

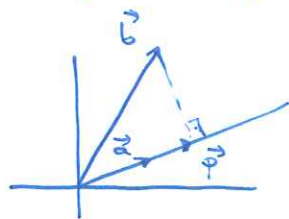
MATH 312
LECTURE 11

Orthogonal basis and
Gram-Schmidt.

(~ Sec. 4.4)

(From last days...)

- Projection of a vector $\vec{b}$ onto a line:



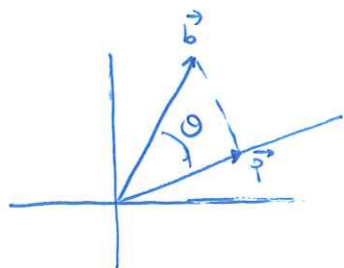
1) We choose $\vec{a}$ (basis for the line)

$$\Rightarrow \text{Then, } \vec{p} = \frac{\vec{a}\vec{a}^T}{\vec{a}^T\vec{a}} \vec{b} = \frac{\vec{a} \cdot \vec{b}}{\|\vec{a}\|^2} \vec{a}$$

Remark: We choose $\vec{a}$. So we can always choose it such that $\|\vec{a}\| = 1$.

In this case, $\vec{p} = (\vec{a} \cdot \vec{b}) \vec{a}$.

We could have obtain this formula with standard trigonometry:



$$\left. \begin{aligned} \cos \theta &= \frac{\|\vec{p}\|}{\|\vec{b}\|} \Rightarrow \|\vec{p}\| = \|\vec{b}\| \cos \theta \\ \vec{a} \cdot \vec{b} &= \|\vec{a}\| \|\vec{b}\| \cos \theta = \|\vec{b}\| \cos \theta \end{aligned} \right\} \Rightarrow$$

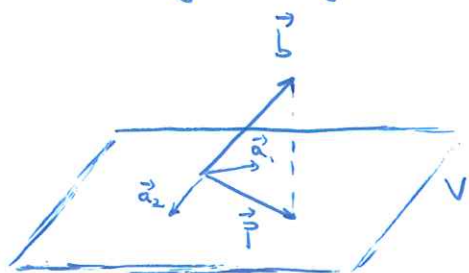
$\uparrow$
 $\|\vec{a}\| = 1$

$$\Rightarrow \|\vec{p}\| = \vec{a} \cdot \vec{b}$$

So the projection vector is $\vec{p} = \|\vec{p}\| \vec{a} = (\vec{a} \cdot \vec{b}) \vec{a}$.

$\uparrow$ again, because $\|\vec{a}\| = 1$.

- Projection of a vector $\vec{b}$ onto a subspace V :



1) Choose a basis of V : e.g. $\{\vec{a}_1, \vec{a}_2\}$ for plane.

2) Create $A = [\vec{a}_1 | \dots | \vec{a}_n]$

3) $\vec{p} = P\vec{b}$ with $P = A(A^T A)^{-1} A^T$.

Remark: We choose the basis $\{\vec{a}_1, \dots, \vec{a}_n\}$. And note that

$$A^T A = \begin{bmatrix} \vec{a}_1^T \\ \vdots \\ \vec{a}_n^T \end{bmatrix} \begin{bmatrix} \vec{a}_1 & \dots & \vec{a}_n \end{bmatrix} = \begin{bmatrix} \vec{a}_1^T \vec{a}_1 & \vec{a}_1^T \vec{a}_2 & \dots & \vec{a}_1^T \vec{a}_n \\ \vec{a}_2^T \vec{a}_1 & \vec{a}_2^T \vec{a}_2 & \dots & \vec{a}_2^T \vec{a}_n \\ \vdots & \vdots & \ddots & \vdots \\ \vec{a}_n^T \vec{a}_1 & \vec{a}_n^T \vec{a}_2 & \dots & \vec{a}_n^T \vec{a}_n \end{bmatrix}$$

In the case of a plane,

$$A^T A = \begin{bmatrix} \vec{a}_1^T \\ \vec{a}_2^T \end{bmatrix} \begin{bmatrix} \vec{a}_1 & \vec{a}_2 \end{bmatrix} = \begin{bmatrix} \vec{a}_1^T \vec{a}_1 & \vec{a}_1^T \vec{a}_2 \\ \vec{a}_2^T \vec{a}_1 & \vec{a}_2^T \vec{a}_2 \end{bmatrix}$$

