

# Matrix Analysis

## Part. 3

Chang Jae Ho

jaehochang@konkuk.ac.kr

February 8, 2020

- 1 Systems of Linear Equations
  - Consistency of a System of Equations
  - Systems of Linear Equations
  - Examples: LS Estimation for Rank Deficient Models
- 2 Special Matrices and Matrix Operators
  - Partitioned Matrices
  - The Kronecker Product
  - The vec Operator
  - Hadamard Product
- 3 Matrix Derivatives and Related Topics
  - Multivariable Differential Calculus

# Contents

## 1 Systems of Linear Equations

- Consistency of a System of Equations
- Systems of Linear Equations
- Examples: LS Estimation for Rank Deficient Models

## 2 Special Matrices and Matrix Operators

- Partitioned Matrices
- The Kronecker Product
- The  $\text{vec}$  Operator
- Hadamard Product

## 3 Matrix Derivatives and Related Topics

- Multivariable Differential Calculus

# Consistency of a System of Equations I

## Theorem

$X\beta = c$  is consistent iff  $\text{rank}\left(\begin{bmatrix} X & c \end{bmatrix}\right) = \text{rank}(X)$ .

## Theorem

$X\beta = c$  is consistent iff  $XX^{-}c = c$ .

## Proof.

Suppose that  $X\beta^* = c$  for some solution  $\beta^*$ . Then,  
 $XX^{-}c = XX^{-}X\beta^* = X\beta^* = c$ . Conversely, if  $XX^{-}c = c$ , then  
 $\beta = X^{-}c$ . □

# Consistency of a System of Equations II

## Corollary

*If the  $n \times p$  matrix  $X$  has a full row rank, then  $X\beta = c$  is consistent.*

## Proof.

Since  $XX^{-} = I_n$ ,  $XX^{-}c = c$  so this equation is consistent. □

# Contents

## 1 Systems of Linear Equations

- Consistency of a System of Equations
- Systems of Linear Equations
- Examples: LS Estimation for Rank Deficient Models

## 2 Special Matrices and Matrix Operators

- Partitioned Matrices
- The Kronecker Product
- The  $\text{vec}$  Operator
- Hadamard Product

## 3 Matrix Derivatives and Related Topics

- Multivariable Differential Calculus

# Systems of Linear Equations I

## Theorem

*Suppose that  $X'X\beta = X'y$  is a consistent system of equations. Then,*

$$\beta = (X'X)^{-}X'y.$$

*is a solution.*

## Proof.

Since  $X'X(X'X)^{-}X'y = X'y$  because of the consistency of the equation,

$$X'X\beta = X'X(X'X)^{-}X'y = X'y$$

so  $\beta$  is a solution.



# Contents

## 1 Systems of Linear Equations

- Consistency of a System of Equations
- Systems of Linear Equations
- Examples: LS Estimation for Rank Deficient Models

## 2 Special Matrices and Matrix Operators

- Partitioned Matrices
- The Kronecker Product
- The  $\text{vec}$  Operator
- Hadamard Product

## 3 Matrix Derivatives and Related Topics

- Multivariable Differential Calculus



# One-way Anova I

Consider a model  $y_{ij} = \mu + \tau_i + \varepsilon_{ij}$  where  $i = 1, \dots, k$  and  $j = 1, \dots, n_i$ . We can reformat this model as  $\beta = (\mu, \tau_1, \dots, \tau_k)'$  and

$$X = \begin{bmatrix} 1_{n_1} & 1_{n_1} & 0 & \cdots & 0 \\ 1_{n_2} & 0 & 1_{n_2} & \cdots & 0 \\ 1_{n_3} & 0 & 0 & \cdots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1_{n_k} & 0 & 0 & \cdots & 1_{n_k} \end{bmatrix}; \quad n \times (k+1).$$

Here,  $\text{rank}(X) = k$  so is not full rank. Now, Let us solve a minimization equation  $(X'X)\beta = X'y$  with the generalized inverse. Since  $X$  is a full row rank, we know that  $\beta = (X'X)^-X'y$ . Therefore, we can pick  $\beta = (X'X)^+X'y = X^+y$  or  $\beta = (X'X)^-X'y = X^L y$ .

