

§5.7 Applications to Differential Equations

System of Linear Differential Equations

$$\begin{cases} x'_1(t) = a_{11}x_1(t) + \dots + a_{1n}x_n(t) \\ x'_2(t) = a_{21}x_1(t) + \dots + a_{2n}x_n(t) \\ \vdots \\ x'_n(t) = a_{n1}x_1(t) + \dots + a_{nn}x_n(t) \end{cases}$$

Let $\vec{x}(t) = \begin{pmatrix} x_1(t) \\ x_2(t) \\ \vdots \\ x_n(t) \end{pmatrix}$, $A = \begin{pmatrix} a_{11} & \dots & a_{1n} \\ \vdots & & \vdots \\ a_{n1} & \dots & a_{nn} \end{pmatrix}$

Then $\vec{x}'(t) = A\vec{x}(t)$

* When A is a diagonal matrix

Example: $\begin{pmatrix} x'_1(t) \\ x'_2(t) \end{pmatrix} = \begin{pmatrix} 3 & 0 \\ 0 & -5 \end{pmatrix} \begin{pmatrix} x_1(t) \\ x_2(t) \end{pmatrix}$ decoupled

That is $\begin{cases} x'_1(t) = 3x_1(t) \\ x'_2(t) = -5x_2(t) \end{cases}$

Solution: $\begin{pmatrix} x_1(t) \\ x_2(t) \end{pmatrix} = \begin{pmatrix} C_1 e^{3t} \\ C_2 e^{-5t} \end{pmatrix} = C_1 \begin{pmatrix} 1 \\ 0 \end{pmatrix} e^{3t} + C_2 \begin{pmatrix} 0 \\ 1 \end{pmatrix} e^{-5t}$

* For general equation $\vec{x}' = A\vec{x}$

The solution is $\vec{x}(t) = C_1 \vec{v}_1 e^{\lambda_1 t} + C_2 \vec{v}_2 e^{\lambda_2 t} + \dots + C_n \vec{v}_n e^{\lambda_n t}$

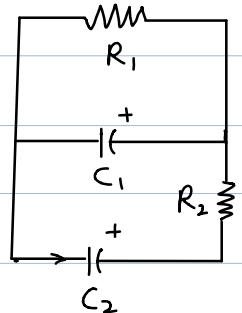
where $\lambda_1, \lambda_2, \dots, \lambda_n$ are eigenvalues of A

and $\vec{v}_1, \vec{v}_2, \dots, \vec{v}_n$ are eigenfunctions of A .

$\vec{v}_i e^{\lambda_i t}$ are called eigenfunction of the differential equation.

Example: The circuit in Figure 1 can be described by the differential equation

$$\begin{pmatrix} x_1'(t) \\ x_2'(t) \end{pmatrix} = \begin{pmatrix} -(1/R_1 + 1/R_2)/C_1 & 1/(R_2 C_2) \\ 1/(R_2 C_2) & -1/(R_1 C_1) \end{pmatrix} \begin{pmatrix} x_1(t) \\ x_2(t) \end{pmatrix}$$



Suppose $R_1 = 1$ ohm, $R_2 = 2$ ohm,
 $C_1 = 1$ farad, $C_2 = 0.5$ farad

$$x_1(0) = 5 \text{ volts}, \quad x_2(0) = 4 \text{ volts}$$

Figure 1

Find formulas for $x_1(t)$ and $x_2(t)$ that describe how the voltages change over time.

