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### 4. Matrix inverses

- left and right inverse
- linear independence
- nonsingular matrices
- matrices with linearly independent columns
- matrices with linearly independent rows

# Left and right inverse

 $AB \neq BA$  in general, so we have to distinguish two types of inverses

**Left inverse:** X is a *left inverse* of A if

$$XA = I$$

A is *left-invertible* if it has at least one left inverse

**Right inverse:** X is a *right inverse* of A if

$$AX = I$$

A is *right-invertible* if it has at least one right inverse

# **Examples**

$$A = \begin{bmatrix} -3 & -4 \\ 4 & 6 \\ 1 & 1 \end{bmatrix}, \qquad B = \begin{bmatrix} 1 & 0 & 1 \\ 0 & 1 & 1 \end{bmatrix}$$

• *A* is left-invertible; the following matrices are left inverses:

$$\frac{1}{9} \begin{bmatrix} -11 & -10 & 16 \\ 7 & 8 & -11 \end{bmatrix}, \qquad \begin{bmatrix} 0 & -1/2 & 3 \\ 0 & 1/2 & -2 \end{bmatrix}$$

• *B* is right-invertible; the following matrices are right inverses:

$$\frac{1}{2} \begin{bmatrix} 1 & -1 \\ -1 & 1 \\ 1 & 1 \end{bmatrix}, \quad \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 0 & 0 \end{bmatrix}, \quad \begin{bmatrix} 1 & -1 \\ 0 & 0 \\ 0 & 1 \end{bmatrix}$$

# Some immediate properties

#### **Dimensions**

a left or right inverse of an  $m \times n$  matrix must have size  $n \times m$ 

### Left and right inverse of (conjugate) transpose

• X is a left inverse of A if and only if  $X^T$  is a right inverse of  $A^T$ 

$$A^T X^T = (XA)^T = I$$

• X is a left inverse of A if and only if  $X^H$  is a right inverse of  $A^H$ 

$$A^H X^H = (XA)^H = I$$

#### **Inverse**

if A has a left **and** a right inverse, then they are equal and unique:

$$XA = I$$
,  $AY = I$   $\Longrightarrow$   $X = X(AY) = (XA)Y = Y$ 

- in this case, we call X = Y the **inverse** of A (notation:  $A^{-1}$ )
- A is *invertible* if its inverse exists

### **Example**

$$A = \begin{bmatrix} -1 & 1 & -3 \\ 1 & -1 & 1 \\ 2 & 2 & 2 \end{bmatrix}, \qquad A^{-1} = \frac{1}{4} \begin{bmatrix} 2 & 4 & 1 \\ 0 & -2 & 1 \\ -2 & -2 & 0 \end{bmatrix}$$

## **Linear equations**

set of m linear equations in n variables

$$A_{11}x_1 + A_{12}x_2 + \dots + A_{1n}x_n = b_1$$

$$A_{21}x_1 + A_{22}x_2 + \dots + A_{2n}x_n = b_2$$

$$\vdots$$

$$A_{m1}x_1 + A_{m2}x_2 + \dots + A_{mn}x_n = b_m$$

- in matrix form: Ax = b
- may have no solution, a unique solution, infinitely many solutions

## **Linear equations and matrix inverse**

**Left-invertible matrix:** if X is a left inverse of A, then

$$Ax = b \implies x = XAx = Xb$$

there is at most one solution (if there is a solution, it must be equal to Xb)

**Right-invertible matrix:** if X is a right inverse of A, then

$$x = Xb \implies Ax = AXb = b$$

there is at least one solution (namely, x = Xb)

**Invertible matrix:** if A is invertible, then

$$Ax = b \iff x = A^{-1}b$$

there is a *unique* solution

### **Outline**

- left and right inverse
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#### **Linear combination**

a linear combination of vectors  $a_1, \ldots, a_n$  is a sum of scalar-vector products

$$x_1a_1 + x_2a_2 + \cdots + x_na_n$$

- the scalars  $x_i$  are the *coefficients* of the linear combination
- can be written as a matrix-vector product

$$x_1a_1 + x_2a_2 + \dots + x_na_n = \begin{bmatrix} a_1 & a_2 & \dots & a_n \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix}$$