# Rank Deficient Generalized LS Models I

Consider the regression model

$$y = X\beta + \varepsilon,$$

where  $\varepsilon|X \sim N_n(0, \sigma^2 C)$  and  $C$  is p.d.. Suppose that  $X$  is not full column rank. Then, the GLS estimator is given by

$$\hat{\beta} = (X' C^{-1} X)^{-1} X' C^{-1} y$$

since the solution equation is given as  $X' C^{-1} X \beta = X' C^{-1} y$ .

# Contents

- 1 Systems of Linear Equations
  - Consistency of a System of Equations
  - Systems of Linear Equations
  - Examples: LS Estimation for Rank Deficient Models
- 2 Special Matrices and Matrix Operators
  - Partitioned Matrices
  - The Kronecker Product
  - The  $\text{vec}$  Operator
  - Hadamard Product
- 3 Matrix Derivatives and Related Topics
  - Multivariable Differential Calculus

# Partitioned Matrices, Kim(2018) pp.191~1

## Theorem

Let  $p \times p$  matrix  $A = \begin{bmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{bmatrix}$  and suppose that  $A$ ,  $A_{11}$  and  $A_{22}$  are nonsingular. Then,

$$A^{-1} = \begin{bmatrix} (A^{-1})_{11} & (A^{-1})_{12} \\ (A^{-1})_{21} & (A^{-1})_{22} \end{bmatrix} = \begin{bmatrix} (A_{11} - A_{12}A_{22}^{-1}A_{21})^{-1} & -A_{11}^{-1}A_{12}(A^{-1})_{22} \\ -A_{22}^{-1}A_{21}(A^{-1})_{11} & (A_{22} - A_{21}A_{11}^{-1}A_{12})^{-1} \end{bmatrix}.$$

Here,  $(A_{11} - A_{12}A_{22}^{-1}A_{21})^{-1} = A_{11}^{-1} + A_{11}^{-1}A_{12}(A^{-1})_{22}A_{21}A_{11}^{-1}$  and  $(A_{22} - A_{21}A_{11}^{-1}A_{12})^{-1} = A_{22}^{-1} + A_{22}^{-1}A_{21}(A^{-1})_{11}A_{12}A_{22}^{-1}$ .

## Proof.

Solve

$$AA^{-1} = \begin{bmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{bmatrix} \begin{bmatrix} (A^{-1})_{11} & (A^{-1})_{12} \\ (A^{-1})_{21} & (A^{-1})_{22} \end{bmatrix} = \begin{bmatrix} I_{p_1} & (0) \\ (0) & I_{p_2} \end{bmatrix} = I_p.$$

# Schur's Identity I

Define a block lower triangular matrix  $L$  and an arbitrary matrix  $M$  as

$$L = \begin{bmatrix} I_p & 0 \\ -D^{-1}C & I_q \end{bmatrix}, M = \begin{bmatrix} A & B \\ C & D \end{bmatrix}.$$

After multiplication with the matrix  $L$  the Schur complement appears in the upper  $p \times p$  block. The product matrix is

$$\begin{aligned} ML &= \begin{bmatrix} A & B \\ C & D \end{bmatrix} \begin{bmatrix} I_p & 0 \\ -D^{-1}C & I_q \end{bmatrix} = \begin{bmatrix} A - BD^{-1}C & B \\ 0 & D \end{bmatrix} \\ &= \begin{bmatrix} I_p & BD^{-1} \\ 0 & I_q \end{bmatrix} \begin{bmatrix} A - BD^{-1}C & 0 \\ 0 & D \end{bmatrix}. \end{aligned}$$

## Schur's Identity II

That is, we have effected a Gaussian decomposition

$$\begin{bmatrix} A & B \\ C & D \end{bmatrix} = \begin{bmatrix} I_p & BD^{-1} \\ 0 & I_q \end{bmatrix} \begin{bmatrix} A - BD^{-1}C & 0 \\ 0 & D \end{bmatrix} \begin{bmatrix} I_p & 0 \\ D^{-1}C & I_q \end{bmatrix},$$

The first and last matrices on the RHS have determinant unity.

