AMES Class Notes – Week 9, Monday: Linear Algebra, cont

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Squaring matrices 1

Consider a matrix $\underset{N\times 1}{V}$ that you'd like to square.

$$V^2 = \underset{1 \times N}{V} \underset{N \times 1}{T} V \tag{1}$$

$$= Z \tag{2}$$

Sometimes you'll have a weighting matrix, $\mathop{M}_{N\times N}$

$$\begin{array}{ccc}
V & M & V & Z \\
1 \times N & N \times NN \times 1 & 1 \times 1
\end{array}$$
(3)

Idempotent matrix

Defn: A matrix that when squared equals itself

$$Z^T Z = Z (4)$$

3 Kronecker product

Kronecker products expand out. You see this is computer algorithms and data management problems. It's a common trick.

Let

$$A = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix} \tag{5}$$

$$Z = \begin{bmatrix} Z_1 \\ Z_3 \end{bmatrix} \tag{6}$$

$$A = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix}$$

$$Z = \begin{bmatrix} Z_1 \\ Z_3 \end{bmatrix}$$

$$A \otimes Z = \begin{bmatrix} a_1 1z_1 & a_1 2z_1 \\ a_1 1z_2 & a_1 2z_2 \\ a_2 1z_1 & a_2 2z_1 \\ a_2 1z_2 & a_2 2z_2 \end{bmatrix}$$

$$(5)$$

$$(6)$$

Trace $\mathbf{4}$

Defn: the product of all diagonal elements. The trace of A is $a_{11} * a_{22}$

5 Linear regression

Consider the equation estimating equation

$$y = a + bx + cz + \epsilon \tag{8}$$

We can rewrite this as

$$Y_{N\times 1} = X \beta_{N\times K_{K\times 1}} + \epsilon_{N\times 1}$$
(9)

(10)

where

$$\beta = \begin{bmatrix} a \\ b \\ c \end{bmatrix} \tag{11}$$

$$X = \begin{bmatrix} 1 & x_1 & z_1 \\ 1 & x_2 & z_2 \\ \vdots & \dots \end{bmatrix}$$
 (12)

$$Y = \begin{bmatrix} y_1 \\ y_2 \\ \vdots \end{bmatrix} \tag{13}$$

Goal: Solve for β by minimizing the sum of square errors.

Solve for the error term:

$$Y_{N\times 1} - X\beta = \epsilon \\
_{N\times K} = N\times 1$$
(14)

Square the error term:

$$\epsilon_{1\times N} {}^{T} \epsilon_{N\times 1} = (Y - X\beta)^{T} (Y - X\beta) \\
{}_{1\times N} {}^{T} (Y - X\beta) \tag{15}$$

$$= \underset{1 \times N}{\overset{T}{Y}} \underset{N \times 1}{\overset{T}{Y}} - \underset{1 \times K}{\overset{T}{X}} \underset{K \times N}{\overset{T}{X}} \underset{N \times 1}{\overset{T}{Y}} - \underset{1 \times N}{\overset{T}{X}} \underset{N \times K}{\overset{K}{K}} + \underset{1 \times K}{\overset{T}{X}} \underset{K \times N}{\overset{T}{X}} \underset{N \times K}{\overset{K}{K}} \times 1$$
(16)

$$=Y_{1\times 1}^T Y - 2\beta^T X^T Y + \beta^T X^T X \beta$$

$$(17)$$

The squared error term is a scalar $\underset{(1\times 1)}{\epsilon^T \epsilon}$.

We want to minimize the squared error term by choosing β . To do so, take derivative of squared error and set equal to zero, solve for β .

$$\frac{\partial \epsilon^T \epsilon}{\partial \beta} = 0 - 2X_{K \times 1}^T Y + 2X_{K \times 1}^T X \beta = 0 \implies (18)$$

$$2X^TY = 2X^TX\beta \implies (19)$$

$$\beta = \left(X^T X\right)^{-1} X^T Y \tag{20}$$

Where $\left(X^TX\right)^{-1}$ is an inversion (because we cannot divide matrices).

6 Life cycle assessment example

Let's consider and input output table (Ag, Transportation, Manufactured)

$$A = \begin{bmatrix} A & T & M \\ A & & & \\ T & & & \\ M & & & \end{bmatrix}$$

$$(21)$$

And our final demand (the demand for goods by consumers)

$$d = \begin{bmatrix} A \\ T \\ M \end{bmatrix} \tag{22}$$

Total amount of goods X is,

$$X = A X + d
 3 \times 33 \times 1 + 3 \times 1$$
(23)

Solve for X by multiplying with the identity matrix

$$\underset{3\times 3}{I}X - AX = d \tag{24}$$

$$(I - A)X = d (25)$$

$$\implies (I - A)^{-1}d = X \tag{26}$$

Total amount of goods is different than final demand because we need input good for the final good. A bunch of intermediate products are required to make a computer.

7 R and Excel

Watch the video! Couple of notes

- to invert a matrix in R you use the solve() command