So if we choose $\vec{a}_1$ and $\vec{a}_2$ with size one and perpendicular to each other, that is, $\vec{a}_1 \cdot \vec{a}_2 = 0$, $\|\vec{a}_1\| = \|\vec{a}_2\| = 1$, then

$$A^T A = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}.$$

• Def: The vectors $\vec{g}_1, \dots, \vec{g}_n$ are orthonormal if

$$\vec{g}_i^T \vec{g}_j = \begin{cases} 0 & i \neq j \text{ (orthogonal)} \\ 1 & i = j \text{ (unit vectors, i.e., } \|\vec{g}_i\| = 1) \end{cases}$$

• Remark: Q is a matrix with orthonormal columns $\Leftrightarrow Q^T Q = I$

$$\hookrightarrow Q = [\vec{g}_1 | \dots | \vec{g}_n]$$

$$Q^T Q = \begin{bmatrix} \vec{g}_1^T \vec{g}_1 & \vec{g}_1^T \vec{g}_2 & \dots & \vec{g}_1^T \vec{g}_n \\ \vdots & \vdots & \ddots & \vdots \\ \vec{g}_n^T \vec{g}_1 & \vec{g}_n^T \vec{g}_2 & \dots & \vec{g}_n^T \vec{g}_n \end{bmatrix} = \begin{bmatrix} 1 & 0 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & 1 \end{bmatrix}$$

• Remark: If Q is in addition square, then $Q^{-1} = Q^T$

$\hookrightarrow$ In this case, the matrix is called orthogonal.

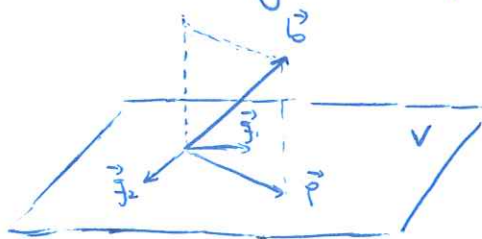
• Remark: For Q rectangular, $Q Q^T$ is not the identity:

Ex:

$$Q = \begin{bmatrix} 1 & 0 \\ 0 & 0 \\ 0 & 1 \end{bmatrix} \rightarrow Q^T Q = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & 0 \\ 0 & 1 \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \text{ (he knew it)}$$

$$Q Q^T = \begin{bmatrix} 1 & 0 \\ 0 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

• Let's go back: projection using orthonormal basis:



1) We chose an orthonormal basis of V : $\{\vec{g}_1, \dots, \vec{g}_n\}$

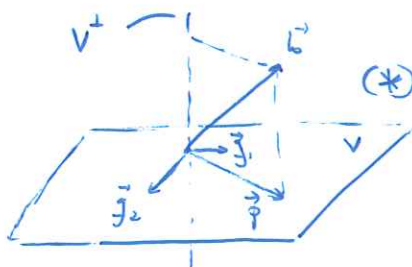
2) Create $Q = [\vec{g}_1 | \dots | \vec{g}_n]$

3) Then, $\vec{p} = P\vec{b}$, with $P = Q(Q^T Q)^{-1} Q^T = Q Q^T$

(This is not the identity in general)

• Particular case of the plane:

$$\vec{p} = \alpha_1 \vec{g}_1 + \alpha_2 \vec{g}_2, \quad \alpha_1, \alpha_2?$$



$$\begin{aligned} \rightarrow \vec{g}_1 \cdot \vec{p} &= \alpha_1 \vec{g}_1 \cdot \vec{g}_1 + \alpha_2 \cancel{\vec{g}_1 \cdot \vec{g}_2} = \alpha_1 \rightarrow \alpha_1 = \vec{g}_1 \cdot \vec{p} \\ \rightarrow \vec{g}_2 \cdot \vec{p} &= \alpha_1 \cancel{\vec{g}_2 \cdot \vec{g}_1} + \alpha_2 \vec{g}_2 \cdot \vec{g}_2 = \alpha_2 \rightarrow \alpha_2 = \vec{g}_2 \cdot \vec{p} \end{aligned} \quad \left. \vphantom{\begin{aligned} \rightarrow \vec{g}_1 \cdot \vec{p} &= \alpha_1 \vec{g}_1 \cdot \vec{g}_1 + \alpha_2 \cancel{\vec{g}_1 \cdot \vec{g}_2} = \alpha_1 \rightarrow \alpha_1 = \vec{g}_1 \cdot \vec{p} \\ \rightarrow \vec{g}_2 \cdot \vec{p} &= \alpha_1 \cancel{\vec{g}_2 \cdot \vec{g}_1} + \alpha_2 \vec{g}_2 \cdot \vec{g}_2 = \alpha_2 \rightarrow \alpha_2 = \vec{g}_2 \cdot \vec{p} \end{aligned}} \right\} \rightarrow$$

