

## Monday, Week 10

Let  $(V, \langle \cdot, \cdot \rangle_V)$  and  $(W, \langle \cdot, \cdot \rangle_W)$  be finite-dimensional inner product spaces over  $\mathbb{F} \in \{\mathbb{R}, \mathbb{C}\}$ .

### Definitions

- **Adjoint:** The adjoint of a map  $T \in \mathcal{L}(V, W)$  is the unique linear map  $T^* \in \mathcal{L}(W, V)$  such that  $\langle Tv, w \rangle_W = \langle v, T^*w \rangle_V$  for all  $v \in V, w \in W$ .
- **Normal Operator:**  $T \in \mathcal{L}(V)$  is *normal* if  $T^*T = TT^*$ .
- **Self-adjoint Operator:**  $T \in \mathcal{L}(V)$  is *self-adjoint* if  $T = T^*$ .
- **Unitary Operator:**  $T \in \mathcal{L}(V, W)$  is *unitary* if:  
 $T^*T = I_V$  and  $TT^* = I_W$ .  
Equivalently,  $T^* = T^{-1}$ .
- **Real and Imaginary Parts:** For any  $T \in \mathcal{L}(V)$ , we define:  
 $\operatorname{Re}(T) := \frac{T+T^*}{2}$ ,  $\operatorname{Im}(T) := \frac{T-T^*}{2i}$ .

**Remark:** When  $V = W$ , unitary operators are always normal.

### Proposition

Let  $T \in \mathcal{L}(V, W)$ . The following statements are equivalent:

- (a)  $T^*T = I_V$
- (b)  $\langle Tv, Tv' \rangle_W = \langle v, v' \rangle_V$  for all  $v, v' \in V$
- (c)  $\|Tv\|_W = \|v\|_V$  for all  $v \in V$

Assume  $T^*T = I_V$ . We want to show that for all  $v, v' \in V$ ,  $\langle Tv, Tv' \rangle_W = \langle v, v' \rangle_V$ .  
Starting with the left-hand side:

$$\langle Tv, Tv' \rangle_W = \langle T^*Tv, v' \rangle_V = \langle I_V v, v' \rangle_V = \langle v, v' \rangle_V.$$

So (a) implies (b).

Assume  $\langle Tv, Tv' \rangle_W = \langle v, v' \rangle_V$  for all  $v, v' \in V$ .

Let  $v = v'$ . Then:

$$\langle Tv, Tv \rangle_W = \langle v, v \rangle_V \Rightarrow \|Tv\|_W^2 = \|v\|_V^2 \Rightarrow \|Tv\|_W = \|v\|_V.$$

Hence, (b) implies (c).

Assume  $\|Tv\|_W = \|v\|_V$  for all  $v \in V$ .

Then:

$$\langle Tv, Tv \rangle_W = \|Tv\|_W^2 = \|v\|_V^2 = \langle v, v \rangle_V.$$

Let's compute:

$$\langle Tv, Tv \rangle_W = \langle T^*Tv, v \rangle_V \Rightarrow \langle (T^*T - I_V)v, v \rangle_V = 0 \text{ for all } v \in V.$$

Since  $T^*T - I_V$  is self-adjoint (from Week 9), the inner product  $\langle (T^*T - I_V)v, v \rangle_V = 0$  for all  $v \in V$  implies that  $T^*T - I_V = 0$ , hence:

$$T^*T = I_V.$$

So (c) implies (a).

## Proposition

Let  $T \in \mathcal{L}(V)$ . Then

$$T \text{ is unitary} \iff T \text{ is normal and } \operatorname{Re}(T)^2 + \operatorname{Im}(T)^2 = I.$$

6/3:

## RECAP:

Let  $V, W$  inner product spaces.  $T \in L(V, W)$ .

We say  $T$  is "unitary" if:

$$T^* = T^{-1}$$

$$(TT^* = I_W \quad T^*T = I_V)$$

equivalently:

① for all  $v, v' \in V$

$$\langle Tv, Tv' \rangle = \langle v, v' \rangle$$

② for all  $v \in V$ ,  $\|Tv\| = \|v\|$

philosophy (why we care)

$T$  being unitary, we know:

