

## Lecture 15

In-class exercise from last time: diagonalize the matrix  $A = \begin{pmatrix} 1 & 1 \\ 1 & 1 \end{pmatrix}$  over  $\mathbb{C}$ .

$$\text{char poly}(A) = \det \begin{pmatrix} -\lambda & 0 & 1 \\ 1 & -\lambda & 0 \\ 0 & 1 & -\lambda \end{pmatrix} = -\lambda^3 + 1$$

Roots of  $-\lambda^3 + 1$ :  $\lambda = 1$  works since  $-1^3 + 1 = 0$ . So divide  $-\lambda^3 + 1$  by  $\lambda - 1 \rightarrow -\lambda^3 + 1 = -(\lambda - 1) \cdot (\lambda^2 + \lambda + 1)$  (Use long division)

Now the roots of  $\lambda^2 + \lambda + 1$  are  $\frac{-1 + i\sqrt{3}}{2}$  and  $\frac{-1 - i\sqrt{3}}{2}$ .

Eigenspaces:  $E_1 = \text{Ker} \begin{pmatrix} -1 & 1 \\ 1 & -1 \end{pmatrix}$

$$\left( \begin{array}{ccc|c} -1 & 0 & 1 & 0 \\ 1 & -1 & 0 & 0 \\ 0 & 1 & -1 & 0 \end{array} \right) \xrightarrow{I \rightarrow -I} \left( \begin{array}{ccc|c} 1 & 0 & -1 & 0 \\ 1 & -1 & 0 & 0 \\ 0 & 1 & -1 & 0 \end{array} \right)$$

$$\xrightarrow{II \rightarrow II - I} \left( \begin{array}{ccc|c} 1 & 0 & -1 & 0 \\ 0 & -1 & 1 & 0 \\ 0 & 1 & -1 & 0 \end{array} \right)$$

$$\xrightarrow{II \rightarrow -II} \left( \begin{array}{ccc|c} 1 & 0 & -1 & 0 \\ 0 & 1 & -1 & 0 \\ 0 & 1 & -1 & 0 \end{array} \right)$$

$$\xrightarrow{III \rightarrow III - II} \left( \begin{array}{ccc|c} 1 & 0 & -1 & 0 \\ 0 & 1 & -1 & 0 \\ 0 & 0 & 0 & 0 \end{array} \right) \quad \text{let}$$

$$\Rightarrow E_1 = \text{Span} \left( \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} \right)$$

$$E_{\frac{-1+i\sqrt{3}}{2}} = \text{Ker} \begin{pmatrix} \frac{1-i\sqrt{3}}{2} & 0 & 1 \\ 1 & \frac{1-i\sqrt{3}}{2} & 0 \\ 0 & 1 & \frac{1-i\sqrt{3}}{2} \end{pmatrix}$$

$$\left( \begin{array}{ccc|c} \frac{1-i\sqrt{3}}{2} & 0 & 1 & 0 \\ 1 & \frac{1-i\sqrt{3}}{2} & 0 & 0 \\ 0 & 1 & \frac{1-i\sqrt{3}}{2} & 0 \end{array} \right) \xrightarrow{I \rightarrow \frac{1-i\sqrt{3}}{2} I} \left( \begin{array}{ccc|c} 1 & 0 & \frac{1+i\sqrt{3}}{2} & 0 \\ 1 & \frac{1-i\sqrt{3}}{2} & 0 & 0 \\ 0 & 1 & \frac{1-i\sqrt{3}}{2} & 0 \end{array} \right)$$

$$\left( \frac{1-i\sqrt{3}}{2} \right)^{-1} = \frac{2}{1-i\sqrt{3}} = \frac{2(1+i\sqrt{3})}{4}$$

$$\text{II} \rightarrow \text{II} - \text{I} \rightarrow \begin{pmatrix} 1 & 0 & \frac{1+i\sqrt{3}}{2} & | & 0 \\ 0 & \frac{1-i\sqrt{3}}{2} & \frac{-1-i\sqrt{3}}{2} & | & 0 \\ 0 & 1 & \frac{1-i\sqrt{3}}{2} & | & 0 \end{pmatrix}$$

