

## Recap of previous lecture

- ▶ Let  $\mathcal{B} = (e_1, e_2, \dots, e_n)$  be a basis for  $\mathbb{V}$  of dimension  $n$ . Any vector  $\mathbf{u} \in \mathbb{V}$  then can be expanded in this basis as

$$\mathbf{u} = \sum_{i=1}^n e_i u_i = \mathcal{B} \cdot \begin{pmatrix} u_1 \\ u_2 \\ \vdots \\ u_n \end{pmatrix} \longleftrightarrow \begin{pmatrix} u_1 \\ u_2 \\ \vdots \\ u_n \end{pmatrix} \in \mathbb{F}^n$$

**Remark:** We **always** write the basis as a row vector.

- ▶ Let  $\mathcal{B} = (e_1, e_2, \dots, e_n)$  and  $\mathcal{B}' = (e'_1, e'_2, \dots, e'_n)$  be two bases for  $\mathbb{V}$ . There is an invertible matrix  $S$  such that

$$e_i = \sum_{j=1}^n e'_j S_{ji} \quad \text{or} \quad \boxed{\mathcal{B} = \mathcal{B}' \cdot S \quad \text{and} \quad \mathcal{B}' = \mathcal{B} \cdot S^{-1}} .$$
$$\implies \begin{pmatrix} u'_1 \\ u'_2 \\ \vdots \\ u'_n \end{pmatrix} = S \cdot \begin{pmatrix} u_1 \\ u_2 \\ \vdots \\ u_n \end{pmatrix}$$

Note that “ $\cdot$ ” here implies matrix multiplication.

# Lecture 6a: Linear Maps and Their Matrices

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# What is a linear map?

Let  $\mathbb{V}$  and  $\mathbb{W}$  be two vector spaces (over  $\mathbb{F}$ ) of dimension  $n$  and  $m$  respectively. A map

$$T : \mathbb{V} \rightarrow \mathbb{W} ,$$

is a linear map if the following holds

$$T(a\mathbf{u} + b\mathbf{v}) = aT(\mathbf{u}) + bT(\mathbf{v})$$

In the LHS, we are adding vectors in  $\mathbb{V}$  before applying the map  $T$ . In the RHS we are adding vectors in  $\mathbb{W}$  after applying  $T$  separately on  $\mathbf{u}$  and  $\mathbf{v}$ .

**Example:** Let  $d/dx$  denote the derivative operator acting on the vector space  $\mathcal{P}_m$  of degree  $m$  polynomials in one variable  $x$ . Show that  $d/dx$  is a linear map with  $\mathbb{V} = \mathbb{W} = \mathcal{P}_m$ .

## Examples of linear maps

- ▶ Let  $\mathbb{V} = \mathbb{W} = \mathbb{R}^3$ . Let  $\mathbf{u} = (u_1, u_2, u_3)^T \in \mathbb{R}^3$ . The following two maps are linear:
  - ▶ "Rotating a vector":

$$T_1(\mathbf{u}) = \begin{pmatrix} \cos \theta & \sin \theta & 0 \\ -\sin \theta & \cos \theta & 0 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} u_1 \\ u_2 \\ u_3 \end{pmatrix}$$

- ▶ "Scaling a vector"  $T_2(\mathbf{u}) = a\mathbf{u}$  for some  $a \in \mathbb{R}$ .
- ▶ A  $\theta$ -polarizer is a linear map with  $\mathbb{V} = \mathbb{W} = \mathbb{C}^2$ .

$$\begin{pmatrix} \mathcal{E}'_z \\ \mathcal{E}'_x \end{pmatrix} = P(\theta) \begin{pmatrix} \mathcal{E}_z \\ \mathcal{E}_x \end{pmatrix}$$

where  $P(\theta)$  is the matrix defined in an earlier lecture.

Later, we will show that we can associate a matrix to every linear map.

# Matrices as linear maps

Consider the case when  $\mathbb{V} = \mathbb{F}^n$  and  $\mathbb{W} = \mathbb{F}^m$ .

