# MATH 311 Topics in Applied Mathematics I

Lecture 18:

Change of basis (continued).

Linear transformations.

#### **Basis and coordinates**

If  $\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n\}$  is a basis for a vector space V, then any vector  $\mathbf{v} \in V$  has a unique representation

$$\mathbf{v} = x_1 \mathbf{v}_1 + x_2 \mathbf{v}_2 + \cdots + x_n \mathbf{v}_n,$$

where  $x_i \in \mathbb{R}$ . The coefficients  $x_1, x_2, \ldots, x_n$  are called the **coordinates** of **v** with respect to the ordered basis  $\mathbf{v}_1, \mathbf{v}_2, \ldots, \mathbf{v}_n$ .

The mapping

vector 
$$\mathbf{v} \mapsto its coordinates (x_1, x_2, \dots, x_n)$$

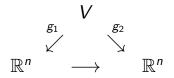
is a one-to-one correspondence between V and  $\mathbb{R}^n$ . This correspondence respects linear operations in V and in  $\mathbb{R}^n$ .

#### Change of coordinates

Let V be a vector space of dimension n.

Let  $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n$  be a basis for V and  $g_1 : V \to \mathbb{R}^n$  be the coordinate mapping corresponding to this basis.

Let  $\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_n$  be another basis for V and  $g_2 : V \to \mathbb{R}^n$  be the coordinate mapping corresponding to this basis.



The composition  $g_2 \circ g_1^{-1}$  is a transformation of  $\mathbb{R}^n$ . It has the form  $\mathbf{x} \mapsto U\mathbf{x}$ , where U is an  $n \times n$  matrix.

U is called the **transition matrix** from  $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n$  to  $\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_n$ . Columns of U are coordinates of the vectors  $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n$  with respect to the basis  $\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_n$ .

# **Problem.** Find the transition matrix from the basis $\mathbf{v}_1 = (1,2,3)$ , $\mathbf{v}_2 = (1,0,1)$ , $\mathbf{v}_3 = (1,2,1)$ to the basis $\mathbf{u}_1 = (1,1,0)$ , $\mathbf{u}_2 = (0,1,1)$ , $\mathbf{u}_3 = (1,1,1)$ .

It is convenient to make a two-step transition: first from  $\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3$  to  $\mathbf{e}_1, \mathbf{e}_2, \mathbf{e}_3$ , and then from  $\mathbf{e}_1, \mathbf{e}_2, \mathbf{e}_3$  to  $\mathbf{u}_1, \mathbf{u}_2, \mathbf{u}_3$ .

Let  $U_1$  be the transition matrix from  $\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3$  to  $\mathbf{e}_1, \mathbf{e}_2, \mathbf{e}_3$  and  $U_2$  be the transition matrix from  $\mathbf{u}_1, \mathbf{u}_2, \mathbf{u}_3$  to  $\mathbf{e}_1, \mathbf{e}_2, \mathbf{e}_3$ :

$$U_1 = \begin{pmatrix} 1 & 1 & 1 \\ 2 & 0 & 2 \\ 3 & 1 & 1 \end{pmatrix}, \qquad U_2 = \begin{pmatrix} 1 & 0 & 1 \\ 1 & 1 & 1 \\ 0 & 1 & 1 \end{pmatrix}.$$

Basis  $\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3 \implies$  coordinates  $\mathbf{x}$ Basis  $\mathbf{e}_1, \mathbf{e}_2, \mathbf{e}_3 \implies$  coordinates  $U_1\mathbf{x}$ 

Basis  $\mathbf{u}_1, \mathbf{u}_2, \mathbf{u}_3 \Longrightarrow \text{coordinates } U_2^{-1}(U_1\mathbf{x}) = (U_2^{-1}U_1)\mathbf{x}$ 

Thus the transition matrix from  $\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3$  to  $\mathbf{u}_1, \mathbf{u}_2, \mathbf{u}_3$  is  $U_2^{-1}U_1$ .

$$U_2^{-1}U_1 = \begin{pmatrix} 1 & 0 & 1 \\ 1 & 1 & 1 \\ 0 & 1 & 1 \end{pmatrix}^{-1} \begin{pmatrix} 1 & 1 & 1 \\ 2 & 0 & 2 \\ 3 & 1 & 1 \end{pmatrix}$$

 $= \begin{pmatrix} 0 & 1 & -1 \\ -1 & 1 & 0 \\ 1 & -1 & 1 \end{pmatrix} \begin{pmatrix} 1 & 1 & 1 \\ 2 & 0 & 2 \\ 3 & 1 & 1 \end{pmatrix} = \begin{pmatrix} -1 & -1 & 1 \\ 1 & -1 & 1 \\ 2 & 2 & 0 \end{pmatrix}.$ 

#### **Linear mapping** = linear transformation = linear function

Definition. Given vector spaces  $V_1$  and  $V_2$ , a mapping  $L: V_1 \to V_2$  is **linear** if  $\boxed{ L(\mathbf{x} + \mathbf{y}) = L(\mathbf{x}) + L(\mathbf{y}),}$   $\boxed{ L(r\mathbf{x}) = rL(\mathbf{x}) }$ 

for any  $\mathbf{x}, \mathbf{y} \in V_1$  and  $r \in \mathbb{R}$ .