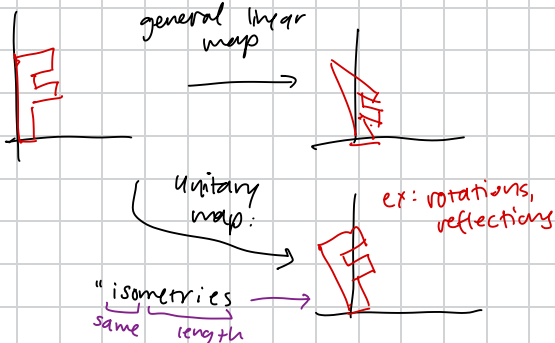
$T$  is linear and invertible: isomorphism of vector space

$[T]$  square

intuitively,  
translation/rename  
between two  
representations  
of the same underlying space

$\langle Tv, Tv' \rangle = \langle v, v' \rangle$ :  $T$  preserves inner products

unitary maps  $\leftrightarrow$  "inner product space isomorphisms"



Ex:  $T = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}$  (reflection across  $x=y$  line)

check  $T$  is unitary:

$$[T][T]^* = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} = I \quad \checkmark$$

Q: let  $v_1, \dots, v_n$  be columns of  $[T]$ .

If  $T^*T = I$ , what does that say about  $v_1, \dots, v_n$ ?

$$\text{pf: let } [T] := \begin{bmatrix} | & | & & | \\ [v_1] & [v_2] & \dots & [v_n] \\ | & | & & | \end{bmatrix}$$

$$\text{Then } T^* := \begin{bmatrix} -[v_1]^T - \\ -[v_2]^T - \\ \vdots \\ -[v_n]^T - \end{bmatrix}$$

$$\text{So } [T^*][T] = \begin{bmatrix} 1 & 0 & \dots & 0 \\ 0 & 1 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & 1 \end{bmatrix}$$

$$[T^*][T] = \begin{bmatrix} \bar{v}_1^T v_1 & \bar{v}_1^T v_2 & \dots & \bar{v}_1^T v_n \\ \bar{v}_2^T v_1 & \bar{v}_2^T v_2 & \dots & \bar{v}_2^T v_n \\ \vdots & \vdots & \ddots & \vdots \\ \bar{v}_n^T v_1 & \bar{v}_n^T v_2 & \dots & \bar{v}_n^T v_n \end{bmatrix}$$

$$\hookrightarrow \begin{bmatrix} \langle v_1, v_1 \rangle & \dots & \langle v_n, v_n \rangle \\ \vdots & \ddots & \vdots \\ \langle v_1, v_n \rangle & \dots & \langle v_n, v_n \rangle \end{bmatrix}$$

multiplying each column by itself to get 1 in diagonal.

$$\Rightarrow \bar{v}_j^T v_i = \begin{cases} 1 & \text{when } i=j \\ 0 & \text{when } i \neq j \end{cases}$$

$\Rightarrow \{v_1, \dots, v_n\}$  are orthonormal.

Take-Home Questions:

Q: what about the rows?

Q: is the converse true?

General thing of  
NORMAL ops

$T \in L(V, V)$  is normal  
 $TT^* = T^*T$ .

ex: self-adjoint, unitary are normal.

Guiding question:

can we say things about eigenvalues  
of normal operators?

Ex:  $T = \begin{bmatrix} a & b \\ 0 & c \end{bmatrix}$

If  $T^*T = TT^*$ ,

what does that tell you more  
specific about  $a, b, c$ ?

$$b = 0$$

$T$  is diagonal

PF:

$$(T)^* = \begin{bmatrix} \bar{a} & 0 \\ \bar{b} & \bar{c} \end{bmatrix}$$

$$TT^* = T^*T$$

$$\begin{bmatrix} a & b \\ 0 & c \end{bmatrix} \begin{bmatrix} \bar{a} & 0 \\ \bar{b} & \bar{c} \end{bmatrix} = \begin{bmatrix} \bar{a} & 0 \\ \bar{b} & \bar{c} \end{bmatrix} \begin{bmatrix} a & b \\ 0 & c \end{bmatrix}$$
$$\begin{bmatrix} a\bar{a} + b\bar{b} & b\bar{c} \\ \bar{c}b & c\bar{c} \end{bmatrix} = \begin{bmatrix} \bar{a}a & \bar{a}b \\ a\bar{b} & \bar{b}b + \bar{c}c \end{bmatrix}$$

$$a\bar{a} + b\bar{b} = \bar{a}a$$

$$b\bar{b} = 0 \Rightarrow b = 0$$

Thus,  $T$  is diagonal.

