

# Math 221: LINEAR ALGEBRA

## Chapter 3. Determinants and Diagonalization

### §3-2. Determinants and Matrix Inverses

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<sup>1</sup>Slides are adapted from those by Karen Seyffarth from University of Calgary.

Determinants and Matrix Inverses

Adjugates

Cramer's Rule

Polynomial Interpolation and Vandermonde Determinant

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# Determinants and Matrix Inverses

## Theorem (Product Theorem)

If  $A$  and  $B$  are  $n \times n$  matrices, then

$$\det(AB) = \det A \det B.$$

### Proof.

If either  $A$  or  $B$  is singular, then both sides are equal to zero.

Now assume that both  $A$  and  $B$  are nonsingular, i.e.,  $\text{rank}(A) = \text{rank}(B) = n$ . Then

$$\text{rref}(A) = \text{rref}(B) = I$$

and

$$A = E_1 E_2 \cdots E_p \quad \text{and} \quad B = F_1 F_2 \cdots F_q.$$

where  $E_i$  and  $F_j$  are elementary matrices. Then by the relation of elementary row operations with determinants (Theorem 3.1.2), we see that

$$\begin{aligned} |AB| &= |E_1 \cdots E_p F_1 \cdots F_q| \\ &= |E_1| \cdots |E_p| |F_1| \cdots |F_q| \\ &= |E_1 \cdots E_p| |F_1 \cdots F_q| \\ &= |A| |B|. \end{aligned}$$



## Theorem (Determinant of Matrix Inverse)

An  $n \times n$  matrix  $A$  is invertible if and only if  $\det A \neq 0$ . In this case,

$$\det(A^{-1}) = (\det A)^{-1} = \frac{1}{\det A}.$$

Proof.

" $\Rightarrow$ ":

$$1 = |\mathbf{I}| = |AA^{-1}| = |A||A^{-1}| \Rightarrow \begin{cases} |A| \neq 0 \\ |A^{-1}| = \frac{1}{|A|}. \end{cases}$$

" $\Leftarrow$ ": If  $|A| \neq 0$ , then  $\text{rref}(A) = \mathbf{I}$  because otherwise one obtains contradiction by Theorem 3.1.2. This is another way to say that  $A$  is invertible: (recall the matrix inverse algorithm)

$$[A|\mathbf{I}] \rightarrow \left[ \underbrace{\text{rref}(A)}_{=\mathbf{I}} \mid A^{-1} \right].$$



### Example

Find all values of  $c$  for which  $A = \begin{bmatrix} c & 1 & 0 \\ 0 & 2 & c \\ -1 & c & 5 \end{bmatrix}$  is invertible.

$$\begin{aligned} \det A &= \begin{vmatrix} c & 1 & 0 \\ 0 & 2 & c \\ -1 & c & 5 \end{vmatrix} = c \begin{vmatrix} 2 & c \\ c & 5 \end{vmatrix} + (-1) \begin{vmatrix} 1 & 0 \\ 2 & c \end{vmatrix} \\ &= c(10 - c^2) - c = c(9 - c^2) = c(3 - c)(3 + c). \end{aligned}$$

Therefore,  $A$  is invertible for all  $c \neq 0, 3, -3$ .

## Theorem (Determinant of Matrix Transpose)

If  $A$  is an  $n \times n$  matrix, then  $\det(A^T) = \det A$ .

Proof.

1. This is trivially true for all elementary matrices.
2. If  $A$  is not invertible, then neither is  $A^T$ . Hence,  $\det A = 0 = \det A^T$ .
3. If  $A$  is invertible, then  $A = E_k E_{k-1} \cdots E_2 E_1$ . Hence, by Case 1,

$$\begin{aligned} |A^T| &= |(E_k E_{k-1} \cdots E_2 E_1)^T| \\ &= |E_1^T E_2^T \cdots E_{k-1}^T E_k^T| \\ &= |E_1^T| |E_2^T| \cdots |E_{k-1}^T| |E_k^T| \\ &= |E_1| |E_2| \cdots |E_{k-1}| |E_k| \\ &= |E_k| |E_{k-1}| \cdots |E_2| |E_1| \\ &= |E_k E_{k-1} \cdots E_2 E_1| \\ &= |A|. \end{aligned}$$



