Linear Mappings

- Linear mappings

 Eigenvalues and Eigenvectors.

Linear Mappings

Linear mappings are functions defined on vector spaces that preserve linear combinations.

Definition

T: mapping, transformation.

Given two vector spaces V and W, we say that T: V-W is a linear mapping if it verifies: Yū, ve V and Yxel? (a) $T(\bar{u}+\bar{v}) = T(\bar{u}) + T(\bar{v})$ $\bar{u}+\bar{v} \in V$ (b) $T(\alpha \bar{u}) = \alpha T(\bar{u})$. $T(\bar{u}), T(\bar{v}) \in W$

Example. The mapping
$$T: \mathbb{R}^2 \to \mathbb{R}^2$$

defined by $T\left(\begin{array}{c} X \\ y \end{array} \right) = \begin{pmatrix} y \\ y \end{array}$ is linear.

 $\bar{u}: \begin{pmatrix} u_1 \\ u_2 \end{pmatrix}$, $\bar{v}: \begin{pmatrix} v_1 \\ v_2 \end{pmatrix}$, $T(\bar{u}) = \begin{pmatrix} v_2 \\ v_1 \end{pmatrix}$
 $T\left(\alpha \bar{u} + \beta \bar{v} \right) = --- = \alpha T(\bar{u}) + \beta T(\bar{v})$
 $T(\alpha \bar{u} + \beta \bar{v}) = T\left(\begin{pmatrix} \alpha u_1 + \beta v_1 \\ \alpha u_2 + \beta v_2 \end{pmatrix} \right) = \begin{pmatrix} \alpha u_2 + \beta v_2 \\ \alpha u_1 + \beta v_1 \end{pmatrix}$
 $= \alpha \begin{pmatrix} u_2 \\ u_1 \end{pmatrix} + \beta \begin{pmatrix} v_2 \\ v_2 \end{pmatrix} = \alpha T(\bar{u}) + \beta T(\bar{v})$

Example. The mapping $D: \mathbb{P} \to \mathbb{P}$ defined by D p(x) = p'(x) is linear.

$$D[\alpha p + \beta q] = (\alpha p + \beta q)' = \alpha p' + \beta q'$$

$$= \alpha Dp + \beta Dq$$

Example. Let A be a matrix of size $m \times n$. The mapping $T: \mathbb{R}^n \to \mathbb{R}^m$ defined by $T(\bar{x}) = A\bar{x}$ is linear.

$$A(\bar{x}+\bar{y}) = A\bar{x}+A\bar{y}$$
 $A(\alpha\bar{x}+\beta\bar{y})$
 $A(\alpha\bar{x}) = \alpha A\bar{x}$ $= \alpha A\bar{x}+\beta A\bar{y}$

$$T[(x)] = (y) = (0 1 (x) 1 0)(y)$$

$$T(x) = Ax$$

- · Using the definition
- · Showing that it can be written as Az

Example. The mapping
$$f: \mathbb{R}^2 \to \mathbb{R}^2$$
 defined by $f\left(\begin{pmatrix} x \\ y \end{pmatrix} \right) = \begin{pmatrix} y \\ x^2 \end{pmatrix}$ is not linear.

scalar multiplication.

$$f\left[\begin{array}{c} 5\left(2\right) \\ 1\end{array}\right] = f\left[\begin{array}{c} 10 \\ 5\end{array}\right] = \left(\begin{array}{c} 5 \\ 100\end{array}\right) = \left(\begin{array}{c} 5 \\ 20\end{array}\right)$$

$$f\left[S\left(\frac{2}{1}\right)\right] \neq 5f\left[\left(\frac{2}{1}\right)\right]$$

Properties of linear mappings Let T:V->W be a linear mapping,

then:

Example. The mapping
$$f: \mathbb{R}^2 \to \mathbb{R}^2$$
 defined

by
$$f\left(\begin{pmatrix} x \end{pmatrix}\right) = \begin{pmatrix} x+1 \end{pmatrix}$$
 is not linear

since
$$f\left(\begin{pmatrix} 0\\0\end{pmatrix}\right) = \begin{pmatrix} 1\\0\end{pmatrix} \neq \begin{pmatrix} 0\\0\end{pmatrix}$$
.

