

1. $[e] = (\vec{v}_1, \vec{v}_2, \dots, \vec{v}_n)$: ordered collection in a vector space V

(a) $[e]$: $\mathbb{R}^n \rightarrow V$ is the linear combination function, defined as follows: (domain \mathbb{R}^n , codomain V)

$$[e] \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix} = x_1 \vec{v}_1 + x_2 \vec{v}_2 + \dots + x_n \vec{v}_n$$

↑ CORRESPONDING L.C. OF e
COEFFICIENT VECTOR

(b) $[e]$ is a linear transformation from \mathbb{R}^n to V :

① (show $[e]\vec{o} = \vec{o}$)

$$[e] \begin{bmatrix} 0 \\ \vdots \\ 0 \end{bmatrix} = 0\vec{v}_1 + 0\vec{v}_2 + \dots + 0\vec{v}_n = \vec{o} + \dots + \vec{o} = \vec{o} \quad \checkmark$$

② (show that $\forall \vec{x} \in \mathbb{R}^n$ and $\alpha \in \mathbb{R}$, $[e](\alpha \vec{x}) = \alpha [e]\vec{x}$)

let $\vec{x} \in \mathbb{R}^n$ and $\alpha \in \mathbb{R}$ be given; then $\vec{x} = \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix}$, where $x_1, \dots, x_n \in \mathbb{R}$.

$$\begin{aligned} \text{now, } [e](\alpha \vec{x}) &= [e](\alpha \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix}) = [e] \begin{bmatrix} \alpha x_1 \\ \vdots \\ \alpha x_n \end{bmatrix} \\ &= (\alpha x_1) \vec{v}_1 + (\alpha x_2) \vec{v}_2 + \dots + (\alpha x_n) \vec{v}_n \\ &= \alpha(x_1 \vec{v}_1 + x_2 \vec{v}_2 + \dots + x_n \vec{v}_n) \\ &= \alpha [e] \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix} = \alpha [e]\vec{x} \quad \checkmark \end{aligned}$$

③ (show that $\forall \vec{x}, \vec{y} \in \mathbb{R}^n$, $[e](\vec{x} + \vec{y}) = [e]\vec{x} + [e]\vec{y}$)

let $\vec{x}, \vec{y} \in \mathbb{R}^n$ be given; then $\vec{x} = \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix}$ and $\vec{y} = \begin{bmatrix} y_1 \\ \vdots \\ y_n \end{bmatrix}$, where $x_i, y_i \in \mathbb{R}$.

$$\begin{aligned} \text{now, } [e](\vec{x} + \vec{y}) &= [e] \left(\begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix} + \begin{bmatrix} y_1 \\ \vdots \\ y_n \end{bmatrix} \right) = [e] \begin{bmatrix} x_1 + y_1 \\ \vdots \\ x_n + y_n \end{bmatrix} \\ &= (x_1 + y_1) \vec{v}_1 + (x_2 + y_2) \vec{v}_2 + \dots + (x_n + y_n) \vec{v}_n \\ &= (x_1 \vec{v}_1 + x_2 \vec{v}_2 + \dots + x_n \vec{v}_n) + (y_1 \vec{v}_1 + y_2 \vec{v}_2 + \dots + y_n \vec{v}_n) \\ &= [e] \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix} + [e] \begin{bmatrix} y_1 \\ \vdots \\ y_n \end{bmatrix} = [e]\vec{x} + [e]\vec{y} \quad \checkmark \end{aligned}$$

By ①, ②, and ③, $[e] : \mathbb{R}^n \rightarrow V$ is a linear transformation ■

2. $[e] = (\vec{v}_1, \vec{v}_2, \dots, \vec{v}_n)$: ordered collection in a vector space V

(a) $\text{IM}[e]$ is just the span of e !

$$\text{why? } \text{IM}[e] \stackrel{\text{def}}{=} \{[e]\vec{x} : \vec{x} \in \mathbb{R}^n\}$$

$$= \{[e] \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix} : x_1, \dots, x_n \in \mathbb{R}\}$$

$$= \{x_1 \vec{v}_1 + x_2 \vec{v}_2 + \dots + x_n \vec{v}_n : x_1, \dots, x_n \in \mathbb{R}\}$$

$\stackrel{\text{def}}{=} \text{SPAN } e$ (THE SET OF ALL L.C.'S OF e !)

