(\vec{x_1}) + bT(\vec{x_2})$$

$$= T(a\vec{x_1} + b\vec{x_2})$$

$$\therefore \vec{x} = a\vec{x_1} + b\vec{x_2}$$

$$Also T(\vec{x}) = a\vec{w_1} + b\vec{w_2}$$

$$\therefore \vec{x} = T^{-1}(a\vec{w_1} + b\vec{w_2}) = a\vec{x_1} + b\vec{x_2}$$

$$= aT^{-1}(\vec{w_1}) + bT^{-1}(\vec{w_2})$$

Theorem $T:V\to W$ is isomorphism, then let α and β are bases of V and W representing then $[T]^{\beta}_{\alpha}$ is invertible, and

$$([T]_{\alpha}^{\beta})^{-1} = [T^{-1}]_{\alpha}^{\beta}$$

Proof. omitted

11.4 Change of basis

What's the effect of a change of basis on coordinate of a vector and matrix of transformation.

Theorem Let α and α' be two bases of V, then

$$[I]^{\alpha'}_{\alpha}[\vec{x}]_{\alpha} = [\vec{x}]_{\alpha'}$$

 $^{^{16}}$ Note: the conclusion could be changed into isomorphism.

Proof.

Let
$$\vec{x} \in V$$

$$I(\vec{x}) = \vec{x}$$

$$[I(\vec{x})]_{\alpha'} = [\vec{x}]_{\alpha'}$$

$$[I]_{\alpha}^{\alpha'}[\vec{x}]_{\alpha} = [\vec{x}]_{\alpha'}$$

 $[I]^{\alpha'}_{\alpha}$ is called the change of basis matrix from α to α' .

Computation Let $\alpha = \{\vec{a_1}, \dots, \vec{a_n}\}$, then

$$[I]_{\alpha}^{\alpha'} = [[\vec{a_1}]_{\alpha'} \mid \dots \mid [\vec{a_n}]_{\alpha'}]$$

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Recall Let α and β be bases for V and $I:V\to V$ is the identity transformation, then

$$[I]^{\beta}_{\alpha}[\vec{x}]_{\alpha} = [\vec{x}]_{\beta}$$

Also,

$$[I]^{\alpha}_{\beta}[\vec{x}]_{\beta} = [\vec{x}]_{\alpha}$$

Example Let $\alpha = \{x^2, 1+x, x+x^2\}$ and β be bases for $P_2(\mathbb{R})$ and

$$[I]_{\alpha}^{\beta} = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 1 \\ -1 & 0 & 1 \end{pmatrix} \text{ and } [\vec{p(x)}]_{\beta} = \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}$$

Solution omitted

Theorem Suppose $[T]_V^W$ is linear, α and α' are any two bases for V and β and β' are any two bases of W, then,

$$[T]_{\alpha'}^{\beta'} = [I]_{\beta}^{\beta'} [T]_{\alpha}^{\beta} [I]_{\alpha'}^{\alpha}$$

Proof.

Recall
$$T = ITI$$

Consider let $\vec{x} \in V$
 $[I]^{\beta'}_{\beta}[T]^{\beta}_{\alpha}[I]^{\alpha}_{\alpha'}[\vec{x}]_{\alpha'}$
 $= [I]^{\beta'}_{\beta}[T]^{\beta}_{\alpha}[\vec{x}]_{\alpha}$
 $= [I]^{\beta'}_{\beta}[T(\vec{x})]_{\beta}$
 $= [T(\vec{x})]_{\beta'}$
 $= [T]^{\alpha'}_{\beta'}[\vec{x}]_{\alpha'}$
 $\Longrightarrow [T]^{\alpha'}_{\beta'} = [I]^{\beta'}_{\beta}[T]^{\beta}_{\alpha}[I]^{\alpha}_{\alpha'}$

Also,

$$[T]^{\beta}_{\alpha} = [I]^{\beta}_{\beta'}[T]^{\beta'}_{\alpha'}[I]^{\alpha'}_{\alpha}$$

Special Case Consider when V = W, $\alpha = \beta$ and $\alpha' = \beta'$. we have

$$[T]_{\alpha'}^{\alpha'} = [I]_{\alpha}^{\alpha'} [T]_{\alpha}^{\alpha} [I]_{\alpha'}^{\alpha}$$

where

$$([I]^{\alpha'}_{\alpha})^{-1} = [I]^{\alpha}_{\alpha'}$$

the equation becomes

$$[T]_{\alpha'}^{\alpha'} = ([I]_{\alpha}^{\alpha'})^{-1} [T]_{\alpha}^{\alpha} [I]_{\alpha'}^{\alpha}$$

and can be written in the form of

$$B = P^{-1}AP$$

Definition Two matrices A and B are **similar** if there exists an <u>invertible</u> matrix P s.t.

$$B = P^{-1}AP$$

A and B representing the same transformation relative to different bases and P is the change of basis matrix if and only if A and B are similar.

