

Announcements

Exam 3

Recitation section

Monday 2:30 Loomis 151

Review

Vdm, conditionij
Runge's phenomena

Goals

Lagrange Polynomials: General Form

$$\varphi_j(x) = \frac{\prod_{k=1, k \neq j}^m (x - x_k)}{\prod_{k=1, k \neq j}^m (x_j - x_k)}$$

$V = \mathbb{F}$

Write down the Lagrange interpolant for nodes $(x_i)_{i=1}^m$ and values $(y_i)_{i=1}^m$.

$$p_{m-1}(x) = \sum_{j=1}^m y_j \varphi_j(x)$$

Newton Interpolation

Find a basis so that V is triangular.

$$\varphi_j(x) = \prod_{k=1}^{j-1} (x - x_k) \quad (j=1 \dots n)$$

forward subst on \triangle vdm: $\mathcal{O}(n^2)$
empty product $(j=1): 1$

Why not Lagrange/Newton?

Newton Interpolation

Find a basis so that V is triangular.

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forward subst on \triangle vdm: $\mathcal{O}(n^2)$
empty product $(j=1): 1$

Why not Lagrange/Newton?

Cheap to form, expensive to evaluate, expensive to do calculus on.

Better conditioning: Orthogonal polynomials

What caused monomials to have a terribly conditioned Vandermonde?



What's a way to make sure two vectors are *not* like that?



But polynomials are functions!

Better conditioning: Orthogonal polynomials

What caused monomials to have a terribly conditioned Vandermonde?

Being close to linearly dependent.

What's a way to make sure two vectors are *not* like that?

Orthogonality

But polynomials are functions!

Orthogonality of Functions

$$f(x), g(x) \quad [1,1]$$

How can functions be orthogonal?

$$\vec{f} \cdot \vec{g} = \sum f_i g_i = (\vec{f}, \vec{g})$$

$$(f, g) = \int_{-1}^1 f(x) g(x) dx$$

$$\|f\| = \sqrt{(f, f)} = \sqrt{\int f^2 dx}$$

Constructing Orthogonal Polynomials

How can we find an orthogonal basis?

Gram-Schmidt

Demo: Orthogonal Polynomials [cleared] — Got: Legendre polynomials.
But how can I practically compute the Legendre polynomials?

Three-term recurrence:

$$P_0 = 1 \quad P_1 = x$$

$$(n+1)P_{n+1} = (2n+1)xP_n - nP_{n-1}$$

Idea: Generalize the inner product to include weights:

$$(f, g)_w = \int_{-1}^1 f(x)g(x)\omega(x)dx$$

Chebyshev Polynomials: Definitions

Three equivalent definitions:

- ▶ Result of Gram-Schmidt with weight $1/\sqrt{1-x^2}$. What is that weight?

half-circle

(Like for Legendre, you won't exactly get the standard normalization if you do this.)

- ▶ $T_k(x) = \cos(k \cos^{-1}(x))$
- ▶ $T_k(x) = 2xT_{k-1}(x) - T_{k-2}(x)$ plus $T_0 = 1$, $T_1 = x$

Chebyshev Interpolation

What is the Vandermonde matrix for Chebyshev polynomials?

What nodes?

$$T_k(x) = \cos(k \cos^{-1}(x))$$

$$x_i = \cos\left(\frac{i}{k} \pi\right) \quad (i=0, 1, \dots, k)$$

These are minima/maxima of T_k

$$V_{ij} = T_j(x_i) = \cos\left(j \cos^{-1}\left(\cos\left(\frac{i}{k} \pi\right)\right)\right)$$

$$= \cos\left(\frac{i \cdot j}{k} \pi\right)$$

Discrete cosine transform

matrix and inverse available in $O(k \log k)$ time.

Chebyshev Nodes

Might also consider roots (instead of extrema) of T_k :

$$x_i = \cos\left(\frac{2i-1}{2k}\pi\right) \quad (i = 1, \dots, k).$$

Vandermonde for these (with T_k) can be applied in $O(N \log N)$ time, too.