$$\Rightarrow \vec{p} = (\vec{g}_1 \cdot \vec{p}) \vec{g}_1 + (\vec{g}_2 \cdot \vec{p}) \vec{g}_2$$

That is, in an orthonormal basis, the projection into a subspace is the sum of the projections in each direction.

$$\left(\begin{aligned} \leadsto \text{Notice in (*) that } \vec{b} &= \text{projection of } \vec{b} \text{ onto } V + \text{projection of } \vec{b} \text{ onto } V^\perp \\ \hookrightarrow \text{so } \vec{p} &= \vec{b} - \text{projection of } \vec{b} \text{ onto } V^\perp \end{aligned} \right)$$

$$V = \{\vec{v}_1, \vec{v}_2\}$$

It is easy to see that

It is easy to see that

$$\vec{b} = -2\vec{u}_1 - 2\vec{u}_2 \rightarrow \vec{b}|_u = (-2, -2) \quad (" " " " " u)$$

$$\vec{b} = -2\vec{u}_1 - 2\vec{u}_2 \rightarrow \vec{b}|_u = (-2, -2) \quad (" " " " " u)$$

$$\vec{b} = \sqrt{8} \hat{z}_1 \rightarrow \vec{b}|_V = (\sqrt{8}, 0) \quad (u, -v)$$

Note that:

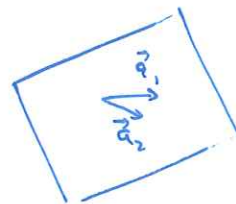
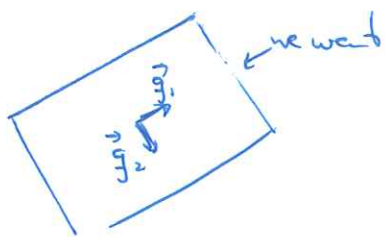
$$\vec{b} = (\vec{b} \cdot \vec{e}_1)\vec{e}_1 + (\vec{b} \cdot \vec{e}_2)\vec{e}_2, \quad \vec{b} = (\vec{b} \cdot \vec{u}_1)\vec{u}_1 + (\vec{b} \cdot \vec{u}_2)\vec{u}_2$$

$$\vec{b} = (\vec{b} \cdot \vec{e}_1) \vec{e}_1 + (\vec{b} \cdot \vec{e}_2) \vec{e}_2$$

This is true in all orthonormal basis: a vector is the sum of its projections onto the basis vectors.

- In summary, orthonormal basis are great, but how to obtain them?

Ex: Let $x+y+z=0$ plane in $\mathbb{R}^3$. Find an orthonormal basis.



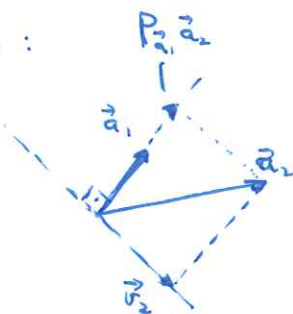
What we know: $\begin{bmatrix} 1 & 1 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \\ z \end{bmatrix} = 0 \sim \left\{ \begin{bmatrix} -1 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix} \right\}$
 $\vec{a}_1, \vec{a}_2$

But $\vec{a}_1 \cdot \vec{a}_2 \neq 0$

From $\{\vec{a}_1, \vec{a}_2\}$, we will obtain a new basis $\{\vec{g}_1, \vec{g}_2\}$ with the same span.

Gram-Schmidt Algorithm

Idea:



1) Take $\vec{v}_1 = \vec{a}_1$.