$$\text{II} \leftrightarrow \text{III} \rightarrow \begin{pmatrix} 1 & 0 & \frac{1+i\sqrt{3}}{2} & | & 0 \\ 0 & 1 & \frac{1-i\sqrt{3}}{2} & | & 0 \\ 0 & \frac{1-i\sqrt{3}}{2} & \frac{-1-i\sqrt{3}}{2} & | & 0 \end{pmatrix}$$

$$\left(\frac{1-i\sqrt{3}}{2}\right)^2 = \frac{1-3-2i\sqrt{3}}{4} = \frac{-1-i\sqrt{3}}{2}$$

$$\text{III} \rightarrow \text{III} - \frac{1-i\sqrt{3}}{2} \text{II} \rightarrow \begin{pmatrix} 1 & 0 & \frac{1+i\sqrt{3}}{2} & | & 0 \\ 0 & 1 & \frac{1-i\sqrt{3}}{2} & | & 0 \\ 0 & 0 & 0 & | & 0 \end{pmatrix}$$

$$\Rightarrow E_{\frac{-1+i\sqrt{3}}{2}} = \text{Span} \left( \begin{pmatrix} \frac{-1-i\sqrt{3}}{2} \\ -1+i\sqrt{3} \\ 1 \end{pmatrix} \right)$$

$$E_{\frac{-1-i\sqrt{3}}{2}} = \text{Ker} \begin{pmatrix} \frac{1+i\sqrt{3}}{2} & 0 & 1 \\ 1 & \frac{1+i\sqrt{3}}{2} & 0 \\ 0 & 1 & \frac{1+i\sqrt{3}}{2} \end{pmatrix}$$

Useful observation: let  $\bar{A} = (\bar{a}_{ij})_{i,j=1,\dots,n}$ , and  $\bar{v} = \begin{pmatrix} \bar{v}_1 \\ \vdots \\ \bar{v}_n \end{pmatrix}$ .

If  $Av = 0$ , then  $\bar{A}\bar{v} = \overline{Av} = \bar{0} = 0$

In other words,  $v \in \text{Ker}(A) \Leftrightarrow \bar{v} \in \text{Ker}(\bar{A})$

$$\text{Therefore } E_{\frac{-1-i\sqrt{3}}{2}} = \text{Span} \left( \begin{pmatrix} \frac{-1+i\sqrt{3}}{2} \\ -1-i\sqrt{3} \\ 1 \end{pmatrix} \right)$$

$$\text{Finally, } S = \begin{pmatrix} 1 & \frac{-1-i\sqrt{3}}{2} & \frac{-1+i\sqrt{3}}{2} \\ 1 & \frac{-1+i\sqrt{3}}{2} & \frac{-1-i\sqrt{3}}{2} \\ 1 & 1 & 1 \end{pmatrix}$$

$$S^{-1} \begin{pmatrix} 1 & \frac{-1-i\sqrt{3}}{2} & \frac{-1+i\sqrt{3}}{2} & | & 1 & 0 & 0 \\ 1 & \frac{-1+i\sqrt{3}}{2} & \frac{-1-i\sqrt{3}}{2} & | & 0 & 1 & 0 \\ 1 & 1 & 1 & | & 0 & 0 & 1 \end{pmatrix} \xrightarrow{\substack{\text{II} \rightarrow \text{II} - \text{I} \\ \text{III} \rightarrow \text{III} - \text{I}}} \begin{pmatrix} 1 & \frac{-1-i\sqrt{3}}{2} & \frac{-1+i\sqrt{3}}{2} & | & 1 & 0 & 0 \\ 0 & i\sqrt{3} & -i\sqrt{3} & | & -1 & 1 & 0 \\ 0 & \frac{3+i\sqrt{3}}{2} & \frac{3-i\sqrt{3}}{2} & | & -1 & 0 & 1 \end{pmatrix}$$