Edge-clustering seemed like a good thing in interpolation nodes. Do these do that?

yes

[Demo: Chebyshev Interpolation \[cleared\]](#) (Part I-IV)

Truncation Error in Interpolation

If f is n times continuously differentiable on a closed interval I and $p_{n-1}(x)$ is a polynomial of degree at most n that interpolates f at n distinct points $\{x_i\}$ ($i = 1, \dots, n$) in that interval, then for each x in the interval there exists ξ in that interval such that

$$f(x) - p_{n-1}(x) = \frac{f^{(n)}(\xi(x))}{n!} (x - x_1)(x - x_2) \cdots (x - x_n).$$

$$R(x) := f(x) - p_{n-1}(x)$$

$$Y_x(f) = R(t) - \frac{R(x)}{W(x)} W(t)$$

$$\text{let } x \in I \setminus \{x_1, \dots, x_n\}.$$

error is zero at nodes!
 $w(x)$

Truncation Error in Interpolation: cont'd.

$$R(t) = f(t) - p_{n-1}(t)$$

$$Y_x(t) = R(t) - \frac{R(x)}{W(x)} W(t) \quad \text{where} \quad W(t) = \prod_{i=1}^n (t - x_i)$$

- x_i are roots of $R(t)$ and $W(t)$
 $Y_x(x) = 0$. Y_x has $n+1$ roots. (x and the nodes)
- Y_x' has n roots ... $Y_x^{(n)}$ has 1 root in I .
 Let's call that root ξ .

$$Y_x^{(n)}(t) = f^{(n)}(t) - \frac{R(x)}{W(x)} n!$$

$$W(x) \frac{f^{(n)}(\xi)}{n!} = R(x)$$

$$0 = Y_x^{(n)}(\xi) = f^{(n)}(\xi) - \frac{R(x)}{W(x)} n!$$

Error Result: Simplified Form

Boil the error result down to a simpler form.

$$f(x) - p_{n-1}(x) = \frac{f^{(n)}(\xi(x))}{n!} (x - x_1)(x - x_2) \cdots (x - x_n).$$

Assume $x_1 < \dots < x_n$

$$|f^{(n)}(x)| \leq M \quad (x \in I)$$

h = length of the interval

Observe: $\forall x \in I, |x - x_i| \leq h$

$$\max_x |f(x) - p_{n-1}(x)| \leq C M h^n$$

► Demo: Interpolation Error [cleared]

$$\mathcal{E}(h) = \mathcal{O}(h^n)$$

n -th order convergence

Going piecewise: Simplest Case

Construct a piecewise linear interpolant at four points.

x_0, y_0		x_1, y_1		x_2, y_2		x_3, y_3
	$f_1 = a_1x + b_1$		$f_2 = a_2x + b_2$		$f_3 = a_3x + b_3$	
	2 unk.		2 unk.		2 unk.	
	$f_1(x_0) = y_0$		$f_2(x_1) = y_1$		$f_3(x_2) = y_2$	
	$f_1(x_1) = y_1$		$f_2(x_2) = y_2$		$f_3(x_3) = y_3$	
	2 eqn.		2 eqn.		2 eqn.	

Why three intervals?

(to be covered after break)



Piecewise Cubic ('Splines')

x_0, y_0		x_1, y_1		x_2, y_2		x_3, y_3
	f_1		f_2		f_3	
	$a_1x^3 + b_1x^2 + c_1x + d_1$		$a_2x^3 + b_2x^2 + c_2x + d_2$		$a_3x^3 + b_3x^2 + c_3x + d_3$	

(to be covered after break)

Piecewise Cubic ('Splines'): Accounting

x_0, y_0		x_1, y_1		x_2, y_2		x_3, y_3
	f_1		f_2		f_3	
	$a_1x^3 + b_1x^2 + c_1x + d_1$		$a_2x^3 + b_2x^2 + c_2x + d_2$		$a_3x^3 + b_3x^2 + c_3x + d_3$	

(to be covered after break)

Outline

Introduction to Scientific Computing

Systems of Linear Equations

Linear Least Squares

Eigenvalue Problems

Nonlinear Equations

Optimization

Interpolation

Numerical Integration and Differentiation

Numerical Integration

Quadrature Methods

Accuracy and Stability

Gaussian Quadrature

Composite Quadrature

Numerical Differentiation

Richardson Extrapolation

Initial Value Problems for ODEs

Boundary Value Problems for ODEs

Partial Differential Equations and Sparse Linear Algebra

Fast Fourier Transform

Additional Topics

Conditioning

Derive the (absolute) condition number for numerical integration.

(to be covered after break)

Interpolatory Quadrature: Examples

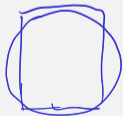
l_i : Lagrange basis

$$f(x) \approx \sum_{i=1}^n f(x_i) l_i(x)$$

$$\int_a^b f(x) dx \approx \sum_{i=1}^n f(x_i) \underbrace{\int_a^b l_i(x) dx}_{\omega_i}$$

x_i : nodes ω_i : weights

"Quadrature rule"



"square" = "Quadrat" in German

Interpolatory Quadrature: Computing Weights

How do the weights in interpolatory quadrature get computed?



[Demo: Newton-Cotes weight finder](#) [\[cleared\]](#)