2) Project $\vec{a}_2$ onto $\vec{a}_1$ → we obtain the component of $\vec{a}_2$ parallel to $\vec{a}_1$:

$$P_{\vec{a}_1} \vec{a}_2 = \frac{\vec{a}_1^T \vec{a}_2}{\vec{a}_1^T \vec{a}_1} \vec{a}_1$$

3) Then $\vec{v}_2 = \vec{a}_2 - P_{\vec{a}_1} \vec{a}_2 = \vec{a}_2 - \frac{\vec{a}_1^T \vec{a}_2}{\vec{a}_1^T \vec{a}_1} \vec{a}_1$ is orthogonal to $\vec{v}_1$.

4) Divide by their length to make them orthonormal:

$$\vec{g}_1 = \frac{\vec{v}_1}{\|\vec{v}_1\|}, \quad \vec{g}_2 = \frac{\vec{v}_2}{\|\vec{v}_2\|}$$

• General: Given basis $\{\vec{a}_1, \dots, \vec{a}_k\}$,

1) Set $\vec{v}_1 = \vec{a}_1$,

2) Set $\vec{v}_2 = \vec{a}_2 - \left(\frac{\vec{v}_1^T \vec{a}_2}{\vec{v}_1^T \vec{v}_1} \right) \vec{v}_1$

3) Set $\vec{v}_3 = \vec{a}_3 - \left(\frac{\vec{v}_1^T \vec{a}_3}{\vec{v}_1^T \vec{v}_1} \right) \vec{v}_1 - \left(\frac{\vec{v}_2^T \vec{a}_3}{\vec{v}_2^T \vec{v}_2} \right) \vec{v}_2$

⋮

k) Set $\vec{v}_k = \vec{a}_k - \left(\frac{\vec{v}_1^T \vec{a}_k}{\vec{v}_1^T \vec{v}_1} \right) \vec{v}_1 - \dots - \left(\frac{\vec{v}_{k-1}^T \vec{a}_k}{\vec{v}_{k-1}^T \vec{v}_{k-1}} \right) \vec{v}_{k-1}$

To conclude, set $\vec{g}_1 = \frac{\vec{v}_1}{\|\vec{v}_1\|}, \dots, \vec{g}_k = \frac{\vec{v}_k}{\|\vec{v}_k\|}$

→ (Continuation of example)

$$\vec{a}_1 = \begin{bmatrix} -1 \\ 1 \\ 0 \end{bmatrix}, \quad \vec{a}_2 = \begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix} \rightarrow$$

$$1) \vec{v}_1 = \begin{bmatrix} -1 \\ 1 \\ 0 \end{bmatrix}$$

$$2) \vec{v}_2 = \begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix} - \frac{\begin{bmatrix} -1 & 1 & 0 \end{bmatrix} \begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix}}{\|\vec{v}_1\|^2} \begin{bmatrix} -1 \\ 1 \\ 0 \end{bmatrix} =$$

$$= \begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix} - \frac{1}{2} \begin{bmatrix} -1 \\ 1 \\ 0 \end{bmatrix} = \begin{bmatrix} -1/2 \\ -1/2 \\ 1 \end{bmatrix} \quad \left(\text{Check that } \vec{v}_1 \cdot \vec{v}_2 = 0 \right)$$

$$\text{And } \vec{g}_1 = \frac{\vec{v}_1}{\|\vec{v}_1\|}$$

$$\vec{g}_2 = \frac{\vec{v}_2}{\|\vec{v}_2\|}$$

MATH 312
 LECTURE 12

 : Linear Transformations (~ Sec. 8.1, 8.2)

I.] Introduction

Def: Let U, V be vector spaces. A function T from U to V ,

$$T: U \rightarrow V \\ \vec{u} \mapsto \vec{v} = T(\vec{u}) \left\{ \begin{array}{l} \text{is called a linear transformation if} \end{array} \right.$$

$$1) T(\vec{u}_1 + \vec{u}_2) = T(\vec{u}_1) + T(\vec{u}_2)$$

$$2) T(\lambda \vec{u}) = \lambda T(\vec{u}) \quad \text{for all } \vec{u}_1, \vec{u}_2 \in U, \lambda \text{ scalar.}$$

