

Return to linear transformations $T: \mathbb{R}^n \rightarrow \mathbb{R}^m \leftrightarrow A: m \times n$ matrix
and discuss important subspaces associated to T .

$$\text{Image}(T) \text{ (or } \text{Im}(T) \text{ for short.)} = \left\{ \underline{b} \in \mathbb{R}^m \mid T(\underline{x}) = \underline{b} \text{ for some } \underline{x} \in \mathbb{R}^n \right\}$$

$$\text{kernel}(T) \text{ (or } \text{Ker}(T) \text{ for short)} = \left\{ \underline{b} \in \mathbb{R}^m \mid A \cdot \underline{x} = \underline{b} \right\}$$

$$= \left\{ \underline{b} \in \mathbb{R}^m \mid \underline{b} \in \text{Span} \left\{ \begin{array}{l} \text{columns} \\ \text{of} \\ A \end{array} \right\} \right\}$$

$$\parallel$$

$$\left\{ \underline{x} \in \mathbb{R}^n \mid T(\underline{x}) = \underline{0} \right\}$$

$$\parallel$$

$$\left\{ \underline{x} \in \mathbb{R}^n \mid A \cdot \underline{x} = \underline{0} \right\}$$

$$\parallel$$

$$\left\{ \underline{x} \in \mathbb{R}^n \mid \underline{x} \text{ mutually perp. to columns of } A \right\}$$

already shown this is subspace.

So $\text{Ker}(T)$ is subspace of \mathbb{R}^n .

makes clear that $\text{Im}(T)$ is a subspace.

Also check this directly.

Note: Subspace of \mathbb{R}^m
range space

Proposition: $T(\underline{x}) = \underline{b}$ (i.e. $A \cdot \underline{x} = \underline{b}$) has ^{at most one} ~~soln~~ soln for every \underline{b}
if and only if $\text{Ker}(T) = \underline{0}$, the trivial subspace of \mathbb{R}^n .

pf: We could use that $\text{Ker}(T)$ determined by pivot columns of \tilde{A}
(whether there is 1 in each column)

But we can give simpler proof applicable more generally...

(\Rightarrow) By contrapositive. If $\text{Ker}(T) \neq \{0\}$, then $T(\underline{x}) = \underline{0}$

has multiple solutions.

(\Leftarrow) By contrapositive. If $T(\underline{x}) = \underline{b}$ has multiple solutions $\underline{x}_1, \underline{x}_2, \dots$

then $\underline{x}_1 - \underline{x}_2 \in \text{Ker}(T)$.

Find basis of $\text{Im}(T)$, $\text{Ker}(T)$. (Remember, bases not unique. Just want an algorithm that gives a basis.)

Im(T): long way: Solve $[A | b]$ with symbolic b 's

if \tilde{A} has rows of 0's, ensure that corresponding entry in $\tilde{b} = 0$.

Gives equations satisfied by b 's, then find basis of these.

easier: find pivots in

\tilde{A} , take those columns of A

as basis! ($\text{Im}(T) = \text{Span} \{ \text{columns of } A \}$, just have to figure out which ones to keep...)

Example: Given $A = \begin{bmatrix} 1 & 2 & 1 \\ 0 & 1 & 1 \\ 1 & 1 & 0 \\ 2 & 1 & -1 \end{bmatrix} \rightsquigarrow \tilde{A} = \begin{bmatrix} 1 & 0 & -1 \\ 0 & 1 & 1 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}$

so $\left\{ \begin{bmatrix} 1 \\ 0 \\ 1 \\ 2 \end{bmatrix}, \begin{bmatrix} 2 \\ 1 \\ 1 \\ 1 \end{bmatrix} \right\}$ is a basis.

(example from wed. showing the column vectors were linearly dep.)

Why does it work? Column vectors are in $\text{Im}(T)$,

since $A \cdot \vec{e}_i = i^{\text{th}}$ column of A .

