Lecture 07 Linear Algebra

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LAST LECTURE REVIEW

- ► Matrices:
 - ► Matrix Operators
 - Rank Trace
 - ► The Determinant
 - ► Positive and Negative Definite Matrices
 - Linear Independence

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REVIEW ASSIGNMENT

- 1. Problem Set 06 solutions are available on Github.
- 2. Any issues or problems **You** would like to discuss?

DAILY ICEBREAKER

- ► Attendance via prompt:
 - ► Name
 - ▶ Program and track
 - ▶ Daily icebreaker subject...



Topic: Linear Algebra

MOTIVATION

- ► General background
 - ► The algebra of manipulating matrices.
 - ► A compact, efficient method for working with systems of equations (e.g., a spreadsheet).
- ▶ Why do economists' care?
 - ▶ These is the foundation of how economists manipulate data.
- ► Application in this career
 - ► Important applications in econometrics.
 - ▶ Useful whenever you are creating or modifying an estimator.

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OVERVIEW

- 1. Systems of Linear Equations
- 2. Gaussian Elimination
- 3. Linear Operators
- 4. Existence of a Solution
- 5. Cramer's Rule
- 6. Eigenvalues and the Characteristic Equation
- 7. Leading Principle Minors
- 8. Regression as a Matrix
- 9. Centering Matrix
- 10. Residual Maker

1. Systems of Linear Equations

Linear Functions: y = Ax with elements y_i of y such that $y_i = a_i^T x$.

$$a_i = \frac{\partial y_i}{\partial x} = \frac{\partial}{\partial x} (a_i^T x)$$

ightharpoonup Linear System: m equations for n unknown variables.

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

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

2. Gaussian Elimination

► Augmented Matrix

$$\begin{pmatrix} a_{11} & a_{12} & \cdots & a_{1n}|b_1 \\ a_{21} & a_{22} & \cdots & a_{2n}|b_2 \\ \vdots & \vdots & \ddots & \vdots|\vdots \\ a_{m1} & a_{m2} & \cdots & a_{mn}|b_m \end{pmatrix}$$

2. Gaussian Elimination

- ► Row Echelon Form
 - ► Interchange any two rows.
 - ► Change row by adding a multiple of another row.
 - ▶ Multiply each element in a row by a non-zero scalar.

$$\begin{pmatrix} 1 & 0 & \cdots & 0|b_1 \\ 0 & 1 & \cdots & 0|b_2 \\ \vdots & \vdots & \ddots & \vdots|\vdots \\ 0 & 0 & \cdots & 1|b_m \end{pmatrix}$$

3. LINEAR OPERATORS

► Product Rule:

$$\frac{\partial a^T b}{\partial x} = \frac{\partial a^T}{\partial x} b + \frac{\partial b^T}{\partial x} a$$

▶ Quadratic Form:

$$x^T A x = \sum_{i=1}^n \sum_{i=1}^n x_i x_j a_{ij}$$

4. Existence of Solution

- ▶ Important to know that a solution for a system exists.
- ▶ A system of linear equations with coefficient matrix A and an augmented matrix \hat{A} has a solution iff:

$$rank\hat{A} = rankA$$

- ▶ There are infinite solutions if #rows(A) < #cols(A).
 - ▶ More unknown variables than observations.
- ► Non-singular Square Matrix: Ensure only one solutions exists iff

$$\#rows(A) = \#cols(A) = rank(A)$$

5. CRAMER'S RULE

A unique solution $x = (x_1, \dots, x_n)$ for $n \times n$ system Ax = b is

$$x_i = \frac{detB_i}{detA} \forall i = 1, \dots, n$$

▶ Where B_i replaces the i'th column of A with the b vector.

6. EIGENVALUES AND THE CHARACTERISTIC EQUATION

► Characteristic Equation

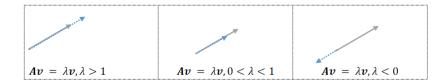
$$Ac = \lambda c$$

- ightharpoonup Characteristic vectors: (c, λ)
- ightharpoonup Eigenvectors: c
- ightharpoonup Eigenvalues: λ

$$Ac = \lambda Ic \iff (A - \lambda I)c = 0$$

► Homogeneous system has non-zero solution if it is singular and has zero determinant: $det(A - \lambda I) = 0$

