

Class 7: 9/27/12

Outline

I. Dimension, basis, independence.

II. All solutions to $AX = \vec{b}$.

Last time: Nullspace = linear combinations of special solutions.

Rank = # pivots = dimension of column space.

Definition: A set of vectors $v_1, v_2, \dots, v_n$ is linearly independent if

$$c_1 v_1 + c_2 v_2 + \dots + c_n v_n = 0 \quad \text{only when } c_1 = c_2 = c_3 = \dots = c_n = 0.$$

Alternate definition: Nullspace of $A = \begin{bmatrix} | & | & & | \\ v_1 & v_2 & & v_n \\ | & | & & | \end{bmatrix}$ is $Z = \{0\}$.

When this happens, we say the columns of A are independent.

Geometrically: Vectors are independent if they "go in different directions." A set of n vectors is independent if all linear combinations span $\mathbb{R}^n$ (biggest space possible).

Example: $\begin{bmatrix} 0 \\ 1 \end{bmatrix}$ and $\begin{bmatrix} 1 \\ 0 \end{bmatrix}$ are independent.

$\begin{bmatrix} 0 \\ 1 \end{bmatrix}$, $\begin{bmatrix} 1 \\ 0 \end{bmatrix}$ and $\begin{bmatrix} 1 \\ 1 \end{bmatrix}$ are dependent.

Note: If $n > m$, a set of n vectors in $\mathbb{R}^m$ is necessarily dependent.

Definition: A set of vectors spans a space if all linear combinations of the vectors fill the space.

E.g. Columns of A span $C(A)$.
Special solutions span $N(A)$.

Definition: Row space of A = span of row vectors (as seen on HW).

* Alternatively, row space = $C(A^T)$.

Note: Vectors don't have to be linearly independent to span a space

e.g. $\begin{bmatrix} 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \end{bmatrix}, \begin{bmatrix} 1 \\ 1 \end{bmatrix}$ span $\mathbb{R}^2$ even though $\begin{bmatrix} 0 \\ 1 \end{bmatrix}$ and $\begin{bmatrix} 1 \\ 1 \end{bmatrix}$ alone already span $\mathbb{R}^2$.

If we don't have redundant vectors, we have a basis.

Definition: A basis of a vector space is a set of vectors that

- 1) span the space
- 2) are linearly independent.

E.g. $\begin{bmatrix} 1 \\ 0 \end{bmatrix}$ and $\begin{bmatrix} 0 \\ 1 \end{bmatrix}$ are a basis for $\mathbb{R}^2$.

$\begin{bmatrix} 0 \\ 1 \end{bmatrix}, \begin{bmatrix} 1 \\ 0 \end{bmatrix}$ and $\begin{bmatrix} 1 \\ 1 \end{bmatrix}$ do not form a basis for $\mathbb{R}^2$.

Fact: If $v_1, v_2, \dots, v_n$ is a basis for V , then there is only one way to write a vector in V as a combination of basis vectors.

Why? Say $a_1 v_1 + a_2 v_2 + \dots + a_n v_n = b_1 v_1 + \dots + b_n v_n$.

Then $(a_1 - b_1)v_1 + (a_2 - b_2)v_2 + \dots + (a_n - b_n)v_n = 0$.

Since the v_i are independent, $a_1 = b_1$

$a_2 = b_2$

$\vdots$

$a_n = b_n$

So the two linear combinations were actually the same.

Facts:

- Pivot columns of A form a basis for the column space (last time: free columns = combination of pivot columns, so those are redundant).

- Pivot rows of A form a basis for the row space.

- $v_1, \dots, v_n$ form a basis for $\mathbb{R}^n$ if ^{and only if} the matrix $A = \begin{bmatrix} | & | & | & | \\ v_1 & v_2 & v_3 & \dots & v_n \\ | & | & | & | \end{bmatrix}$ is an $n \times n$ invertible matrix.

Definition: Dimension of a vector space = # vectors in a basis.

I. Use our nullspace experience to find all solutions $A\vec{x} = \vec{b}$.

Note: If $\vec{x}_n \in N(A)$ and $\vec{x}_p$ is a vector with $A\vec{x}_p = \vec{b}$, then

$$A(\vec{x}_p + \vec{x}_n) = A\vec{x}_p + A\vec{x}_n = \vec{b} + \vec{0} = \vec{b}.$$

So we can add any vector in the nullspace to any particular solution to get another solution.