Kernel and Image of a linear mapping

Definition

Let $T: V \rightarrow W$ be a linear mapping. We define the kernel and image of T, respectively, as follows: $\ker T = \{ \bar{x} \in V, \ T(\bar{x}) = \bar{o} \} \subseteq V$ $\operatorname{Im} T = \{ T(\bar{x}) \in W, \ \bar{x} \in V \} \cdot \subseteq W$

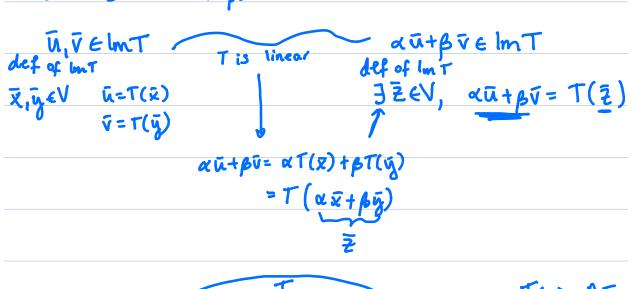
= { y ∈ W, y=T(x) for some x ∈ V}

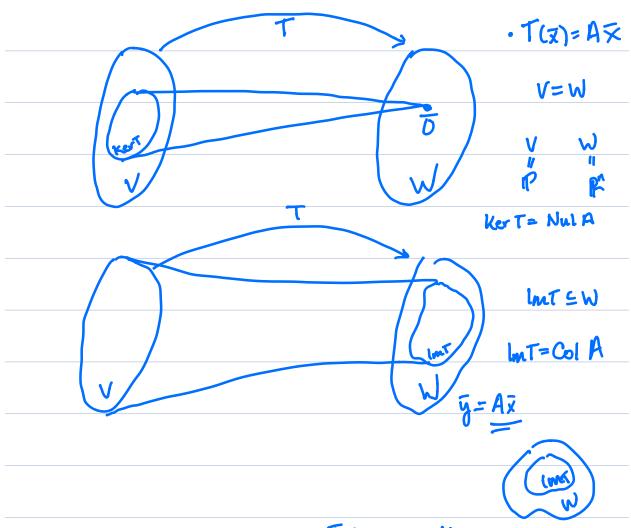
- ▶ Ker T is a subspace of V
- In T is a subspace of W.

KerT is a subspace:

$$\bar{x}, \bar{y} \in \ker T$$
 $\times \bar{x} + \beta \bar{y} \in \ker T$
 $\cdot T(\bar{x}) = \bar{0}$ $T(\alpha \bar{x} + \beta \bar{y}) = \kappa T(\bar{x}) + \beta T(\bar{y})$
 $\cdot T(\bar{y}) = \bar{0}$ $= \kappa \bar{0} + \beta \bar{0} = \bar{0}$

InT is a subspace.





T is surjectives InT=W

Theorem

Let $T:V \rightarrow W$ be a linear mapping and let $\{\bar{u}_1, \bar{u}_2, ..., \bar{u}_n\}$ be a system of generators of V. Then $\{T(\bar{u}_i), T(\bar{u}_2), ..., T(\bar{u}_n)\}$ is a system of generators of Im.

V=Span { ūnūz,..., ūn } -> lm T=Span { T(ū1), T(ūz),..., Ttuiw}

Example. Find the kernel and image of the linear mapping $f: \mathbb{R}^3 \to \mathbb{R}^3$

defined by
$$f\left(\begin{array}{c} X \\ Y \\ Z \end{array}\right) = \begin{pmatrix} X + Z \\ Y \\ X + 2y + Z \end{pmatrix}$$

$$f(x) = \begin{cases} X \\ Y \\ Y \\ Z \end{cases}$$

$$f(x) = \begin{cases} X + Z \\ Y \\ X + 2y + Z \end{cases}$$

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$$A \sim \begin{pmatrix} 1 & 0 & 1 \\ 0 & 2 & 0 \\ 0 & 0 & 0 \end{pmatrix} \qquad \begin{array}{c} x + z = 0 \\ y = 0 \\ z = \lambda \end{array}$$

Nul A = Ker
$$f = Span$$

Col A = Imf = Span | L | 6 | 1 |

Span $\{f(\bar{u}_1), f(\bar{u}_2), f(\bar{u}_2)\}$

= Span | Col A = Imf = Span | L | 6 | 1 |

Span $\{f(\bar{u}_1), f(\bar{u}_2), f(\bar{u}_2)\}$

= Span | Col A = Imf = Imf = Span | Col A = Imf = Imf = Span | Col A = Imf = Imf = Span | Col A = Imf = I

If we apply the previous theorem
$$\bar{u}_1 = \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}, \quad \bar{u}_2 = \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix}, \quad \bar{u}_3 = \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}$$