## Problem

Suppose  $A$  is a  $3 \times 3$  matrix. Find  $\det A$  and  $\det B$  if

$$\det(2A^{-1}) = -4 = \det(A^3(B^{-1})^T).$$

## Solution

First,

$$\begin{aligned}\det(2A^{-1}) &= -4 \\ 2^3 \det(A^{-1}) &= -4 \\ \frac{1}{\det A} &= \frac{-4}{8} = -\frac{1}{2}\end{aligned}$$

Therefore,  $\det A = -2$ .

## Solution (continued)

Now,

$$\begin{aligned}\det(A^3(B^{-1})^T) &= -4 \\ (\det A)^3 \det(B^{-1}) &= -4 \\ (-2)^3 \det(B^{-1}) &= -4 \\ (-8) \det(B^{-1}) &= -4 \\ \frac{1}{\det B} &= \frac{-4}{-8} = \frac{1}{2}\end{aligned}$$

Therefore,  $\det B = 2$ .



## Problem

Suppose  $A$ ,  $B$  and  $C$  are  $4 \times 4$  matrices with

$$\det A = -1, \det B = 2, \quad \text{and} \quad \det C = 1.$$

Find  $\det(2A^2(B^{-1})(C^T)^3B(A^{-1}))$ .

## Solution

$$\begin{aligned}\det(2A^2(B^{-1})(C^T)^3B(A^{-1})) &= 2^4(\det A)^2 \frac{1}{\det B} (\det C)^3 (\det B) \frac{1}{\det A} \\ &= 16(\det A)(\det C)^3 \\ &= 16 \times (-1) \times 1^3 \\ &= -16.\end{aligned}$$



## Problem

A square matrix  $A$  is **orthogonal** if and only if  $A^T = A^{-1}$ . What are the possible values of  $\det A$  if  $A$  is orthogonal?

## Solution

Since  $A^T = A^{-1}$ ,

$$\begin{aligned}\det A^T &= \det(A^{-1}) \\ \det A &= \frac{1}{\det A} \\ (\det A)^2 &= 1\end{aligned}$$

Assuming  $A$  is a **real** matrix, this implies that  $\det A = \pm 1$ , i.e.,  $\det A = 1$  or  $\det A = -1$ . ■

Determinants and Matrix Inverses

**Adjugates**

Cramer's Rule

Polynomial Interpolation and Vandermonde Determinant

# Adjugates

For a  $2 \times 2$  matrix  $A = \begin{bmatrix} a & b \\ c & d \end{bmatrix}$ , we have already seen the **adjugate** of  $A$  defined as

$$\text{adj}(A) = \begin{bmatrix} d & -b \\ -c & a \end{bmatrix},$$

and observed that

$$\begin{aligned} A \text{adj}(A) &= \begin{bmatrix} a & b \\ c & d \end{bmatrix} \begin{bmatrix} d & -b \\ -c & a \end{bmatrix} \\ &= \begin{bmatrix} ad - bc & 0 \\ 0 & ad - bc \end{bmatrix} \\ &= (\det A)I_2 \end{aligned}$$

Furthermore, if  $\det A \neq 0$ , then  $A$  is invertible and

$$A^{-1} = \frac{1}{\det A} \text{adj}(A).$$

## Definition (Adjugate Matrix)

If  $A$  is an  $n \times n$  matrix, then the **adjugate matrix of  $A$**  is defined to be

$$\text{adj}(A) \stackrel{\text{def}}{=} [ c_{ij}(A) ]^T = [ (-1)^{i+j} \det(A_{ij}) ]^T,$$

where  $c_{ij}(A)$  is the  $(i, j)$ -cofactor of  $A$ , i.e.,  $\text{adj}(A)$  is the transpose of the **cofactor matrix** (matrix of cofactors).

## Problem

Find  $\text{adj}(A)$  when  $A = \begin{bmatrix} 2 & 1 & 3 \\ 5 & -7 & 1 \\ 3 & 0 & -6 \end{bmatrix}$  and compute  $A \text{adj}(A)$ .

