

Ch 3: Introduction to Interpolation

Monday, October 6, 2025

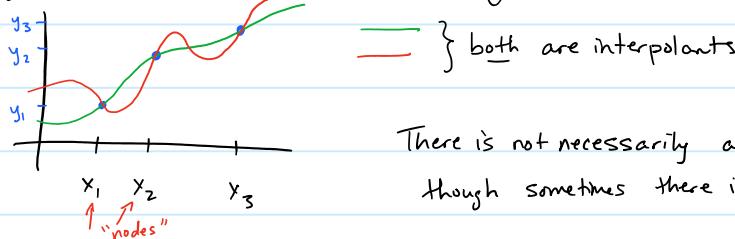
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- Topics:
- Interpolation definitions
 - Interpolating polynomials: existence / uniqueness,
 - vis-a-vis Weierstrass
 - Vandermonde matrix
 - Overfitting
 - Fix: local / composite interpolation
 - Runge phenomenon
 - Fix: don't use equally spaced points!
 - eg Chebyshev nodes
- to be discussed in more detail later in the course (or next semester)

Interpolation

Given a dataset $\{(x_i, y_i)\}$, an interpolant is a function g

that agrees with the dataset, meaning $g(x_i) = y_i$



There is not necessarily a "true" function f though sometimes there is.

Usually we want the interpolating function to be nice,
i.e., both easy to work with (ex: integrate)
and have nice properties like differentiability, etc.

Traditionally, "interpolation" means filling in the blanks inside an interval

Ex: it was 73°F at 8 AM
and 82°F at 11 AM.

What's a guess for the temperature at 9 AM?

(one approach: find an interpolant g such that
 $g(8) = 73$, $g(11) = 82$, then look at $g(9)$)

...in contrast to "extrapolation" which is about prediction outside your interval,
i.e., the future.

In above example, extrapolation might ask what is the
temperature at noon?

Interpolation isn't the only approach

- another class of methods is "least-squares" and many variants
i.e. "best approximation"

- as you'll see, we'll focus on polynomials. The first thing you
might think of is the Taylor polynomial but this is not a good idea
usually, as these are only good very locally, and not on an interval

Polynomials

Often we require the interpolant to be a polynomial.

Why?

- Easy to work with, e.g. derivatives & integrals are very easy
- They are smooth ($C^\infty(\mathbb{R})$)
- Nice theory, and hundreds of years of experience

Ex:

"interpolants" are not unique, but

"polynomial interpolants" of low-degree are unique, as made precise below:

Refresher: Polynomials

Degree 0 polynomials: $p(x) = c$. If $c \neq 0$, no roots.
If $c = 0$, ∞ roots.

"Fundamental Theorem of Algebra"

(1) Every polynomial h of degree 1 or higher has at least 1 (possibly complex) root

(2) (Corollary) A n^{th} degree polynomial has n (possibly complex) roots
if you count with multiplicity *

* Except 0 polynomial has ∞ roots

} i.e. at most n roots if you only count unique roots (i.e., not w/ multiplicity)

Ex: $f(x) = (x-1)^2(x-4)$ has "3" roots: $\{1, 1, 4\}$
if we count w/ multiplicity.

In particular, $f(x) = a_n(x-x_1)^{m_1} \cdot (x-x_2)^{m_2} \cdots (x-x_k)^{m_k}$
 m_i unique, $\sum_{i=1}^k m_i = n$

Corollary (2.18)

If $p(x)$ and $g(x)$ are polynomials, both of degree n
or less, then if we have a set of k ^{distinct} points $\{x_1, x_2, \dots, x_k\}$,
then if

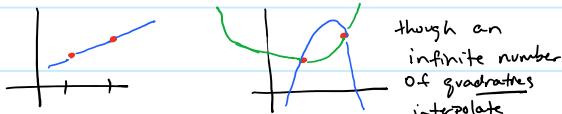
$$p(x_i) = g(x_i) \quad \forall i=1, \dots, k$$

then $k > n \Rightarrow p = g$

Proof: If $k > n$, this means $p-g$ is a n degree polynomial
with more than n roots, which is impossible unless $p-g = 0$. \square

Theorem: Uniqueness
 If two polynomials $p(x)$ and $g(x)$ agree on K data points $\{x_i\}_{i=1}^K$ and both polynomials are degree n or less,
 then $p = g$ (ie, the interpolating polynomial of degree n is unique)
 if $K > n$.