(general) Conjecture:

Assume  $T \in \mathbb{C}^{n \times n}$  is upper-triangular

If  $T$  is normal, then  $TT^*$  is diagonal.

Take-home Q: try on  $3 \times 3$ , try to prove  
in general.

punchline: Atiyah says all operators  
 $V \rightarrow V$  are triangulizable (in an.B)

So if  $T$  normal, then it's  
diagonalizable in orthonormal basis

### The spectral theorem

Let  $(V, \langle \cdot, \cdot \rangle)$  be an n-dim inner product space over  $F = \mathbb{C}$ .

Thm (Spectral Theorem)

If  $T \in L(V)$  is normal, then there exists an ONB of  $V$  such that  ${}_B[T]_B$  is diagonal ("unitarily diagonalizable").

### Claim

This is equivalent to the statement: If  $A \in \mathbb{C}^{n \times n}$  is normal, then there exists a unitary  $U \in \mathbb{C}^{n \times n}$  such that  $U^{-1}AU$  is diagonal.

### Pf

Suppose that there exists a unitary matrix  $U \in \mathbb{C}^{n \times n}$  such that  $U^{-1}AU$  is diagonal.

${}_B[T]_B := {}_B[I]_E ({}_E[T]_E) {}_E[I]_B$ . Let  $b_j := \sum_{i=1}^n u_{ij} e_i$  so that  ${}_E[E]_B = ({}_E[b_1] \dots {}_E[b_n])$ . Then  ${}_B[T]_B = {}_B[I]_E ({}_E[T]_E) {}_E[I]$ , is diagonal. It remains to be shown that  $B := \{b_1, \dots, b_n\}$  is an ONB.

$$\begin{aligned} \text{Indeed, } \langle b_i, b_j \rangle &= \left\langle \sum_{i'=1}^n u_{i'i} e_{i'}, \sum_{j'=1}^n u_{j'j} e_{j'} \right\rangle \\ &= \sum_{i'=1}^n u_{i'i} \langle e_{i'}, \sum_{j'=1}^n u_{j'j} e_{j'} \rangle = \sum_{i'=1}^n u_{i'i} \langle e_{i'}, \sum_{j'=1}^n u_{j'j} e_{j'} \rangle \\ &= \sum_{i'=1}^n u_{i'i} \overline{u_{j'i}} \langle e_{i'}, e_{j'} \rangle = \sum_{i'=1}^n (U^* U)_{j'i} \\ &= I_{j'i} = \{1, j = i, 0, j \neq i\}. \end{aligned}$$

It remains to prove the spectral theorem for matrices. By HW 9 A4, every  $A \in \mathbb{C}^{n \times n}$  is unitarily triangularizable.

If  $A$  is also normal, then in fact, the triangularization is a diagonalization.

### The singular-value decomposition (SVD)

Let  $A \in \mathbb{C}^{m \times n}$  and suppose  $m \geq n$ . Then,  $A^* A$  is self-adjoint, so by the spectral theorem,  
 $A^* A = V \Lambda V^{-1}$  for some diagonal  $\Lambda \in \mathbb{C}^{n \times n} =$

$$\begin{vmatrix} \lambda_1 & & \\ & \dots & \\ & & \lambda_n \end{vmatrix}$$

For some unitary  $V \in \mathbb{C}^{n \times n} = [v_1 \dots v_n] \in \mathbb{C}^{n \times n}$ .

Then  $\lambda_i \in \mathbb{R}$  and  $\lambda_i \geq 0$  because  $A^* A$  is self-adjoint.  $A^* A v_i = \lambda_i v_i \Rightarrow$

$$\langle A^* A v_i, v_i \rangle = \langle \lambda_i v_i, v_i \rangle = \lambda_i \langle v_i, v_i \rangle$$

$$\Rightarrow \|A v_i\|^2 = \lambda_i \|v_i\|^2$$

$$\Rightarrow \lambda_i = \frac{\|A v_i\|^2}{\|v_i\|^2} > 0.$$

## Discussion 6/5

### SVD Recap

We are given any matrix  $A \in \mathbb{C}^{m \times n}$  where  $m \geq n$ .

**Note:**  $A^*A$  is self-adjoint so there exists an orthonormal basis  $\{v_1, \dots, v_n\}$  of eigenvectors of  $A^*A$  by the Spectral Theorem.

Let  $\lambda_i$  be an eigenvalue corresponding to  $v_i$ , then  $\sigma_i = \sqrt{\lambda_i}$ . We refer to these  $\sigma_i$  as singular values.