## Solution

$$\text{adj}(A) = \begin{bmatrix} 42 & 6 & 22 \\ 33 & -21 & 13 \\ 21 & 3 & -19 \end{bmatrix}.$$

Notice that

$$A \text{adj}(A) = \begin{bmatrix} 2 & 1 & 3 \\ 5 & -7 & 1 \\ 3 & 0 & -6 \end{bmatrix} \begin{bmatrix} 42 & 6 & 22 \\ 33 & -21 & 13 \\ 21 & 3 & -19 \end{bmatrix} = \begin{bmatrix} 180 & 0 & 0 \\ 0 & 180 & 0 \\ 0 & 0 & 180 \end{bmatrix}$$

$$\det A = \begin{vmatrix} 2 & 1 & 3 \\ 5 & -7 & 1 \\ 3 & 0 & -6 \end{vmatrix} = \begin{vmatrix} 2 & 1 & 3 \\ 19 & 0 & 22 \\ 3 & 0 & -6 \end{vmatrix} = (-1) \begin{vmatrix} 19 & 22 \\ 3 & -6 \end{vmatrix} = 180,$$

Therefore,

$$A \text{adj}(A) = (\det A)I.$$



## Theorem (The Adjugate Formula)

If  $A$  is an  $n \times n$  matrix, then

$$A \operatorname{adj}(A) = (\det A)I = \operatorname{adj}(A)A.$$

Furthermore, if  $\det A \neq 0$ , then

$$A^{-1} = \frac{1}{\det A} \operatorname{adj}(A).$$

### Proof.

We only prove the case when  $n = 3$ .

$$A \operatorname{adj}(A) = \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{bmatrix} \begin{bmatrix} c_{11} & c_{21} & c_{31} \\ c_{12} & c_{22} & c_{32} \\ c_{13} & c_{23} & c_{33} \end{bmatrix} = \begin{bmatrix} |A| & 0 & 0 \\ 0 & |A| & 0 \\ 0 & 0 & |A| \end{bmatrix}$$

where, for example,

$$\begin{aligned} (3,2)\text{-th entry} &= a_{31}c_{21} + a_{32}c_{22} + a_{33}c_{23} \\ &= \det \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{31} & a_{32} & a_{33} \\ a_{31} & a_{32} & a_{33} \end{bmatrix} = 0. \end{aligned}$$



### Example

For an  $n \times n$  matrix  $A$ , show that  $\det \operatorname{adj}(A) = (\det A)^{n-1}$ .

Using the adjugate formula,

$$\begin{aligned}A \operatorname{adj}(A) &= (\det A)I \\ \det(A \operatorname{adj}(A)) &= \det((\det A)I) \\ (\det A) \times \det \operatorname{adj}(A) &= (\det A)^n (\det I) \\ (\det A) \times \det \operatorname{adj}(A) &= (\det A)^n\end{aligned}$$

If  $\det A \neq 0$ , then divide both sides of the last equation by  $\det A$ :

$$\det \operatorname{adj}(A) = (\det A)^{n-1}.$$

## Example (continued)

For the case  $\det A = 0$ , we claim that

$$\det A = 0 \quad \Rightarrow \quad \det \operatorname{adj}(A) = 0, \quad (\star)$$

which implies that

$$\det \operatorname{adj}(A) = 0 = 0^{n-1} = (\det A)^{n-1}.$$

## Proof. (of $(\star)$ )

We will prove  $(\star)$  by contradiction. Indeed, if  $\det A = 0$ , then

$$A \operatorname{adj}(A) = (\det A)I = (0)I = O,$$

i.e.,  $A \operatorname{adj}(A)$  is the zero matrix. If  $\det \operatorname{adj}(A) \neq 0$ , then  $\operatorname{adj}(A)$  would be invertible, and  $A \operatorname{adj}(A) = O$  would imply  $A = O$ . However, if  $A = O$ , then  $\operatorname{adj}(A) = O$  and is not invertible, and thus has determinant equal to zero, i.e.,  $\det \operatorname{adj}(A) = 0$ , (a contradiction!) Therefore,  $\det \operatorname{adj}(A) = 0$ , i.e.,  $(\star)$  is true. ■

## Problem

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

## Solution

Notice that

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

Since a matrix and its transpose have the same determinant

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



## Problem

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

1. If  $\text{adj}(A)$  exists, then  $A$  is invertible.
2. If  $A$  and  $B$  are  $n \times n$  matrices, then  $\det(AB) = \det(B^T A)$ .