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$$f(\bar{u}_1) = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$$
 $f(\bar{u}_2) = \begin{pmatrix} 0 \\ 1 \end{pmatrix}$ $f(\bar{u}_3) = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$

T(2)=Ax

$$| \mathbf{m} T = \left\{ T(\bar{\mathbf{x}}) : \bar{\mathbf{x}} \in V \right\} = \left\{ A \bar{\mathbf{x}} : \bar{\mathbf{x}} \in V \right\}$$

$$= \left\{ x_1 \bar{a}_1 + x_2 \bar{a}_2 + \dots + x_n \bar{a}_n : \bar{\mathbf{x}} = \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix} \in V$$

= Col A

$$Ker T = \left\{ \overline{X} \in V : T(\overline{X}) = \overline{0} \right\} = \left\{ \overline{X} \in V : A\overline{X} = \overline{0} \right\}$$

$$= Nul A.$$

Injective, surjective, and bijective mapping

A function f: A→B is injective
if and only if

 $\forall x_1 y \in A$, if $x \neq y$ then $f(x) \neq f(y)$, or equivalently

 $\forall x_1 y \in A$, if f(x) = f(y) then x = y.

It is injective iif f(x) = y has at most 1 sol. $\forall y$

A function $f:A \rightarrow B$ is surjective if and only if

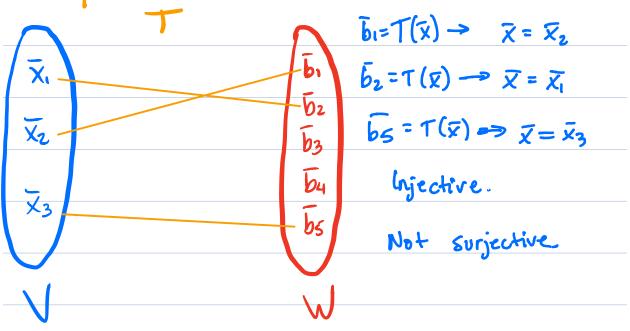
Y be B, there exists a eA such that b=f(a).

f is surjective iff f(z)= g has at least 1 sol. Yg

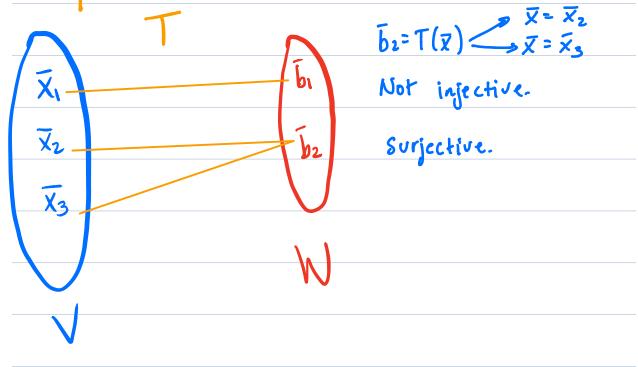
A function is bijective if and only if it is both injective and surjective.

f is bijective iff f(x)=y always has a unique solution.

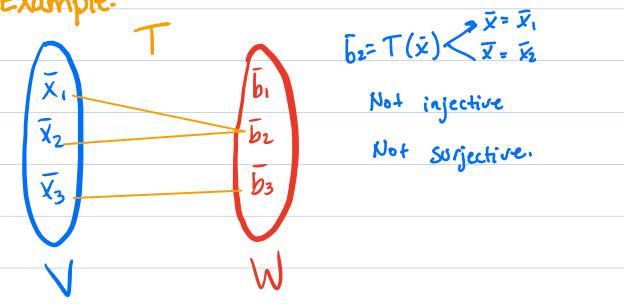
Example.



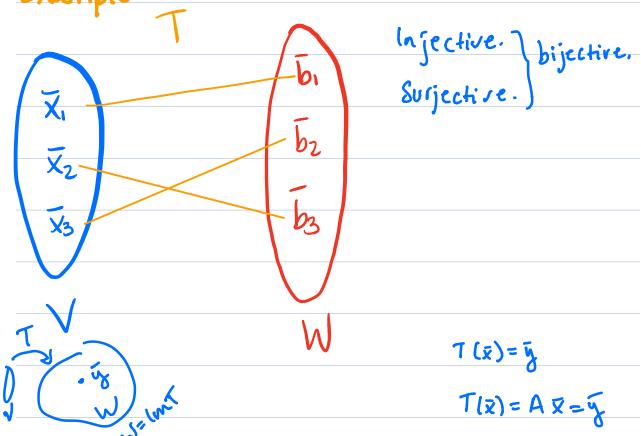
Example.