• the *trivial* linear combination has coefficients  $x_1 = \cdots = x_n = 0$ 

(same definition holds for real and complex vectors/scalars)

## Linear dependence

a collection of vectors  $a_1, a_2, \ldots, a_n$  is *linearly dependent* if

$$x_1a_1 + x_2a_2 + \dots + x_na_n = 0$$

for some scalars  $x_1, \ldots, x_n$ , not all zero

- the vector 0 can be written as a nontrivial linear combination of  $a_1, \ldots, a_n$
- equivalently, at least one vector  $a_i$  is a linear combination of the other vectors:

$$a_i = -\frac{x_1}{x_i}a_1 - \dots - \frac{x_{i-1}}{x_i}a_{i-1} - \frac{x_{i+1}}{x_i}a_{i+1} - \dots - \frac{x_n}{x_i}a_n$$

if  $x_i \neq 0$ 

## **Example**

the vectors

$$a_1 = \begin{bmatrix} 0.2 \\ -7 \\ 8.6 \end{bmatrix}, \qquad a_2 = \begin{bmatrix} -0.1 \\ 2 \\ -1 \end{bmatrix}, \qquad a_3 = \begin{bmatrix} 0 \\ -1 \\ 2.2 \end{bmatrix}$$

are linearly dependent

• 0 can be expressed as a nontrivial linear combination of  $a_1$ ,  $a_2$ ,  $a_3$ :

$$0 = a_1 + 2a_2 - 3a_3$$

•  $a_1$  can be expressed as a linear combination of  $a_2$ ,  $a_3$ :

$$a_1 = -2a_2 + 3a_3$$

(and similarly  $a_2$  and  $a_3$ )

# **Linear independence**

vectors  $a_1, \ldots, a_n$  are *linearly independent* if they are not linearly dependent

the zero vector cannot be written as a nontrivial linear combination:

$$x_1a_1 + x_2a_2 + \cdots + x_na_n = 0 \implies x_1 = x_2 = \cdots = x_n = 0$$

• none of the vectors  $a_i$  is a linear combination of the other vectors

#### Matrix with linearly independent columns

$$A = [a_1 \quad a_2 \quad \cdots \quad a_n]$$

has linearly independent columns if

$$Ax = 0 \implies x = 0$$

## **Example**

the vectors

$$a_1 = \begin{bmatrix} 1 \\ -2 \\ 0 \end{bmatrix}, \qquad a_2 = \begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix}, \qquad a_3 = \begin{bmatrix} 0 \\ 1 \\ 1 \end{bmatrix}$$

are linearly independent:

$$x_1 a_1 + x_2 a_2 + x_3 a_3 = \begin{bmatrix} x_1 - x_2 \\ -2x_1 + x_3 \\ x_2 + x_3 \end{bmatrix} = 0$$

holds only if  $x_1 = x_2 = x_3 = 0$ 

# **Dimension inequality**

if n vectors  $a_1, a_2, \ldots, a_n$  of length m are linearly independent, then

$$n \leq m$$

(proof is in textbook)

- if an  $m \times n$  matrix has linearly independent columns then  $m \ge n$
- if an  $m \times n$  matrix has linearly independent rows then  $m \leq n$

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## **Nonsingular matrix**

for a **square** matrix *A* the following four properties are equivalent

- 1. A is left-invertible
- 2. the columns of A are linearly independent
- 3. *A* is right-invertible
- 4. the rows of *A* are linearly independent

a square matrix with these properties is called nonsingular

### Nonsingular = invertible

- if properties 1 and 3 hold, then A is invertible (page 4.5)
- if A is invertible, properties 1 and 3 hold (by definition of invertibility)

### **Proof**

- we show that (a) holds in general
- we show that (b) holds for square matrices
- (a') and (b') follow from (a) and (b) applied to  $A^T$

