

Section 1 - Matrix Algebra and Row Reduction

• Elementary Row Operations (ERO's):

- ① Add a row to another row
- ② Multiply a row by some real number
- ③ Swap 2 rows.

These ERO's preserve the solution of a matrix.

• REF & RREF? More like WTF & WWTF!

Actually, no. More like:

A matrix A is in REF if:

- ① All $\vec{0}$ rows are at the bottom
- ② Every row's **leading** non-zero entry is to the right of the row above it

A Matrix A in RREF satisfies ① & ② and

- ③ The leading non-zero entry in each row is 1

- ④ If a column contains a leading 1, all other entries in that column are 0

Every matrix has exactly 1 RREF

Rank of A: $\text{Rank}(A) =$

- (1) # of nonzero rows in RREF
- (2) # of fixed variables
- (3) # of leading 1's in rref

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Matrix * Vector

A matrix A in $\mathbb{R}^{m \times n}$ can only be multiplied by a vector in $\mathbb{R}^n$

(*) m rows & n columns

Suppose $A = \begin{bmatrix} \vec{k}_1 & \vec{k}_2 & \dots & \vec{k}_n \end{bmatrix}$ and $\vec{v} = \begin{bmatrix} v_1 \\ v_2 \\ \vdots \\ v_n \end{bmatrix}$ $v_i \in \mathbb{R}$
 $\vec{k}_i \in \mathbb{R}^m$

Then $A\vec{v} = v_1\vec{k}_1 + v_2\vec{k}_2 + \dots + v_n\vec{k}_n$

And $A\vec{v} \in \mathbb{R}^m$ ($A\vec{v}$ gives a vector in $\mathbb{R}^m$)

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Matrix * Matrix

For two matrices $A \in \mathbb{R}^{m \times n}$ and $B \in \mathbb{R}^{q \times r}$

*Note: if $q \neq n$, then you can't multiply AB .

Suppose $M = AB$

*Note: $M = AB \nrightarrow M = BA$ (If $AB = BA$,
 A "commutes w/"
 B)

Section 2: Linear Transformations

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- A LITw/ domain $\mathbb{R}^n$ & codomain/range $\mathbb{R}^m$ is written " $T: \mathbb{R}^n \rightarrow \mathbb{R}^m$ " and can be given by a matrix $A \in \mathbb{R}^{m \times n}$.
- A transformation is linear $\Leftrightarrow$
 - ① $T(\vec{x} + \vec{v}) = T(\vec{x}) + T(\vec{v})$
 - ② $T(c\vec{x}) = cT(\vec{x})$ for all $c \in \mathbb{R}$
 - ③ $T(\vec{0}) = \vec{0}$

How to find the matrix of a linear transformation T

If $\vec{x} = \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix}$, then $T(\vec{x}) = T\left(\begin{bmatrix} x_1 \\ \vdots \\ 0 \end{bmatrix}\right) + T\left(\begin{bmatrix} 0 \\ \vdots \\ x_2 \end{bmatrix}\right) + \dots + T\left(\begin{bmatrix} 0 \\ \vdots \\ x_n \end{bmatrix}\right)$

$= x_1 T(\vec{e}_1) + \dots + x_n T(\vec{e}_n)$ (B/c T is linear)

★ This looks a lot like Matrix * Vector

$$= \begin{bmatrix} | & | & & | \\ T(\vec{e}_1) & T(\vec{e}_2) & \dots & T(\vec{e}_n) \\ | & | & & | \end{bmatrix} \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix}$$

This is A !

$$\text{So } A = \begin{bmatrix} | & | & & | \\ T(\vec{e}_1) & T(\vec{e}_2) & \dots & T(\vec{e}_n) \\ | & | & & | \end{bmatrix}$$

* $\vec{e}_i$ is the unit vector in the direction of the i th axis. $\vec{e}_1 = \begin{bmatrix} 1 \\ 0 \\ \vdots \end{bmatrix}$ $\vec{e}_2 = \begin{bmatrix} 0 \\ 1 \\ \vdots \end{bmatrix}$ $\vec{e}_3 = \begin{bmatrix} 0 \\ 0 \\ 1 \\ \vdots \end{bmatrix}$ etc.