- ▶ Linear maps are  $m \times n$  matrices.
- ▶ The set of  $m \times n$  matrices will be denoted by  $\mathbb{F}^{m,n}$ .
- ▶ This set can be made to an LVS with the following two operations:
  - ▶ Vector addition corresponds to addition of matrices.
  - ▶ Scalar multiplication corresponds to multiplying all entries of the matrix by the same scalar.
- ▶ The dimension of this vector space is  $mn$  as there are precisely that many linearly independent elements in an arbitrary  $m \times n$  matrix.
- ▶ Let  $a \in [1, m]$  and  $b \in [1, n]$ . Let  $e_{a,i}$  be the  $m \times n$  matrix which has only one non-zero element at location  $(a, i)$  in the matrix. These  $mn$  matrices provide a basis for  $\mathbb{F}^{m,n}$ .

# The space of linear maps is an LVS.

Fix the vector spaces  $\mathbb{V}$  and  $\mathbb{W}$  and consider **all** linear maps between them. Call that space  $\mathbb{T}$ . The following two (obvious?) operations convert it into an LVS.

- ▶ Let  $T_1, T_2 \in \mathbb{T}$  be two linear maps. Define the linear map, “ $T_1 + T_2$ ”, that is the addition of linear maps as follows:

$$“T_1 + T_2”(u) := T_1(u) + T_2(u) \text{ for all } u \in \mathbb{V} .$$

- ▶ Let  $T \in \mathbb{T}$  and  $a \in \mathbb{F}$ . The linear map corresponding to scalar multiplication, “ $a \cdot T$ ” is defined to be

$$“a \cdot T”(u) := aT(u) \text{ for all } u \in \mathbb{V} .$$

Check that  $\mathbb{T}$  is indeed a vector space of  $\mathbb{F}$ .

# The matrix of a linear operator

Recall that if we pick a basis for a vector space  $\mathbb{V}$  (resp.  $\mathbb{W}$ ), we can identify it with  $\mathbb{F}^n$  (resp.  $\mathbb{F}^m$ ). Given a linear map,  $T$ , we can then associate a  $m \times n$  matrix with it.

- ▶ Let  $\mathcal{B}_V = (e_1, e_2, \dots, e_n)$  be a basis for  $\mathbb{V}$  and  $\mathcal{B}_W = (f_1, f_2, \dots, f_m)$  be a basis for  $\mathbb{W}$ .
- ▶ Let  $T(\mathcal{B}_V) = (T(e_1), T(e_2), \dots, T(e_n))$ . This is a set of vectors in  $\mathbb{W}$ . Then, we can write

$$T(\mathcal{B}_V) = \mathcal{B}_W \cdot \begin{pmatrix} T_{11} & T_{12} & \cdots & T_{1n} \\ T_{21} & T_{22} & \cdots & T_{2n} \\ \vdots & \ddots & \ddots & \vdots \\ T_{m1} & T_{m2} & \cdots & T_{mn} \end{pmatrix} = \mathcal{B}_W \cdot \mathbf{T}$$

where  $T(e_j) = \sum_{a=1}^m f_a T_{aj}$ .

- ▶ Thus we obtain a matrix  $\mathbf{T} \in \mathbb{F}^{m,n}$  for every linear map.

# Change of bases

- ▶ Let  $\mathcal{B}_V = \mathcal{B}'_V \cdot S$  and  $\mathcal{B}_W = \mathcal{B}'_W \cdot \tilde{S}$  be basis changes in  $\mathbb{V}$  and  $\mathbb{W}$ .
- ▶ How does the matrix  $\mathbf{T}$  of a linear map change?

$$T(\mathcal{B}_V) = \mathcal{B}_W \cdot \mathbf{T} \quad \text{definition of } \mathbf{T}$$

$$T(\mathcal{B}'_V \cdot S) = \mathcal{B}'_W \cdot \tilde{S} \cdot \mathbf{T} \quad \text{change of bases}$$

$$T(\mathcal{B}'_V) \cdot S = \mathcal{B}'_W \cdot \tilde{S} \cdot \mathbf{T} \quad \text{linearity of } T$$