### **Part a:** suppose A is left-invertible

• if B is a left inverse of A (satisfies BA = I), then

$$Ax = 0 \implies BAx = 0$$
$$\implies x = 0$$

• this means that the columns of A are linearly independent: if

$$A = [a_1 \quad a_2 \quad \cdots \quad a_n]$$

then

$$x_1a_1 + x_2a_2 + \dots + x_na_n = 0$$

holds only for the trivial linear combination  $x_1 = x_2 = \cdots = x_n = 0$ 

**Part b:** suppose A is square with linearly independent columns  $a_1, \ldots, a_n$ 

- for every n-vector b the vectors  $a_1, \ldots, a_n, b$  are linearly dependent (from dimension inequality on page 4.13)
- hence for every b there exists a nontrivial linear combination

$$x_1a_1 + x_2a_2 + \cdots + x_na_n + x_{n+1}b = 0$$

- we must have  $x_{n+1} \neq 0$  because  $a_1, \ldots, a_n$  are linearly independent
- hence every b can be written as a linear combination of  $a_1, \ldots, a_n$
- in particular, there exist n-vectors  $c_1, \ldots, c_n$  such that

$$Ac_1 = e_1, \qquad Ac_2 = e_2, \qquad \dots, \qquad Ac_n = e_n,$$

• the matrix  $C = [c_1 \ c_2 \ \cdots \ c_n]$  is a right inverse of A:

$$A \begin{bmatrix} c_1 & c_2 & \cdots & c_n \end{bmatrix} = \begin{bmatrix} e_1 & e_2 & \cdots & e_n \end{bmatrix} = I$$

# **Examples**

• *A* is nonsingular because its columns are linearly independent:

$$x_1 - x_2 + x_3 = 0$$
,  $-x_1 + x_2 + x_3 = 0$ ,  $x_1 + x_2 - x_3 = 0$ 

is only possible if  $x_1 = x_2 = x_3 = 0$ 

• *B* is singular because its columns are linearly dependent:

$$Bx = 0$$
 for  $x = (1, 1, 1, 1)$ 

## **Example: Vandermonde matrix**

$$A = \begin{bmatrix} 1 & t_1 & t_1^2 & \cdots & t_1^{n-1} \\ 1 & t_2 & t_2^2 & \cdots & t_2^{n-1} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & t_n & t_n^2 & \cdots & t_n^{n-1} \end{bmatrix} \quad \text{with } t_i \neq t_j \text{ for } i \neq j$$

we show that A is nonsingular by showing that Ax = 0 only if x = 0

• Ax = 0 means  $p(t_1) = p(t_2) = \cdots = p(t_n) = 0$  where

$$p(t) = x_1 + x_2t + x_3t^2 + \dots + x_nt^{n-1}$$

p(t) is a polynomial of degree n-1 or less

- if  $x \neq 0$ , then p(t) can not have more than n-1 distinct real roots
- therefore  $p(t_1) = \cdots = p(t_n) = 0$  is only possible if x = 0

# Inverse of transpose and product

#### Transpose and conjugate transpose

if A is nonsingular, then  $A^T$  and  $A^H$  are nonsingular and

$$(A^T)^{-1} = (A^{-1})^T, \qquad (A^H)^{-1} = (A^{-1})^H$$

we write these as  $A^{-T}$  and  $A^{-H}$ 

#### **Product**

if A and B are nonsingular and of equal size, then AB is nonsingular with

$$(AB)^{-1} = B^{-1}A^{-1}$$

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#### **Gram matrix**

the *Gram matrix* associated with a matrix

$$A = [a_1 \quad a_2 \quad \cdots \quad a_n]$$

is the matrix of column inner products

for real matrices:

$$A^{T}A = \begin{bmatrix} a_{1}^{T}a_{1} & a_{1}^{T}a_{2} & \cdots & a_{1}^{T}a_{n} \\ a_{2}^{T}a_{1} & a_{2}^{T}a_{2} & \cdots & a_{2}^{T}a_{n} \\ \vdots & \vdots & & \vdots \\ a_{n}^{T}a_{1} & a_{n}^{T}a_{2} & \cdots & a_{n}^{T}a_{n} \end{bmatrix}$$