(b) $\text{Ker}[e]$ gives us the coefficient vectors for all possible linear relations on e .

$$\text{why? } \text{Ker}[e] \stackrel{\text{def}}{=} \{\vec{x} \in \mathbb{R}^n : [e]\vec{x} = \vec{o}\}$$

$$= \left\{ \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix} : x_1, \dots, x_n \in \mathbb{R} \text{ AND } [e] \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix} = \vec{o} \right\}$$

$$= \left\{ \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix} : x_1, \dots, x_n \in \mathbb{R} \text{ AND } x_1 \vec{v}_1 + x_2 \vec{v}_2 + \dots + x_n \vec{v}_n = \vec{o} \right\}$$

↑
THE SET OF ALL COEFFICIENT VECTORS THAT GIVE LINEAR RELATIONS ON e .

3. $[e = (\vec{v}_1, \vec{v}_2, \dots, \vec{v}_n)]$: ordered collection in a vector space V ?

(a) If e is L.I. in V , then $[e]$ is injective,
because $\ker [e]$ will be $\{\vec{0}\}$.

WHY? As noted in 2(b), the vectors of $\ker [e]$ correspond to the linear relations on e ; if e is L.I., then the only L.R. on e is the trivial one, so $\ker [e] = \{\vec{0}\}$, which from the previous problem set we know means that $[e]$ is injective.

ALTERNATIVE EXPLANATION (NOT AS GOOD OF PRACTICE WITH USING OUR NEW TERMINOLOGY): If e is L.I. in V , then each vector of V can be written as a L.C. of e in at most one way, so $[e]$ can hit each $\vec{v} \in V$ at most once — i.e., $[e]$ is injective.

(b) If e spans V , then $[e]$ is surjective,
because $\text{im } [e]$ will be all of V .

WHY? As noted in 2(a), $\text{im } [e] = \text{SPAN } e$; if e spans V , then $\text{im } [e] = \text{SPAN } e = V$, so by definition, $[e]$ is surjective.

ALTERNATIVE EXPLANATION: If e spans V , then each vector of V can be written as a L.C. of e in at least one way, so $[e]$ hits each $\vec{v} \in V$ at least once — i.e., $[e]$ is surjective.

(c) If e is a basis for V , then $[e]$ is bijective.

WHY? If e is a basis for V , then e is L.I. in V and spans V , so by (a) and (b), $[e]$ is both injective and surjective — i.e., $[e]$ is bijective.

ALTERNATIVE EXPLANATION: If e is a basis for V , then each vector of V can be written as a L.C. of e in exactly one way, so $[e]$ pairs up each $\vec{x} \in \mathbb{R}^n$ with one $\vec{v} \in V$ and vice-versa — i.e., $[e]$ is bijective.

* A bijective linear transformation is, by definition, an isomorphism.

4. $[e = (\vec{v}_1, \vec{v}_2, \dots, \vec{v}_n)]$: ordered collection in \mathbb{R}^m ; $[e] \rightsquigarrow$ matrix A

E.g., $e = ([\begin{matrix} 1 \\ 0 \end{matrix}], [\begin{matrix} 4 \\ 1 \end{matrix}], [\begin{matrix} 0 \\ 2 \end{matrix}], [\begin{matrix} -1 \\ 0 \end{matrix}])$ in \mathbb{R}^3 : $[e] \rightsquigarrow \begin{bmatrix} 1 & 4 & 0 & -1 \\ 0 & 1 & 2 & 1 \\ 2 & 5 & 0 & -1 \end{bmatrix} = A$

(a) A has m rows and n columns;

ROWS: SIZE OF
COLUMN VECTORS

COLS: # OF
COLUMN VECTORS

E.G. AS ABOVE: A IS A 3×4 MATRIX.

THUS, IT IS AN m BY n MATRIX.

DIMENSIONS OF THE MATRIX:
ROWS \times COLUMNS

(b) If $\vec{x} \in \mathbb{R}^n$, $A\vec{x} \stackrel{\text{def}}{=} [e]\vec{x}$ is a linear combination of e , i.e., a linear combination of the columns of A (with coefficients given by the entries of \vec{x}).