## Problem

Prove or give a counterexample to the following statement:

$$\text{If } \det A = 1, \text{ then } \text{adj}(A) = A.$$

Determinants and Matrix Inverses

Adjugates

**Cramer's Rule**

Polynomial Interpolation and Vandermonde Determinant

## Cramer's Rule

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

### Theorem (Cramer's Rule)

Let  $A$  be an  $n \times n$  invertible matrix, the solution to the system  $A\vec{x} = \vec{b}$  of  $n$  equations in the variables  $x_1, x_2, \dots, x_n$  is given by

$$x_1 = \frac{\det(A_1(\vec{b}))}{\det A}, \quad x_2 = \frac{\det(A_2(\vec{b}))}{\det A}, \quad \dots, \quad x_n = \frac{\det(A_n(\vec{b}))}{\det A}$$

where, for each  $j$ , the matrix  $A_j(\vec{b})$  is obtained from  $A$  by replacing column  $j$  with  $\vec{b}$ :

$$A_j(\vec{b}) = \begin{bmatrix} \vec{a}_1 & \cdots & \vec{a}_{j-1} & \vec{b} & \vec{a}_{j+1} & \cdots & \vec{a}_n \end{bmatrix}$$

Proof.

► Notice that

$$\begin{aligned} A_j(\vec{b}) &= \begin{bmatrix} \vec{a}_1 & \cdots & \vec{a}_{j-1} & \vec{b} & \vec{a}_{j+1} & \cdots & \vec{a}_n \end{bmatrix} \\ &= \begin{bmatrix} A\vec{e}_1 & \cdots & A\vec{e}_{j-1} & A\vec{x} & A\vec{e}_{j+1} & \cdots & A\vec{e}_n \end{bmatrix} \\ &= A \begin{bmatrix} \vec{e}_1 & \cdots & \vec{e}_{j-1} & \vec{x} & \vec{e}_{j+1} & \cdots & \vec{e}_n \end{bmatrix} \\ &= A I_j(\vec{x}) \end{aligned}$$

where

$$\begin{aligned} I_j(\vec{x}) &= \begin{bmatrix} \vec{e}_1 & \cdots & \vec{e}_{j-1} & \vec{x} & \vec{e}_{j+1} & \cdots & \vec{e}_n \end{bmatrix} \\ &= \begin{bmatrix} 1 & & & x_1 & & & \\ & \ddots & & \vdots & & & \\ & & 1 & x_{j-1} & & & \\ & & & x_j & & & \\ & & & x_{j+1} & 1 & & \\ & & & \vdots & & \ddots & \\ & & & x_n & & & 1 \end{bmatrix} \end{aligned}$$

Proof. (continued)

- ▶ Hence, by taking the determinants on both sides, we have

$$\begin{aligned}\det(A_j(\vec{b})) &= \det(A I_j(\vec{x})) \\ &= \det(A) \det(I_j(\vec{x}))\end{aligned}$$

- ▶ And because  $\det(A) \neq 0$ , we can then write:

$$\det(I_j(\vec{x})) = \frac{\det(A_j(\vec{b}))}{\det(A)}$$

- ▶ Finally, notice that  $\det(I_j(\vec{x})) = \cdots = x_j$ .



## Problem

Find  $x_3$  such that

$$\begin{array}{rccccrcr} 3x_1 & + & x_2 & - & x_3 & = & -1 \\ 5x_1 & + & 2x_2 & & & = & 2 \\ x_1 & + & x_2 & - & x_3 & = & 1 \end{array}$$

## Solution

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



### Remark

For practice, you should 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  $x_2$ .