Solution: Let $A = \begin{pmatrix} -(1/R_1 + 1/R_2)/C_1 & 1/(R_2 C_2) \\ 1/(R_2 C_2) & -1/(R_1 C_1) \end{pmatrix} = \begin{pmatrix} -1.5 & 0.5 \\ 1 & -1 \end{pmatrix}$.

$$\vec{x}(0) = \begin{pmatrix} 5 \\ 4 \end{pmatrix}.$$

The eigenvalues of A are $\lambda_1 = -0.5$, $\lambda_2 = -2$
with corresponding eigenvectors

$$\vec{v}_1 = \begin{pmatrix} 1 \\ 2 \end{pmatrix} \text{ and } \vec{v}_2 = \begin{pmatrix} -1 \\ 1 \end{pmatrix}$$

The eigenfunctions $\vec{x}^{(1)}(t) = \vec{v}_1 e^{\lambda_1 t}$, $\vec{x}^{(2)}(t) = \vec{v}_2 e^{\lambda_2 t}$
both satisfy $\vec{x}' = A\vec{x}$.

$$\vec{x}(t) = C_1 \vec{x}^{(1)}(t) + C_2 \vec{x}^{(2)}(t)$$

$$= C_1 \begin{pmatrix} 1 \\ 2 \end{pmatrix} e^{-0.5t} + C_2 \begin{pmatrix} -1 \\ 1 \end{pmatrix} e^{-2t}$$

Let $t=0$, $\begin{pmatrix} 5 \\ 4 \end{pmatrix} = C_1 \begin{pmatrix} 1 \\ 2 \end{pmatrix} + C_2 \begin{pmatrix} -1 \\ 1 \end{pmatrix}$

We get $c_1 = 3$, $c_2 = -2$.

Hence $\vec{x}(t) = 3 \begin{pmatrix} 1 \\ 2 \end{pmatrix} e^{-0.5t} - 2 \begin{pmatrix} -1 \\ 1 \end{pmatrix} e^{-2t}$

Example: Suppose a particle is moving in a planar force field and its position vector \vec{x} satisfies $\vec{x}' = A\vec{x}$ and $\vec{x}(0) = \vec{x}_0$, where

$$A = \begin{pmatrix} 4 & -5 \\ -2 & 1 \end{pmatrix}, \quad \vec{x}_0 = \begin{pmatrix} 2.9 \\ 2.6 \end{pmatrix}.$$

Solution: The eigenvalues of A is $\lambda_1 = 6$ and $\lambda_2 = -1$ with corresponding eigenvectors $\vec{v}_1 = \begin{pmatrix} -5 \\ 2 \end{pmatrix}$ and $\vec{v}_2 = \begin{pmatrix} 1 \\ 1 \end{pmatrix}$.

For any constants c_1 and c_2 , the function

$$\vec{x}(t) = c_1 \vec{v}_1 e^{\lambda_1 t} + c_2 \vec{v}_2 e^{\lambda_2 t}$$

$$= c_1 \begin{pmatrix} -5 \\ 2 \end{pmatrix} e^{6t} + c_2 \begin{pmatrix} 1 \\ 1 \end{pmatrix} e^{-t} \quad \dots (*)$$

is a solution of $\vec{x}' = A\vec{x}$.

Since $\vec{x}(0) = \begin{pmatrix} 2.9 \\ 2.6 \end{pmatrix}$, we let $t=0$ in $(*)$ and

get $\begin{pmatrix} 2.9 \\ 2.6 \end{pmatrix} = c_1 \begin{pmatrix} -5 \\ 2 \end{pmatrix} + c_2 \begin{pmatrix} 1 \\ 1 \end{pmatrix}$.

$$c_1 = -\frac{3}{70}, \quad c_2 = \frac{188}{70}.$$

Hence $\vec{x}(t) = -\frac{3}{70} \begin{pmatrix} -5 \\ 2 \end{pmatrix} e^{6t} + \frac{188}{70} \begin{pmatrix} 1 \\ 1 \end{pmatrix} e^{-t}$.

* Decoupling a Dynamical System

Consider a dynamical system $\vec{x}'(t) = A\vec{x}(t)$,

If A is $n \times n$ and has n linearly independent eigenvectors, then A is diagonalizable.

Suppose $\lambda_1, \dots, \lambda_n$ are eigenvalues with corresponding eigenvectors $\vec{v}_1, \dots, \vec{v}_n$.