E.g. AS ABOVE: $\begin{bmatrix} 1 & 4 & 0 & -1 \\ 0 & 1 & 2 & 1 \\ 2 & 5 & 0 & -1 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \\ 2 \\ -1 \end{bmatrix} = 1 \begin{bmatrix} 1 \\ 0 \\ 2 \end{bmatrix} + 0 \begin{bmatrix} 4 \\ 1 \\ 5 \end{bmatrix} + 2 \begin{bmatrix} 0 \\ 2 \\ 0 \end{bmatrix} - 1 \begin{bmatrix} -1 \\ 1 \\ -1 \end{bmatrix} = \begin{bmatrix} 2 \\ 3 \\ 3 \end{bmatrix}$

(c) Just as with $[e]$,
 \downarrow DOMAIN
 \downarrow CODOMAIN
 $A: \mathbb{R}^n \rightarrow \mathbb{R}^m$
 \downarrow
COEFFICIENTS
IN \mathbb{R}^n \mapsto L.C. IN \mathbb{R}^m

E.G. AS ABOVE, $A: \mathbb{R}^4 \rightarrow \mathbb{R}^3$.

[A : $m \times n$ MATRIX ARISING FROM $\mathbf{c} = (\mathbf{v}_1, \dots, \mathbf{v}_n)$ IN \mathbb{R}^m]

5. $N(A) = \{\mathbf{x} \in \mathbb{R}^n : A\mathbf{x} = \mathbf{0}\}$ (i.e., THE KERNEL OF THE L.T. A)

- $N(A)$ IS A SUBSPACE OF \mathbb{R}^n (THE DOMAIN OF THE L.T. A)
- WE CAN FIND A BASIS FOR $N(A)$ SIMPLY BY SOLVING THE HOMOGENEOUS SYSTEM $A\mathbf{x} = \mathbf{0}$ AND SPLITTING OUR SOLUTION AS THE SPAN OF ITS FREE VARIABLES' CONTRIBUTIONS.

E.G. AS ABOVE, TO FIND A BASIS FOR $N(A)$, SOLVE:

$$\begin{bmatrix} q_1 & q_2 & q_3 & q_4 \\ 1 & 4 & 0 & -1 \\ 0 & 1 & 2 & 1 \\ 2 & 5 & 0 & -1 \end{bmatrix} \rightsquigarrow \begin{bmatrix} 1 & 0 & 0 & q_3 \\ 0 & 1 & 0 & -q_3 \\ 0 & 0 & 1 & q_4 \\ 0 & 0 & 0 & q_4 \end{bmatrix} \quad \begin{array}{l} q_4: \text{FREE} \\ q_1 = -4q_3 \\ q_2 = -2q_3 \\ q_3 = -2q_4 \end{array}$$

$$\begin{bmatrix} q_1 \\ q_2 \\ q_3 \\ q_4 \end{bmatrix} = \begin{bmatrix} -4q_4 \\ -2q_4 \\ q_4 \\ q_4 \end{bmatrix} = q_4 \begin{bmatrix} -4 \\ -2 \\ 1 \\ 1 \end{bmatrix}$$

SO $\left\{ \begin{bmatrix} -4 \\ -2 \\ 1 \\ 1 \end{bmatrix} \right\}$ IS A BASIS FOR $N(A)$.

6. $C(A) = \{\mathbf{Ax} : \mathbf{x} \in \mathbb{R}^n\}$ (i.e., THE IMAGE OF THE L.T. A)

- $C(A)$ IS A SUBSPACE OF \mathbb{R}^m (THE CODOMAIN OF THE L.T. A)
- WE CAN FIND A BASIS FOR $C(A)$ BY REDUCING A AND TAKING THE COLUMNS (OF OUR ORIGINAL MATRIX A !) THAT GAVE PIVOTS (JUST AS WE DO WHEN FINDING THE BASIS OF A SPAN).

E.G. AS ABOVE, A BASIS FOR $C(A)$ IS $\left\{ \begin{bmatrix} 1 \\ 0 \\ 2 \end{bmatrix}, \begin{bmatrix} 4 \\ 1 \\ 5 \end{bmatrix}, \begin{bmatrix} 0 \\ 2 \\ 0 \end{bmatrix} \right\}$.