$$\xrightarrow{\text{I} \rightarrow \frac{1}{i\sqrt{3}} \text{II}} \begin{pmatrix} 1 & \frac{-1-i\sqrt{3}}{2} & \frac{-1+i\sqrt{3}}{2} & | & 1 & 0 & 0 \\ 0 & 1 & -1 & | & \frac{-i\sqrt{3}}{3} & \frac{i\sqrt{3}}{3} & 0 \\ 0 & \frac{3+i\sqrt{3}}{2} & \frac{3-i\sqrt{3}}{2} & | & -1 & 0 & 1 \end{pmatrix}$$

$$\begin{array}{l} I \rightarrow I + \frac{1+i\sqrt{3}}{2} II \\ III \rightarrow III - \frac{3-i\sqrt{3}}{2} II \end{array} \rightarrow \left( \begin{array}{ccc|cc} 1 & 0 & -1 & \frac{3-i\sqrt{3}}{6} & \frac{3+i\sqrt{3}}{6} & 0 \\ 0 & 1 & -1 & \frac{1+i\sqrt{3}}{3} & \frac{-i\sqrt{3}}{3} & 0 \\ 0 & 0 & 3 & \frac{-1-i\sqrt{3}}{2} & \frac{-1+i\sqrt{3}}{2} & 1 \end{array} \right)$$

$$\rightarrow \begin{array}{l} III \rightarrow \frac{1}{3} III \end{array} \rightarrow \left( \begin{array}{ccc|cc} 1 & 0 & -1 & \frac{3-i\sqrt{3}}{6} & \frac{3+i\sqrt{3}}{6} & 0 \\ 0 & 1 & -1 & \frac{1+i\sqrt{3}}{3} & \frac{-i\sqrt{3}}{3} & 0 \\ 0 & 0 & 1 & \frac{-1-i\sqrt{3}}{6} & \frac{-1+i\sqrt{3}}{6} & \frac{1}{3} \end{array} \right)$$

$$\begin{array}{l} I \rightarrow I + III \\ II \rightarrow II + III \end{array} \rightarrow \left( \begin{array}{ccc|cc} 1 & 0 & 0 & \frac{1}{3} & \frac{1}{3} & \frac{1}{3} \\ 0 & 1 & 0 & \frac{1+i\sqrt{3}}{6} & \frac{-1-i\sqrt{3}}{6} & \frac{1}{3} \\ 0 & 0 & 1 & \frac{-1-i\sqrt{3}}{6} & \frac{-1+i\sqrt{3}}{6} & \frac{1}{3} \end{array} \right)$$

So  $\begin{pmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 1 & 0 & 0 \end{pmatrix} = \begin{pmatrix} 1 & \frac{-1-i\sqrt{3}}{2} & \frac{-1+i\sqrt{3}}{2} \\ 1 & \frac{-1+i\sqrt{3}}{2} & \frac{-1-i\sqrt{3}}{2} \\ 1 & 1 & 1 \end{pmatrix} \begin{pmatrix} 1 & & \\ & -\frac{1-i\sqrt{3}}{2} & \\ & & -\frac{1+i\sqrt{3}}{2} \end{pmatrix} \begin{pmatrix} \frac{1}{3} & & \\ \frac{1+i\sqrt{3}}{6} & \frac{-1-i\sqrt{3}}{6} & \\ \frac{-1-i\sqrt{3}}{6} & \frac{-1+i\sqrt{3}}{6} & \frac{1}{3} \end{pmatrix}$

### Finishing up: Assorted questions

- The algebraic multiplicity of an eigenvalue cannot exceed its geometric multiplicity.
- If an  $n \times n$  matrix  $A$  is diagonalizable (over  $\mathbb{R}$ ), then there must be a basis of  $\mathbb{R}^n$  consisting of eigenvectors of  $A$ .
- If the standard vectors  $\vec{e}_1, \vec{e}_2, \dots, \vec{e}_n$  are eigenvectors of an  $n \times n$  matrix  $A$ , then  $A$  must be diagonal.
- If  $\vec{v}$  is an eigenvector of  $A$ , then  $\vec{v}$  must be an eigenvector of  $A^3$  as well.
- There exists a diagonalizable  $5 \times 5$  matrix with only two distinct eigenvalues (over  $\mathbb{C}$ ).
- There exists a real  $5 \times 5$  matrix without any real eigenvalues.