Example.



Example.



Theorem

Let T: V->W be a linear mapping. Then:

1. T is injective if and only if KerT= 203.

2. T is surjective if and only if ImT= W.

Other characterizations

For any given linear mapping $f:V \to W$, we have:

1. f is injective if and only if for each linearly independent set {ū,ū, ..., ūn}, the set {f(ū,), f(ūz), ..., f(ūn)} is linearly independent.

f is injective (=> f preserves lin. independence

> Suppose that f is injective.

NTS: if ¿ūi,...,ūn j is lin. ind.

then ¿f(ūi),...,f(ūn) } lin. ind.

 $c_1f(\bar{u}_1) + c_2f(\bar{u}_2) + --- + c_nf(\bar{u}_n) = \bar{b}_w$ $f\left(c_1\bar{u}_1 + c_2\bar{u}_2 + --\cdot + c_n\bar{u}_n\right) = \bar{b}_w$ \bar{b}_v

fis injective > CuitCzuzt --- + Cnun = ov {u, --, u, } is liniad > C= Cz = Cz = --- Cn = 0

X= 5

- 2. f is surjective if and only if for each system of generators of $v \{\bar{u}_1, \bar{u}_2, ..., \bar{u}_n\}$, the set $\{f(\bar{u}_1), f(\bar{u}_2), ..., f(\bar{u}_n)\}$ is a system of generators of W.
- 3. f is bijective if and only if for each basis of V lū,ūz,...,ūm], the set f(ū,), f(ūz),..., f(ūm)] is a basis of W.

Operations with linear mappings

Given two linear mappings $f,g:V\to W$ and a scalar $\lambda \in \mathbb{R}$, we define the

operations $f+g:V\to W$; $(f+g)(\bar{x})=f(\bar{x})+g(\bar{x})$ $\lambda f:V\to W$; $(\lambda f)(\bar{x})=\lambda f(\bar{x})$

V, W { T linear mapping: T: V->w}

With the operations defined above, the set of linear mappings between two vector spaces V and W is itself a vector space.

Moreover, the set of linear mappings between V and W has additional structure:

丁(文)=百

Theorem

T(v) = 0x

Given two linear mappings $f:V\to W$ and $g:W\to U$, their composition $g\circ f:V\to U$ defined by $(g\circ f)(x)=g(f(x))$ is also a linear mapping. Recall that if a function is bijective then it has an inverse function.

In particular, if $f:V\to W$ is a bijective linear mapping, then there exists a function $f^{-1}:W\to V$ such that $V\bar{v}\in V$, $f^{-1}(f(\bar{v}))=\bar{v}=Id_{V}$ $V\to W\to V$ $V\bar{w}\in W$, $f(f^{-1}(\bar{w}))=\bar{w}$. $W\to V\to W\to V$

Theorem

For each bijective linear function $f:V\to W$, its inverse mapping $f^{-1}:W\to V$ is also a linear mapping. That is, $f^{-1}(\alpha\bar{u}+\beta\bar{v})=\alpha f^{-1}(\bar{u})+\beta f^{-1}(\bar{v})$, $\forall \bar{u},\bar{v}\in W$ $\forall \alpha,\beta\in \mathbb{R}$.



Linear mappings and matrices.

In this section, we will see that linear mappings between finite-dimensional vector spaces have a matrix representation. The mechanism to achieve this matrix representation consists in using coordinates with respect to some basis of V and W.

More concretely, let $T:V\to W$ be a linear mapping, let B be a basis of V, and let C be a mapping of W. For any vector $\overline{x} \in V$, the image of \overline{x} under T is $T\overline{x} \in W$. The coordinates of \overline{x} and $T\overline{x}$ with respect to the corresponding bases are $[\overline{x}]_B$ and $[T\overline{x}]_C$. We will see that there exists a matrix $M_T^{C,B}$ such that

 $[T\bar{x}]_c = M_T^{GB} [\bar{x}]_B. \qquad T(\bar{x})$

The matrix $M_T^{C,B}$ is the matrix representation of T when we fix the bases B and C. This matrix converts the coordinates $[\bar{x}]_B$ into the coordinates $[\bar{x}]_C$.

The notation MT has been chosen to remark that the matrix represention of T depends of the bases B and C. In other words, it changes with the choice of bases.

Matrix associated to a linear mapping.

In this section, we will construct the matrix associated to the linear mapping T: V->W.