Let $P = [\vec{v}_1, \dots, \vec{v}_n]$

$$D = \begin{pmatrix} \lambda_1 & & & \\ & \lambda_2 & & \\ & & \ddots & \\ & & & \lambda_n \end{pmatrix}$$

Define $\vec{y}(t) = P^{-1}\vec{x}(t)$.

Then $x(t) = P\vec{y}(t)$.

$$\frac{d}{dt}P\vec{y}(t) = A(P\vec{y}) = (PDP^{-1})P\vec{y}(t) = P D \vec{y}(t)$$

Left-multiply both sides by P^{-1} , we have

$$y'(t) = Dy(t).$$

or

$$\begin{pmatrix} y_1'(t) \\ y_2'(t) \\ \vdots \\ y_n'(t) \end{pmatrix} = \begin{pmatrix} \lambda_1 & 0 & \cdots & 0 \\ 0 & \lambda_2 & \cdots & 0 \\ \ddots & \ddots & \ddots & \ddots \\ 0 & 0 & \cdots & \lambda_n \end{pmatrix} \begin{pmatrix} y_1(t) \\ y_2(t) \\ \vdots \\ y_n(t) \end{pmatrix}$$

We know $y_1(t) = c_1 e^{\lambda_1 t}, \dots, y_n(t) = c_n e^{\lambda_n t}$.

$$y(t) = \begin{pmatrix} c_1 e^{\lambda_1 t} \\ \vdots \\ c_n e^{\lambda_n t} \end{pmatrix}$$

$$\text{where } \begin{pmatrix} c_1 \\ \vdots \\ c_n \end{pmatrix} = \vec{y}(0) = P^{-1}\vec{x}(0) = P^{-1}\vec{x}_0$$

$$x(t) = P y(t) = [\vec{v}_1, \dots, \vec{v}_n] \vec{y}(t)$$

$$= c_1 \vec{v}_1 e^{\lambda_1 t} + \dots + c_n \vec{v}_n e^{\lambda_n t}$$

This is consistent with the eigenfunction expansion discussed earlier.

§6.1 Inner Product, Length and Orthogonality

$\mathbb{R}^2, \mathbb{R}^3$: length, distance, perpendicularity

* The Inner Product

$$\vec{u} = \begin{pmatrix} u_1 \\ u_2 \\ \vdots \\ u_n \end{pmatrix}, \quad \vec{v} = \begin{pmatrix} v_1 \\ v_2 \\ \vdots \\ v_n \end{pmatrix}$$

$$\vec{u}^\top \vec{v} = [u_1, u_2, \dots, u_n] \begin{pmatrix} v_1 \\ v_2 \\ \vdots \\ v_n \end{pmatrix} = u_1 v_1 + u_2 v_2 + \dots + u_n v_n = \vec{u} \cdot \vec{v}$$

The number $\vec{u}^\top \vec{v}$ ($\vec{u} \cdot \vec{v}$) is called the inner product (or dot product) of \vec{u} and \vec{v} .

Example: Compute $\vec{u} \cdot \vec{v}$ and $\vec{v} \cdot \vec{u}$ for $\vec{u} = \begin{pmatrix} 2 \\ -5 \\ -1 \end{pmatrix}$ and

$$\vec{v} = \begin{pmatrix} 3 \\ 2 \\ -3 \end{pmatrix}.$$

Solution: $\vec{u} \cdot \vec{v} = \vec{u}^\top \vec{v} = [2 \ -5 \ -1] \begin{pmatrix} 3 \\ 2 \\ -3 \end{pmatrix}$

$$= 2 \cdot 3 + (-5) \cdot 2 + (-1) \cdot (-3)$$

$$= -1$$

$$\vec{v} \cdot \vec{u} = \vec{v}^T \vec{u} = [3 \ 2 \ -3] \begin{pmatrix} 2 \\ -5 \\ -1 \end{pmatrix}$$