Remarks:

$$1) T(\vec{0}) = \vec{0}$$

$$2) T(\alpha \vec{u}_1 + \beta \vec{u}_2) = \alpha T(\vec{u}_1) + \beta T(\vec{u}_2)$$

Example: Which of these transformations are linear? $a, b \in \mathbb{R}$

$$a) T\left(\begin{bmatrix} a \\ b \end{bmatrix}\right) = \begin{bmatrix} a \\ a \end{bmatrix} \quad \checkmark$$

$$b) T\left(\begin{bmatrix} a \\ b \end{bmatrix}\right) = ab \quad \times$$

$$c) T\left(\begin{bmatrix} a \\ b \end{bmatrix}\right) = \begin{bmatrix} 0 \\ 1 \end{bmatrix} \quad \times$$

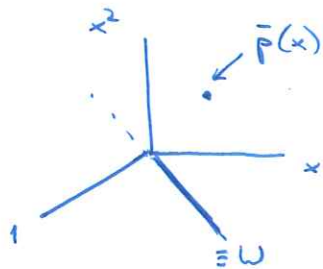
$$d) T\left(\begin{bmatrix} a \\ b \end{bmatrix}\right) = a - b \quad \checkmark$$

• Notice that in a) and d) T is given by a matrix:

$$a) T\left(\begin{bmatrix} a \\ b \end{bmatrix}\right) = \begin{bmatrix} 1 & 0 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} a \\ b \end{bmatrix} = \begin{bmatrix} a \\ a \end{bmatrix} \quad , \quad d) T\left(\begin{bmatrix} a \\ b \end{bmatrix}\right) = \begin{bmatrix} 1 & -1 \end{bmatrix} \begin{bmatrix} a \\ b \end{bmatrix} = a - b$$

• So now we can "draw" polynomials!

$$p(x) = a + bx + cx^2 \sim (a, b, c)$$



$$\bar{p}(x) = x + x^2 \sim (0, 1, 1)$$

V is the plane "1, x" (subspace of U).

$$W = \{ p \text{ pol} : p(x) = a + ax \}$$

Let's consider the derivative of polynomials in U :

$$\left. \begin{aligned} T: U &\rightarrow U \\ p(x) &\mapsto T(p(x)) = \frac{d}{dx} p(x) \end{aligned} \right\}$$

$$T(a + bx + cx^2) = b + 2cx$$

We can write this using matrices:

$$p(x) = a + bx + cx^2 \rightarrow p|_{B_U} = \begin{bmatrix} a \\ b \\ c \end{bmatrix}$$

$$T(p(x)) = b + 2cx \rightarrow T(p(x))|_{B_U} = \begin{bmatrix} b \\ 2c \\ 0 \end{bmatrix}$$

} that is

$$T\left(\begin{bmatrix} a \\ b \\ c \end{bmatrix}\right) = \begin{bmatrix} 0 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 2 \end{bmatrix} \begin{bmatrix} a \\ b \\ c \end{bmatrix} = \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 2 \\ 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} a \\ b \\ c \end{bmatrix} = \begin{bmatrix} b \\ 2c \\ 0 \end{bmatrix}$$

Let's check it: $p(x) = 4x^2 + 3$

$$\frac{d}{dx} p(x) = 8x \checkmark ; T\left(\begin{bmatrix} 3 \\ 0 \\ 4 \end{bmatrix}\right) = \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 2 \\ 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} 3 \\ 0 \\ 4 \end{bmatrix} = \begin{bmatrix} 0 \\ 8 \\ 0 \end{bmatrix} \rightarrow 8x //$$

Remark: The matrix that represents the transformation is different if we choose different basis for U

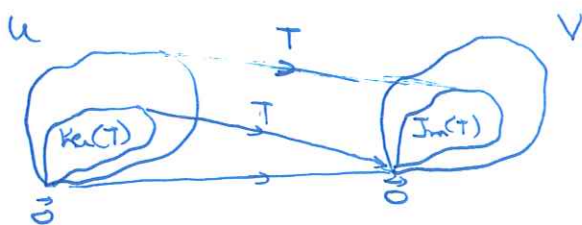
Def: Kernel of a linear transformation.

