

# Introduction to Square Linear Systems

# Contents

① Opening Example: Polynomial Interpolation

② Square Linear Systems

# Opening Example: Polynomial Interpolation

# Polynomial Interpolation

$\{1, 2, 3, \dots, n\}$

## Formal Statement

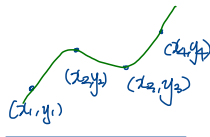
Given a set of  $n$  data points  $\{(x_j, y_j) \mid j \in \mathbb{N}[1, n]\}$  with distinct  $x_j$ 's, not necessarily sorted, find a polynomial of degree  $n - 1$ ,

$$p(x) = c_1 + c_2x + c_3x^2 + \dots + c_nx^{n-1}, \quad (\star)$$

(ascending)

which interpolates the given points, i.e.,

$$p(x_j) = y_j, \quad \text{for } j = 1, 2, \dots, n.$$



- The goal is to determine the coefficients  $c_1, c_2, \dots, c_n$ .
- Note that the total number of data point is 1 larger than the degree of the interpolating polynomial.

# Why Do We Care?

- to find the values between the discrete data points;
- to approximate a (complicated) function by a polynomial, which makes such computations as differentiation or integration easier.

# Interpolation to Linear System

Convert to a linear algebra problem.

Writing out the  $n$  interpolating conditions  $p(x_j) = y_j$ :

## Equations

$$\left. \begin{aligned} p(x_1) &= c_1 + c_2 x_1 + \cdots + c_n x_1^{n-1} = y_1 \\ p(x_2) &= c_1 + c_2 x_2 + \cdots + c_n x_2^{n-1} = y_2 \\ &\vdots \\ p(x_n) &= c_1 + c_2 x_n + \cdots + c_n x_n^{n-1} = y_n \end{aligned} \right\}$$

$x_{.1,0} \quad x_{.1,1} \quad \dots \quad x_{.1,(n-1)}$   
 $\downarrow \quad \downarrow \quad \quad \quad \downarrow$

Matrix equation

$$\underbrace{\begin{bmatrix} 1 & x_1 & \cdots & x_1^{n-1} \\ 1 & x_2 & \cdots & x_2^{n-1} \\ \vdots & \vdots & & \vdots \\ 1 & x_n & \cdots & x_n^{n-1} \end{bmatrix}}_V \underbrace{\begin{bmatrix} c_1 \\ c_2 \\ \vdots \\ c_n \end{bmatrix}}_c = \underbrace{\begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{bmatrix}}_y$$

- This is a linear system of  $n$  equations with  $n$  unknowns.
- The matrix  $V$  is called a **Vandermonde matrix**.

Let  $x = \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix}$  be a column vector.

Then

$$V = \begin{bmatrix} x_{.1,0} & x_{.1,1} & \cdots & x_{.1,(n-1)} \end{bmatrix}$$

$x$ -data

unknown

$y$ -data

$$\gg V = X.^{(0:n-1)}$$

## Example: Fitting Population Data

U.S. Census data are collected every 10 years.

Year	Population (millions)
1980	226.546
1990	248.710
2000	281.422
2010	308.746
2020	332.639

**Question.** How do we estimate population in other years?

- Interpolate available data to compute population in intervening years.

## Example: Fitting Population Data (cont')

- Input data.
- Match up notation (optional).
- Note the shift in Line 7.
- Construct the Vandermonde matrix  $V$  by *broadcasting*.
- Solve the system using the backslash ( $\backslash$ ) operator.

$$V\vec{c} = \vec{y}$$

$$V\vec{c} = \mathbf{I}$$

$$\vec{c} = V^{-1}\mathbf{I} = V^{-1}$$

Analytically:

$$\vec{c} = V^{-1}\vec{y}$$

Numerically:

$$c = V \backslash y;$$

```
1 year = (1980:10:2020)';  
2 pop = [226.546;  
3        248.710;  
4        281.422;  
5        308.746;  
6        332.639];  
7 x = year - 1980;  
8 y = pop;  
9 n = length(x);  
10 V = x.^(0:n-1);  
11 c = V \ y;
```

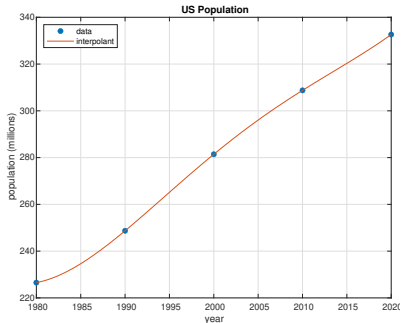
→ numerical stability



# Post-Processing

```
1 xx = linspace(0, 40, 100)';  
2 yy = polyval(flip(c), xx);  
3 clf  
4 plot(1980+xx, y, '.', 1980+xx, yy)  
5 title('US Population'),  
6 xlabel('year'), ylabel('population (millions)')  
7 legend('data', 'interpolant', 'location', 'northwest')
```

- Use the `polyval` function to evaluate the polynomial.
- MATLAB expects coefficients to be in descending order. (`flip`)



## Square Linear Systems

Given  $A \in \mathbb{R}^{n \times n}$ ,  $\vec{y} \in \mathbb{R}^n$  ( $\in \mathbb{R}^{n \times 1}$  column vector),

find  $\vec{x} \in \mathbb{R}^n$  such that

$$A \vec{x} = \vec{y}.$$

# Overview

← square

Let  $A \in \mathbb{R}^{n \times n}$  and  $\mathbf{b} \in \mathbb{R}^n$ . Then the equation  $A\mathbf{x} = \mathbf{b}$  has the following possibilities:

- If  $A$  is invertible (or nonsingular), then  $A\mathbf{x} = \mathbf{b}$  has a unique solution  $\mathbf{x} = A^{-1}\mathbf{b}$ , or  $\det(A) \neq 0$
- If  $A$  is not invertible (or singular), then  $A\mathbf{x} = \mathbf{b}$  has either no solution or infinitely many solutions.  $\det(A) = 0$

## The Backslash Operator “ \ ”

To solve for  $\mathbf{x}$  in MATLAB, we use the backslash symbol “ \ ”:

```
>> x = A \ b
```

This produces the solution without explicitly forming the inverse of  $A$ .

