

Lecture 6: Subspaces of a matrix

Thursday, September 10, 2015

9:30 AM

Admin: Reading: Start Ch. 4 of Meyer

Definition: For an $m \times n$ matrix A ,

- The **nullspace** of A is

$$N(A) = \{ \vec{x} \in \mathbb{R}^n \mid A\vec{x} = \vec{0} \}.$$

(everything A sends to $\vec{0}$)

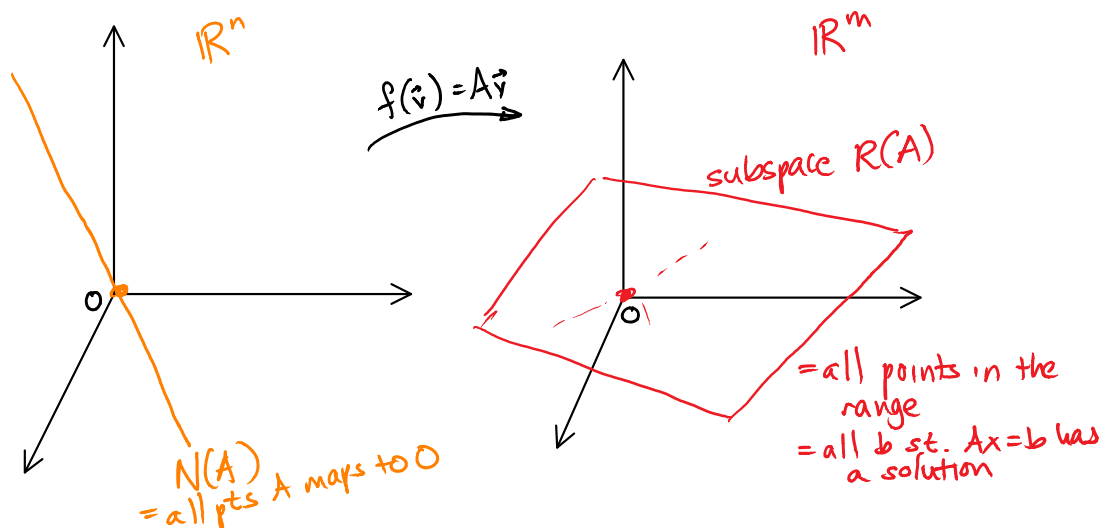
- The **range** (or **column space**) of A is

$$R(A) = \{ A\vec{x} \mid \vec{x} \in \mathbb{R}^n \} \\ = \text{Span}(\text{columns of } A)$$

(everything reachable by A)

- The **row space** of A is $R(A^T)$

(span of the rows of A)



Observe: $\vec{b} \in \text{Range}(A) \iff A\vec{x} = \vec{b}$ has a solution

Example: Let $\vec{u} = \begin{pmatrix} 1 \\ 2 \end{pmatrix}$, $\vec{v} = \begin{pmatrix} 3 \\ 4 \end{pmatrix}$, and let

$$A = \vec{u} \vec{v}^T = \begin{pmatrix} 1 \\ 2 \end{pmatrix} (3 \ 4) = \begin{pmatrix} 3 & 4 \\ 6 & 8 \end{pmatrix}.$$

What are $R(A)$, $R(A^T)$, $N(A)$, $N(A^T)$?

Answer: Yes, we could apply Gaussian elimination.

But we can also just observe that for any vector $\vec{x}$,

$$A\vec{x} = \vec{u}\vec{v}^T\vec{x} = (\vec{v}^T\vec{x})\cdot\vec{u}$$

$$\Rightarrow R(A) = \text{Span}(\{\vec{u}\}), \text{ a line.}$$

Similarly, $R(A^T) = \text{Span}(\{\vec{v}\})$, a line.

$$\text{Also } A\vec{x} = \vec{0} \Leftrightarrow \vec{v}^T\vec{x} = 0$$

$$3x_1 + 4x_2$$

$$\Leftrightarrow x \propto (-4, 3)$$

$$\Rightarrow N(A) = \text{Span}(\{(-4, 3)\}), \text{ the line perpendicular to } \vec{u}$$

$$\text{Similarly } N(A^T) = \text{Span}(\{(-2, 1)\}). \checkmark$$

NULLSPACE $\{x \in \mathbb{R}^n \mid Ax = \vec{0}\}$

ie., the set of solutions to the homogeneous equations specified by A

Claim 1: This is a subspace!