False:  $\begin{pmatrix} 1 & \\ & 0 \end{pmatrix}$

True (By definition)

True:  $A = \begin{pmatrix} \lambda_1 & & \\ & \ddots & \\ & & \lambda_n \end{pmatrix}$  where  $\lambda_i$  is the eig. of  $e_i$ .

True:  $A^3 v = \lambda^3 v$

False: char poly has deg 5, say  $p(\lambda)$ . Then  $\lim_{\lambda \rightarrow \infty} p(\lambda)$  and  $\lim_{p \rightarrow -\infty} p(\lambda)$  are  $\pm \infty \Rightarrow$  There is some  $\lambda \in \mathbb{R}$  s.t.  $p(\lambda) = 0$ .

- If 1 is the only eigenvalue of an  $n \times n$  matrix  $A$ , then  $A$  must be  $I_n$ .
- If  $A$  and  $B$  are  $n \times n$  matrices, if  $\alpha$  is an eigenvalue of  $A$ , and if  $\beta$  is an eigenvalue of  $B$ , then  $\alpha\beta$  must be an eigenvalue of  $AB$ .
- If 3 is an eigenvalue of an  $n \times n$  matrix  $A$ , then 9 must be an eigenvalue of  $A^2$ .
- The matrix of any ~~orthogonal~~ projection onto a subspace  $V$  of  $\mathbb{R}^n$  is diagonalizable.
- All diagonalizable matrices are invertible.
- If vector  $\vec{v}$  is an eigenvector of both  $A$  and  $B$ , then  $\vec{v}$  must be an eigenvector of  $A+B$ .
- If matrix  $A^2$  is diagonalizable, then matrix  $A$  must be diagonalizable as well.
- The determinant of a matrix is the product of its eigenvalues (over  $\mathbb{C}$ ), counted with their algebraic multiplicities.
- All lower triangular matrices are diagonalizable (over  $\mathbb{C}$ ).

False:  $A = \begin{pmatrix} 2 & 1 \\ 0 & 1 \end{pmatrix}$

False:  $A = \begin{pmatrix} 0 & 2 \\ 1 & 0 \end{pmatrix}$ ,  $B = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix}$ ,  $AB = \begin{pmatrix} 2 & 0 \\ 0 & 1 \end{pmatrix}$

True:  $Av = 3v \Rightarrow A^2 v = 9v$

False:  $\begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix}$

True:  $(A+B)v = Av + Bv = \lambda v + \mu v = (\lambda + \mu)v$

False:  $\begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix}^2 = \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix}$  is diagonalizable, but  $\begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix}$  is in JNF and is not diagonal

False:  $\begin{pmatrix} 1 & 0 \\ 1 & 1 \end{pmatrix}$  is not diagonalizable:  $a_2 = 2$  whereas  $g_2 = \dim(E_1)$

$$= \dim(\text{Ker} \begin{pmatrix} 0 & 0 \\ 1 & 0 \end{pmatrix})$$

$$= 2 - \text{rank} \begin{pmatrix} 0 & 0 \\ 1 & 0 \end{pmatrix} = 1$$

Today and next week: orthogonality

Idea: length and angles

Let  $v = \begin{pmatrix} v_1 \\ \vdots \\ v_n \end{pmatrix}$ . Then one can define the length of  $v$  as  $\sqrt{v_1^2 + \dots + v_n^2}$ .