7. (COMPUTING THE RESULT OF A MATRIX ACTING ON A COLUMN VECTOR IS JUST COMPUTING A LINEAR COMBINATION!)

$$(a) \begin{bmatrix} 2 & 1 \\ -3 & 0 \\ -1 & 5 \\ 0 & 3 \end{bmatrix} \begin{bmatrix} 1 \\ -2 \end{bmatrix} = 1 \begin{bmatrix} 2 \\ -3 \\ -1 \\ 0 \end{bmatrix} - 2 \begin{bmatrix} 1 \\ 5 \\ 3 \end{bmatrix} = \begin{bmatrix} 0 \\ -3 \\ -11 \\ -6 \end{bmatrix}$$

$$(b) \begin{bmatrix} 3 & 0 & 1 & 4 \\ -2 & -2 & 1 & 5 \end{bmatrix} \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} = 0 \begin{bmatrix} 3 \\ -2 \end{bmatrix} + 1 \begin{bmatrix} 0 \\ -2 \end{bmatrix} + 2 \begin{bmatrix} 1 \\ 1 \end{bmatrix} + 3 \begin{bmatrix} 4 \\ 5 \end{bmatrix} = \begin{bmatrix} 14 \\ 15 \end{bmatrix}$$

$$(c) \begin{bmatrix} 4 & 1 & 3 & 0 \\ 2 & -2 & 0 & 5 \\ 1 & 0 & 4 & 2 \end{bmatrix} \begin{bmatrix} 1 \\ -1 \\ 1 \\ -1 \end{bmatrix} = 1 \begin{bmatrix} 4 \\ 2 \\ 1 \end{bmatrix} - 1 \begin{bmatrix} 1 \\ -2 \\ 0 \end{bmatrix} + 1 \begin{bmatrix} 3 \\ 0 \\ 4 \end{bmatrix} - 1 \begin{bmatrix} 0 \\ 5 \\ 2 \end{bmatrix} = \begin{bmatrix} 6 \\ -1 \\ 3 \end{bmatrix}$$

8. Ae_j IS JUST THE j^{TH} COLUMN OF A .

↳ ALL 0'S EXCEPT FOR
A 1 IN THE j^{TH} ENTRY

$$(b) A = \begin{bmatrix} 3 & 0 & 1 & 4 \\ -2 & -2 & 1 & 5 \end{bmatrix}; \mathbf{x} = \begin{bmatrix} 0 \\ 1 \\ 2 \\ 3 \end{bmatrix} = 0\vec{e}_1 + 1\vec{e}_2 + 2\vec{e}_3 + 3\vec{e}_4$$

$$A\mathbf{x} = A(0\vec{e}_1 + 1\vec{e}_2 + 2\vec{e}_3 + 3\vec{e}_4)$$

$$= 0 \cdot A\vec{e}_1 + 1 \cdot A\vec{e}_2 + 2 \cdot A\vec{e}_3 + 3 \cdot A\vec{e}_4 \quad \text{BY LINEARITY OF } A$$

$$= 0 \begin{bmatrix} 3 \\ -2 \end{bmatrix} + 1 \begin{bmatrix} 0 \\ -2 \end{bmatrix} + 2 \begin{bmatrix} 1 \\ 1 \end{bmatrix} + 3 \begin{bmatrix} 4 \\ 5 \end{bmatrix} = \begin{bmatrix} 14 \\ 15 \end{bmatrix}, \text{ JUST AS IN 7(b).}$$

9. $[A : mxn \text{ MATRIX}]$

RECALL THAT:

- $N(A) = \{\vec{x} : A\vec{x} = \vec{0}\} \leftrightarrow \text{L.R. is on the columns of } A$
- $C(A) = \{A\vec{x} : \vec{x} \in \mathbb{R}^n\} = \text{SPAN of the columns of } A$

(a) $A\vec{x} = \vec{0}$ only for $\vec{x} = \vec{0}$ tells us that $N(A) = \{\vec{0}\}$, so the only L.R. on the columns of A is the trivial one — i.e., the columns of A are linearly independent.

(b) $A\vec{x} = \vec{0}$ for some $\vec{x} \neq \vec{0}$ tells us that $N(A) \neq \{\vec{0}\}$, so there is a nontrivial L.R. on the columns of A — i.e., the columns of A are linearly dependent.

(c) $A\vec{x} = \vec{b}$ has a solution tells us that $\vec{b} \in C(A)$ — i.e., that \vec{b} is in the span of the columns of A .

(d) $A\vec{x} = \vec{b}$ has no solutions tells us that $\vec{b} \notin C(A)$ — i.e., that \vec{b} is not in the span of the columns of A .

(e) for each $\vec{b} \in \mathbb{R}^m$, the equation $A\vec{x} = \vec{b}$ has a unique solution tells us that every $\vec{b} \in \mathbb{R}^m$ can be written as a L.C. of the columns of A in exactly one way, so the columns of A are a basis for \mathbb{R}^m . (Alternatively, this tells us that when A is reduced, we obtain a pivot in every row + column... so, again, the columns of A form a basis for \mathbb{R}^m).