(The solutions to a set of homogeneous equations form a subspace.)

Proof:

• closure under multiplication:

If x is a solution and α any scalar, then

$$A(\alpha x) = \alpha Ax = \alpha \cdot \vec{0} = \vec{0}$$

$\Rightarrow \alpha x$ is a solution $\checkmark$

• closure under addition:

If x and y are both solutions, then

$$A(x+y) = Ax + Ay = \vec{0} + \vec{0} = \vec{0}$$

$\Rightarrow x+y$ is a solution $\checkmark$

□

Problem:

$$\text{Let } M = \begin{pmatrix} 1 & 2 & 1 & 1 & 5 \\ -2 & -4 & 0 & 4 & -2 \\ 1 & 2 & 2 & 4 & 9 \end{pmatrix}$$

$$\text{Let } M = \begin{pmatrix} 1 & 2 & 1 & 1 & 5 \\ -2 & -4 & 0 & 4 & -2 \\ 1 & 2 & 2 & 4 & 9 \end{pmatrix}$$

What is $N(M)$?

Answer: Solving for $N(M) \Leftrightarrow$ Solving $M\vec{x} = \vec{0}$

$$M \xrightarrow{GE} \begin{pmatrix} 1 & 2 & 1 & 1 & 5 \\ 0 & 0 & 2 & 6 & 8 \\ 0 & 0 & 1 & 3 & 4 \end{pmatrix} \xrightarrow{GE} \begin{pmatrix} 1 & 2 & 1 & 1 & 5 \\ 0 & 0 & 1 & 3 & 4 \\ 0 & 0 & 0 & 0 & 0 \end{pmatrix}$$

$$\xrightarrow{GE} \begin{pmatrix} 1 & 2 & 0 & -2 & 1 \\ 0 & 0 & 1 & 3 & 4 \\ 0 & 0 & 0 & 0 & 0 \end{pmatrix}$$

$$\Rightarrow x_1 = -2x_2 + 2x_4 - x_5$$

$$x_3 = -3x_4 - 4x_5$$

$\Rightarrow$ General solution

$$\begin{pmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \end{pmatrix} = x_2 \begin{pmatrix} -2 \\ 1 \\ 0 \\ 0 \\ 0 \end{pmatrix} + x_4 \begin{pmatrix} 2 \\ 0 \\ -3 \\ 1 \\ 0 \end{pmatrix} + x_5 \begin{pmatrix} -1 \\ 0 \\ -4 \\ 0 \\ 1 \end{pmatrix}$$

$$\Rightarrow N(M) = \text{Span} \left(\left\{ \begin{pmatrix} -2 \\ 1 \\ 0 \\ 0 \\ 0 \end{pmatrix}, \begin{pmatrix} 2 \\ 0 \\ -3 \\ 1 \\ 0 \end{pmatrix}, \begin{pmatrix} -1 \\ 0 \\ -4 \\ 0 \\ 1 \end{pmatrix} \right\} \right) \checkmark$$

Observe: Row operations don't change the nullspace.

Claim 2: Consider the equations $Ax = b$, with $b \neq 0$.

Let y be a solution ($Ay = b$). Then

$$\begin{aligned} \{x \mid Ax = b\} &= y + \{x \mid Ax = 0\} \\ &= y + N(A) \end{aligned}$$

Thus, the solutions to a set of non-homogeneous equations form an affine subspace.

form an affine subspace.

(In English: The general solution of the nonhomogeneous system is given by a particular solution (y) plus the general solution of the associated homogeneous system. [We claimed this in Lecture 2, but didn't prove it.])

Proof: Let $V_b = \{x \mid Ax = b\}$.

To show equality, we need to verify $V_b \subseteq y + N(A)$
and $V_b \supseteq y + N(A)$.

• $V_b \supseteq y + N(A)$:

Indeed, if $z \in N(A)$, then $A(y+z) = Ay + Az = b + 0 = b$

$\Rightarrow y+z \in V_b \quad \checkmark$

• $V_b \subseteq y + N(A)$:

Indeed, if $x \in V_b$, then $Ax = b$, so

$$A(y-b) = Ay - Ax = b - b = 0$$

$\Rightarrow y-b \in N(A) \quad \checkmark$

□

Corollary: The equations $Ax=b$ have infinitely many solutions if and only if $N(A) = \{0\}$.