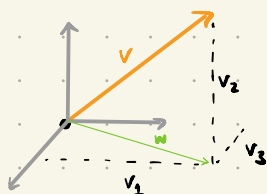
This matches our intuition:

• In  $\mathbb{R}^2$ ,



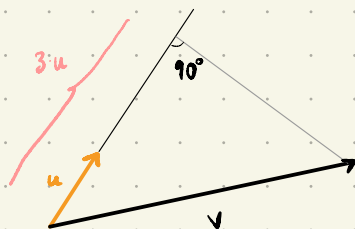
$$\Rightarrow \text{length}(v) = \sqrt{v_1^2 + v_2^2} \quad (\text{Pythagorean theorem})$$

• In  $\mathbb{R}^3$ ,



$$\begin{aligned} \text{Pythagoras: } \text{length}(v) &= \sqrt{v_2^2 + \text{length}(w)^2} \\ &= \sqrt{v_1^2 + v_2^2 + v_3^2} \end{aligned}$$

Next, fix a vector  $u = \begin{pmatrix} u_1 \\ \vdots \\ u_n \end{pmatrix}$  of length 1. One can consider the transformation  $T: \mathbb{R}^n \rightarrow \mathbb{R}$  given by the matrix  $(u_1 \dots u_n)$ . This represents an "orthogonal projection" onto the line spanned by  $u$ :



$$\Rightarrow T(v) = 3$$

(This is explained in 3Blue1Brown's video)

Observe that  $u$  and  $v$  are "perpendicular" iff  $T(v) = 0$  i.e. iff  $u^T v = 0$ .

Definition 1: • Let  $v, w \in \mathbb{R}^n$ ,  $v = \begin{pmatrix} v_1 \\ \vdots \\ v_n \end{pmatrix}$ ,  $w = \begin{pmatrix} w_1 \\ \vdots \\ w_n \end{pmatrix}$ . Then the dot product of  $v$  and  $w$  is

$$v \cdot w = v^T w = v_1 w_1 + \dots + v_n w_n.$$

• The length of  $v$  is  $\|v\| = \sqrt{v \cdot v}$ . If  $\|v\| = 1$ , then  $v$  is a unit vector.

• The vectors  $v$  and  $w$  are orthogonal iff  $v \cdot w = 0$ .

Let  $S$  be a subspace of  $\mathbb{R}^n$ . In what follows, we will be interested in finding bases of  $S$  consisting of unit vectors, all pairwise orthogonal.



Definition 2: Let  $u_1, \dots, u_k \in \mathbb{R}^n$ . Then  $u_1, \dots, u_k$  are orthonormal iff  $\|v_i\| = 1$  for all  $i$  and

$$u_i \cdot u_j = 0 \text{ for all } i, j \text{ s.t. } i \neq j. \quad (\text{More succinctly: } v_i \cdot v_j = \begin{cases} 1 & i=j \\ 0 & i \neq j \end{cases})$$

Example 1: the vectors  $\begin{pmatrix} 0 \\ 1 \\ 0 \\ 0 \end{pmatrix}, \begin{pmatrix} 0 \\ 0 \\ 1 \\ 0 \end{pmatrix}, \begin{pmatrix} 0 \\ 0 \\ 0 \\ 1 \end{pmatrix}$  are orthonormal.

Theorem 1: Orthonormal vectors are linearly independent.

Proof: Suppose  $u_1, \dots, u_k$  are orthonormal. Take a linear dependence  $\lambda_1 u_1 + \dots + \lambda_k u_k$ . We want to show that  $\lambda_1 = \dots = \lambda_k = 0$ .

$$\text{Now } 0 = u_i \cdot (0) = u_i \cdot (\lambda_1 u_1 + \dots + \lambda_k u_k) = \lambda_i \quad \text{for all } i, \text{ as desired. } \square$$

$u_i \cdot u_j = \begin{cases} 1 \\ 0 \end{cases}$

Corollary: Let  $S \subseteq \mathbb{R}^n$  be a subspace of dimension  $s$ . Then, any set of  $s$  orthonormal vectors in  $S$  forms a basis of  $S$ .