**Goal:** find  $U, \Sigma, V$  such that  $A = U\Sigma V^*$

$$V = \begin{bmatrix} v_1 & v_2 & \dots & v_n \end{bmatrix}$$

**Note:**  $\{v_1, \dots, v_n\}$  is an orthonormal basis, so  $V$  is unitary.

We want to construct  $U$  such that it is unitary and  $\Sigma$  such that it is diagonal(ish). Specifically, we want to find that

$$\Sigma = \begin{bmatrix} \sigma_1 & 0 & \dots & 0 \\ 0 & \sigma_2 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & \sigma_n \\ & \mathbf{0}_{(m-n) \times n} & & \end{bmatrix}$$

and

$$U^*A = \Sigma V^*$$

in which case we can define  $u$  by

$$u_i = \frac{Av_i}{\sigma_i}$$

It remains to show:

$$U = \begin{bmatrix} u_1 & u_2 & \dots & u_n \end{bmatrix}$$

Other details to consider:

- What if  $\sigma_i = 0$ ?
- Is  $U$  actually unitary?
- $A = U\Sigma V^*$

# Final Review

## Week 1

### Key Terms

- Vector spaces
- Groups
- Commutativity
- Associativity
- Identity
- Inverse
- Closure
- Fields
- Set
- Addition
- Multiplication
- Distributivity
- Compatibility
- Subspaces
- Intersection of subspaces
- Sum of subspaces

### Key Facts

- Subspaces are vector spaces
- Unique identity and inverse
- Subspace check: closure properties, existence of 0
- Intersections and sums of subspaces are subspaces
- To prove  $A = B$ , show  $A \subseteq B$  and  $B \subseteq A$

### Aside:

If we have subspaces  $V_1, \dots, V_n$  where each  $V_i = \text{span}\{v_i\}$ , then

$$W = V_1 + \dots + V_n \iff W = \text{span}\{v_1, \dots, v_n\}.$$

Moreover,

$$W = V_1 \oplus V_2 \oplus \dots \oplus V_n \iff W = \text{span}\{v_1, \dots, v_n\} \quad \text{and} \quad \{v_1, \dots, v_n\} \text{ is linearly independent.}$$



## Week 2

### Key Terms

- Linear Independence
- Basis
- Span
- Dimensions

### Key Facts

- Linear dependence lemma
- Steinitz Exchange
- all bases same size
- bases are maximally independent and minimally spanning

## Week 3

### Key Terms

- Linear Maps
- Kernel
- Image
- Nullity
- Rank
- Injectivity
- Surjectivity
- Bijectivity

### Key Facts

- a map is linear if it satisfies additivity and homogeneity, can show both at once using  $\alpha T(v) + T(w)$
- $T(0) = 0$  if  $T \in L(V, W)$
- Rank-Nullity Theorem
- Bijective  $\iff$  Surjective and Injective
- $T \in L(V, W)$  is injective  $\iff \ker(T) = \{0\}$

- $T \in L(V, W)$  is surjective  $\iff W = \text{im}(T)$
- kernel and image are subspaces

## Week 4

### Key Terms

- Inverse
- Isomorphism
- Matrices
- Matrix Addition
- Matrix Multiplication
- Column Space
- Column Rank
- Row Space
- Row Rank

### Key Facts

- matrix multiplication is a representation of function composition
- column  $k$  of matrix product  $AB$  is  $A$  times column  $k$  of  $B$
- columns of matrix product  $AB$  are linear combinations of columns of  $A$
- rows of matrix product  $AB$  are linear combinations of row of  $B$
- transpose of a matrix is the matrix obtained by interchanging the rows and columns
- Change of Basis: columns of a transformation matrix represent transformed coordinates of basis vectors,
 
$${}_W[I]_V = [{}_W[v_1] \dots {}_W[v_n]]$$

$${}_V[I]_W = [{}_V[w_1] \dots {}_V[w_n]]$$
- If  $T \in L(V, W)$  is invertible, then  $T^{-1}$  is its inverse.  $TT^{-1} = I_W$ .  $T^{-1}T = I_V$ .
- invertibility  $\iff$  bijectivity
- injectivity is equivalent to surjectivity if  $\dim(V) = \dim(W)$  for  $T \in L(V, W)$
- two vector spaces are isomorphic  $\iff$  have the same dimension