RANGE $\{A\vec{x} \mid \vec{x} \in \mathbb{R}^n\}$

Claim 3: For $S \subseteq \mathbb{R}^n$, let

$$A(S) = \{A\vec{x} \mid \vec{x} \in S\}$$

(map the set S forward by applying A)

1. If S is a subspace, so is $A(S)$.

2. If $s_1, \dots, s_k$ span S ,

then $As_1, \dots, As_k$ span $A(S)$.

Proof:

1) Closure under addition:

Let $y, z \in A(S)$. Is $y+z \in A(S)$?
 Yes! $y = Ax, z = Ax'$
 $y+z = A(x+x') \in A(S) \checkmark$

Closure under multiplication is similar $\checkmark$

2) Any vector $y \in A(S)$ can be written $y = Ax$ for some $x \in S$.
 $s_1, \dots, s_k$ span $S \Rightarrow x = \sum_{j=1}^k \alpha_j s_j$ for some scalars

$\alpha_1, \dots, \alpha_k$.

$\Rightarrow$ by linearity, $y = Ax = \sum_{j=1}^k \alpha_j (As_j)$

$\Rightarrow y \in \text{Span}(As_1, \dots, As_k) \checkmark \quad \square$

Observe: ① $b \in R(A) \Leftrightarrow Ax = b$ has a solution.

② $R(A) = A(\mathbb{R}^n)$
 $= \text{Span}(\text{columns of } A)$

Why?

$\mathbb{R}^n$ is spanned by $(1, 0, \dots, 0), \dots, (0, 0, \dots, 1)$

By the exercise, $A(\mathbb{R}^n)$ is spanned by

$A\begin{pmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{pmatrix} = \text{1st column}, \dots, A\begin{pmatrix} 0 \\ \vdots \\ 1 \end{pmatrix} = \text{last column of } A \checkmark$

Thus the range is also known as the column space of A .

Observe: ③ $A_{n \times n}$ is invertible (A^{-1} exists)

$\Leftrightarrow$ every point $b \in \mathbb{R}^n$ has a preimage x , $Ax = b$

$\Leftrightarrow R(A) = \mathbb{R}^n$ (everything)

ROW SPACE $R(A^T) = \text{Span}(\text{rows of } A)$

Problem:

Let $M = \begin{pmatrix} 1 & 2 & 1 & 1 & 5 \\ -2 & -4 & 0 & 4 & -2 \\ 1 & 2 & 2 & 4 & 9 \end{pmatrix}$ (as before)

What are the row and column spaces of M ?

Answer:

$M \text{ GE } / 1 \ 2 \ 1 \ 1 \ 5 \backslash \text{ as } / 1 \ 2 \ 1 \ 1 \ 5 \backslash$

ANSWER:

$$M \xrightarrow{GE} \begin{pmatrix} 1 & 2 & 1 & 1 & 5 \\ 0 & 0 & 2 & 6 & 8 \\ 0 & 0 & 1 & 3 & 4 \end{pmatrix} \xrightarrow{GE} \begin{pmatrix} 1 & 2 & 1 & 1 & 5 \\ 0 & 0 & 1 & 3 & 4 \\ 0 & 0 & 0 & 0 & 0 \end{pmatrix}$$

$$\xrightarrow{GE} \begin{pmatrix} 1 & 2 & 0 & -2 & 1 \\ 0 & 0 & 1 & 3 & 4 \\ 0 & 0 & 0 & 0 & 0 \end{pmatrix}$$

Observe: The row operations of Gaussian elimination do not change the row space.

(Because $\text{Span}(x_1, x_2, x_3, \dots, x_k) = \text{Span}(x_1, x_2 + \alpha x_1, x_3, \dots, x_k)$.)

$$\Rightarrow R(M^T) = \text{Span}\{(1, 2, 0, -2, 1), (0, 0, 1, 3, 4)\}.$$

row space

How can we compute $R(M)$?