A linear mapping  $\ell: V \to \mathbb{R}$  is called a **linear** functional on V.

If  $V_1 = V_2$  (or if both  $V_1$  and  $V_2$  are functional spaces) then a linear mapping  $L: V_1 \to V_2$  is called a **linear operator**.

#### **Linear mapping** = linear transformation = linear function

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for any  $\mathbf{x}, \mathbf{y} \in V_1$  and  $r \in \mathbb{R}$ .

Remark. A function  $f : \mathbb{R} \to \mathbb{R}$  given by f(x) = ax + b is a linear transformation of the vector space  $\mathbb{R}$  if and only if b = 0.

# Basic properties of linear transformations

Let  $L: V_1 \to V_2$  be a linear mapping.

- $L(r_1\mathbf{v}_1 + \cdots + r_k\mathbf{v}_k) = r_1L(\mathbf{v}_1) + \cdots + r_kL(\mathbf{v}_k)$ for all k > 1,  $\mathbf{v}_1, \dots, \mathbf{v}_k \in V_1$ , and  $r_1, \dots, r_k \in \mathbb{R}$ .
- $L(r_1\mathbf{v}_1 + r_2\mathbf{v}_2) = L(r_1\mathbf{v}_1) + L(r_2\mathbf{v}_2) = r_1L(\mathbf{v}_1) + r_2L(\mathbf{v}_2),$   $L(r_1\mathbf{v}_1 + r_2\mathbf{v}_2 + r_3\mathbf{v}_3) = L(r_1\mathbf{v}_1 + r_2\mathbf{v}_2) + L(r_3\mathbf{v}_3) =$ 
  - =  $r_1L(\mathbf{v}_1) + r_2L(\mathbf{v}_2) + r_3L(\mathbf{v}_3)$ , and so on. •  $L(\mathbf{0}_1) = \mathbf{0}_2$ , where  $\mathbf{0}_1$  and  $\mathbf{0}_2$  are zero vectors in
  - $V_1$  and  $V_2$ , respectively.  $L(\mathbf{0}_1) = L(0\mathbf{0}_1) = 0L(\mathbf{0}_1) = \mathbf{0}_2$ .
    - $L(-\mathbf{v}) = -L(\mathbf{v})$  for any  $\mathbf{v} \in V_1$ .

$$L(-\mathbf{v}) = L((-1)\mathbf{v}) = (-1)L(\mathbf{v}) = -L(\mathbf{v}).$$

# **Examples of linear mappings**

- Scaling  $L: V \to V$ ,  $L(\mathbf{v}) = s\mathbf{v}$ , where  $s \in \mathbb{R}$ .  $L(\mathbf{x} + \mathbf{y}) = s(\mathbf{x} + \mathbf{y}) = s\mathbf{x} + s\mathbf{y} = L(\mathbf{x}) + L(\mathbf{y})$ ,  $L(r\mathbf{x}) = s(r\mathbf{x}) = r(s\mathbf{x}) = rL(\mathbf{x})$ .
  - Dot product with a fixed vector  $\ell : \mathbb{R}^n \to \mathbb{R}, \ \ell(\mathbf{v}) = \mathbf{v} \cdot \mathbf{v}_0, \ \text{where } \mathbf{v}_0 \in \mathbb{R}^n.$

$$\ell(\mathbf{x} + \mathbf{y}) = (\mathbf{x} + \mathbf{y}) \cdot \mathbf{v}_0 = \mathbf{x} \cdot \mathbf{v}_0 + \mathbf{y} \cdot \mathbf{v}_0 = \ell(\mathbf{x}) + \ell(\mathbf{y}),$$
  

$$\ell(r\mathbf{x}) = (r\mathbf{x}) \cdot \mathbf{v}_0 = r(\mathbf{x} \cdot \mathbf{v}_0) = r\ell(\mathbf{x}).$$

- Cross product with a fixed vector  $L: \mathbb{R}^3 \to \mathbb{R}^3$ ,  $L(\mathbf{v}) = \mathbf{v} \times \mathbf{v}_0$ , where  $\mathbf{v}_0 \in \mathbb{R}^3$ .
- Multiplication by a fixed matrix  $L: \mathbb{R}^n \to \mathbb{R}^m$ ,  $L(\mathbf{v}) = A\mathbf{v}$ , where A is an  $m \times n$  matrix and all vectors are column vectors.