Let $T: U \rightarrow V$
 $\vec{u} \mapsto \vec{v} = T(\vec{u})$ } lin. transf. Its kernel is defined as

$$\text{Ker}(T) = \{ \vec{u} \in U : T(\vec{u}) = \vec{0} \}$$

Def: Range of a lin. transf. (or Image).

$$\text{Im}(T) = \{ \vec{v} \in V : \exists \vec{u} \in U \text{ with } T(\vec{u}) = \vec{v} \}$$



→ So if the linear transformation is given by a matrix, $T(\vec{x}) = A\vec{x}$,
then $\text{Ker}(T) \equiv N(A)$
 $\text{Im}(T) \equiv C(A)$.

- For the derivative, what is the kernel?

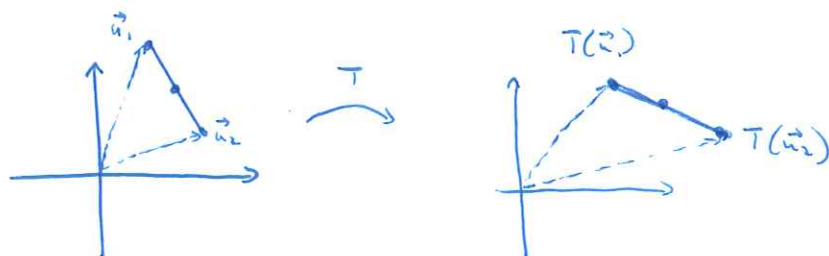
$\text{Ker}(T) = \text{polynomials of degree zero (constants)}$.

$$\hookrightarrow A = \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 2 \\ 0 & 0 & 0 \end{bmatrix} \xrightarrow{\text{basis of}} N(A) = \left\{ \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} \right\}$$

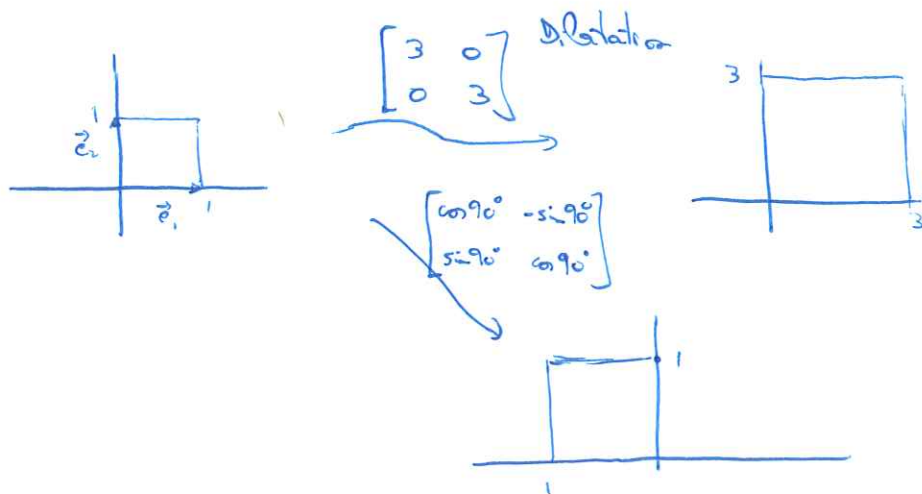
$\text{Im}(T) = \text{polynomials of degree one or less} \rightarrow V$

$$\hookrightarrow C(A) \rightarrow \left\{ \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} \right\}$$

Example: Image transformation.



The middle points goes to the middle point: $T\left(\frac{1}{2}(\vec{u}_1 + \vec{u}_2)\right) = \frac{1}{2}(T(\vec{u}_1) + T(\vec{u}_2))$.



You only need to see where $\vec{e}_1$ and $\vec{e}_2$ goes!