Solution.  $x_1 = -1$ ,  $x_2 = 7/2$ .



Determinants and Matrix Inverses

Adjugates

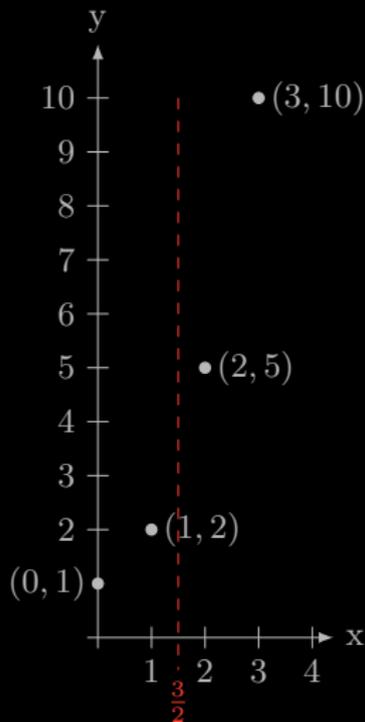
Cramer's Rule

Polynomial Interpolation and Vandermonde Determinant

# Polynomial Interpolation and Vandermonde Determinant

## Problem

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  $x = 3/2$ .



## Solution

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  $p(3) = 10$ .

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

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

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

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

### Solution (continued)

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

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

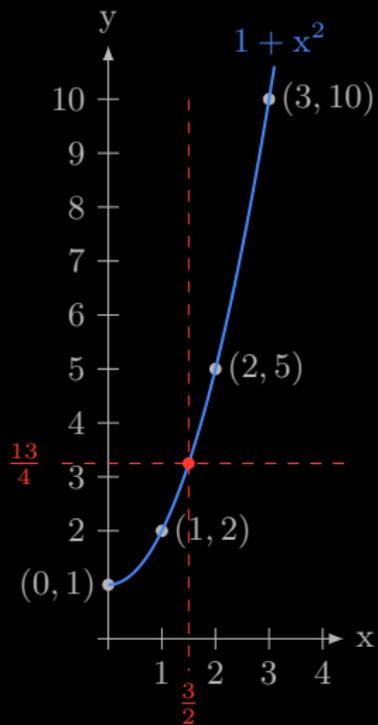
Therefore,  $r_0 = 1$ ,  $r_1 = 0$ ,  $r_2 = 1$ ,  $r_3 = 0$ , and so

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

Finally, the estimate is

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





## Theorem (Polynomial Interpolation)

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

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

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.

To find  $p(x) = r_0 + r_1x + r_2x^2 + \cdots + r_{n-1}x^{n-1}$ , set up a system of  $n$  linear equations in the  $n$  variables  $r_0, r_1, r_2, \dots, r_{n-1}$ .

$$\begin{aligned} r_0 + r_1x_1 + r_2x_1^2 + \cdots + r_{n-1}x_1^{n-1} &= y_1 \\ r_0 + r_1x_2 + r_2x_2^2 + \cdots + r_{n-1}x_2^{n-1} &= y_2 \\ r_0 + r_1x_3 + r_2x_3^2 + \cdots + r_{n-1}x_3^{n-1} &= y_3 \\ &\vdots \\ &\vdots \\ r_0 + r_1x_n + r_2x_n^2 + \cdots + r_{n-1}x_n^{n-1} &= y_n \end{aligned}$$

The coefficient matrix for this system is

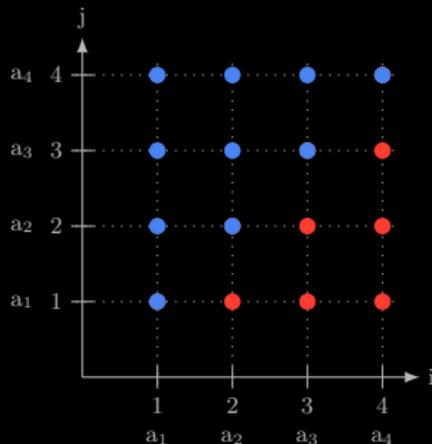
$$\begin{bmatrix} 1 & x_1 & x_1^2 & \cdots & x_1^{n-1} \\ 1 & x_2 & x_2^2 & \cdots & x_2^{n-1} \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ 1 & x_n & x_n^2 & \cdots & x_n^{n-1} \end{bmatrix}$$

- ▶ Such matrix is called **Vandermonde matrix**.
- ▶ Its determinant is called **Vandermonde determinant**.