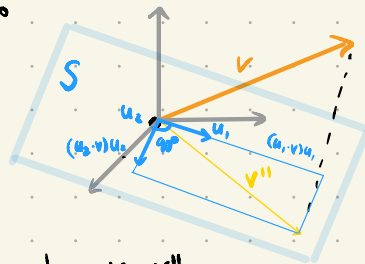
## Orthogonal projections

Fix a subspace  $S \subseteq \mathbb{R}^n$ , and assume  $u_1, \dots, u_s$  is an orthonormal basis of  $S$ .

Definition 3: Let  $v \in \mathbb{R}^n$ . Then the orthogonal projection of  $v$  onto  $S$  is given by:

$$v'' = (u_1 \cdot v)u_1 + (u_2 \cdot v)u_2 + \dots + (u_s \cdot v)u_s$$

Picture:  $u_i \cdot v$  is the "coordinate in the direction of  $u_i$ ", so



Definition 4: The perpendicular component of  $v$  is  $v^\perp = v - v''$ .

Theorem 2: The vector  $v^\perp$  is orthogonal to the subspace  $S$  (i.e. orthogonal to every vector in  $S$ ).

Proof: First note that  $u_i \cdot v^\perp = u_i \cdot (v - v'')$

$$= (u_i \cdot v) - u_i \cdot ((u_1 \cdot v)u_1 + (u_2 \cdot v)u_2 + \dots + (u_s \cdot v)u_s)$$

$$= (u_i \cdot v) - (u_i \cdot v) \cdot \underbrace{(u_i \cdot u_i)}_1 = 0.$$

Now, if  $w \in S$ , we can write it as  $w = \lambda_1 u_1 + \dots + \lambda_s u_s$ , so  $v^\perp \cdot w = \lambda_1 \underbrace{(v^\perp \cdot u_1)}_0 + \dots + \lambda_s \underbrace{(v^\perp \cdot u_s)}_0 = 0$ .

Example 2: Consider  $S = \text{Span} \left( \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix}, \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix} \right)$ , and  $v = \begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix}$ .

Note that  $u_1 = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix}$ ,  $u_2 = \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}$  form an orthonormal basis.

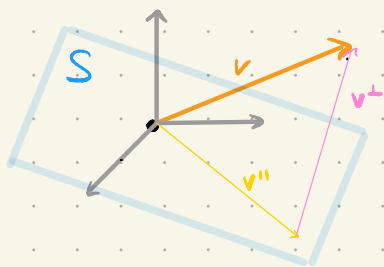
Then  $v'' = (v \cdot u_1) \cdot u_1 + (v \cdot u_2) \cdot u_2 = \frac{3}{\sqrt{2}} \cdot u_1 + 3 \cdot u_2 = \begin{pmatrix} 3/\sqrt{2} \\ 3/\sqrt{2} \\ 3 \end{pmatrix}$

$$v^\perp = v - v'' = \begin{pmatrix} -1/2 \\ 1/2 \\ 0 \end{pmatrix}$$

$$\text{Observe: } v^\perp \cdot u_1 = -\frac{1}{\sqrt{2}} + \frac{1}{\sqrt{2}} = 0$$

$$v^\perp \cdot u_2 = 0 + 0 = 0$$

Remark: observe that every  $v \in \mathbb{R}^n$  can be written uniquely as  $v = v'' + v^\perp$ .



Important remark:  $v''$  and  $v^\perp$  are always in reference to some subspace  $S \subseteq \mathbb{R}^n$ . Sometimes, we write  $v'' = \text{proj}_S(v)$  every time, to be less ambiguous.

Observation: If  $S = \mathbb{R}^n$ ,  $\text{proj}_S = \text{identity map}$ , so if  $u_1, \dots, u_n$  is an orthonormal basis of  $\mathbb{R}^n$ , every  $v \in \mathbb{R}^n$  can be written as  $v = (u_1 \cdot v) u_1 + \dots + (u_n \cdot v) u_n$ .