## **Week 5**

### **Key Terms**

- Determinant
- Polynomials
- Polynomial division

### **Key Facts**

- determinant is multilinear, alternating, and normalized
- determinant of an upper triangular matrix is the product of its diagonal entries
- polynomial division theorem

# Math 115A Lecture Notes

June 6, 2025

## Matrix of a linear map

Let  $T \in L(V)$  be a linear map on a vector space  $V$ . Let  $B = \{v_1, \dots, v_n\}$  be a basis for  $V$ . The matrix of  $T$  with respect to the basis  $B$  is denoted as  $[T]_B$ .

$$[T]_B = \begin{pmatrix} a_{11} & \dots & a_{1n} \\ \vdots & \ddots & \vdots \\ a_{m1} & \dots & a_{mn} \end{pmatrix}$$

This matrix is constructed by applying  $T$  to each basis vector and expressing the result in the same basis:  $[T]_B = ([Tv_1]_B \quad [Tv_2]_B \quad \dots \quad [Tv_n]_B)$

## Triangularization and Diagonalization

If  $[T]_B$  is an upper triangular matrix, then the eigenvalues of  $T$  are the diagonal entries  $a_{11}, \dots, a_{nn}$ .

If  $[T]_B$  is a diagonal matrix:  $[T]_B = \begin{pmatrix} a_{11} & & 0 \\ & \ddots & \\ 0 & & a_{nn} \end{pmatrix}$  then the eigenvalues

of  $T$  are the diagonal entries  $a_{11}, \dots, a_{nn}$ , and the corresponding eigenvectors are the basis vectors  $v_1, \dots, v_n$ .

For a general matrix representation, we have  $Tv_j = \sum_{i=1}^n a_{ij}v_i$ . In the case of a diagonal matrix, this simplifies to  $Tv_j = a_{jj}v_j$ .

The Spectral Theorem relates to the case where the basis  $B$  is also orthonormal.

## Topics Discussed

- Eigenvalues of linear maps, diagonalization, and triangularization
- The Spectral Theorem and orthonormal diagonalization
- Determinants using the permutation formula
- Minimal Polynomials
- Direct Sums

## Direct Sums

Let  $V$  be a vector space, and let  $U_1, \dots, U_k$  be subspaces of  $V$ . The sum  $W = U_1 + \dots + U_k$  is a direct sum, denoted  $W = U_1 \oplus \dots \oplus U_k$  or  $W = \bigoplus_{i=1}^k U_i$ , if every vector  $w \in W$  can be written uniquely as a sum  $w = u_1 + \dots + u_k$ , where each  $u_i \in U_i$ .

Reminder: The sum of two subspaces is defined as  $W = U_1 + U_2 = \{u_1 + u_2 \mid u_1 \in U_1, u_2 \in U_2\}$ .

A condition for a sum to be direct is that for any  $i \neq j$ , the intersection of the subspaces is the zero vector:  $U_i \cap U_j = \{0\}$ .

### Example in $\mathbb{R}^3$

Let  $V = \mathbb{R}^3$ . Consider the subspaces:  $U_1 = \text{span} \left\{ \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix}, \begin{pmatrix} 0 \\ 1 \\ 1 \end{pmatrix} \right\}$   $U_2 = \text{span} \left\{ \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}, \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix} \right\}$

(Note: The original notes have some ambiguity here, this is one interpretation).

Let's check if  $\mathbb{R}^3 = U_1 \oplus U_2$ . A vector in  $\mathbb{R}^3$  is written as a sum of vectors from  $U_1$  and  $U_2$ .

Another example from the notes: Let  $U_1 = \text{span} \left\{ \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix}, \begin{pmatrix} 0 \\ 1 \\ 1 \end{pmatrix} \right\}$ . Let's

define  $U_2$  based on Branden's idea. Pick a vector  $v$  not in  $U_1$ . Let  $v = \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}$ ,

and set  $U_2 = \text{span}\{v\}$ . To check if  $\mathbb{R}^3 = U_1 \oplus U_2$ , we can form a matrix with the basis vectors of  $U_1$  and  $U_2$  and check if they are linearly independent. The

vectors are  $\begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix}$ ,  $\begin{pmatrix} 0 \\ 1 \\ 1 \end{pmatrix}$ , and  $\begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}$ . These are linearly independent, so their span is  $\mathbb{R}^3$  and the sum is direct.

The notes also show a calculation for a cross product which results in the vector  $\begin{pmatrix} 0 \\ 1 \\ 1 \end{pmatrix}$ .