II] The matrix of a linear transformation.

$$\text{Let } T_A: \mathbb{R}^n \rightarrow \mathbb{R}^m \left\{ \begin{array}{l} \vec{x} \mapsto T(\vec{x}) = A\vec{x} \end{array} \right. \text{ with } A \text{ } m \times n \text{ matrix.}$$

For example,

$$A = \begin{bmatrix} 1 & 2 & 0 & 0 \\ 4 & 0 & 1 & 1 \\ 5 & 0 & 1 & 0 \end{bmatrix}, \text{ so } T_A: \mathbb{R}^4 \rightarrow \mathbb{R}^3$$

Choose the canonical basis for $\mathbb{R}^4$ and $\mathbb{R}^3$ (i.e., $\mathbb{R}^4 \sim \left\{ \underbrace{\begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \end{bmatrix}}_{\vec{e}_1}, \underbrace{\begin{bmatrix} 0 \\ 1 \\ 0 \\ 0 \end{bmatrix}}_{\vec{e}_2}, \dots \right\}$).

What is $T_A(\vec{e}_1)$?

$$T_A(\vec{e}_1) = \begin{bmatrix} 1 & 2 & 0 & 0 \\ 4 & 0 & 1 & 1 \\ 5 & 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \end{bmatrix} = \begin{bmatrix} 1 \\ 4 \\ 5 \end{bmatrix} \quad \downarrow \text{1st column of } A.$$

So,

$$A = \left[T(\vec{e}_1) \mid \dots \mid T(\vec{e}_4) \right]$$

Example. $a + b x + c x^2 \rightsquigarrow \begin{bmatrix} a \\ b \\ c \end{bmatrix}$ is basis $\{1, x, x^2\}$, $T \equiv$ deriv.

$$\left. \begin{array}{l} 1 \xrightarrow{T} 0 \rightsquigarrow \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} \\ x \xrightarrow{T} 1 \rightsquigarrow \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} \\ x^2 \xrightarrow{T} 2x \rightsquigarrow \begin{bmatrix} 0 \\ 2 \\ 0 \end{bmatrix} \end{array} \right\} \rightarrow A = \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 2 \\ 0 & 0 & 0 \end{bmatrix}.$$

Change of basis:

Let $T: U \rightarrow U$
 $u \mapsto T(u) = u$ } but choose a different basis of U
 for the input and output spaces.

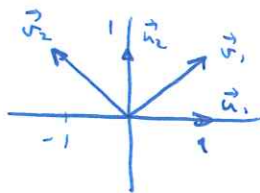
Input basis $B_i = \{\vec{u}_1, \dots, \vec{u}_n\}$ (same number of)
 Output basis $B_o = \{\vec{v}_1, \dots, \vec{v}_n\}$ vectors!

Assume we know

$$\left. \begin{aligned} \vec{u}_1 &= m_{11}\vec{v}_1 + \dots + m_{1n}\vec{v}_n \\ \vec{u}_2 &= m_{21}\vec{v}_1 + \dots + m_{2n}\vec{v}_n \\ &\vdots \\ \vec{u}_n &= m_{n1}\vec{v}_1 + \dots + m_{nn}\vec{v}_n \end{aligned} \right\} \rightarrow \text{Define } M = \begin{bmatrix} m_{11} & m_{12} & \dots & m_{1n} \\ m_{21} & & & \\ \vdots & & & \\ m_{n1} & & & m_{nn} \end{bmatrix} \begin{array}{l} \text{change of} \\ \text{basis} \\ \text{matrix} \end{array}$$

$\uparrow$ coordinates of $\vec{u}_i$ in B_o
 $\uparrow$ coordinates of $\vec{u}_n$ in B_o

Example:



~~$\vec{u}_1 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}, \vec{u}_2 = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$~~

Given $\vec{b} = 3\vec{u}_1 + 2\vec{u}_2$ (i.e., $\vec{b}|_u = (3, 2)$),
 what are the coordinates of $\vec{b}$ in V ?

$$\left. \begin{aligned} \vec{v}_1 &= \vec{u}_1 + \vec{u}_2 \\ \vec{v}_2 &= -\vec{u}_1 + \vec{u}_2 \end{aligned} \right\} \rightarrow \begin{aligned} \vec{u}_1 &= \frac{1}{2}\vec{v}_1 - \frac{1}{2}\vec{v}_2 \\ \vec{u}_2 &= \frac{1}{2}\vec{v}_1 + \frac{1}{2}\vec{v}_2 \end{aligned} \rightarrow M = \begin{bmatrix} 1/2 & 1/2 \\ -1/2 & 1/2 \end{bmatrix}$$

$$\text{So } \vec{b}|_V = \begin{bmatrix} 1/2 & 1/2 \\ -1/2 & 1/2 \end{bmatrix} \begin{bmatrix} 3 \\ 2 \end{bmatrix} = \begin{bmatrix} 5/2 \\ -1/2 \end{bmatrix} \rightarrow \vec{b} = \frac{5}{2}\vec{v}_1 - \frac{1}{2}\vec{v}_2$$

Remark: $\vec{v}_1|_V = \begin{bmatrix} 1/2 & 1/2 \\ -1/2 & 1/2 \end{bmatrix} \begin{bmatrix} 1 \\ 1 \end{bmatrix} = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$ of course!