## Theorem (Vandermonde Determinant)

Let  $a_1, a_2, \dots, a_n$  be real numbers,  $n \geq 2$ . The corresponding Vandermonde determinant is

$$\det \begin{bmatrix} 1 & a_1 & a_1^2 & \cdots & a_1^{n-1} \\ 1 & a_2 & a_2^2 & \cdots & a_2^{n-1} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & a_n & a_n^2 & \cdots & a_n^{n-1} \end{bmatrix} = \prod_{1 \leq j < i \leq n} (a_i - a_j).$$



Proof.

We will prove this by induction. It is clear that when  $n = 2$ ,

$$\det \begin{pmatrix} 1 & a_1 \\ 1 & a_2 \end{pmatrix} = a_2 - a_1 = \prod_{1 \leq j < i \leq 2} (a_i - a_j).$$

Assume that it is true for  $n - 1$ . Now let's consider the case  $n$ . Denote

$$p(\mathbf{x}) := \det \begin{bmatrix} 1 & a_1 & a_1^2 & \cdots & a_1^{n-1} \\ 1 & a_2 & a_2^2 & \cdots & a_2^{n-1} \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ 1 & a_{n-1} & a_{n-1}^2 & \cdots & a_{n-1}^{n-1} \\ 1 & \mathbf{x} & \mathbf{x}^2 & \cdots & \mathbf{x}^{n-1} \end{bmatrix}.$$

Proof. (continued)

Because  $p(a_1) = \cdots = p(a_{n-1}) = 0$  (why?),  $p(x)$  has to take the following form:

$$p(x) = c(x - a_1)(x - a_2) \cdots (x - a_{n-1}).$$

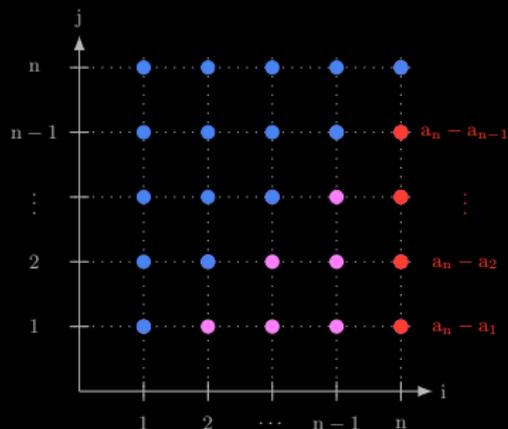
To identify the constant  $c$ , notice that  $c$  is the coefficient for  $x^{n-1}$ . By cofactor expansion of the determinant along the last row,

$$\begin{aligned} c &= (-1)^{n+n} \det \begin{bmatrix} 1 & a_1 & a_1^2 & \cdots & a_1^{n-1} \\ 1 & a_2 & a_2^2 & \cdots & a_2^{n-1} \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ 1 & a_{n-1} & a_{n-1}^2 & \cdots & a_{n-1}^{n-1} \end{bmatrix} \\ &= \prod_{1 \leq j < i \leq n-1} (a_i - a_j). \end{aligned}$$

Proof. (continued)

Hence,

$$p(a_n) = \left( \prod_{1 \leq j < i \leq n-1} (a_i - a_j) \right) \times (a_n - a_1)(a_n - a_2) \cdots (a_n - a_{n-1})$$



$$p(a_n) = \prod_{1 \leq j < i \leq n} (a_i - a_j).$$



## Example

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

$$a_1 = 0, \quad a_2 = 1, \quad a_3 = 2, \quad a_4 = 3$$

giving us the Vandermonde determinant

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

According to the previous theorem, this determinant is equal to

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

## Corollary

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

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