

Chapter 3: Linear transformations

Linear transformations, Algebra of
linear transformations, matrices,
dual spaces, double duals

Linear transformations

- V, W vector spaces with same fields F .
 - **Definition:** $T:V \rightarrow W$ s.t. $T(ca+db)=c(Ta)+dT(b)$ for all a,b in V . c,d in F . Then T is **linear**.
 - Property: $T(O)=O$. $T(ca+db)=cT(a)+dT(b)$, a,b in V , c,d in F . (equivalent to the def.)
 - **Example:** A $m \times n$ matrix over F . Define T by $Y=AX$. $T:F^n \rightarrow F^m$ is linear.
 - Proof: $T(aX+bY)=A(aX+bY)=aAX+bAY = aT(X)+bT(Y)$.

- $U: F^{1 \times m} \rightarrow F^{1 \times n}$ defined by $U(a) = aA$ is linear.
- Notation: $F^m = F^{m \times 1}$ (not like the book)
- Remark: $L(F^{m \times 1}, F^{n \times 1})$ is same as $M_{m \times n}(F)$.
 - For each $m \times n$ matrix A we define a unique linear transformation T given by $T(X) = AX$.
 - For each a linear transformation T has A such that $T(X) = AX$. We will discuss this in section 3.3.
 - Actually the two spaces are isomorphic as vector spaces.
 - If $m = n$, then compositions correspond to matrix multiplications exactly.

- Example: $T(x)=x+4$. $F=\mathbb{R}$. $V=\mathbb{R}$. This is not linear.
- Example: $V = \{f \text{ polynomial}: F \rightarrow F\}$
 $T:V \rightarrow V$ defined by $T(f)=Df$.

$$\begin{aligned} f(x) &= c_0 + c_1x + c_2x^2 + \cdots + c_kx^k \\ Df(x) &= c_1 + 2c_2x + \cdots + kc_kx^{k-1} \end{aligned}$$

- $V=\{f:\mathbb{R}\rightarrow\mathbb{R} \text{ continuous}\}$

$$Tf(x) = \int_0^x f(t)dt$$

- **Theorem 1:** V vector space over F .
basis $\alpha_1, \dots, \alpha_n$. W another one with
vectors β_1, \dots, β_m (any kind $m \geq n$). Then
exists a **unique** linear transformation
 $T: V \rightarrow W$ s.t. $T(\alpha_j) = \beta_j, j = 1, \dots, n$
- **Proof:** Check the following map is linear.

$$\begin{aligned}\alpha &= x_1\alpha_1 + \dots + x_n\alpha_n \\ T\alpha &= x_1\beta_1 + \dots + x_n\beta_n\end{aligned}$$

- **Null space** of $T : V \rightarrow W := \{ v \text{ in } V \mid Tv = 0 \}$.
- **Rank** $T := \dim\{Tv \mid v \text{ in } V\}$ in W . = $\dim \text{range } T$.
- Null space is a vector subspace of V .
- Range T is a vector subspace of W .

- Example:

$$\begin{pmatrix} 1 & 2 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} x \\ y \\ z \\ t \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}$$

- Null space $z=t=0$. $x+2y=0$ $\dim = 1$
- Range = W . $\dim = 3$

- Theorem: $\text{rank } T + \text{nullity } T = \dim V$.
- Proof: $\mathbf{a}_1, \dots, \mathbf{a}_k$ basis of N . $\dim N = k$.
Extend to a basis of V : $\mathbf{a}_1, \dots, \mathbf{a}_k, \mathbf{a}_{k+1}, \dots, \mathbf{a}_n$.
 - We show $T \mathbf{a}_{k+1}, \dots, T \mathbf{a}_n$ is a basis of R . Thus $n-k = \dim R$. $n-k+k=n$.
 - Spans R :

$$\begin{aligned} v &= x_1 \alpha_1 + \dots + x_n \alpha_n \\ Tv &= x_{k+1} T(\alpha_{k+1}) + \dots + x_n T(\alpha_n) \end{aligned}$$
 - Independence:

$$\begin{aligned} \sum_{i=k+1}^n c_i T \alpha_i &= 0 \\ T(\sum_{i=k+1}^n c_i \alpha_i) &= 0 \\ \sum_{i=k+1}^n c_i \alpha_i &\in N \\ \sum_{i=k+1}^n c_i \alpha_i &= \sum_{i=1}^k c_i \alpha_i \\ c_i &= 0, i = k+1, \dots, n \end{aligned}$$

- Theorem 3: A $m \times n$ matrix.
Row rank $A =$ Column rank A .
- Proof:
 - column rank $A =$ rank T where $T: \mathbb{R}^n \rightarrow \mathbb{R}^m$ is defined by $Y=AX$. e_i goes to i -th column. So range is spanned by column vectors.
 - rank $T +$ nullity $T = n$ by above theorem.
 - column rank $A +$ dim $S = n$ where $S = \{X | AX=O\}$ is the null space.
 - dim $S = n -$ row rank A (Example 15 Ch. 2 p.42)
 - row rank = column rank.