Definition 5: The orthogonal complement of a subspace  $S \subseteq \mathbb{R}^n$  is  $S^\perp = \{v \in \mathbb{R}^n : v \text{ is orthogonal to } S\}$   
 $= \{v \in \mathbb{R}^n : v \cdot w = 0 \text{ for all } w \in S\}$

Theorem 3: Let  $S \subseteq \mathbb{R}^n$  be a linear subspace and let  $u_1, \dots, u_s$  be an orthonormal basis of  $S$ .

Then the transformation  $\text{proj}_S : \mathbb{R}^n \rightarrow \mathbb{R}^n$  is linear, its image is  $S$  and its kernel is  $S^\perp$ .

Proof: To see that  $\text{proj}_S$  is linear, we need to show:

- $\text{proj}_S(v+w) = \text{proj}_S(v) + \text{proj}_S(w)$  (In-class exercise)

- $\text{proj}_S(\lambda v) = (u_1 \cdot \lambda v)u_1 + \dots + (u_n \cdot \lambda v)u_n$

$$\begin{aligned} &= \lambda(u_1 \cdot v)u_1 + \dots + \lambda(u_n \cdot v)u_n \\ &\quad \downarrow \\ &\text{dot product} \\ &\text{is linear} \\ &= \lambda((u_1 \cdot v)u_1 + \dots + (u_n \cdot v)u_n) \\ &= \lambda \text{proj}_S(v). \end{aligned}$$

$\text{Im}(S) = S$  (In-class exercise).

$\text{Ker}(S) = \{v \in \mathbb{R}^n : (u_1 \cdot v)u_1 + \dots + (u_n \cdot v)u_n = 0\}$

$$\begin{aligned} &= \{v \in \mathbb{R}^n : u_1 \cdot v = u_2 \cdot v = \dots = u_n \cdot v = 0\} \\ &\quad \downarrow \\ &\quad u_1, \dots, u_n \text{ l.i.} \\ &= \{v \in \mathbb{R}^n : w \cdot v = 0 \text{ for all } w \in S\} \\ &\quad \downarrow \\ &\quad u_1, \dots, u_n \text{ span } S \\ &= S^\perp. \quad \square \end{aligned}$$

Corollary:  $S^\perp$  is a subspace of  $\mathbb{R}^n$

Natural question: what is the matrix of  $\text{proj}_S$ ?

Example 3: Consider  $S = \text{Span} \left( \begin{pmatrix} 1 \\ 0 \end{pmatrix}, \begin{pmatrix} 0 \\ 1 \end{pmatrix} \right)$ , and  $v = \begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix}$  again.

Then  $\text{proj}_S \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} = (e_1 \cdot u_1) \cdot u_1 + (e_1 \cdot u_2) \cdot u_2$   
 $= \frac{1}{\sqrt{2}} u_1 + 0 \cdot u_2 = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix}$

$\text{proj}_S \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix} = (e_2 \cdot u_1) \cdot u_1 + (e_2 \cdot u_2) \cdot u_2$   
 $= \frac{1}{\sqrt{2}} u_1 = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix}$

$\text{proj} \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix} = (e_3 \cdot u_1)u_1 + (e_3 \cdot u_2)u_2$   
 $= 0u_1 + u_2 = \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}$

$\text{proj}_S$  is given by the matrix

$$\begin{pmatrix} \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} & 0 \\ \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} & 0 \\ 0 & 0 & 1 \end{pmatrix}$$

Next time: how to find orthonormal bases?

In-class exercises:

Let  $S = \text{Span} \left( \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}, \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix} \right)$ . Let  $v = \begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix}$ . Find  $v^{\parallel}$  and  $v^{\perp}$ . Verify that  $v^{\perp}$  is orthogonal to  $S$ .