EIGENVECTORS SCALED BY EIGENVALUES



7. LEADING PRINCIPLE MINORS

- ► A way to test for matrix definiteness.
- ▶ Leading Principal Sub-matrix: Let A be a $N \times N$ matrix. The K'th order principal sub-matrix of A obtained by deleting the last N - K rows and the last N - K columns.
- ▶ Leading Principal Minor: The determinant of the *K*'th order leading principal sub-matrix.
- \triangleright Ex., for a 3 × 3 matrix, the leading principal minors are:

$$|a_{11}|, \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix}, \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{bmatrix}$$

7. LEADING PRINCIPLE MINORS

- ▶ Positive Definite: Iff all N leading principal minors are ≥ 0 .
- ightharpoonup Negative Definite: Iff the N leading principal minors alternate signs

$$\mbox{\it det}(A_1) < 0, \mbox{\it det}(A_2) > 0, \mbox{\it det}(A_3) < 0$$
 , etc.

▶ Indefinite: Leading principal minors follow any other order.

8. REGRESSION AS A MATRIX

► A linear OLS model:

$$Y_{N\times 1} = X_{N\times K}\beta_{K\times 1} + e_{N\times 1}$$

▶ With the goal of selecting $\hat{\beta}$ that minimizes squared predicted errors.

$$\hat{e}^T\hat{e} = (Y - X\hat{\beta})^T(Y - X\hat{\beta})$$

8. Regression as a Matrix

▶ So, taking the first derivative we can get:

$$X^{T}X\hat{\beta} - X^{T}Y = 0$$

$$\Longrightarrow$$

$$-X^{T}(Y - X\hat{\beta}) = -X^{T}\hat{e} = 0$$

► And the solution to OLS is:

$$\hat{\beta} = (X^T X)^{-1} X^T Y$$

9. CENTERING MATRIX

- ► It is common in statistics to transform data to be 'deviations from the mean'.
- ▶ This can be done by creating the "centering matrix".
- ▶ First, create a multiplier $\frac{1}{N}ii^T$.

$$\frac{1}{N}ii^{T} = \frac{1}{N} \begin{pmatrix} 1\\1\\\vdots\\1 \end{pmatrix} \begin{pmatrix} 1 & 1 & \cdots & 1 \end{pmatrix} = \begin{pmatrix} \frac{1}{N} & \frac{1}{N} & \cdots & \frac{1}{N}\\\vdots & \vdots & \ddots & \vdots\\\frac{1}{N} & \frac{1}{N} & \cdots & \frac{1}{N} \end{pmatrix}$$

► Then define vectors of means.

$$i\bar{x} = \frac{1}{N}ii^{T}x = \begin{pmatrix} \frac{1}{N} & \frac{1}{N} & \cdots & \frac{1}{N} \\ \vdots & \vdots & \ddots & \vdots \\ \frac{1}{N} & \frac{1}{N} & \cdots & \frac{1}{N} \end{pmatrix} \begin{pmatrix} x_{1} \\ x_{2} \\ \vdots \\ x_{N} \end{pmatrix} = \begin{pmatrix} \bar{x} \\ \bar{x} \\ \vdots \\ \bar{x} \end{pmatrix}$$

9. CENTERING MATRIX

► The vector of derivatives can be expressed as:

$$x - i\bar{x} = \begin{pmatrix} x_1 - \bar{x} \\ x_2 - \bar{x} \\ \vdots \\ x_N - \bar{x} \end{pmatrix} = x - \frac{1}{N}ii^Tx = (I - \frac{1}{N}ii^T)x = M^0x$$

▶ Summing these derivatives can be written as follows:

$$\sum_{i=1}^{N} (x_i - x)^2 = (x - i\bar{x})^T (x - i\bar{x}) = (M^0 x)^T (M^0 x) = x^T M^{0T} M^0 x$$

10. RESIDUAL MAKER

▶ Define

$$M = I - X(X^T X)^{-1} X^T$$

► Then

$$\hat{e} = Y - X\hat{\beta} = (I - X(X^TX)^{-1}X^T)Y = MY$$

► Residual Maker: M

PRACTICE: LINEAR ALGEBRA

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REVIEW OF LINEAR ALGEBRA

- 1. Systems of Linear Equations
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ASSIGNMENT

- ► Readings on Functions before Lecture 08:
- ► Assignment:
 - ► Problem Set 07 (PS07)
 - ► Solution set will be available following end of Lecture 08
- ► Struggling?
 - 1. Read the 'Encouraged Reading'
 - 2. Review 'Supplementary material'
 - 3. Reach out directly