Example Omitted

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13.1 Recall

Omitted.

13.2 Diagonalization

Definition Consider a linear operator $T: V \to V$ is **diagonalizable** if and only \exists a basis β for V s.t.

$$[T]^{\beta}_{\beta}$$

is diagonal.

Note Let $\beta = \{\vec{v_1}, \dots, \vec{v_n}\}$ be a basis, $T: V \to V$ is diagonalizabel if and only it's in form $\begin{bmatrix} \lambda_1 & \dots & 0 \\ \dots & \dots & \dots \\ 0 & \dots & \lambda_n \end{bmatrix}$

Definition $T: V \to V$ is a linear operator on V, a non-zero vector $\vec{x} \in V$ is an **eigenvector** of $T(\vec{x}) = \lambda \vec{x}$ for some $\lambda \in \mathbb{R}$. λ is called the **eigenvalue** of T corresponding to vector \vec{x} .

Theorem Linear operator $T: V \to V$ is diagonalizable if and only exist a basis of V consisting of eigenvectors of T. If T is diagonalizable, the diagonal entries of $[T]^{\beta}_{\beta}$ are corresponding eigenvalues of T, in the same order.

13.3 How to find eigenvalues and eigenvectors of T

Definition The **determinant** of T is defined as $det([T]^{\alpha}_{\alpha})$ for any basis α for V.

Note The determinant of linear operator T does <u>not</u> depends on the choice of basis of α for V, since similar matrices have the same determinant.

Theorem $\lambda \in \mathbb{R}$ is an eigenvalue of T if and only if

$$det(T - \lambda I) = 0$$

 $Proof.\lambda$ is an eigenvalue of T

$$\iff \exists \vec{x} \in V, \ \vec{x} \neq \vec{0}, \ s.t. \ T(\vec{x}) = \lambda \vec{x}$$

$$\iff T(\vec{x}) - \lambda \vec{x} = \vec{0}$$

$$\iff (T - \lambda I)(\vec{x}) = \vec{0}$$

$$\iff \vec{x} \in Ker(T - \lambda I)$$

$$\therefore Ker(T - \lambda I) \neq \{\vec{0}\}$$

$$\iff (T - \lambda I) \text{ is not injective}$$

$$\iff [T - \lambda I]^{\alpha}_{\alpha} \text{ is not injective}$$

$$\iff det([T - \lambda I]^{\alpha}_{\alpha}) = det(T - \lambda I) = 0$$

Note $det(T - \lambda I) = 0$ is called the **characteristic polynomial** of T, written as $P_T(\lambda) := det(T - \lambda I)$, the degree os $P_T(\lambda)$ is the dimension of V.

Note λ is an eigenvalue $\iff \lambda$ is a root of $P_T(\lambda)$.

Theorem $T: V \to V$ is a linear operator and λ is an eigenvalue of T, \vec{x} is an eigenvector of T corresponding to eigenvalue λ , if and only if

$$\vec{x} \neq \vec{0} \land \vec{x} \in Ker(T - \lambda I)$$

Proof.

Exercise, using definition

Definition $Ker(T - \lambda I)$ is called the **eigenspace** of T corresponding to eigenvalue λ , noted as $E_{\lambda}(T)$, which is a subspace of V.

Note To find eigenvalues and eigenvectors of $T: V \to V$, choose any basis β for V, \vec{x} is an eigenvector with corresponding eigenvalue λ if and only if $[\vec{x}]_{\beta}$ is an eigenvector of $[T]_{\beta}^{\beta}$ with corresponding eigenvalue λ .