- (Example 15 Ch. 2 p.42) $A^{m \times n}$. S solution space. R r - r - e matrix
- r = number of nonzero rows of R .
- $RX=0$ $k_1 < k_2 < \dots < k_r$. $J = \{1, \dots, n\} - \{k_1, k_2, \dots, k_r\}$.

$$\begin{array}{rcl}
 x_{k_1} & + & \sum_{j=1}^{n-r} C_{1j} u_j = 0 \\
 & & x_{k_2} + \sum_{j=1}^{n-r} C_{2j} u_j = 0 \\
 & & \dots + \vdots = \vdots \\
 & & x_{k_r} + \sum_{j=1}^{n-r} C_{rj} u_j = 0
 \end{array}$$

- Solution spaces parameter u_1, \dots, u_{n-r} .
- Or basis E_j given by setting $u_j = 1$ and other $u_i = 0$ and $x_{k_i} = c_{ij}$.

Algebra of linear transformations

- Linear transformations can be added, and multiplied by scalars. Hence they form a vector space themselves.
- Theorem 4: $T, U: V \rightarrow W$ linear.
 - Define $T+U: V \rightarrow W$ by $(T+U)(a) = T(a) + U(a)$.
 - Define $cT: V \rightarrow W$ by $cT(a) = c(T(a))$.
 - Then they are linear transformations.

- **Definition:** $L(V,W)=\{T:V\rightarrow W \mid T \text{ is linear}\}$.
- **Theorem 5:** $L(V,W)$ is a finite dim vector space if so are V,W . $\dim L = \dim V \dim W$.
- **Proof:** We find a basis: $\mathcal{B} = \{\alpha_1, \dots, \alpha_n\} \subset V$
 $\mathcal{B}' = \{\beta_1, \dots, \beta_m\} \subset W$
 - Define a linear transformation $V \rightarrow W$:

$$E^{p,q}(\alpha_i) = \begin{cases} 0, & i \neq q \\ \beta_p, & i = q \end{cases} = \delta_{iq} \beta_p, \quad 1 \leq p \leq m, 1 \leq q \leq n$$

- The basis:

$$\begin{matrix} E^{1,1}, & \dots, & E^{1,n} \\ \vdots & \ddots & \vdots \\ E^{m,1}, & \dots, & E^{m,n} \end{matrix}$$

– Spans: $T:V \rightarrow W$. $T\alpha_j = \sum_{p=1}^m A_{pj}\beta_p$

- We show

$$T = U = \sum_{p=1}^m \sum_{q=1}^n A_{pq} E^{p,q}$$

$$U(\alpha_j) = \sum_{p=1}^m \sum_{q=1}^n A_{p,q} E^{p,q}(\alpha_j)$$

$$= \sum_{p=1}^m \left(\sum_{q=1}^n A_{p,q} \delta_{j,q} \right) (\beta_p)$$

$$= \sum_{p=1}^m A_{pj} \beta_p = T\alpha_j, j = 1, \dots, m$$

$$T = U$$

– Independence

- Suppose

$$\begin{aligned}
 U &= \sum_p \sum_q A_{pq} E^{p,q} = 0 \\
 U \alpha_j &= 0 \\
 \sum_p A_{pj} \beta_p &= 0 \\
 \{\beta_p\} & \text{ independent} \\
 A_{pj} &= 0 \text{ for all } p, j
 \end{aligned}$$

- **Example:** $V = F^m$ $W = F^n$. Then

- $M_{m \times n}(F)$ is isomorphic to $L(F^m, F^n)$ as vector spaces. Both dimensions equal mn .
- $E^{p,q}$ is the $m \times n$ matrix with 1 at (p,q) and 0 everywhere else.
- Any matrix is a linear combination of $E^{p,q}$.