We start by fixing a basis for V and W. Let B be a basis for V and let C be a basis for W. In particular, let $B = \{\bar{b}_1, \bar{b}_2, \dots, \bar{b}_n\}$.

Then we can represent every xeV as a linear combination of B using the coordinates [x]. That is,

 $\bar{x} = x_1 \bar{b}_1 + x_2 \bar{b}_2 + x_3 \bar{b}_3 + \cdots + x_n \bar{b}_n$; $[\bar{x}]_{g} = \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix}$.

Now, using the linearity of T, we have, $T\bar{x}=T(x_1\bar{b}_1+x_2\bar{b}_2+x_3\bar{b}_3+\cdots+x_n\bar{b}_n)$

= $X_1T(b_1) + X_2T(b_2) + X_3T(b_3) + \cdots + X_nT(b_n)$.

Since Tx is a vector of W, we can take the coordinates of Tx with respect to the basis C.

Therefore,

[$T\bar{x}$] $c = x_1 [T(\bar{b}_1)]_c + x_2 [T(\bar{b}_2)]_c + \cdots + x_n [T(\bar{b}_n)]_c$ This last equation can be written as a matrix equation: [$T\bar{x}$] $c = ([T(\bar{b}_1)]_c [T(\bar{b}_2)]_c ... [T(\bar{b}_n)]_c)[\bar{x}]_B$. Comparing with the equation $[T\bar{x}]_c = M_T^{c,tb}[\bar{x}]_B$ at the beginning of the section, we see that $M_T^{c,tb} = ([T(\bar{b}_1)]_c [T(\bar{b}_2)]_c ... [T(\bar{b}_n)]_c)$.

Matrix equation of a linear mapping.

Let V and W be two vector spaces of dimensions n and m, respectively, and let B and C be bases of V and W, respectively.

Given a linear mapping $T:V\to W$, the matrix equation of T with respect to B and C is $[\vec{y}]_c = [T\vec{x}]_c = M_T^{GB}[\vec{x}]_B \implies \vec{T}(\vec{x}) = \vec{y}$

which, given the coordinates [x] of a vector xeV with respect to B, computes the coordinates [Tx]c of its image Tx with respect to C.

 M_T^{GB} is the matrix associated to T with respect to B and C_1 that is: the matrix of size $M \times M_T^{GB} = (T(\bar{b}_1)]_c [T(\bar{b}_2)]_c \dots [T(\bar{b}_n)]_c)$.

The notation $M_T^{c,tb}$ has been chosen so that, in the equation $[T\bar{x}]_{c} = M_T^{c,b} [\bar{x}]_{b}$, everything indexed with B is collected on the right, and everything indexed with C is collected on the left.

$$T: \mathbb{R}^2 \to \mathbb{R}^2$$

$$T_{\lambda}(\overline{x}) = \overline{x} \qquad \begin{bmatrix} x \\ y \end{bmatrix} \to \begin{bmatrix} y \\ x \end{bmatrix} T : V \to V$$

$$\varepsilon_2 \qquad \varepsilon_2 \qquad \varepsilon_3$$

A special case is when V=W. Nevertheless,

B and C may not be the same bases.

So a second special case when V=W and B=C. $T(\bar{x})=A\bar{x}$ E=C

Example. Let $f: \mathbb{R}^2 \to \mathbb{R}^3$ be a linear mapping such that

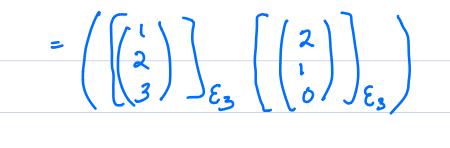
$$\frac{f(1)}{0} = \begin{pmatrix} 1 \\ 2 \end{pmatrix}, \quad f(0) = \begin{pmatrix} 2 \\ 1 \end{pmatrix}.$$

$$\begin{pmatrix} 3 \\ 3 \end{pmatrix}$$

We would like to write the matrix associated to f with respect to Ez and Ez, the canonical bases of R² and R³, respectively.

$$\mathcal{E}_{3}, \mathcal{E}_{2}$$

$$\mathsf{M}_{p} = \left(\left[f(\hat{0}) \right]_{\mathcal{E}_{3}} \left[f(\hat{0}) \right]_{\mathcal{E}_{3}} \right)$$





Example. Let $D: \mathbb{P}_3 \to \mathbb{P}_3$ be the linear mapping defined by Dp(x) = p'(x). Consider the canonical basis $B = \{1, x, x^2, x^3\}$. Write the matrix associated with D with respect to B (the second special case discussed above).