That's

$$T(\vec{x}) = \lambda \vec{x}$$

$$\implies [T(\vec{x})]_{\beta} = [\lambda \vec{x}]_{\beta}$$

$$\iff [T]_{\beta}^{\beta} [\vec{x}]_{\beta} = \lambda [\vec{x}]_{\beta}$$

Note Consider diagonalization in MAT223,

$$D = P^{-1}AP$$

Let D and A representing the same linear operator $[T]_V^V$ and let β be a basis of V consisting of eigenvectors of T and α is another basis of V. Then, the above equation is

$$[T]^{\beta}_{\beta} = ([I]^{\alpha}_{\beta})^{-1} [T]^{\alpha}_{\alpha} [I]^{\alpha}_{\beta}$$

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Theorem Suppose λ_0 is an eigenvalue of linear operator $T: V \to V$, let $dim(E_{\lambda 0}) = k$, then $(\lambda - \lambda_0)^k$ divides $P_T(\lambda)$

Proof.

Let
$$\{\vec{v_1}, \dots, \vec{v_k}\}$$
 be basis for E_{λ_0}
Let $dim(V) = n$
Extend basis of E_{λ_0} to basis of V .
 $\alpha = \{\vec{v_1}, \dots, \vec{v_k}\} \cup \{\vec{v_{k+1}}, \dots, \vec{v_n}\}$
Since $\vec{v_i} \in E_{\lambda_0}$,
Therefore $T(\vec{v_i}) = \lambda_0 \vec{v_i}$
 $[T]^{\alpha}_{\alpha} = \begin{bmatrix} A & B \\ 0 & D \end{bmatrix}$
Where $A = diag(\lambda_0, \dots, \lambda_0) \in \mathbb{M}_{k \times k}(\mathbb{R})$
And $B \in \mathbb{M}_{k \times n-k}(\mathbb{R}), D \in \mathbb{M}_{n-k \times n-k}(\mathbb{R})$
 $P_T(\lambda) = det(A - \lambda I) * det(D - \lambda I)$
 $= (\lambda_0 - \lambda)^k * det(D - \lambda I)$
Therefore $(\lambda - \lambda_0)^k \mid P_T(\lambda)$

Definition The **multiplicity** of eigenvalue λ_0 is the number of times $(\lambda - \lambda_0)$ appears as a factor in $P_T(\lambda)$.

Note If eigenvalue λ has multiplicity m, the above theorem says

$$1 \leq dim(E_{\lambda}) \leq m$$

if m = 1, then $dim(E_{\lambda}) = 1$.

Theorem If $\lambda_1, \ldots, \lambda_k$ are distinct eigenvalues of $T: V \to V$ and $\alpha = \{\vec{x_1}, \ldots, \vec{x_k}\}$ are corresponding eigenvectors, then the set α is linearly independent.

Proof.

Exercise

(*)**Theorem** Let $\lambda_1, \ldots, \lambda_k$ be distinct eigenvalues of T, suppose the characteristic polynomial is in form

$$P_T(\lambda) = \prod_{i=1}^k (\lambda - \lambda_i)^{m_i}$$

the T is diagonalizable if and only if

$$dim(E_{\lambda_i}) = m_i, \ \forall i$$

Note Also, $\sum_{i=1}^{k} m_i = dim(V) = n$

Proof.

Take basis for E_{λ_i} , as $\{v^{\vec{i}}_1, \dots, v^{\vec{i}}_{m_i}\}$

Claim: the union of bases of $E_{\lambda_i} \, \forall i$ gives a basis consisting of eigenvectors of T.

Note
$$|\bigcup_{i=1}^{k} \{\vec{v_{i_1}}, \dots, \vec{v_{m_i}}\}| = \sum_{i=1}^{k} m_i = dim(V)$$

All we need to show is linear independence.

Consider
$$\sum_{i=1}^{k} \sum_{j=1}^{m_i} c_{ij} \vec{v_j^i} = \vec{0}(\star)$$
Consider
$$\sum_{j=1}^{m_i} c_{ij} \vec{v_j^i} \in E_{\lambda_i} = \vec{x_i}$$

So
$$(\star)$$
 becomes $\sum_{i=1}^{k} \vec{x_i} = \vec{0}$ where $\vec{x_i} \in E_{\lambda_i}$, $\forall i$

Since $\vec{x_i}$ is eigenvectors for T, corresponding to different eigenvalues,

Therefore, $\{\vec{x_{i1}}, \dots, \vec{x_{ik}}\}$ is linearly independent

So
$$\vec{x_i} = \vec{0} \ \forall i$$

That's $\sum_{j=1}^{m_i} c_{ij} \vec{v_j^i} = \vec{x} = \vec{0} \ \forall i$
 $\implies c_{ij} = 0 \ \forall i, j$

Therefore linearly independent, so exists basis for V consisting of eigenvectors, Therefore T is diagonalizable.

