

Linear Methods (Math 211)

Lecture 23 - §3.2

(with slides adapted from K. Seyffarth)

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Recall

- 1 Determinants - Products, inverses and transpose
- 2 Adjugates

Today

- 1 Determinants and Transpose
- 2 Cramer's Rule
- 3 Polynomial interpolation
- 4 Vandermonde Determinants

Example (§3.2 Exercise 17)

Let A and B be $n \times n$ matrices. Show that $\det(A + B^T) = \det(A^T + B)$.

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Notice that

$$(A + B^T)^T = A^T + (B^T)^T = A^T + B.$$

Since a matrix and its transpose have the same determinant

$$\begin{aligned}\det(A + B^T) &= \det((A + B^T)^T) \\ &= \det(A^T + B).\end{aligned}$$

Example

For each of the following statements, determine if it is true or false, and supply a proof or a counterexample.

- If $\text{adj } A$ exists, then A is invertible.
- If A and B are $n \times n$ matrices, then $\det(AB) = \det(B^T A)$.
- Prove or give a counterexample to the following statement: if $\det A = 1$, then $\text{adj } A = A$.

Example

For each of the following statements, determine if it is true or false, and supply a proof or a counterexample.

- If $\text{adj } A$ exists, then A is invertible.

False. $\text{adj } 0$ exists, but 0 is not invertible.

- If A and B are $n \times n$ matrices, then $\det(AB) = \det(B^T A)$.

True. $\det AB = \det A \cdot \det B = \det B^T \cdot \det A = \det B^T A$.

- Prove or give a counterexample to the following statement: if $\det A = 1$, then $\text{adj } A = A$.

False. Note that if $\det A = 1$ then $\text{adj } A = A^{-1}$. There are plenty of matrices of determinant 1 for which $A \neq A^{-1}$:

$\begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix}$ for example.

Cramer's Rule

If A is an $n \times n$ invertible matrix, then the solution to $A\mathbf{x} = \mathbf{b}$ can be given in terms of determinants of matrices.

Theorem (§3.2 Theorem 5)

Let A be an $n \times n$ invertible matrix, and consider the system $A\mathbf{x} = \mathbf{b}$, where $\mathbf{x} = [x_1 \ x_2 \ \cdots \ x_n]^T$. We define A_i to be the matrix obtained from A by replacing column i with \mathbf{b} . Then setting

$$x_i = \frac{\det A_i}{\det A}$$

gives a solution to $A\mathbf{x} = \mathbf{b}$.

Note: This is a very inefficient method for large systems.

Example

Solve for x_3 :

$$3x_1 + x_2 - x_3 = -1$$

$$5x_1 + 2x_2 = 2$$

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$$x_1 + x_2 - x_3 = 1$$

By Cramer's rule, $x_3 = \frac{\det A_3}{\det A}$, where

$$A = \begin{bmatrix} 3 & 1 & -1 \\ 5 & 2 & 0 \\ 1 & 1 & -1 \end{bmatrix} \quad \text{and} \quad A_3 = \begin{bmatrix} 3 & 1 & -1 \\ 5 & 2 & 2 \\ 1 & 1 & 1 \end{bmatrix}.$$

Computing the determinants of these two matrices,

$$\det A = -4 \quad \text{and} \quad \det A_3 = -6.$$

Therefore, $x_3 = \frac{-6}{-4} = \frac{3}{2}$.

Example (continued)

We can also compute $\det A_1$ and $\det A_2$, where

$$A_1 = \begin{bmatrix} -1 & 1 & -1 \\ 2 & 2 & 0 \\ 1 & 1 & -1 \end{bmatrix} \quad \text{and} \quad A_2 = \begin{bmatrix} 3 & -1 & -1 \\ 5 & 2 & 0 \\ 1 & 1 & -1 \end{bmatrix},$$

and then solve for x_1 and x_2 .

We get $x_1 = -1$, $x_2 = \frac{7}{2}$.

Polynomial Interpolation

Example

Given data points $(0, 1)$, $(1, 2)$, $(2, 5)$ and $(3, 10)$, find an interpolating polynomial $p(x)$ of degree at most three, and then estimate the value of y corresponding to $x = \frac{3}{2}$.

Polynomial Interpolation

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Given data points $(0, 1)$, $(1, 2)$, $(2, 5)$ and $(3, 10)$, find an interpolating polynomial $p(x)$ of degree at most three, and then estimate the value of y corresponding to $x = \frac{3}{2}$.