Multiplying both sides by  $S^{-1}$  to the right, we get

$$T(\mathcal{B}'_V) = \mathcal{B}'_W \cdot \tilde{S} \cdot \mathbf{T} \cdot S^{-1} := \mathcal{B}'_W \cdot \mathbf{T}'$$

- ▶ Under change of bases, the matrix transforms as

$$\boxed{\mathbf{T}' = \tilde{S} \cdot \mathbf{T} \cdot S^{-1}} .$$



# Two linear maps of interest

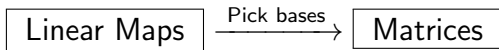
There are two linear maps that are of interest in QM.

- ▶ The first one is when  $\mathbb{W} = \mathbb{F}$ .
  - ▶ Such linear maps are called **linear functions**.
  - ▶ The LVS of linear functions on  $\mathbb{V}$  is called the dual space (to  $\mathbb{V}$ ) and is denoted by  $\mathbb{V}^*$ .
  - ▶  $\dim(\mathbb{V}^*) = \dim(\mathbb{V}) = n$ .
- ▶ The second one is when  $\mathbb{W} = \mathbb{V}$ .
  - ▶ We will call such maps **linear operators**. The  $\theta$ -polarizer is linear operator with  $\mathbb{V} = \mathbb{C}^2$ .
  - ▶ The dimension of the LVS of linear operator is  $n^2$  and the matrix is a square  $n \times n$  matrix.
  - ▶ Under a change of basis with  $\tilde{S} = S$ , one has

$$\mathbf{T}' = S \cdot \mathbf{T} \cdot S^{-1} .$$

- ▶ Note that  $\det(\mathbf{T})$  and  $\text{Tr}(\mathbf{T})$  are invariant under change of basis.

# Summary and Outlook



- ▶ Under change of bases, the matrix transforms as

$$\mathbf{T}' = \tilde{S} \cdot \mathbf{T} \cdot S^{-1} .$$

**Remark:** In your maths course, you must have studied row operations that bring a matrix  $\mathbf{T}$  to its row reduced echelon form,  $\mathbf{T}'$ . Verify that row operations correspond to  $S = 1$  with non-trivial  $\tilde{S}$ .

- ▶  $\mathbb{V}^*$  is the space of linear functions on  $\mathbb{V}$ .
- ▶ Linear operators are linear maps from a LVS to itself. Under change of basis,

$$\mathbf{T}' = S \cdot \mathbf{T} \cdot S^{-1} .$$

In quantum mechanics, the matrix  $S$  is used to make  $\mathbf{T}'$  a diagonal matrix for a certain class of linear operators.

Out of syllabus material!

# Subspaces from linear maps

Given a linear map,  $T : \mathbb{V} \rightarrow \mathbb{W}$ , we can construct two subspaces, one each of  $\mathbb{V}$  and  $\mathbb{W}$ .

- ▶ The first one is called  $\ker(T)$  (short for kernel of  $T$ ). It is a subspace of  $\mathbb{V}$  and defined as follows:

$$\ker(T) = \left\{ \mathbf{u} \in \mathbb{V} \mid T(\mathbf{u}) = \mathbf{0} \right\}$$

**Example** Consider  $d/dx : \mathcal{P}_n \rightarrow \mathcal{P}_m$  for some  $m \geq (n - 1)$ . It is easy to see that  $\ker(d/dx) = \text{Span}(1) = \mathbb{F}$

- ▶ The second one is called  $\text{Im}(T)$  (short for image of  $T$ ). It is a subspace of  $\mathbb{W}$  and is defined as follows:

$$\text{Im}(T) = \left\{ \mathbf{w} \in \mathbb{W} \mid T(\mathbf{u}) = \mathbf{w} \text{ for some } \mathbf{u} \in \mathbb{V} \right\}$$

**Example:**  $\text{Im}(d/dx) = \text{Span}(1, x, x^2, \dots, x^{n-1}) = \mathcal{P}_{n-1}$ .