To show pivot columns of A are linearly independent,

solutions to $A\underline{x} = \underline{0}$ are given by ^(uniquely) choice of non-pivot vars. in \tilde{A} .

Suppose we set them all equal to 0.

E.g. if x_1, x_3, x_5 are pivot vars, x_2, x_4 non-pivots.

Set $x_2, x_4 = 0$ consider resulting equation:

$$A \cdot \begin{bmatrix} x_1 \\ 0 \\ x_3 \\ 0 \\ x_5 \end{bmatrix} = \underline{0}.$$

Only sol'n is $x_1 = x_3 = x_5 = 0$
since, having specified non-pivot vars,
sol'n is unique.

linear combinations
of pivot columns
of A

So ^{pivot} column vectors of A are
linearly independent.

Why do they span $\text{Im}(T)$? $\text{Im}(T) = \text{Span} \left\{ \begin{matrix} \text{columns} \\ \text{of } A \end{matrix} \right\}$, so
need to show non-pivot columns are linear comb. of pivot columns (hence redundant)
clear from echelon form (in fact, only need columns to left of
non-pivot column: $\begin{matrix} * & * & * & * & * \\ & * & * & * & * \\ & & * & * & * \\ & & & * & * \\ & & & & * \end{matrix}$)

For $\text{ker}(T)$, reduce A to \tilde{A} ,
write sol'ns in terms of non-pivot vars. This gives basis of
size equal to # of non-pivot vars.

Back to our example with $\tilde{A} = \begin{bmatrix} 1 & 0 & -1 \\ 0 & 1 & 1 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}$: then $x_1 - x_3 = 0$
 $x_2 + x_3 = 0$
so $x_1 = +x_3$
 $x_2 = -x_3$

General sol'n: $x_3 \cdot \begin{bmatrix} 1 \\ -1 \\ 1 \end{bmatrix}$ so take $\begin{bmatrix} 1 \\ -1 \\ 1 \end{bmatrix}$ as basis.

(No magic here. Just usual approach...)

notice then, $\dim(\text{Im}(T)) = \#$ of pivot columns

$\dim(\text{Ker}(T)) = \#$ of non-pivot columns

so if $T: \mathbb{R}^n \rightarrow \mathbb{R}^m$ then $A: m \times n$ matrix and

$$\dim(\text{Im}(T)) + \dim(\text{Ker}(T)) = n.$$

"rank"

"nullity"

"Dimension Formula"

or

"Rank-Nullity Theorem"

How I picture this?

Solutions of $A\underline{x} = \underline{b}$ are given by one particular

solution \underline{x}_0 plus all elements of $\text{Ker}(T)$:

$$\text{If } \underline{y} \in \text{Ker}(T) \text{ then } A(\underline{x}_0 + \underline{y}) = A(\underline{x}_0) = \underline{b}$$

$$\text{If } \underline{x}_0, \underline{x}_1 \text{ are two solutions, } A(\underline{x}_0 - \underline{x}_1) = A(\underline{x}_0) - A(\underline{x}_1)$$

$$= \underline{b} - \underline{b} = \underline{0}$$

$$\text{So } \underline{x}_0 - \underline{x}_1 \in \text{Ker}(T).$$

$$\text{i.e. } \underline{x}_1 = \underline{x}_0 + \underline{y}$$

for some

$\underline{y} \in \text{Ker}(T)$.

Corollary: $T: \mathbb{R}^n \rightarrow \mathbb{R}^n$ then

~~with~~ $\text{Im}(T) = \mathbb{R}^n$ if

and only if $\text{Ker}(T) = \underline{0}$.

~~XXXXXXXXXXXXXXXXXXXX~~

Final result: $\text{rank}(A) = \text{rank}(A^T)$ T : transpose.

so span of columns has same dimension as span of rows.

\ddagger : ~~can~~ show $\text{rank}(A^T)$ is preserved under row operations.

then know $\text{rank}(A^T)$ given by rows with pivots = # columns with pivots