$$M_{D}^{B_{1}B} = \begin{pmatrix} \begin{bmatrix} D \downarrow \end{bmatrix}_{B} & \begin{bmatrix} D \chi \end{bmatrix}_{B} & \begin{bmatrix} D \chi^{2} \end{bmatrix}_{B} & \begin{bmatrix} D \chi^{3} \end{bmatrix}_{B} \end{pmatrix}$$

$$= \begin{pmatrix} \begin{bmatrix} 0 \end{bmatrix}_{B} & \begin{bmatrix} 1 \end{bmatrix}_{B} & \begin{bmatrix} 2 \chi \end{bmatrix}_{B} & \begin{bmatrix} 3 \chi^{2} \end{bmatrix}_{B} \end{pmatrix}$$

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

$$\bar{x} = 1 + 3x^2$$
 $T\bar{x} = D(1 + 3x^2) = 6x$

$$[Tx]_B = M_0^{B,B} [\bar{x}]_B$$
derivative of $\bar{x} \in \mathbb{P}_3$

$[T_{\overline{x}}]_{B} =$	Ó	1	0	0 \	/1 \	
	0	0	2	0	0	
	0	D	0	3	3	
	0	D	0	0)	0	

$$\begin{bmatrix}
T\bar{x} \\
B = 0
\end{bmatrix} \Rightarrow 6x$$

Example. Let T: B > P4 be the linear mo	pping						
defined by $Tp(x) = xp(x)$. Consider the bases $B = \{1, 1+x, x+x^2, x^2+x^3\}$ and							
	•						

Associated matrix: Injectivity and Surjectivity
In this section, we will relate some properties
of a linear mapping with some properties
of its associated matrix.

Consider the linear mapping $T:V \to W$ between to finite-dimensional vector spaces. Since we can associate a matrix with T, we will use the tools developed for matrices to study the mapping T.

For this, we need to fix a basis for V: B=?b, bz,...,bn3.

We will also need to fix a basis for W: C.

With these two bases, we can construct MT, the matrix associated with T, such that

[$T\bar{x}$]_c= M_T^{GB} [\bar{x}]_B, $\forall \bar{x} \in V$.

Recall that the matrix M_T^{GB} is given by: $M_T^{GB} = ([T(\bar{b}_1)]_c [T(\bar{b}_2)]_c ... [T(\bar{b}_n)]_c)$

We have the following facts strated previously:

- Studying a set of vectors is equivalent to studying its coordinates.
- Since B is a linearly independent set,

 T is injective if and only if

 {T(b_1), T(b_2), ..., T(b_n)} is a linearly

 independent set.

- Since B is a generating system of V,

 T is surjective if and only if

 IT(bi), T(bi), ..., T(bn)} is a generating

 system of W.
- First, we will study the relationship between the injectivity of T and the pivot columns of M_T. We know that T is injective if and only if $\{[T(\bar{b}_1)]_c, [T(\bar{b}_2)]_c, [T(\bar{b}_n)]_c\}$ is linearly independent. But these vectors are the columns of M_T. Therefore, we have

T is injective if and only if the columns of Mr are linearly independent, equivalently, all the columns are pivots.

- Tis injective if and only if [b]c=MTEx]B,
 belmT, has a unique solution.
- Clearly, the number of pivots in MTB is independent of the chosen bases B and C.
- We can relate the injectivity of M7,8 with the rank of M7

rank $M_T^{c,B} = \#$ pivots = # columns of $M_T^{c,B} = \dim V$ by injectivity of T,

all the columns are pivots.

Example. Let $f: \mathbb{R}^2 \to \mathbb{R}^3$ be a linear mapping such that

$$\frac{f\left(1\right)}{0} = \begin{pmatrix} 1\\2\\3 \end{pmatrix}, \quad f\left(0\right) = \begin{pmatrix} 2\\1\\0 \end{pmatrix}.$$

We have already found the associated matrix with respect to the canonical bases:

We can see that all the columns of $M_{f}^{\epsilon_{3},\epsilon_{2}}$ are pivots (equivalently, rank $M_{f}^{\epsilon_{3},\epsilon_{2}}$ =dim \mathbb{R}^{2} =2), therefore f is injective.