① Apply Gaussian elimination to M^T

$$M^T = \begin{pmatrix} 1 & -2 & 1 \\ 2 & -4 & 2 \\ 1 & 0 & 2 \\ 1 & 4 & 4 \\ 5 & -2 & 7 \end{pmatrix} \xrightarrow{\substack{R_2 \leftarrow R_2 - 2R_1 \\ R_3 \leftarrow R_3 - R_1 \\ R_4 \leftarrow R_4 - R_1 \\ R_5 \leftarrow R_5 - 5R_1}} \begin{pmatrix} 1 & -2 & 1 \\ 0 & 0 & 0 \\ 0 & 2 & 1 \\ 0 & 6 & 3 \\ 0 & 8 & 4 \end{pmatrix} \xrightarrow{\substack{R_4 \leftarrow R_4 - 3R_3 \\ R_5 \leftarrow R_5 - 4R_3}} \begin{pmatrix} 1 & -2 & 1 \\ 0 & 0 & 0 \\ 0 & 2 & 1 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix}$$

$$\Rightarrow R(M) = \text{Span}\{(1, -2, 1), (0, 2, 1)\}$$

② But if we have already applied G.E. to M , there is no need to apply it again to M^T .

Observe: The row operations of G.E. finish with a matrix like

$$\begin{pmatrix} \boxed{\times} & ? & ? & ? & ? & ? & ? \\ 0 & 0 & 0 & \boxed{\times} & ? & ? & ? \\ 0 & 0 & 0 & 0 & \boxed{\times} & ? & ? \\ 0 & 0 & 0 & 0 & 0 & \boxed{\times} & ? \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ 0 & 0 & 0 & 0 & 0 & 0 & \boxed{\times} \end{pmatrix}$$

nonzero

all those the "basic columns"

Claim: Every column is a linear combination of the basic columns.

Proof: Starting with a non-basic column, use the last basic column to cancel out its last coordinate, etc., up to using the first basic column to cancel out the first coordinate. $\square$

$$M = \begin{pmatrix} 1 & 2 & 1 & 1 & 5 \\ -2 & -4 & 0 & 4 & -2 \\ 1 & 2 & 2 & 4 & 9 \end{pmatrix} \xrightarrow{\text{G.E.}} G = \begin{pmatrix} 1 & 2 & 1 & 1 & 5 \\ 0 & 0 & 1 & 3 & 4 \\ 0 & 0 & 0 & 0 & 0 \end{pmatrix}$$

basic columns

eg. $(1, 3, 0) = 3(1, 1, 0) - 2(1, 0, 0) \checkmark$

The same linear combination of columns works for M :

$$(1, 4, 4) = 3(1, -2, 1) - 2(1, -2, 1)$$

column 4 column 3 column 1
 $M\vec{e}_4$ $M\vec{e}_3$ $M\vec{e}_1$

Why?

$M = QG$, where Q is the matrix that implements the row operations in Gaussian Elim.

$$\Rightarrow \text{If } G\vec{e}_j = \sum_k \alpha_k G\vec{e}_k$$

jth col kth column

$$\text{then } QG\vec{e}_j = Q\left(\sum_k \alpha_k G\vec{e}_k\right)$$

$M\vec{e}_j$ $\sum_k \alpha_k M\vec{e}_k$

$$\Rightarrow \text{Range}(M) = \text{Span}\{(1, -2, 1), (1, 0, 2)\}$$

In general:

After Gaussian elimination, identify the basic columns.

$$R(M) = \text{Span}(\text{those same basic columns in } M).$$

Exercise: Prove

① $R(AB) \subseteq R(A)$

(Multiplying on the right can only reduce the range.)

$R(AB) = R(A)$ if B is invertible.

② $N(BA) \supseteq N(A)$

(Multiplying on the left can only increase the nullspace.)

$N(BA) = N(A)$ if B is invertible

Easy example: Take $B = 0$, so $AB = 0$, $BA = 0$

$R(AB) = \{0\} \subseteq R(A)$, $N(BA) = \mathbb{R}^n \supseteq N(A)$ ✓

Note: These are all sensitive to perturbations/numerical errors.

E.g.