We want to find the coefficients r_0 , r_1 , r_2 and r_3 of

$$p(x) = r_0 + r_1x + r_2x^2 + r_3x^3$$

so that $p(0) = 1$, $p(1) = 2$, $p(2) = 5$, and $p(3) = 10$.

$$p(0) = r_0 = 1$$

$$p(1) = r_0 + r_1 + r_2 + r_3 = 2$$

$$p(2) = r_0 + 2r_1 + 4r_2 + 8r_3 = 5$$

$$p(3) = r_0 + 3r_1 + 9r_2 + 27r_3 = 10$$

Example (continued)

Solve this system of four equations in the four variables r_0 , r_1 , r_2 and r_3 .

$$\left[\begin{array}{cccc|c} 1 & 0 & 0 & 0 & 1 \\ 1 & 1 & 1 & 1 & 2 \\ 1 & 2 & 4 & 8 & 5 \\ 1 & 3 & 9 & 27 & 10 \end{array} \right] \rightarrow \cdots \rightarrow \left[\begin{array}{cccc|c} 1 & 0 & 0 & 0 & 1 \\ 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 1 \\ 0 & 0 & 0 & 1 & 0 \end{array} \right]$$

Therefore $r_0 = 1$, $r_1 = 0$, $r_2 = 1$, $r_3 = 0$, and so

$$p(x) = 1 + x^2.$$

The estimate is

$$y = p\left(\frac{3}{2}\right) = 1 + \left(\frac{3}{2}\right)^2 = \frac{13}{4}.$$

Theorem (§3.2 Theorem 6)

Given n data points $(x_1, y_1), (x_2, y_2), \dots, (x_n, y_n)$ with the x_i *distinct*, there is a unique polynomial

$$p(x) = r_0 + r_1x + r_2x^2 + \cdots + r_{n-1}x^{n-1}$$

such that $p(x_i) = y_i$ for $i = 1, 2, \dots, n$.

The polynomial $p(x)$ is called the **interpolating polynomial** for the data. We will prove that interpolating polynomials exist and are unique in the next few slides.

To find $p(x)$, set up a system of n linear equations in the n variables $r_0, r_1, r_2, \dots, r_{n-1}$.

$$p(x) = r_0 + r_1x + r_2x^2 + \cdots + r_{n-1}x^{n-1}:$$

$$r_0 + r_1x_1 + r_2x_1^2 + \cdots + r_{n-1}x_1^{n-1} = y_1$$

$$r_0 + r_1x_2 + r_2x_2^2 + \cdots + r_{n-1}x_2^{n-1} = y_2$$

$$r_0 + r_1x_3 + r_2x_3^2 + \cdots + r_{n-1}x_3^{n-1} = y_3$$

$$\vdots$$

$$r_0 + r_1x_n + r_2x_n^2 + \cdots + r_{n-1}x_n^{n-1} = y_n$$

The coefficient matrix for this system is

$$\begin{bmatrix} 1 & x_1 & x_1^2 & \cdots & x_1^{n-1} \\ 1 & x_2 & x_2^2 & \cdots & x_2^{n-1} \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ 1 & x_n & x_n^2 & \cdots & x_n^{n-1} \end{bmatrix}$$

The determinant of a matrix of this form is called a **Vandermonde** determinant.

The Vandermonde Determinant

Theorem (§3.2 Theorem 7)

Let x_1, x_2, \dots, x_n be real numbers, $n \geq 2$. The the corresponding Vandermonde determinant is

$$\det \begin{bmatrix} 1 & x_1 & x_1^2 & \cdots & x_1^{n-1} \\ 1 & x_2 & x_2^2 & \cdots & x_2^{n-1} \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ 1 & x_n & x_n^2 & \cdots & x_n^{n-1} \end{bmatrix} = \prod_{1 \leq j < i \leq n} (x_i - x_j).$$

Example

In our earlier example with the data points $(0, 1)$, $(1, 2)$, $(2, 5)$ and $(3, 10)$, we have

$$x_1 = 0, x_2 = 1, x_3 = 2, x_4 = 3$$

giving us the Vandermonde determinant

$$\begin{vmatrix} 1 & 0 & 0 & 0 \\ 1 & 1 & 1 & 1 \\ 1 & 2 & 4 & 8 \\ 1 & 3 & 9 & 27 \end{vmatrix}$$

Example

In our earlier example with the data points $(0, 1)$, $(1, 2)$, $(2, 5)$ and $(3, 10)$, we have

$$x_1 = 0, x_2 = 1, x_3 = 2, x_4 = 3$$

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$$\begin{vmatrix} 1 & 0 & 0 & 0 \\ 1 & 1 & 1 & 1 \\ 1 & 2 & 4 & 8 \\ 1 & 3 & 9 & 27 \end{vmatrix}$$

According to Theorem 7, this determinant is equal to

$$\begin{aligned} & (a_2 - a_1)(a_3 - a_1)(a_3 - a_2)(a_4 - a_1)(a_4 - a_2)(a_4 - a_3) \\ &= (1 - 0)(2 - 0)(2 - 1)(3 - 0)(3 - 1)(3 - 2) = 2 \times 3 \times 2 \\ &= 12. \end{aligned}$$

As a consequence of Theorem 7, the Vandermonde determinant is nonzero if a_1, a_2, \dots, a_n are distinct.

This means that given n data points $(x_1, y_1), (x_2, y_2), \dots, (x_n, y_n)$ with **distinct** x_i , then there is a unique interpolating polynomial

$$p(x) = r_0 + r_1x + r_2x^2 + \cdots + r_{n-1}x^{n-1}.$$

Summary

- 1 Determinants and Transpose
- 2 Cramer's Rule
- 3 Polynomial interpolation
- 4 Vandermonde Determinants