Math Bootcamp

UC San Diego

Umberto Mignozzetti

Inverse Matrix

Determinants

Vector Spaces

Eigenvectors and Eigenvalues

Applications

Inverse 3

Definition: Inverse Matrix

Let A a matrix $n \times n$. The matrix B of $n \times n$ is an called the **inverse** for A if AB = BA = I (where I is the Identity Matrix).

If AB = I them B is a right inverse.

If BA = I them B is a left inverse.

Example: Calculate the inverse of:

$$\begin{pmatrix} a & b \\ c & d \end{pmatrix}^{-1} = \frac{1}{ad - bc} \begin{pmatrix} d & -b \\ -c & a \end{pmatrix}$$

Important: These matrices generally are non-singular (and therefore invertible) if, and only if, ad - bc differs from 0.

Inverse matrices are very useful. See the theorem below:

Theorem For any square matrix A, the following statements are equivalent:

- (a) A is invertible.
- (b) A has a right inverse.
- (c) A has a left inverse.
- (d) Every system Ax = b has at least one solution for every b.
- (e) Every system Ax = b has at most one solution for every b.
- (f) A is nonsingular.
- (g) A has maximal rank n.

If A is nonsingular (invertible) them for Ax = b $x = A^{-1}b$. Solve the following system using the inverse of the matrix.

$$x - 2y = 8$$

$$3x + y = 3$$

Your turn! Compute letter (c) of the first set and letter (b) of the second set:

a)
$$\begin{pmatrix} 2 & 1 \\ 1 & 1 \end{pmatrix}$$
, b) $\begin{pmatrix} 4 & 5 \\ 2 & 4 \end{pmatrix}$, c) $\begin{pmatrix} 2 & 1 \\ -4 & -2 \end{pmatrix}$,

d) $\begin{pmatrix} 2 & 4 & 0 \\ 4 & 6 & 3 \\ -6 & -10 & 0 \end{pmatrix}$, e) $\begin{pmatrix} 2 & 1 & 0 \\ 6 & 2 & 6 \\ -4 & -3 & 9 \end{pmatrix}$,

f) $\begin{pmatrix} 2 & 6 & 0 & 5 \\ 6 & 21 & 8 & 17 \\ 4 & 12 & -4 & 13 \\ 0 & -3 & -12 & 2 \end{pmatrix}$.

Invert the coefficient matrix to solve the following systems of equations:

a)
$$2x_1 + x_2 = 5$$

 $x_1 + x_2 = 3;$

$$2x_1 + x_2 = 4$$

b) $6x_1 + 2x_2 + 6x_3 = 20$
 $-4x_1 - 3x_2 + 9x_3 = 3;$

Determinants

What are determinants?

- ▶ Determinants: a number that determines whether a square-matrix is singular or not.
- ► Singular matrix: has no inverse.
- ▶ If the determinant is zero, then the matrix is singular.
- ► It also has an important role in optimization theory (but let's leave that for the next classes)

We define the Determinant computations inductively:

- ightharpoonup Starting with a constant, we have: det(a) = a.
- ▶ Now let $A_{2\times 2}$. We define the determinant as:

$$det(A) = a_{11}det(a_{22}) - a_{12}det(a_{21})$$

▶ But what about higher dimension matrices?

We need a generalization of this technique.

Definition: Minor of a Matrix

Let $A_{n\times n}$. We define the matrix's minor M_{ij} as the determinant of the matrix obtained by deleting the ith row and the jth column.

Example:

$$A = \begin{pmatrix} 1 & 2 & 4 \\ 2 & 3 & 1 \\ 1 & 1 & 2 \end{pmatrix}$$

Then:

$$M_{11} = \det \begin{pmatrix} 3 & 1 \\ 1 & 2 \end{pmatrix}$$

What about M_{12} and M_{33} ?

Definition: Cofactor

The cofactor C_{ij} is defined as:

$$Cij = (-1)^{i+j} M_{ij}$$

Examples:

$$C_{11} = (-1)^{1+1} M_{11} = M_{11}$$

$$C_{12} = (-1)^{1+2} M_{12} = -M_{12}$$

And the determinant is defined as the sum of the cofactors along the row or column of our choice. This technique is called **cofactor expansion**.