### Theorem

If  $\exists D^{-1}$ , then

$$\begin{vmatrix} A & B \\ C & D \end{vmatrix} = |D| \times |A - BD^{-1}C|.$$

# Contents

- 1 Systems of Linear Equations
  - Consistency of a System of Equations
  - Systems of Linear Equations
  - Examples: LS Estimation for Rank Deficient Models
- 2 Special Matrices and Matrix Operators
  - Partitioned Matrices
  - The Kronecker Product
  - The  $\text{vec}$  Operator
  - Hadamard Product
- 3 Matrix Derivatives and Related Topics
  - Multivariable Differential Calculus

# The Kronecker Product I

## Definition

For  $m \times n$  matrix  $A$  and  $p \times q$  matrix  $B$ , define  $\otimes$  as

$$A \otimes B = \begin{bmatrix} a_{11}B & a_{12}B & \cdots & a_{1n}B \\ a_{21}B & a_{22}B & \cdots & a_{2n}B \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1}B & a_{m2}B & \cdots & a_{mn}B \end{bmatrix}; mp \times nq.$$



# The Kronecker Product II

## Theorem

*Let  $A, B, C$  be any matrices and  $a, b$  be any two vectors. Then,*

- a**  $a \otimes A = A \otimes a = aA$  for any scalar  $a$
- b**  $(aA) \otimes (\beta B) = a\beta(A \otimes B)$  for any scalars  $a, \beta$
- c**  $(A \otimes B) \otimes C = A \otimes (B \otimes C)$
- d**  $(A + B) \otimes C = (A \otimes C) + (B \otimes C)$
- e**  $A \otimes (B + C) = (A \otimes B) + (A \otimes C)$
- f**  $(A \otimes B)' = A' \otimes B'$
- g**  $ab' = a \otimes b' = b' \otimes a$

Prove for exercises.

# The Kronecker Product III

## Theorem

*Pick any  $A, B, C, D$  so that  $AC, BD$  exist. Then,*

$$(A \otimes B)(C \otimes D) = AC \otimes BD$$

## Proof.

$$\Gamma := \begin{bmatrix} a_{11}B & \cdots & a_{1n}B \\ \vdots & \ddots & \vdots \\ a_{m1}B & \cdots & a_{mn}B \end{bmatrix} \begin{bmatrix} c_{11}D & \cdots & c_{1c}D \\ \vdots & \ddots & \vdots \\ c_{n1}D & \cdots & c_{nc}D \end{bmatrix}$$

where  $\Gamma_{ij} = \sum_{k=1}^n a_{ik}c_{kj}BD$  for  $i = 1, \dots, m$  and  $j = 1, \dots, c$ . □

# The Kronecker Product IV

Proof.

Here,  $\Gamma_{ij} = (AC)_{ij}BD$  so

$$\Gamma = \begin{bmatrix} (AC)_{11}D & \cdots & (AC)_{1c}D \\ \vdots & \ddots & \vdots \\ (AC)_{m1}D & \cdots & (AC)_{mc}D \end{bmatrix} = AC \otimes BD.$$

□

Theorem

*For square matrices  $A, B$ ,  $\text{tr}(A \otimes B) = \text{tr}(A)\text{tr}(B)$ .*

Proof is easy for this one.