#### Linear mappings of functional vector spaces

- Evaluation at a fixed point
- $\ell: F(\mathbb{R}) \to \mathbb{R}, \ \ell(f) = f(a), \text{ where } a \in \mathbb{R}.$ 
  - Multiplication by a fixed function
- $L:F(\mathbb{R}) o F(\mathbb{R}),\ L(f)=gf,\ ext{where}\ g\in F(\mathbb{R}).$
- Differentiation  $D: C^1(\mathbb{R}) \to C(\mathbb{R})$ , L(f) = f'. D(f+g) = (f+g)' = f' + g' = D(f) + D(g), D(rf) = (rf)' = rf' = rD(f).
  - Integration over a finite interval
- $\ell: C(\mathbb{R}) \to \mathbb{R}, \ \ell(f) = \int_a^b f(x) \, dx$ , where  $a, b \in \mathbb{R}, \ a < b$ .

### More properties of linear mappings

- If a linear mapping  $L:V\to W$  is invertible then the inverse mapping  $L^{-1}:W\to V$  is also linear.
- If  $L: V \to W$  and  $M: W \to X$  are linear mappings then the composition  $M \circ L: V \to X$  is also linear.
- If  $L_1: V \to W$  and  $L_2: V \to W$  are linear mappings then the sum  $L_1 + L_2$  is also linear.

## **Linear differential operators**

• an ordinary differential operator

$$L: C^{\infty}(\mathbb{R}) \to C^{\infty}(\mathbb{R}), \quad L=g_0\frac{d^2}{dx^2}+g_1\frac{d}{dx}+g_2,$$

where  $g_0, g_1, g_2$  are smooth functions on  $\mathbb{R}$ .

That is,  $L(f) = g_0 f'' + g_1 f' + g_2 f$ .

• Laplace's operator  $\Delta: C^\infty(\mathbb{R}^2) \to C^\infty(\mathbb{R}^2)$ ,

$$\Delta f = \frac{\partial^2 f}{\partial x^2} + \frac{\partial^2 f}{\partial y^2}$$

(a.k.a. the Laplacian; also denoted by  $\nabla^2$ ).

### Linear integral operators

anti-derivative

$$L: C[a,b] \to C^{1}[a,b], \ (Lf)(x) = \int_{-\infty}^{\infty} f(y) \, dy.$$

Hilbert-Schmidt operator

$$L: C[a,b] \rightarrow C[c,d], \ (Lf)(x) = \int_a^b K(x,y)f(y) \, dy,$$
 where  $K \in C([c,d] \times [a,b]).$ 

Laplace transform

$$\mathcal{L}:BC(0,\infty)\to C(0,\infty),\ (\mathcal{L}f)(x)=\int_0^\infty e^{-xy}f(y)\,dy.$$

*Examples.*  $\mathcal{M}_{m,n}(\mathbb{R})$ : the space of  $m \times n$  matrices.

• 
$$\alpha: \mathcal{M}_{m,n}(\mathbb{R}) \to \mathcal{M}_{n,m}(\mathbb{R}), \quad \alpha(A) = A^T.$$

$$\alpha(A+B) = \alpha(A) + \alpha(B) \iff (A+B)^T = A^T + B^T.$$
  
 $\alpha(rA) = r \alpha(A) \iff (rA)^T = rA^T.$ 

Hence  $\alpha$  is linear.

• 
$$\beta: \mathcal{M}_{2,2}(\mathbb{R}) \to \mathbb{R}, \ \beta(A) = \det A.$$

Let 
$$A = \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix}$$
 and  $B = \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix}$ .

Then 
$$A + B = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$$
.

We have  $\det(A) = \det(B) = 0$  while  $\det(A + B) = 1$ . Hence  $\beta(A + B) \neq \beta(A) + \beta(B)$  so that  $\beta$  is not linear.

#### Range and kernel

Let V, W be vector spaces and  $L: V \rightarrow W$  be a linear mapping.

*Definition.* The **range** (or **image**) of L is the set of all vectors  $\mathbf{w} \in W$  such that  $\mathbf{w} = L(\mathbf{v})$  for some  $\mathbf{v} \in V$ . The range of L is denoted L(V).

The **kernel** of L, denoted ker L, is the set of all vectors  $\mathbf{v} \in V$  such that  $L(\mathbf{v}) = \mathbf{0}$ .

**Theorem** (i) The range of L is a subspace of W. (ii) The kernel of L is a subspace of V.