Definition: A matrix has full column rank if $r = n$

Then 1. All columns are pivot columns

2. No free variables or special solutions, so $N(A) = \{0\}$

3. If $A\vec{x} = \vec{b}$ has a solution, the solution is unique.

4. Columns are independent.

Definition: A matrix has full row rank if $r = m$

Then 1. All rows have pivots, and R has no zero rows.

2. $A\vec{x} = \vec{b}$ is always solvable.

3. $C(A) = \mathbb{R}^m$

4. $N(A)$ has $n - r = n - m$ special solutions.

5. rows of A are independent

Possibilities:

1) $R = [I]$; A is square, invertible $A\vec{x} = \vec{b}$ has exactly 1 solution.
 $r = m = n$

2) $R = [I | F]$; $A\vec{x} = \vec{b}$ has ∞ -ly many solutions.
 $r = m < n$

3) $R = \begin{bmatrix} I \\ 0 \end{bmatrix}$; $A\vec{x} = \vec{b}$ has 0 or 1 solution.
 $r = n < m$

4) $R = \begin{bmatrix} I & F \\ 0 & 0 \end{bmatrix}$; $A\vec{x} = \vec{b}$ has 0 or ∞ -ly many solutions.
 $r < m, r < n$

Example: Find all solutions to $\begin{bmatrix} 1 & 1 & 2 & 3 \\ 2 & 2 & 8 & 10 \\ 3 & 3 & 10 & 13 \end{bmatrix} \vec{x} = \begin{bmatrix} 2 \\ 1 \\ 3 \end{bmatrix}$

Eliminate $\left[\begin{array}{cccc|c} 1 & 1 & 2 & 3 & 2 \\ 2 & 2 & 8 & 10 & 1 \\ 3 & 3 & 10 & 13 & 3 \end{array} \right] \rightarrow \left[\begin{array}{cccc|c} 1 & 1 & 2 & 3 & 2 \\ 0 & 0 & 4 & 4 & -3 \\ 0 & 0 & 4 & 4 & -3 \end{array} \right]$

$$\rightarrow \left[\begin{array}{cccc|c} 1 & 1 & 2 & 3 & 2 \\ 0 & 0 & 4 & 4 & -3 \\ 0 & 0 & 0 & 0 & 0 \end{array} \right] \rightarrow \left[\begin{array}{cccc|c} 1 & 1 & 2 & 3 & 2 \\ 0 & 0 & 1 & 1 & -3/4 \\ 0 & 0 & 0 & 0 & 0 \end{array} \right]$$

$$\rightarrow \left[\begin{array}{cccc|c} 1 & 1 & 0 & 1 & 7/2 \\ 0 & 0 & 1 & 1 & -3/4 \\ 0 & 0 & 0 & 0 & 0 \end{array} \right]$$

Particular solution?

Choose values for free variables, say $x_2 = x_4 = 0$.

Then $x_1 = 7/2$ and $x_3 = -3/4$.

$$\begin{bmatrix} 7/2 \\ 0 \\ -3/4 \\ 0 \end{bmatrix} \text{ is a solution.}$$

All solutions are $\begin{bmatrix} 7/2 \\ 0 \\ -3/4 \\ 0 \end{bmatrix} + \text{nullspace vectors}$

$$N = \begin{bmatrix} -1 & 1 \\ 1 & 0 \\ 0 & -1 \\ 0 & 1 \end{bmatrix} \text{ from } N = \begin{bmatrix} -F \\ I \end{bmatrix}$$

So All solutions are

$$\begin{bmatrix} 7/2 \\ 0 \\ -3/4 \\ 0 \end{bmatrix} + c \begin{bmatrix} -1 \\ 1 \\ 0 \\ 0 \end{bmatrix} + d \begin{bmatrix} 1 \\ 0 \\ -1 \\ 1 \end{bmatrix}$$

Note: If $A\vec{x}_1 = \vec{b}$ and $A\vec{x}_2 = \vec{b}$, then $\vec{x}_1 = \vec{x}_2 + \vec{x}_n$ ← in nullspace. since

$$A\vec{x}_1 - A\vec{x}_2 = \vec{b} - \vec{b} = 0 \Rightarrow A(\vec{x}_1 - \vec{x}_2) = 0, \text{ so } \vec{x}_1 - \vec{x}_2 \in N(A)$$