Example. Let $D: \mathbb{P}_3 \to \mathbb{P}_3$ be the linear mapping defined by Dp(x) = p'(x). As we have previously seen, the associated matrix with respect to the standard basis B is

$$M_{D}^{20/18} = \begin{pmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 2 & 0 \\ 0 & 0 & 0 & 3 \\ 0 & 0 & 0 & 0 \end{pmatrix}$$

This matrix only has 3 pivots and one nonpivot column. Therefore D is not injective. Equivalently, D is not injective because rank $M_D^{8,8} = 3 \neq dim P_3 = 4$

Example. Let $T: \mathbb{P}_3 \to \mathbb{P}_4$ be the linear mapping defined by Tp(x) = xp(x). Consider the bases $\mathcal{B} = \{1, 1+x, x+x^2, x^2+x^3\}$ and $C = \{1, x, x^2, x^3, x^4\}$ of \mathbb{P}_3 and \mathbb{P}_4 , respectively.

The associated matrix is

In this case, every column of $M_T^{C_1B}$ is a pinot and therefore T is injective. Equivalently,

T is injective because

rank $M_T^{C_1B} = 4 = dim P_3$

Now, we will study the relationship between the surjectivity of T and the pivots of M_T^{C,B}. We know that T is surjective if and only if I[T(b̄1)]c, [T(b̄2)]c,..., [T(b̄n)]c] is a system of generators of R^m, where m=dim W. In other words, the matrix equation

[b]_c = $M_T^{c,B}$ [x]_B, be W Should always have a solution for any beW. For this to be true, each row of $M_T^{c,B}$ must have a pivot in order to avoid contradictions of the form 0=1. T is surjective if and only if the columns of $M_T^{C,B}$ constitutes a system of generators of \mathbb{R}^m , $m = \dim W$, equivalently, each row of $M_T^{C,B}$ has a pivot.

- T is surjective if and only if the equation [b]c=Mtc,B[x]B always has a solution for any beW.
- We can relate the surjectivity of T with the rank of MT

rank Mr = # pivots = # rows of Mr = dim W.

by the surjectivity

of T, each row has

a pivot.

Example. Let $f: \mathbb{R}^2 \to \mathbb{R}^3$ be a linear mapping such that

$$\frac{f\left(\frac{1}{0}\right) = \begin{pmatrix} 1\\2\\3 \end{pmatrix}}{f\left(\frac{0}{1}\right) = \begin{pmatrix} 2\\1\\0 \end{pmatrix}}.$$

We have already found the associated matrix with respect to the canonical bases:

$$M_{\frac{2}{4}}^{\epsilon_{3},\epsilon_{2}} = \begin{pmatrix} 1 & 2 & 1 & 0 \\ 2 & 1 & \sim & 0 & 1 \\ 3 & 0 & 0 & 0 \end{pmatrix}$$

We can see that not every row of $M_T^{E_3,E_2}$ has a pivot, therefore f is not surjective.

Equivalently, f is not surjective because rank $M_T^{E_3,E_2} = 2 \neq \dim \mathbb{R}^3 = 3$

Example. Consider the linear mapping $T: \mathbb{R}^5 \to \mathbb{R}^3$ defined by $T(\bar{x}) = A\bar{x}$ where

$$A = \begin{pmatrix} 1 & 4 & 7 & 5 & 11 \\ 2 & 5 & 8 & 7 & 13 \\ 3 & 0 & 10 & 9 & 10 \end{pmatrix}.$$

Clearly, the matrix associated to Twith respect to the canonical bases is A.

Since every now of A has a pivot, we have that T is surjective. Equivalently, T is surjective since

rank A= 3 = dim R3

- To finish this section, it must be remarked that T is bijective if and only if every row and every column of MTB has a pivot. For this, it is necessary that MTB is square and invertible.
- Moreover, a mapping can be both not injective and not surjective. An example of this is the mapping $D: \mathbb{R}_3 \to \mathbb{R}_3$ defined by Dp(x) = p'(x).

Associated matrix and change of bases.

Let V and W be two vector spaces, and let T: V > W be a fixed, but arbitrary, linear mapping.

Recall that the matrix associated with T depends on the bases chosen for V and W.

Nevertheless, the linear mapping should not depend on our choice of bases.

This implies that there is a relation between two matrices associated with T but with respect to distinct bases.

Indeed, these relations exist and constitute the center of attention of this section.

Recall that V and W are vector spaces and the linear mapping T: V -> W is fixed but arbitrary.

Let's choose two distinct bases for V: B and \widetilde{B} , and two distinct bases for W: C and \widetilde{C} .

Then, we have the following relations between coordinates:

where bet and cec are the corresponding matrices of change of bases.