**Warning:** Even though  $\mathbf{x} = A^{-1}\mathbf{b}$  analytically, don't use  $\mathbf{x} = \text{inv}(A) * \mathbf{b}$ !

# Triangular Systems

Systems involving triangular matrices are easy to solve.

- A matrix  $U \in \mathbb{R}^{n \times n}$  is **upper triangular** if all entries below main diagonal are zero:

$$U = \begin{bmatrix} u_{11} & u_{12} & u_{13} & \cdots & u_{1n} \\ 0 & u_{22} & u_{23} & \cdots & u_{2n} \\ 0 & 0 & u_{33} & \cdots & u_{3n} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \cdots & u_{nn} \end{bmatrix}.$$

- A matrix  $L \in \mathbb{R}^{n \times n}$  is **lower triangular** if all entries above main diagonal are zero:

$$L = \begin{bmatrix} \ell_{11} & 0 & 0 & \cdots & 0 \\ \ell_{21} & \ell_{22} & 0 & \cdots & 0 \\ \ell_{31} & \ell_{32} & \ell_{33} & \cdots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ \ell_{n1} & \ell_{n2} & \ell_{n3} & \cdots & \ell_{nn} \end{bmatrix}.$$

## Example: Upper Triangular Systems

Solve the following  $4 \times 4$  system

$$x_4 = \frac{b_4}{u_{44}} \quad \begin{bmatrix} u_{11} & u_{12} & u_{13} & u_{14} \\ 0 & u_{22} & u_{23} & u_{24} \\ 0 & 0 & u_{33} & u_{34} \\ 0 & 0 & 0 & u_{44} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} = \begin{bmatrix} b_1 \\ b_2 \\ b_3 \\ b_4 \end{bmatrix}$$

$$x_3 = \frac{b_3 - u_{34} x_4}{u_{33}}$$

$$x_2 = \frac{b_2 - u_{23} x_3 - u_{24} x_4}{u_{22}}$$

$$x_1 = \frac{b_1 - (u_{12} x_2 + u_{13} x_3 + u_{14} x_4)}{u_{11}}$$

complex

$$\begin{aligned} u_{11} x_1 + u_{12} x_2 + u_{13} x_3 + u_{14} x_4 &= b_1 \\ u_{22} x_2 + u_{23} x_3 + u_{24} x_4 &= b_2 \\ u_{33} x_3 + u_{34} x_4 &= b_3 \\ u_{44} x_4 &= b_4 \end{aligned}$$

Simple

Backward Substitution

$$x_i = \frac{b_i - \sum_{j=i+1}^n u_{ij} x_j}{u_{ii}}$$

# General Results

$0 : 1 : -2 \rightarrow$  empty array

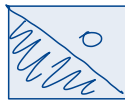
- **Backward Substitution.** To solve a general  $n \times n$  upper triangular system  $U\mathbf{x} = \mathbf{b}$ :

$$\left\{ \begin{array}{l} x_n = \frac{b_n}{u_{nn}} \text{ and} \\ x_i = \frac{1}{u_{ii}} \left( b_i - \sum_{j=i+1}^n u_{ij} x_j \right) \end{array} \right. \quad \left. \begin{array}{l} \text{can combine the two} \\ \text{w/ convention} \\ \sum_{j=n+1}^n \text{ as } 0 \end{array} \right\}$$

for  $i = n - 1, n - 2, \dots, 1$ .

- **Forward Elimination.** To solve a general  $n \times n$  lower triangular system  $L\mathbf{x} = \mathbf{b}$ :

$$\left\{ \begin{array}{l} x_1 = \frac{b_1}{l_{11}} \text{ and} \\ x_i = \frac{1}{l_{ii}} \left( b_i - \sum_{j=1}^{i-1} l_{ij} x_j \right) \end{array} \right.$$



• sample  
↓  
complex

for  $i = 2, 3, \dots, n$ .

## Summation as an inner product

$$\sum_{j=i+1}^n u_{ij} x_j = \begin{bmatrix} u_{i,i+1} & u_{i,i+2} & \cdots & u_{i,n} \end{bmatrix} \begin{bmatrix} x_{i+1} \\ x_{i+2} \\ \vdots \\ x_n \end{bmatrix}$$

# Implementation: Backward Substitution

```
function x = backsub(U,b)
% BACKSUB x = backsub(U,b)
% Solve an upper triangular linear system.
% Input:
%   U    upper triangular square matrix (n by n)
%   b    right-hand side vector (n by 1)
% Output:
%   x    solution of Ux=b (n by 1 vector)
n = length(U);
x = zeros(n,1); % preallocate
for i = n:-1:1
    x(i) = ( b(i) - U(i,i+1:n)*x(i+1:n) ) / U(i,i);
end
end
```

*Sum as inner prod.*



# Implementation: Forward Elimination

**Exercise.** Complete the code below.

```
function x = forelim(L,b)
% FORELIM x = forelim(L,b)
% Solve a lower triangular linear system.
% Input:
%   L    lower triangular square matrix (n by n)
%   b    right-hand side vector (n by 1)
% Output:
%   x    solution of Lx=b (n by 1 vector)

end
```

# Does It Always Work?

## Singularity of Triangular Matrix

A triangular matrix is singular if and only if at least one of its diagonal elements is zero.