# The Kronecker Product V

## Theorem

*Let  $A, B$  be  $m \times n, p \times q$  matrices. Then,  $(A \otimes B)^- = A^- \otimes B^-$  for any generalized inverses  $A^-, B^-$ .*

## Proof.

$$(A \otimes B)(A^- \otimes B^-)(A \otimes B) = (AA^- \otimes BB^-)(A \otimes B) = AA^-A \otimes BB^-B = A \otimes B.$$

□

# The Kronecker Product VI

## Theorem

*Let  $\{\lambda_i\}_{i \in I}$ ,  $\{\partial_j\}_{j \in J}$  be eigenvalues of the  $m \times m$  matrix  $A$  and  $p \times p$  matrix  $B$ . Then the eigenvalues of  $A \otimes B$  is given by  $\{\lambda_i \partial_j\}_{(i,j) \in I \times J}$ .*

## Theorem

$$|A \otimes B| = |A|^p |B|^m$$

## Proof.

$$|A \otimes B| = \prod_{i \in I} \prod_{j \in J} (\lambda_i \partial_j) = \prod_{i \in I} \left( \lambda_i^p \prod_{j \in J} \partial_j \right) = \prod_{i \in I} (\lambda_i^p |B|) = |A|^p |B|^m.$$



# The Kronecker Product VII

## Theorem

Let  $A, B$  be  $m \times n, p \times q$  matrices. Then,

$$\text{rank}(A \otimes B) = \text{rank}(A)\text{rank}(B).$$

## Proof.

Let  $\{\lambda_i\}_{i \in I}, \{\partial_j\}_{j \in J}$  be eigenvalues of  $AA', BB'$ . Since  $(A \otimes B)(A \otimes B)'$  is symmetric,

$$\text{rank}(A \otimes B) = \text{rank} \{ (A \otimes B)(A \otimes B)' \} = \text{rank}(AA' \otimes BB').$$

Since  $AA', BB'$ , and their Kronecker product are all symmetric,

$$\begin{aligned} \text{rank}(AA' \otimes BB') &= \#\{\lambda_i \partial_j \neq 0; i \in I, j \in J\} = \#\{\lambda_i \neq 0; i \in I\} \times \#\{\partial_j \neq 0; j \in J\} \\ &= \text{rank}(AA')\text{rank}(BB') = \text{rank}(A)\text{rank}(B). \end{aligned}$$

□

# An Example: Balanced One-way ANOVA I

For the balanced One-way ANOVA model with an intercept, we can find an LS estimator as

$$\begin{aligned}
 \hat{\beta} &= (X'X)^{-1}X'y = \left\{ \begin{bmatrix} 1'_k \otimes 1'_n \\ l_k \otimes 1'_n \end{bmatrix} \begin{bmatrix} 1_k \otimes 1_n & l_k \otimes 1_n \end{bmatrix} \right\}^{-1} \begin{bmatrix} 1'_k \otimes 1'_n \\ l_k \otimes 1'_n \end{bmatrix} y \\
 &= \begin{bmatrix} 1'_k 1_k \otimes 1'_n 1_n & 1'_k \otimes 1'_n 1_n \\ 1_k \otimes 1'_n 1_n & l_k \otimes 1'_n 1_n \end{bmatrix}^{-1} \begin{bmatrix} 1'_k \otimes 1'_n \\ l_k \otimes 1'_n \end{bmatrix} y \\
 &= \begin{bmatrix} nk & n1'_k \\ n1_k & nl_k \end{bmatrix}^{-1} \begin{bmatrix} 1'_k \otimes 1'_n \\ l_k \otimes 1'_n \end{bmatrix} y = \begin{bmatrix} 1/(nk) & 0' \\ 0 & \frac{l_k - P_{1_k}}{n} \end{bmatrix} \begin{bmatrix} 1'_k \otimes 1'_n \\ l_k \otimes 1'_n \end{bmatrix} y \\
 &= \begin{bmatrix} (1'_k \otimes 1'_k)/(nk) \\ (l_k \otimes 1'_n)/n - (1_k 1'_k \otimes 1'_n)/(nk) \end{bmatrix} y.
 \end{aligned}$$

# Contents

- 1 Systems of Linear Equations
  - Consistency of a System of Equations
  - Systems of Linear Equations
  - Examples: LS Estimation for Rank Deficient Models
- 2 Special Matrices and Matrix Operators
  - Partitioned Matrices
  - The Kronecker Product
  - The vec Operator
  - Hadamard Product
- 3 Matrix Derivatives and Related Topics
  - Multivariable Differential Calculus