Definition The determinant of a 3×3 matrix is given by

$$\det\begin{pmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{pmatrix} = a_{11}C_{11} + a_{12}C_{12} + a_{13}C_{13}$$

$$= a_{11}M_{11} - a_{12}M_{12} + a_{13}M_{13}$$

$$= a_{11} \cdot \det\begin{pmatrix} a_{22} & a_{23} \\ a_{32} & a_{33} \end{pmatrix} - a_{12} \cdot \det\begin{pmatrix} a_{21} & a_{23} \\ a_{31} & a_{33} \end{pmatrix}$$

$$+ a_{13} \cdot \det\begin{pmatrix} a_{21} & a_{22} \\ a_{31} & a_{32} \end{pmatrix}.$$

Compute the determinant of A:

$$A = \begin{pmatrix} 1 & 2 & 0 \\ 0 & 3 & 1 \\ 1 & 0 & 2 \end{pmatrix}$$

Exercise 13

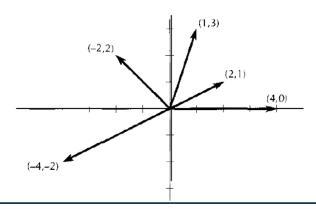
Your turn!

Compute the determinant of:

$$\begin{pmatrix} 1 & 0 & 0 \\ 2 & 3 & 0 \\ 3 & 4 & 1 \end{pmatrix}$$

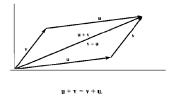
In Linear Algebra we frequently deal with vectors.

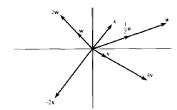
Vectors, for social scientists, are simply a collection of multidimensional data.



The algebra of vectors is mostly done component-wise.

- ► Sum?
- ► Scalar multiplication?





Definition: Length of a Vector

Let the vector $\mathbf{x} = (\mathbf{x_1}, \mathbf{x_2}, \cdots, \mathbf{x_n})$. The length of \mathbf{x} is defined as:

$$||\mathbf{x}|| = \sqrt{\mathbf{x_1^2 + \cdots + x_n^2}}$$

Compute the length of vector $\mathbf{x} = (\mathbf{1}, \mathbf{2}, \mathbf{3})$

Definition: Unit Vector Vector with length equal to 1.

We have sum and scalar product. Is there a way to multiply two vectors?

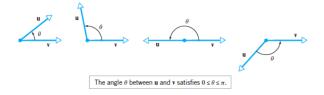
Definition: Inner Product (dot product)

Let $\mathbf{u} = (\mathbf{u_1}, \mathbf{u_2}, \cdots, \mathbf{u_n})$ and $\mathbf{v} = (\mathbf{v_1}, \mathbf{v_2}, \cdots, \mathbf{v_n})$ vectors in \mathbb{R}^n . The euclidean inner product of \mathbf{u} and \mathbf{v} , written as $\mathbf{u} \cdot \mathbf{v}$, is:

$$u\cdot v=u_1v_1+u_2v_2+\cdots+u_nv_n$$

Example: Let $\mathbf{u}=(\mathbf{4},-\mathbf{1},\mathbf{2})$ and $\mathbf{v}=(\mathbf{6},\mathbf{3},-\mathbf{4}).$ Then, $\mathbf{u}\cdot\mathbf{v}=\mathbf{13}$

A few facts about inner products:



DEFINITION 3 If \mathbf{u} and \mathbf{v} are nonzero vectors in R^2 or R^3 , and if θ is the angle between \mathbf{u} and \mathbf{v} , then the *dot product* (also called the *Euclidean inner product*) of \mathbf{u} and \mathbf{v} is denoted by $\mathbf{u} \cdot \mathbf{v}$ and is defined as

$$\mathbf{u} \cdot \mathbf{v} = \|\mathbf{u}\| \|\mathbf{v}\| \cos \theta \tag{12}$$

If $\mathbf{u} = \mathbf{0}$ or $\mathbf{v} = \mathbf{0}$, then we define $\mathbf{u} \cdot \mathbf{v}$ to be 0.