$$= 3 \cdot 2 + 2 \cdot (-5) + (-3) \cdot (-1)$$

$$= -1$$

Theorem: Let \vec{u}, \vec{v} and \vec{w} be vectors in \mathbb{R}^n , and let c be a scalar. Then

a) $\vec{u} \cdot \vec{v} = \vec{v} \cdot \vec{u}$

b) $(\vec{u} + \vec{v}) \cdot \vec{w} = \vec{u} \cdot \vec{w} + \vec{v} \cdot \vec{w}$

c) $(c\vec{u}) \cdot \vec{v} = c(\vec{u} \cdot \vec{v}) = \vec{u} \cdot (c\vec{v})$

d) $\vec{u} \cdot \vec{u} \geq 0$ and $\vec{u} \cdot \vec{u} = 0$ iff $\vec{u} = \vec{0}$.

Based on b) and c), we have

$$(c_1 \vec{u}_1 + \dots + c_p \vec{u}_p) \cdot \vec{w} = c_1 (\vec{u}_1 \cdot \vec{w}) + \dots + c_p (\vec{u}_p \cdot \vec{w})$$

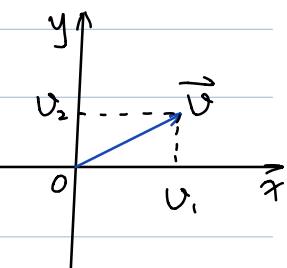
* The length of a vector

Def: The length (or norm) of \vec{v} is the nonnegative scalar $\|\vec{v}\|$ defined by

$$\|\vec{v}\| = \sqrt{\vec{v} \cdot \vec{v}} = \sqrt{v_1^2 + v_2^2 + \dots + v_n^2}.$$

For any scalar c ,

$$\|c\vec{v}\| = |c| \|\vec{v}\|$$



$$\|\vec{v}\| = \sqrt{v_1^2 + v_2^2}$$

Proof: $\|c\vec{v}\|^2 = c\vec{v} \cdot (c\vec{v}) = c^2 \vec{v} \cdot \vec{v} = c^2 \|\vec{v}\|^2$

$$\text{So } \|c\vec{v}\| = \sqrt{c^2 \|\vec{v}\|^2} = |c| \|\vec{v}\|$$

Unit vector: \vec{u} with $\|\vec{u}\| = 1$

For a vector $\vec{v} \neq \vec{0}$, $\vec{u} = \frac{\vec{v}}{\|\vec{v}\|}$ is a unit vector parallel to \vec{v} . This process is called normalizing \vec{v} and we say \vec{u} is in the same directions as \vec{v} .

Example: Let $\vec{v} = (1, -2, 2, 0)$. Find a unit vector \vec{u} in the same direction as \vec{v} .

Solution: $\|\vec{v}\| = \sqrt{\vec{v} \cdot \vec{v}} = \sqrt{1^2 + (-2)^2 + 2^2 + 0^2} = \sqrt{9} = 3$

$$\vec{u} = \frac{\vec{v}}{\|\vec{v}\|} = \frac{(1, -2, 2, 0)}{3} = \left(\frac{1}{3}, -\frac{2}{3}, \frac{2}{3}, 0\right)$$

Example: Let W be the subspace of \mathbb{R}^2 spanned by $\vec{x} = \left(\frac{2}{3}, 1\right)$. Find a unit vector \vec{z} that is a basis for W .