Recall that $M_T^{c,B}$ denotes the matrix associated with T with respect to B and C, and $M_T^{c,B}$ denotes the matrix associated with T with respect to B and C.

We can conveniently summarize the relation between these matrices with the following diagram

 $W \leftarrow T V$

$$\begin{bmatrix} \overline{w} \end{bmatrix}_{c} = M_{T}^{c,B} \begin{bmatrix} \overline{v} \end{bmatrix}_{B}$$

$$\tilde{c} \leftarrow c$$

Our goal is to find a relation between $M_T^{c,B}$ and $M_T^{\varepsilon,B}$. To do this, we proceed as follow:

Consider the matrix equation associated with T with respect to the bases B and C:

Note that $[T\overline{v}]_{c} = cP\overline{c} [T\overline{v}]_{c}$ an $[\overline{v}]_{b} = pP\overline{c} [\overline{v}]$. Substituting these relations in the above equation, we obtain

Lastly, multiplying both sides of this equation by P' = P we obtain $c \in \tilde{c} \in \tilde{c} \in c$

$$[T\bar{v}]_{\tilde{c}} = P M_{T}^{c,\delta} P_{[\bar{v}]_{\tilde{g}}}.$$

Comparing this last equation with [To]= MT [v]E, we deduce

$$M_T^{\widetilde{c},\widetilde{b}} = P \quad M_T^{CB} \quad P$$

$$\widetilde{c} \leftarrow c \quad B \leftarrow \widetilde{b}$$

Change of bases formula for the matrix associated with a linear mapping T.

This formula is the relation between $M_T^{C,B}$ and $M_T^{C,B}$ that we were booking for.

Observe how the notation can help us remember the formula above as follows:

On the left hand side of the formula we have $M_T^{\tilde{c},\tilde{b}}$ with superindeces \tilde{c} and \tilde{b} . On the right hand side, these same indices are on the exterior, while B and C appear on the interior. Moreover, notice that the subindex B y below the superindex C, and similarly, the subindex C y below the superindex C.

We can also use the diagram

to remember how to write the change of bases formula. Notice that the arrow connecting directly B and C is labeled MT. Also notice that B and C can be connected alternatively via: $\widetilde{B} \rightarrow B \rightarrow C \rightarrow \widetilde{C}$. Writing (from right to left) the labels of this alternative path we get the MT P Since the direct path and the alternative path connect the same two vertices B and c we will say that they are equal: MT = P MT P, obtaining the desired change of bases formula.

Example. Let $f: \mathbb{R}^2 \to \mathbb{R}^3$ be a linear mapping such that

$$\frac{f\left(1\right)}{0} = \begin{pmatrix} 1\\2\\3 \end{pmatrix}, \quad f\left(0\right) = \begin{pmatrix} 2\\1\\1 \end{pmatrix}.$$

We have already found the associated matrix with respect to the canonical bases:

$$M_{+}^{\varepsilon_{3},\varepsilon_{2}} = \begin{pmatrix} 1 & 2 \\ 2 & 1 \\ 3 & 0 \end{pmatrix}$$

Now consider the following bases for \mathbb{R}^2 and \mathbb{R}^3 :

$$B = \begin{cases} (1) & (1) & (1) & (2) & (1) & (1) & (0) & (1)$$

We	will	compu	ite the	. matri	ix associated	with
1 W	oith r	respect	to B	and	C .	

Example. Consider the linear mapping $T: \mathbb{R}^5 \to \mathbb{R}^3$ defined by $T(\overline{x}) = A \overline{x}$ where

$$A = \begin{pmatrix} 1 & 4 & 7 & 5 & 11 \\ 2 & 5 & 8 & 7 & 13 \\ 3 & 6 & 10 & 9 & 16 \end{pmatrix}.$$

Clearly, the matrix associated to Twith respect to the canonical bases \mathcal{E}_{S} and \mathcal{E}_{S} is $M_{T}^{\mathcal{E}_{S},\mathcal{E}_{S}}A$. Consider the bases of \mathbb{R}^{S} and \mathbb{R}^{3}

$$B = \begin{cases} 1 & 0 & 0 & -1 & 0 \\ 0 & 1 & 0 & -1 & -1 \\ 0 & 0 & 1 & 0 & -1 \\ 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 1 \end{cases}$$

$$C = \begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix} \begin{pmatrix} 4 \\ 5 \\ 6 \end{pmatrix} \begin{pmatrix} 7 \\ 8 \\ 10 \end{pmatrix}$$

The matrix associated with T with
respect to B and C is given by