Question: what does it mean for a vector to be perpendicular (orthogonal)?

A few facts about inner products:

Theorem Let \mathbf{u} , \mathbf{v} , \mathbf{w} be arbitrary vectors in $\mathbf{R}^{\mathbf{n}}$ and let \mathbf{r} be an arbitrary scalar. Then,

- (a) $\mathbf{u} \mathbf{v} = \mathbf{v} \mathbf{u}$,
- (b) $\mathbf{u} \cdot (\mathbf{v} + \mathbf{w}) = \mathbf{u} \cdot \mathbf{v} + \mathbf{u} \cdot \mathbf{w}$,
- (c) $\mathbf{u}(r\mathbf{v}) = r(\mathbf{u} \ \mathbf{v}) = (r\mathbf{u}) \ \mathbf{v}$,
- (d) $\mathbf{u} \cdot \mathbf{u} \geq 0$,
- (e) $\mathbf{u} \cdot \mathbf{u} = 0$ implies $\mathbf{u} = 0$, and
- (f) (u + v) (u + v) = u u + 2(u v) + v v.

Your turn! Find $\mathbf{u} \cdot \mathbf{v}$, $\mathbf{u} \cdot \mathbf{u}$, $\mathbf{v} \cdot \mathbf{v}$, $||\mathbf{u}||$, and $||\mathbf{v}||$ for:

(a)
$$\mathbf{u} = (3, 1, 4), \ \mathbf{v} = (2, 2, -4)$$

(b)
$$\mathbf{u} = (1, 1, 4, 6), \ \mathbf{v} = (2, -2, 3, -2)$$

Eigenvectors and Eigenvalues are important to change the representation of the problem we are working on.

- r is eigenvalues of A if, and only if, A rI is singular (det(A rI) = 0).
- ightharpoonup det(A-rI) is call characteristic polynomial of A.
- ightharpoonup r is eigenvalues of A if and only if is the root of the characteristic polynomial of A.

Example:

$$A = \begin{pmatrix} a & b \\ c & d \end{pmatrix} \qquad A - rI = \begin{pmatrix} a - r & b \\ c & d - r \end{pmatrix}$$
$$det(A - rI) = r^2 - (a + d)r + (ad - bc)$$

When r is a eigenvalue of A, an nonzero vector \mathbf{v} such as

$$(A - rI)\mathbf{v} = \mathbf{0}$$

Then \mathbf{v} is called an of eigenvector of A corresponding to eigenvalue r.

Find the eigenvalues and eigenvector of:

$$A = \begin{pmatrix} 1 & 3 \\ 2 & 0 \end{pmatrix}$$

Solution:

$$det(A - rI) = det\begin{pmatrix} -1 - r & 3\\ 2 & 0 - r \end{pmatrix} = r^2 + r - 6 = (r + 3)(r - 2)$$

The eigenvalues of A are the roots of the characteristic polynomial: -3 and 2.

For r = -3

$$(A - (-3)I)\mathbf{v} = \begin{pmatrix} 2 & 3 \\ 2 & 3 \end{pmatrix} \begin{pmatrix} u \\ v \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

(3, -2) is eigenvector of A associated to -3 (eigenvalue).

Important facts:

Theorem Let A be a $k \times k$ matrix. Let r_1, r_2, \ldots, r_k be eigenvalues of A, and $\mathbf{v}_1, \mathbf{v}_2, \ldots, \mathbf{v}_k$ the corresponding eigenvectors. Form the matrix

$$P = [\mathbf{v}_1 \mathbf{v}_2 \cdots \mathbf{v}_k]$$

whose columns are these k eigenvectors. If P is invertible, then

$$P^{-1}AP = \begin{pmatrix} r_1 & 0 & \cdots & 0 \\ 0 & r_2 & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & r_k \end{pmatrix}.$$

Conversely, if $P^{-1}AP$ is a diagonal matrix D, the columns of P must be eigenvectors of A and the diagonal entries of D must be eigenvalues of A.