Solution: $W = \text{span}\{\vec{x}\} = \{c\vec{x} \mid c \in \mathbb{R}\}$

$$\vec{z} = \frac{\vec{x}}{\|\vec{x}\|} = \frac{\left(\frac{2}{3}, 1\right)}{\sqrt{\left(\frac{2}{3}\right)^2 + 1^2}} = \frac{\left(\frac{2}{3}, 1\right)}{\sqrt{\frac{2^2 + 3^2}{3^2}}} = \frac{3 \times \left(\frac{2}{3}, 1\right)}{\sqrt{13}}$$

$$= \left(\frac{2}{\sqrt{13}}, \frac{3}{\sqrt{13}}\right)$$

* Distance in \mathbb{R}^n

Def: For \vec{u} and \vec{v} in \mathbb{R}^n , the distance between \vec{u} and \vec{v} , written as $\text{dist}(\vec{u}, \vec{v})$, is the length of the vector $\vec{u} - \vec{v}$. That is,

$$\text{dist}(\vec{u}, \vec{v}) = \|\vec{u} - \vec{v}\|$$

Example: Compute the distance between the vectors $\vec{u} = (7, 1)^T$ and $\vec{v} = (3, 2)^T$.

Solution: $\vec{u} - \vec{v} = \begin{pmatrix} 7 \\ 1 \end{pmatrix} - \begin{pmatrix} 3 \\ 2 \end{pmatrix} = \begin{pmatrix} 4 \\ -1 \end{pmatrix}$

$$\|\vec{u} - \vec{v}\| = \sqrt{4^2 + (-1)^2} = \sqrt{17}$$

Example: If $\vec{u} = (u_1, u_2, u_3)$ and $\vec{v} = (v_1, v_2, v_3)$, then
 $\text{dist}(\vec{u}, \vec{v}) = \|\vec{u} - \vec{v}\| = \sqrt{(\vec{u} - \vec{v}) \cdot (\vec{u} - \vec{v})}$
 $= \sqrt{(u_1 - v_1)^2 + (u_2 - v_2)^2 + (u_3 - v_3)^2}$

* Orthogonal Vectors (perpendicular)

Def: Two vectors \vec{u} and \vec{v} in \mathbb{R}^n are orthogonal (to each other) if $\vec{u} \cdot \vec{v} = 0$.

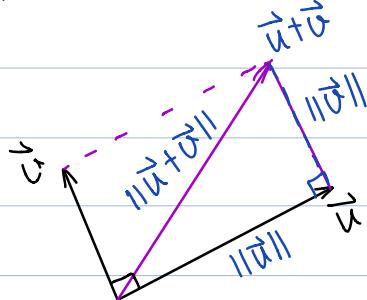
Remark: $\vec{0}$ is orthogonal to every vector in \mathbb{R}^n because $\vec{0}^T \vec{v} = 0$.

|Pythagorean|

Thm: The Pythagorean Theorem

Two vectors \vec{u} and \vec{v} are orthogonal iff

$$\|\vec{u} + \vec{v}\|^2 = \|\vec{u}\|^2 + \|\vec{v}\|^2$$



$$\text{Proof: } \|\vec{u} + \vec{v}\|^2 = (\vec{u} + \vec{v}) \cdot (\vec{u} + \vec{v})$$

$$= \vec{u} \cdot \vec{u} + \vec{v} \cdot \vec{u} + \vec{u} \cdot \vec{v} + \vec{v} \cdot \vec{v}$$

$$= \|\vec{u}\|^2 + \|\vec{v}\|^2 + 2\vec{u} \cdot \vec{v}$$

$$= \|\vec{u}\|^2 + \|\vec{v}\|^2 + 2\vec{u} \cdot \vec{v}$$

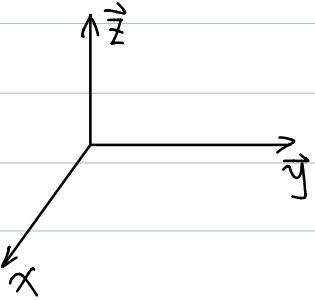
$$\text{Thus } \|\vec{u} + \vec{v}\|^2 = \|\vec{u}\|^2 + \|\vec{v}\|^2 \text{ iff } \vec{u} \cdot \vec{v} = 0$$

* Orthogonal Complements

Def: Let $\mathbb{Z} \in \mathbb{R}^n$, W a subspace of \mathbb{R}^n .

If \vec{z} is orthogonal to every vector in W , we say
 \vec{z} is orthogonal to W .