# The $\text{vec}$ Operator I

For an  $m \times n$  matrix  $A = \begin{bmatrix} A_1 & A_2 & \cdots & A_n \end{bmatrix}$ , define the  $\text{vec}$  operator as

$$\text{vec}(A) = \begin{bmatrix} A_1 \\ A_2 \\ \vdots \\ A_n \end{bmatrix}.$$

## Theorem

*Let  $a, b$  be any vectors, while  $A, B$  are two matrices of the same size. Then,*

- a**  $\text{vec}(a) = \text{vec}(a') = a,$
- b**  $\text{vec}(ab') = b \otimes a,$
- c**  $\text{vec}(aA + \beta B) = a\text{vec}(A) + \beta\text{vec}(B)$  for scalars  $a, \beta.$

# The vec Operator II

## Theorem

Let  $A, B$  be  $m \times n, m \times n$  matrices. Then,

$$\text{tr}(A'B) = \text{vec}(A)' \text{vec}(B).$$

## Proof.

$$\text{tr}(A'B) = \sum_{i=1}^n a'_i b_i = \begin{bmatrix} a'_1 & \cdots & a'_n \end{bmatrix} \begin{bmatrix} b_1 \\ \vdots \\ b_n \end{bmatrix} = \text{vec}(A)' \text{vec}(B).$$



# The vec Operator III

## Theorem

Let  $A, B, C$  be  $m \times n, n \times p, p \times q$ , respectively. Then,  $\text{vec}(ABC) = (C' \otimes A)\text{vec}(B)$ .

## Proof.

Note that  $B = \begin{bmatrix} B_1 & \cdots & B_p \end{bmatrix}$  can be written as  $B = \sum_{i=1}^p B_i e_i'$ . Thus,

$$\begin{aligned} \text{vec}(ABC) &= \text{vec} \left\{ A \left( \sum_{i=1}^p B_i e_i' \right) C \right\} = \sum_{i=1}^p \text{vec}(AB_i e_i' C) \\ &= \sum_{i=1}^p \text{vec} \{ AB_i (C' e_i)' \} = \sum_{i=1}^p C' e_i \otimes AB_i \\ &= (C' \otimes A) \sum_{i=1}^p (e_i \otimes b_i) = (C' \otimes A) \text{vec}(B). \end{aligned}$$



# The vec Operator IV

## Theorem

*Let  $A, B, C, D$  be matrices of sizes such that  $ABCD$  exists and is square. Then,*

$$\text{tr}(ABCD) = \text{vec}(A')'(D' \otimes B)\text{vec}(C).$$

## Proof.

We know that  $\text{tr}(ABCD) = \text{vec}(A')'\text{vec}(BCD)$ . Therefore,

$$\text{tr}(ABCD) = \text{vec}(A')'(D' \otimes B)\text{vec}(C).$$



# Contents

- 1 Systems of Linear Equations
  - Consistency of a System of Equations
  - Systems of Linear Equations
  - Examples: LS Estimation for Rank Deficient Models
- 2 Special Matrices and Matrix Operators
  - Partitioned Matrices
  - The Kronecker Product
  - The  $\text{vec}$  Operator
  - Hadamard Product
- 3 Matrix Derivatives and Related Topics
  - Multivariable Differential Calculus

# Hadamard Product I

## Definition

For  $m \times n$  matrices  $A, B$ , define  $\odot$  as

$$A \odot B = \begin{bmatrix} a_{11}b_{11} & \cdots & a_{1n}b_{1n} \\ \vdots & \ddots & \vdots \\ a_{m1}b_{m1} & \cdots & a_{mn}b_{mn} \end{bmatrix},$$

that is, an elementwise product of the identically sized matrices.