Theorem Let r_1, \ldots, r_h be h distinct eigenvalues of the $k \times k$ matrix A. Let $\mathbf{v}_1, \ldots, \mathbf{v}_h$ be corresponding eigenvectors. Then, $\mathbf{v}_1, \ldots, \mathbf{v}_h$ are linearly independent, that is, no one of them can be written as a linear combination of the others.

$$A = \begin{pmatrix} 1 & 3 \\ 2 & 0 \end{pmatrix}$$

$$A = \begin{pmatrix} 1 & 3 \\ 2 & 0 \end{pmatrix}$$

Solution:

The eigenvalues of A are the roots of the characteristic polynomial: -3 and 2.

(3, -2) is eigenvector of A associated to 3 (eigenvalue). (1, 1) is eigenvector of A associated to -2 (eigenvalue).

$$D = P^{-1}AP \quad (A = PDP^{-1})$$

$$\begin{pmatrix} -3 & 0 \\ 0 & 2 \end{pmatrix} = \begin{pmatrix} 3 & 1 \\ -2 & 1 \end{pmatrix}^{-1} \begin{pmatrix} 1 & 3 \\ 2 & 0 \end{pmatrix} \begin{pmatrix} 3 & 1 \\ -2 & 1 \end{pmatrix}$$

This decomposition helps us calculate matrix powers without making the products using the following property:

$$A^n = PD^nP^{-1}$$

We know that D is a diagonal matrix, and its power is the power of the diagonal elements.

Example:

$$\begin{pmatrix} -1 & 3 \\ 2 & 0 \end{pmatrix}^n = \begin{pmatrix} 3 & 1 \\ -2 & 1 \end{pmatrix} \begin{pmatrix} (-3)^n & 0 \\ 0 & 2^n \end{pmatrix} \begin{pmatrix} 3 & 1 \\ -2 & 1 \end{pmatrix}^{-1}$$

There are three possibilities for the roots of the characteristic polynomial p(r):

- \triangleright k distinct, real roots.
- ► Some repeated roots.
- ► Some complex roots.

Definition: Trace

Theorem

Trace(A)) is the sum of the diagonal elements of A.

And one important property of the eigenanalysis is:

Let A be a $k \times k$ matrix with eigenvalues r_1, \ldots, r_k . Then,

- (a) $r_1 + r_2 + \cdots + r_k = \text{trace of } A$, and (b) $r_1 \cdot r_2 \cdots r_k = \det A$.

1.

$$A = \begin{pmatrix} 4 & 0 & 2 \\ 0 & 3 & 0 \\ 3 & 0 & -1 \end{pmatrix}$$

$$Trace(A) = 4 + 3 - 1 = 6$$

Eigenvalues are $r_1 = 1, r_2 = 2$ and $r_3 = 3$,

$$r_1 + r_2 + r_3 = 6.$$

2.

$$\begin{pmatrix} 4 & 1 \\ -1 & 2 \end{pmatrix}$$

$$Trace(A) = 4 + 2 = 6$$

Eigenvalues are $r_1 = 3, r_2 = 3, r_1 + r_2 = 6$.

Exercise 30

Your turn! Compute A^{10} for the matrix:

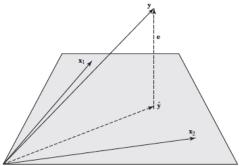
$$A = \begin{pmatrix} 0 & 0 & -2 \\ 1 & 2 & 1 \\ 1 & 0 & 3 \end{pmatrix}$$

► Interpolation:

Find a cubic polynomial whose graph passes through the points

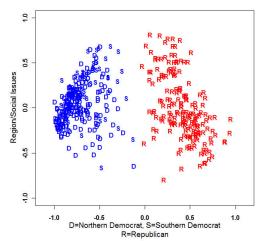
$$(1,3), (2,-2), (3,-5), (4,0)$$

► Regression Analysis



▶ Dimension Reduction

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Questions?

See you in the next class!