Ex:



$$W = xy\text{-plane} = \{(x, y, 0)\}$$

$$\vec{z} = (0, 0, 2)$$

Then \vec{z} is orthogonal to W .

Def: The set of all vectors \vec{z} that are orthogonal to W
is called the orthogonal complement of W , and is
denoted by W^\perp ("W perp")

Ex: $W = xy\text{-plane in } \mathbb{R}^3$

$$W^\perp = z\text{-axis}$$

$V = x_1, x_2\text{-plane in } \mathbb{R}^4$,

$$= \{(x_1, x_2, 0, 0) \mid x_1, x_2 \in \mathbb{R}\}$$

$V^\perp = \{x_3, x_4\} - \text{plane in } \mathbb{R}^4$

$$= \{(0, 0, x_3, x_4) \mid x_3, x_4 \in \mathbb{R}\}$$

Fact: 1) A vector \vec{x} is in W^\perp iff \vec{x} is orthogonal to every vector in a set that spans W .

2) W^\perp is a subspace of \mathbb{R}^n .

Proof: 1) obvious

2) Take \vec{u}, \vec{v} in W^\perp and \vec{x} in W

$$\text{Then } (a\vec{u} + b\vec{v}) \cdot \vec{x} = a\vec{u} \cdot \vec{x} + b\vec{v} \cdot \vec{x} = 0$$

Thus $a\vec{u} + b\vec{v}$ is orthogonal to every vector in W . Hence $a\vec{u} + b\vec{v}$ is in W^\perp .

Theorem: Let A be an $m \times n$ matrix. The orthogonal complement of the row space of A is the null space of A , and the orthogonal complement of the column space of A is the null space of A^T :

$$(\text{Row } A)^\perp = \text{Nul } A$$

$$(\text{Col } A)^\perp = \text{Nul } A^T$$

Proof: $A = (\vec{a}_1, \dots, \vec{a}_n) = \begin{pmatrix} \vec{b}_1 \\ \vdots \\ \vec{b}_m \end{pmatrix}$

$$\text{Row } A = \text{Span } \{\vec{b}_1, \dots, \vec{b}_m\}$$

$$\vec{x} \in (\text{Row } A)^\perp \Leftrightarrow \vec{x} \cdot \vec{b}_1 = 0, \dots, \vec{x} \cdot \vec{b}_m = 0$$

$$\Leftrightarrow 0 = \begin{pmatrix} \vec{b}_1 \\ \vdots \\ \vec{b}_m \end{pmatrix} \cdot \vec{x} = \begin{pmatrix} \vec{b}_1 \cdot \vec{x} \\ \vdots \\ \vec{b}_m \cdot \vec{x} \end{pmatrix} \Leftrightarrow \vec{x} \in \text{Nul } A$$

Since $\text{col } A = \text{Row } A^T$,

$$(\text{col } A)^\perp = (\text{Row } A^T)^\perp = \text{Nul } A^T$$

Exercise: Let W be a subspace of \mathbb{R}^n . Then W^\perp is also a subspace of \mathbb{R}^n . Prove that

$$\dim W + \dim W^\perp = n$$

Proof: If $W \neq \{\vec{0}\}$. Let $\{\vec{b}_1, \dots, \vec{b}_p\}$ be a basis for W , where $1 \leq p \leq n$. Let A be the $p \times n$ matrix having rows $\vec{b}_1^T, \dots, \vec{b}_p^T$. Then $W = \text{Row } A$.

$$\text{Hence } W^\perp = (\text{Row } A)^\perp = \text{Nul } A.$$

$$\text{Therefore, } \dim W + \dim W^\perp$$

$$= \dim(\text{Row } A) + \dim(\text{Nul } A)$$

$$= \text{rank } A + \dim(\text{Nul } A)$$

$$= n$$

If $W = \{\vec{0}\}$, then $W^\perp = \mathbb{R}^n$. The result is obviously true.