Example: $\{\vec{u}_1, \vec{u}_2, \vec{u}_3\} = \{1, x, x^2\} \rightarrow U \text{ basis}$

$\{\vec{v}_1, \vec{v}_2, \vec{v}_3\} = \{1, x-1, (x-1)^2\} \rightarrow V \text{ basis}$

Given $p(x) = 2 + x + 2x^2$, how to write it in the V basis?

$$\vec{u}_1 = \vec{u}_1$$

$$\vec{u}_2 = \vec{u}_1 + \vec{u}_2$$

$$\vec{u}_3 = \vec{u}_1 + 2\vec{u}_2 + \vec{u}_3$$

$$\} \rightarrow P = \begin{bmatrix} 1 & 1 & 1 \\ 0 & 1 & 2 \\ 0 & 0 & 1 \end{bmatrix}$$

$$\begin{aligned} (x-1)^2 &= \\ &= x^2 - 2x + 1 \end{aligned}$$

$$\text{And therefore } p(x)|_V = \begin{bmatrix} 1 & 1 & 1 \\ 0 & 1 & 2 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 2 \\ 1 \\ 2 \end{bmatrix} = \begin{bmatrix} 5 \\ 5 \\ 2 \end{bmatrix}$$

that is, $2 + x + 2x^2 = 5 + 5(x-1) + 2(x-1)^2$ Taylor expansion at $x=1$!!

Exercise: $U = \{\vec{u}_1, \vec{u}_2, \vec{u}_3\}$, $V = \{\vec{v}_1, \vec{v}_2, \vec{v}_3\}$ $\left| \begin{array}{l} T \text{ linear.} \\ T(\vec{u}_1) = \vec{v}_1, T(\vec{u}_2) = T(\vec{u}_3) = \vec{v}_1 + \vec{v}_3 \end{array} \right.$

$\left\{ \begin{array}{l} \text{Find the matrix } A \text{ such that } T(\vec{u}) = A\vec{u} \\ \text{Find } T(\vec{u}_1 + \vec{u}_2 + \vec{u}_3) \end{array} \right.$

Sol:

We know that $A = \begin{bmatrix} T(\vec{u}_1) & T(\vec{u}_2) & T(\vec{u}_3) \end{bmatrix}$ with $T(\vec{u}_1), T(\vec{u}_2), T(\vec{u}_3)$

written in the output basis V .

So,

$$T(\vec{u}_1)|_V = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, T(\vec{u}_2)|_V = T(\vec{u}_3)|_V = \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix} \rightarrow A = \begin{bmatrix} 1 & 1 & 1 \\ 0 & 0 & 0 \\ 0 & 1 & 1 \end{bmatrix}.$$

Therefore, since

$$(\vec{u}_1 + \vec{u}_2 + \vec{u}_3)|_U = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} \rightarrow T(\vec{u}_1 + \vec{u}_2 + \vec{u}_3)|_V = \begin{bmatrix} 1 & 1 & 1 \\ 0 & 0 & 0 \\ 0 & 1 & 1 \end{bmatrix} \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} = \begin{bmatrix} 3 \\ 0 \\ 2 \end{bmatrix} \leftarrow$$

• We could have done this without the matrix:

$$\begin{aligned} T(\vec{u}_1 + \vec{u}_2 + \vec{u}_3) &= T(\vec{u}_1) + T(\vec{u}_2) + T(\vec{u}_3) = \vec{v}_1 + (\vec{v}_1 + \vec{v}_3) + (\vec{v}_1 + \vec{v}_3) = \\ &= \underset{=}{3\vec{v}_1} + \underset{=}{2\vec{v}_3}. \end{aligned}$$