 \rightarrow

Suppose T is diagonalizable,

Since T is diagonalizable, then exists basis for V consisting of eigenvectors, say α

Consider
$$[T]^{\alpha}_{\alpha} = \begin{bmatrix} \lambda_1 & 0 & \dots & 0 \\ 0 & \lambda_1 & \dots & 0 \\ 0 & \dots & \lambda_2 & \dots & 0 \\ \dots & \dots & \dots & \dots \end{bmatrix}$$

Where λ_1 takes first m_1 rows, λ_2 takes the next m_2 rows, etc.

$$P_T(\lambda) = \det([T]_{\alpha}^{\alpha} - \lambda I)$$

$$= \prod_{i=1}^{k} (\lambda_i - \lambda)^{m_i}$$
Since $1 \le \dim(E_{\lambda_i}) \le m_i \ \forall i$

$$\implies \dim(E_{\lambda_i}) = m_i \ \forall i$$

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15.1 Vector space over a field

Definition A vector space over field F is a set V together with two operations, addition and scalar multiplication s.t. [Very similar to those those defining properties for real vector space.]

15.2 Complex vector space

Complex vector space $\mathbb{C}^n = \{(z_1, \dots, z_n) | z_1, \dots, z_n \in \mathbb{C}\}$ is a vector space over \mathbb{C} , with dimension n and standard basis $\{\vec{e_1}, \dots, \vec{e_n}\}$

Let F be a field, then

$$F^n = \{(x_1, \dots, x_n) | x_1, \dots, x_n \in F\}$$

and

$$dim(F^n) = n$$

 F^n is a vector space over field F w.r.t. usual coordinate wise addition and scalar multiplication.

Definition Let V vector space over field F, then $\{\vec{x_1}, \ldots, \vec{x_n}\}$ is **linearly independent** if and only if

$$\sum_{i=1}^{i} c_i \vec{x_i} = \vec{0}, \ c_1, \dots, c_n \in F \implies c_1 = \dots = c_2 = 0 \in F$$

Definition span $\{\vec{x_1}, \dots, \vec{x_n}\}$ is defined as

$$\{\sum_{i=1}^{n} c_i \vec{x_i} | c_1, \dots, c_n \in F\}$$

Definition Consider V, W as two vector spaces over fields F then transformation $T: V \to F$ is **linear** if and only if

$$\forall \vec{v_1}, \vec{v_2} \in V, c, d \in F, T(c\vec{v_1} + d\vec{v_2}) = cT(\vec{v_1}) + dT(\vec{v_2})$$

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Theorem Let $T: V \to V$ be a linear operator, and β is a basis for vector space V. Let W_i be the span of first i vectors in β , then $[T]^{\beta}_{\beta}$ is upper-triangular if and only if

$$T(W_i) \subset W_i, \ \forall i$$

Definition Let $T: V \to V$ be a linear operator, a subspace W of V is called **invariant** under T (T-invariant) if $T(W) \subset W$.

Examples For linear operator T, 17

- 1. (Trivial invariant subspaces) V, $\{\vec{0}\}$
- 2. Ker(T)
- 3. Im(T)
- 4. $E_{\lambda}(T)$ for any eigenvalue λ of T
- 5. $T: \mathbb{R}^3 \to \mathbb{R}^3$ defined as

$$T((x, y, z)) = (3x + 2y, y - z, 4x + 2y - z)$$

Then subspace of \mathbb{R}^3 : $W = \{(x, y, x) \mid x, y \in \mathbb{R}\}$ is T-invariant.

Theorem Let $T: V \to V$ be a linear operator, $\beta = \{\vec{x_1}, \dots, \vec{x_k}\}$ is a basis for V, then $[T]^{\beta}_{\beta}$ is upper-triangular if and only if W_i is T-invariant for all $i \leq k$.

Note
$$\{\vec{0}\} \subset W_1 \subset W_2 \subset W_3 \cdots \subset W_k = V$$

Definition Linear operator $T:V\to V$ is said to be **triangularizable** if there exists a basis β for V such that $[T]^{\beta}_{\beta}$ is upper-triangular.

¹⁷Proofs are omitted.

Note If $[T]^{\beta}_{\beta}$ is upper-triangular, the characteristic polynomial $P_T(\lambda) = (\lambda - \lambda_1)(\lambda - \lambda_2) \dots (\lambda - \lambda_n)$ where λ_i are entries on the main diagonal.

Problem Entries above main diagonal are **not** uniquely determined by T, it's also depends on the choice of basis β .