How to find A^{-1}

$A \in \mathbb{R}^{m \times n}$ is invertible $\Leftrightarrow$

① $m=n$

② $\text{rank}(A) = m = n \Leftrightarrow \text{ref}(A) = I_n^*$

* $I_n \in \mathbb{R}^{n \times n}$ such that $I_n = \begin{bmatrix} 1 & & 0 \\ & \ddots & \\ 0 & & 1 \end{bmatrix}$

$A = I_n A = A I_n$

To find A^{-1}

- Augment A by I_n to get $[A | I_n]$
- Row reduce (everything done to the left should also be done to the right, as always)
- You should get $[I_n | ?]$
- $? = A^{-1}$

$I_n \in \mathbb{R}^{2 \times 2}$

$$A = \begin{bmatrix} a & b \\ c & d \end{bmatrix} \Rightarrow A^{-1} = \frac{1}{ad-bc} \begin{bmatrix} d & -b \\ -c & a \end{bmatrix}$$

What does it mean?

$$A^{-1}A = AA^{-1} = I_n$$

So that's pretty cool...



Subspaces of $\mathbb{R}^n$

Def: $W \subset \mathbb{R}^n$ is a subspace $\Leftrightarrow$

- ① If $\vec{x} \in W$, then $c\vec{x} \in W$ for all $c \in \mathbb{R}$
- ② If $x, y \in W$ then $x+y \in W$
- ③ $\vec{0} \in W$

What is this "span" you speak of?

$$\begin{aligned} \text{Span} \{ \vec{x}_1, \vec{x}_2, \dots, \vec{x}_n \} &= \{ c_1 \vec{x}_1 + \dots + c_n \vec{x}_n \mid c_i \in \mathbb{R} \} \\ &= \{ \text{set of } \underline{\text{ALL}} \text{ linear combos of } \vec{x}_1, \dots, \vec{x}_n \} \end{aligned}$$

Kernel

$$\begin{aligned} \text{Ker}(A) &= \{ \vec{x} \in \mathbb{R}^m \mid A\vec{x} = T(\vec{x}) = \vec{0} \} \\ &= \text{set of } \underline{\text{ALL}} \vec{x}'\text{s in } T\text{'s domain} \\ &\quad \text{that return } \vec{0} \text{ as output} \end{aligned}$$

How to find it:

- ① Since ERO's preserve sol'n, find $\text{RREF}(A)$
- ② Find general sol'n of $[\text{RREF}(A) \mid \vec{0}]$
 $= c_1 \vec{x}_1 + \dots + c_k \vec{x}_k$ (See pg 3)
- ③ $\text{ker}(A) = \text{Span} \{ \vec{x}_1, \vec{x}_2, \dots, \vec{x}_k \}$

Image of $T(\vec{x}) = A\vec{x}$

$$\text{Image} = \{ \vec{x} \in \mathbb{R}^n \mid \text{For some } \vec{v} \in \mathbb{R}^m, T(\vec{v}) = \vec{x} \}$$

Theory

$A(\vec{x})$ is always some linear combination of the column vectors of A , so strictly speaking

$\text{Im}(A) = \text{span} \{ \text{col vectors of } A \}$
is correct, BUT we want a more economical way.

How-To

- Like normal, find $\text{rref}(A)$
- The leading 1's point out independent columns, so every column w/ a leading 1 is ind. in both A & $\text{rref}(A)$

So

$$\text{Im}(A) = \text{span} \{ \text{columns of } A \text{ w/ a leading 1 in } \text{rref}(A) \}$$

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Rank Nullity Thm / Fundamental Thm of Linear Algebra (FTOLA)

$$\dim(\text{Im}(A)) + \dim(\text{ker}(A)) = \dim(A)$$

of Ind. vectors in $\text{Im}(A)$

of ind. vectors in $\text{ker}(A)$

$\text{domain}(A)$
of columns in A
= # of variables