• for complex matrices

$$A^{H}A = \begin{bmatrix} a_{1}^{H}a_{1} & a_{1}^{H}a_{2} & \cdots & a_{1}^{H}a_{n} \\ a_{2}^{H}a_{1} & a_{2}^{H}a_{2} & \cdots & a_{2}^{H}a_{n} \\ \vdots & \vdots & & \vdots \\ a_{n}^{H}a_{1} & a_{n}^{H}a_{2} & \cdots & a_{n}^{H}a_{n} \end{bmatrix}$$

## **Nonsingular Gram matrix**

the Gram matrix is nonsingular if only if A has linearly independent columns

• suppose  $A \in \mathbb{R}^{m \times n}$  has linearly independent columns:

$$A^{T}Ax = 0 \implies x^{T}A^{T}Ax = (Ax)^{T}(Ax) = ||Ax||^{2} = 0$$

$$\implies Ax = 0$$

$$\implies x = 0$$

therefore  $A^TA$  is nonsingular

• suppose the columns of  $A \in \mathbf{R}^{m \times n}$  are linearly dependent

$$\exists x \neq 0, \ Ax = 0 \implies \exists x \neq 0, \ A^T Ax = 0$$

therefore  $A^TA$  is singular

(for  $A \in \mathbb{C}^{m \times n}$ , replace  $A^T$  with  $A^H$  and  $x^T$  with  $x^H$ )

### **Pseudo-inverse**

- suppose  $A \in \mathbf{R}^{m \times n}$  has linearly independent columns
- this implies that A is tall or square  $(m \ge n)$ ; see page 4.13

the *pseudo-inverse* of A is defined as

$$A^{\dagger} = (A^T A)^{-1} A^T$$

- this matrix exists, because the Gram matrix  $A^TA$  is nonsingular
- $A^{\dagger}$  is a left inverse of A:

$$A^{\dagger}A = (A^T A)^{-1}(A^T A) = I$$

(for complex A with linearly independent columns,  $A^{\dagger} = (A^H A)^{-1} A^H$ )

# **Summary**

the following three properties are equivalent for a real matrix A

- 1. A is left-invertible
- 2. the columns of A are linearly independent
- 3.  $A^T A$  is nonsingular

- $1 \Rightarrow 2$  was already proved on page 4.16
- $2 \Rightarrow 1$ : we have seen that the pseudo-inverse is a left inverse
- $2 \Leftrightarrow 3$ : proved on page 4.22
- a matrix with these properties must be tall or square
- for complex matrices, replace  $A^TA$  in property 3 by  $A^HA$

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#### **Pseudo-inverse**

- suppose  $A \in \mathbb{R}^{m \times n}$  has linearly independent rows
- this implies that A is wide or square  $(m \le n)$ ; see page 4.13

the *pseudo-inverse* of A is defined as

$$A^{\dagger} = A^T (AA^T)^{-1}$$

- $A^T$  has linearly independent columns
- hence its Gram matrix  $AA^T$  is nonsingular, so  $A^{\dagger}$  exists
- $A^{\dagger}$  is a right inverse of A:

$$AA^{\dagger} = (AA^T)(AA^T)^{-1} = I$$

(for complex A with linearly independent rows,  $A^{\dagger} = A^H (AA^H)^{-1}$ )

# **Summary**

the following three properties are equivalent

- 1. *A* is right-invertible
- 2. the rows of *A* are linearly independent
- 3.  $AA^T$  is nonsingular

- $1 \Rightarrow 2$  and  $2 \Leftrightarrow 3$ : by transposing result on page 4.24
- $2 \Rightarrow 1$ : we have seen that the pseudo-inverse is a right inverse
- a matrix with these properties must be wide or square
- for complex matrices, replace  $AA^T$  in property 3 by  $AA^H$