**Remark** In  $\mathbb{R}$ ,  $A \star B$  is the Hadamard product for  $m \times n$  matrices  $A, B$ .

# Hadamard Product II

## Theorem

For the identically sized matrices  $A, B, C$ ,

**a**  $A \odot B = B \odot A,$

**b**  $(A \odot B) \odot C = A \odot (B \odot C)$

**c**  $(A + B) \odot C = A \odot C + B \odot C$

**d**  $(A \odot B)' = A' \odot B'$

**e**  $A \odot (0) = (0)$

**f**  $A \odot 1_m 1_n' = A,$

**g**  $A \odot I_m = \text{diag}(a_{11}, \dots, a_{mm})$  if  $A$  is square,

**h**  $C(A \odot B) = (CA) \odot B = A \odot (CB)$  and  $(A \odot B)C = (AC) \odot B = A \odot (BC)$  if  $A, B, C$  are square and  $C$  is diagonal,

**i**  $ab' \odot cd' = (a \odot c)(b \odot d)'$  for the identically sized vectors  $a, c$  and  $b, d$ .

# Hadamard Product III

*pp.268 ~ of Schott(1997)*

## Theorem

*For  $m \times n$  matrices  $A, B$ ,  $\text{rank}(A \odot B) \leq \text{rank}(A)\text{rank}(B)$ .*

## Theorem

*For  $m \times n$  matrices  $A, B$  and  $m \times 1, n \times 1$  vectors  $x, y$ ,*

$$\text{1} \quad 1'_m (A \odot B) 1_n = \text{tr}(AB'),$$

$$\text{2} \quad x'(A \odot B)y = \text{tr}(D_x A D_y B')$$

*where  $D_x$  denotes a diagonal matrix with  $x = (x_1, \dots, x_m)$ .*

## Theorem

*Let  $A, B$  be symmetric matrices. Then,*

**1**  $A \odot B$  is semi-p.d. if  $A, B$  are semi-p.d..

**2**  $A \odot B$  is p.d. if  $A, B$  are p.d..



# Hadamard Product IV

## Theorem

*Let  $A, B$  be symmetric matrices. If  $B$  is p.d. and  $A$  is semi-p.d. with positive diagonal elements, then  $A \odot B$  is p.d.*

## Theorem

*If  $A$  is an  $m \times m$  p.d. matrix, then*

$$|A| \leq \prod_{i=1}^m a_{ii}$$

*with equality iff  $A$  is a diagonal matrix.*

# Hadamard Product V

## Proof.

By the constrained Rayleigh Quotient in *pp.105 ~ 106* of Schott(1997), if we partition  $A$  as  $A = \begin{bmatrix} A_1 & \cdots & A_i & A_{i+1} & \cdots & A_m \end{bmatrix} = \begin{bmatrix} A_{\sim i} & A_{i+1 \sim} \end{bmatrix}$ ,

$$\hat{\lambda}_i = \min_{x \in S_h; \|x\|_2=1} x'Ax$$

where  $S_h = \{A_{\sim i}x; x \in \mathbb{R}^i\}$  for  $i = 1, \dots, m$ . So, we know that  $\hat{\lambda}_i \leq a_{ii}$  for  $x = e_i$ . Therefore,  $\prod_{i=1}^m \hat{\lambda}_i \leq \prod_{i=1}^m a_{ii}$  since  $A$  is p.d.. □

The extremal properties of  $\hat{\lambda}(A \odot B)$  is given in *pp.274 ~* of Schott(1997).

# Hadamard Product VI

## Corollary

*Let  $B$  be an  $m \times m$  nonsingular matrix. Then,*

$$|B|^2 \leq \prod_{i=1}^m \left( \sum_{j=1}^m b_{ij}^2 \right)$$

*with equality iff the rows of  $B$  are orthogonal.*

This is obvious because  $|BB'| = |B| \cdot |B'| = |B|^2$  and  $(BB')_{ii} = \sum_{j=1}^m b_{ij}^2$ .

## Theorem

*Let  $A, B$  be  $m \times m$  semi-p.d. matrices. Then,*

$$|A| \prod_{i=1}^m b_{ii} \leq |A \odot B|.$$