

Matrix Inverses

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October 21, 2016

Matrix Inverse - Definition

Definition

Let A be an $n \times n$ matrix. The inverse of A , denoted A^{-1} is the matrix which satisfies

$$A \cdot A^{-1} = A^{-1} \cdot A = I_n.$$

Note that in general $AB \neq BA$ for matrices, but for the inverse matrix $AA^{-1} = A^{-1}A$.

Matrix Inverse - Example

Define the matrix

$$M = \begin{pmatrix} 1 & 0 & 3 \\ 6 & -2 & 8 \\ -1 & 1 & 4 \end{pmatrix}$$

Using the MATLAB command `inv(M)`, we obtain:

$$M^{-1} = \begin{pmatrix} 4 & -0.75 & -1.5 \\ 8 & -1.75 & -2.5 \\ -1 & 0.25 & 0.5 \end{pmatrix}$$

Algorithm for Computing Inverses

Note that the inverse matrix takes input $(1, 0, 0)^T$ and gives output the solution to $Ax_1 = (1, 0, 0)^T$, similarly,

$$(0, 1, 0)^T \rightarrow \text{solution of } Ax_2 = (0, 1, 0)^T$$

$$(0, 0, 1)^T \rightarrow \text{solution of } Ax_3 = (0, 0, 1)^T$$

This will be enough to determine our inverse matrix

$$A^{-1} = [x_1 \mid x_2 \mid x_3] .$$

Vector Norms

A *vector norm* is a mapping from the set of vectors to the set of nonnegative real numbers. It has the following properties:

1. $\|a\mathbf{v}\| = |a| \cdot \|\mathbf{v}\|$ for real scalars a ,
2. $\|\mathbf{v} + \mathbf{w}\| \leq \|\mathbf{v}\| + \|\mathbf{w}\|$ (triangle inequality), and
3. $\|\mathbf{v}\| = 0$ if and only if \mathbf{v} is the zero vector.

$\|\mathbf{v}\|_e$: The Euclidean norm.

$\|\mathbf{v}\|_p$: The p -norm, for $p \geq 1$.

$\|\mathbf{v}\|_1$: The 1-norm.

$\|\mathbf{v}\|_\infty$: The infinity-norm.

Matrix Norms

A *matrix norm* is a mapping from the set of matrices to the set of nonnegative real numbers, with all the same properties as the vector norm.

A matrix norm is induced by a vector norm if you define it as $\|A\| = \max_{x \in \mathbb{R}^n} \|Ax\| / \|x\|$, using the vector norm on the right-hand side. In each of the following,

$\|M\|_f$: The Frobenius norm. (not induced)

$\|M\|_1$: The column-sum norm. (induced by vector 1-norm)

$\|M\|_\infty$: The row-sum norm. (induced by vector ∞ -norm)

$\|M\|_2$: The spectral norm. (induced by vector Euclidean norm)

The Spectral Norm

The spectral norm is the square root of the largest eigenvalue of $A^T A$.

Condition Number

Definition

The **condition number** of a function is a measurement of how much the output changes relative to small changes in the input.

Definition

The **matrix condition number** of A is the condition number of the function which takes b as input and then solves for $Ax = b$.

Precisely, if $Ax = b$ and $A(x + \Delta x) = b + \Delta b$, it is equal to

$$\max_{b, \Delta b} \frac{\|\Delta x\| / \|x\|}{\|\Delta b\| / \|b\|}.$$

Matrix Condition Number

Let's begin from this relationship:

$$A(x + \Delta x) = b + \Delta b \iff A(\Delta x) = \Delta b \iff \Delta x = A^{-1}\Delta b.$$

This means that:

$$\frac{\|\Delta x\|/\|x\|}{\|\Delta b\|/\|b\|} = \frac{\|A^{-1}\Delta b\|/\|A^{-1}b\|}{\|\Delta b\|/\|b\|} = \frac{\|A^{-1}\Delta b\|}{\|\Delta b\|} \cdot \frac{\|b\|}{\|A^{-1}b\|}.$$

Substitute $b = Ac$ for some vector c .

$$= \frac{\|A^{-1}\Delta b\|}{\|\Delta b\|} \cdot \frac{\|Ac\|}{\|c\|}.$$

Maximizing over Δb and c gives the product $\|A^{-1}\| \cdot \|A\|$, using the operator norm.

Condition number - Example

$$M = \begin{pmatrix} 1 & 0 & 3 \\ 6 & -2 & 8 \\ -1 & 1 & 4 \end{pmatrix}, M^{-1} = \begin{pmatrix} 4 & -0.75 & -1.5 \\ 8 & -1.75 & -2.5 \\ -1 & 0.25 & 0.5 \end{pmatrix}$$

$\text{norm}(M) \cdot \text{norm}(M^{-1})$, computed in MATLAB, gives 105.6137. The matrix is not well-conditioned.