- **Theorem.** $T:V \rightarrow W$, $U:W \rightarrow Z$.
 $UT:V \rightarrow Z$ defined by $UT(a) = U(T(a))$ is linear.
- **Definition:** Linear operator $T:V \rightarrow V$.
- $L(V, V)$ has a multiplication.
 - **Define** $T^0 = I$, $T^n = T \dots T$. n times.
 - **Example:** A $m \times n$ matrix B $p \times m$ matrix
 T defined by $T(X) = AX$. U defined by $U(Y) = BY$.
 Then $UT(X) = BAX$. Thus
 UT is defined by BA if T is defined by A and U by
 B .
 - **Matrix multiplication is defined to mimic composition.**

- Lemma:
 - $IU=UI=U$
 - $U(T_1+T_2)=UT_1+UT_2, (T_1+T_2)U=T_1U+T_2U.$
 - $c(UT_1)= (cU)T_1=U(cT_1).$
- Remark: This make $L(V,V)$ into linear algebra (i.e., vector space with multiplications) in fact same as the matrix algebra $M_{n \times n}(F)$ if $V=F^n$ or more generally $\dim V = n.$ (Example 10. P.78)

- **Example:** $V = \{f: F \rightarrow F \mid f \text{ is a polynomial}\}$.
 - $D: V \rightarrow V$ differentiation.

$$\begin{aligned} f(x) &= c_0 + c_1x + \cdots + c_nx^n \\ Df(x) &= c_1 + \cdots + nc_nx^{n-1} \end{aligned}$$

- $T: V \rightarrow V$: T sends $f(x)$ to $xf(x)$
- $DT - TD = I$. We need to show $DT - TD(f) = f$ for each polynomial f .
- (QP-PQ=ih! In quantum mechanics.)

Invertible transformations

- $T:V \rightarrow W$ is **invertible** if there exists $U:W \rightarrow V$ such that $UT=I_V$ $TU=I_W$. U is denoted by T^{-1} .
- **Theorem 7:** If T is linear, then T^{-1} is linear.
- **Definition:** $T:V \rightarrow W$ is nonsingular if $Tc=0$ implies $c=0$
 - Equivalently the null space of T is $\{0\}$.
 - T is one to one.
- **Theorem 8:** T is nonsingular iff T carries each linearly independent set to a linearly independent set.

- **Theorem 9:** V, W $\dim V = \dim W$.
 $T:V \rightarrow W$ is linear. TFAE:
 - T is invertible.
 - T is nonsingular
 - T is onto.
- **Proof:** We use $n = \dim V = \dim W$.
 $\text{rank } T + \text{nullity } T = n$.
 - (ii) iff (iii): T is nonsingular iff nullity $T = 0$ iff rank $T = n$ iff T is onto.
 - (I) \rightarrow (ii): $TX=0, T^{-1}TX=0, X=0$.
 - (ii) \rightarrow (i): T is nonsingular. T is onto. T is 1-1 onto. The inverse function exists and is linear. T^{-1} exists.

Groups

- A **group** (G, \cdot) :
 - A set G and an operation $G \times G \rightarrow G$:
 - $x(yz) = (xy)z$
 - There exists e s.t. $xe = ex = x$
 - To each x , there exists x^{-1} s.t. $xx^{-1} = e$ and $x^{-1}x = e$.
- Example: The set of all 1-1 maps of $\{1, 2, \dots, n\}$ to itself.
- Example: The set of nonsingular maps $GL(V, V)$ form a group.

Isomorphisms

- V, W $T:V \rightarrow W$ one-to-one and onto (invertible). Then T is an **isomorphism**. V, W are **isomorphic**.
- Isomorphic relation is an equivalence relation: $V \sim V$, $V \sim W \Leftrightarrow W \sim V$, $V \sim W$, $W \sim U \rightarrow V \sim W$.

- **Theorem 10:** Every n-dim vector space over F is isomorphic to F^n .
(noncanonical)
- Proof: V n-dimensional
 - Let $B = \{a_1, \dots, a_n\}$ be a basis.
 - Define $T: V \rightarrow F^n$ by
$$\alpha = x_1 a_1 + \dots + x_n a_n \mapsto (x_1, \dots, x_n) \in F^n$$
 - One-to-one
 - Onto

- Example: isomorphisms (F a subfield of R)

$$F^n = \{(x_1, \dots, x_n) \mid x_i \in F\}$$

$$\cong \left\{ \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix} \mid x_i \in F \right\}$$

$$P^n(F) = \{f : F \rightarrow F \mid f(x) = c_0 + c_1x + \dots + c_nx^n\}$$

$$\cong F^{n+1}$$

Basis $\{1, x, x^2, \dots, x^n\}$

$$c_0 + c_1x + \dots + c_nx^n \mapsto (c_0, c_1, \dots, c_n)$$

There will be advantages in looking this way!