	<u>Range</u>	<u>Nullspace</u>
$\begin{pmatrix} 1 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix}$	x-axis $\{(x, 0, 0) \mid x \in \mathbb{R}\}$	yz-plane $\{(0, y, z) \mid y, z \in \mathbb{R}\}$
$\begin{pmatrix} 1 & 0 & 0 \\ 0 & \epsilon & 0 \\ 0 & 0 & \epsilon \end{pmatrix}$ $\epsilon \neq 0$	$\mathbb{R}^3$	$\{0\}$

Example: What is the nullspace of

$$A = \begin{pmatrix} -1 & 1 & & & & \\ 1 & -2 & 1 & & & \\ & 1 & -2 & 1 & & \\ & & 1 & -2 & 1 & \\ & & & 1 & -2 & 1 \\ \bigcirc & & & & 1 & -2 & 1 \\ & & & & & 1 & -1 \end{pmatrix} \quad ?$$

Answer:

Intuition: Recall that A is the matrix we got from discretizing the second derivative operator

$$f''(t) = \lim_{\delta t \rightarrow 0} \frac{1}{(\delta t)^2} (f(t+\delta t) - 2f(t) + f(t-\delta t)).$$

The functions with second derivative 0 are exactly constant functions $f(t) = \text{constant}$.

$\Rightarrow$ We expect the nullspace of A to be the set

of vectors with all-equal coordinates.

Claim: $N(A) = \text{Span}\left\{\begin{pmatrix} 1 \\ 1 \\ \vdots \\ 1 \end{pmatrix}\right\}$

= set of all vectors with all-equal coordinates.

Proof:

$$A \begin{pmatrix} 1 \\ 1 \\ \vdots \\ 1 \end{pmatrix} = \begin{pmatrix} \text{sum of } a_{ij} \text{ across row 1} \\ \vdots \\ \text{sum across row } n \end{pmatrix} = \mathbf{0}$$

$$\Rightarrow (1, 1, \dots, 1) \in N(A)$$

$$\Rightarrow \text{Span}(\{(1, \dots, 1)\}) \subseteq N(A).$$

But are there other vectors in $N(A)$?

$$A = \begin{pmatrix} -1 & 1 & & & & & \\ 1 & -2 & 1 & & & & \\ & 1 & -2 & 1 & & & \\ & & 1 & -2 & 1 & & \\ & & & 1 & -2 & 1 & \\ & & & & 1 & -2 & 1 \\ & & & & & 1 & -2 & 1 \\ & & & & & & 1 & -2 & 1 \\ & & & & & & & 1 & -2 & 1 \\ & & & & & & & & 1 & -2 & 1 \end{pmatrix} \xrightarrow{R_1} \begin{pmatrix} -1 & 1 & & & & & \\ 0 & -1 & 1 & & & & \\ & 1 & -2 & 1 & & & \\ & & 1 & -2 & 1 & & \\ & & & 1 & -2 & 1 & \\ & & & & 1 & -2 & 1 \\ & & & & & 1 & -2 & 1 \\ & & & & & & 1 & -2 & 1 \\ & & & & & & & 1 & -2 & 1 \\ & & & & & & & & 1 & -2 & 1 \end{pmatrix} \xrightarrow{R_1} \begin{pmatrix} -1 & 1 & & & & & \\ 0 & -1 & 1 & & & & \\ & 0 & -1 & 1 & & & \\ & & 0 & -1 & 1 & & \\ & & & 0 & -1 & 1 & \\ & & & & 0 & -1 & 1 \\ & & & & & 0 & -1 & 1 \\ & & & & & & 0 & -1 & 1 \\ & & & & & & & 0 & -1 & 1 \\ & & & & & & & & 0 & -1 & 1 \end{pmatrix} \xrightarrow{B} \begin{pmatrix} -1 & 1 & & & & & \\ 0 & -1 & 1 & & & & \\ & 0 & -1 & 1 & & & \\ & & 0 & -1 & 1 & & \\ & & & 0 & -1 & 1 & \\ & & & & 0 & -1 & 1 \\ & & & & & 0 & -1 & 1 \\ & & & & & & 0 & -1 & 1 \\ & & & & & & & 0 & -1 & 1 \\ & & & & & & & & 0 & -1 & 1 \end{pmatrix}$$

$$A\vec{x} = \mathbf{0} \iff B\vec{x} = \mathbf{0}$$

$$\iff \begin{matrix} x_1 = x_2 & \text{(first row)} \\ x_2 = x_3 & \text{(2nd row)} \\ \vdots & \\ \dots & \end{matrix}$$

$$\begin{array}{c}
 x_2 = x_3 \quad (\text{2nd row}) \\
 \vdots \\
 x_{n-1} = x_n \\
 \Leftrightarrow \vec{x} \in \text{Span}\{(1, 1, 1, \dots, 1)\}
 \end{array}$$

Question: What is $R(A)$? (Note $A = A^T$)