Ex: "two points determine a line"
 $k=2$ $n=1$



though an infinite number of quadratics interpolate 2 points.

Theorem: existence

If $\{x_i\}_{i=1}^K$ are distinct points, and $\{y_i\}_{i=1}^K$ are any y -values,
 then there exists a polynomial of degree $n = K-1$ that interpolates the data.

Proof: a $n = K-1$ degree polynomial looks like

$$p(x) = c_K x^{K-1} + c_{K-1} x^{K-2} + \dots + c_2 x^1 + c_1$$

so it has K degrees of freedom (parameters) $\{c_i\}_{i=1}^K$

Solve:

$$\begin{aligned} c_K x_1^{K-1} + \dots + c_2 x_1 + c_1 &= y_1 && \text{plug in } x=x_1 \\ \vdots & & & \vdots \\ c_K x_K^{K-1} + \dots + c_2 x_K + c_1 &= y_K && x=x_K \end{aligned}$$

linear system

$$\underbrace{\begin{pmatrix} 1 & x_1 & x_1^2 & \dots & x_1^{K-1} \\ 1 & x_2 & x_2^2 & \dots & x_2^{K-1} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & x_K & x_K^2 & \dots & x_K^{K-1} \end{pmatrix}}_{V} \begin{pmatrix} c_1 \\ c_2 \\ \vdots \\ c_K \end{pmatrix} = \begin{pmatrix} y_1 \\ y_2 \\ \vdots \\ y_K \end{pmatrix}, \text{ i.e. } \boxed{V \cdot \vec{c} = \vec{y}}$$

a matrix of this form is called a Vandermonde matrix
 and you can prove $\det(V) = \prod_{1 \leq i < j \leq K} (x_i - x_j)$

so $\{x_i\}$ distinct $\Rightarrow \det(V) \neq 0 \Rightarrow V$ is invertible
 $\Rightarrow \vec{c} = V^{-1} \vec{y}$ is the (unique) sol'n. \square

don't confuse with
the Wronskian

ex $\prod_{i=1}^3 z^i = z^1 \cdot z^2 \cdot z^3$

notation: " \prod " means product
 like " \sum " for sum

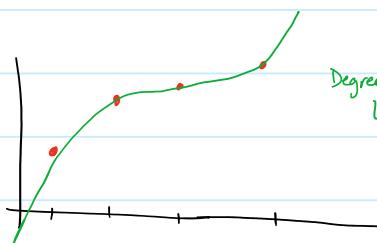
Construction

i.e. how to find coefficients $\{c_i\}$.

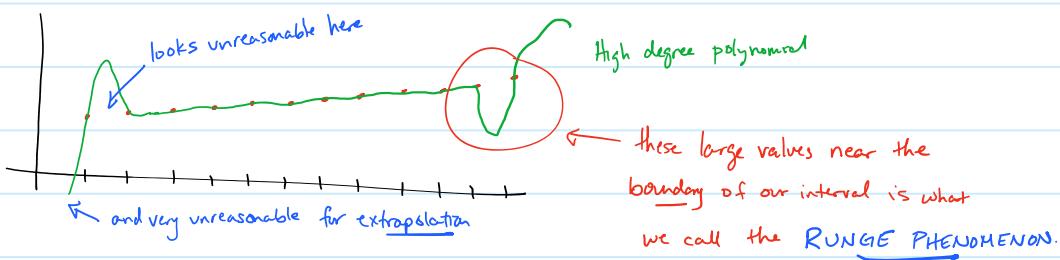
Do not invert the Vandermonde matrix, since this is ill-conditioned

Instead, see the other videos (ch 31, 3.3), so we'll come back to this

We actually proved
existence and uniqueness.

Overfitting

Degree 3 polynomial.
Looks ok



What to do? (it's a fundamental problem, independent of the algorithm)

① With many nodes, we can't interpolate globally with a low-degree polynomial. But, we could interpolate locally w/ a low-degree polynomial, and then stitch these together (ie, create a piecewise polynomial)

We'll later discuss **Splines** and **composite quadrature**, which are both techniques based on this general idea

② If we have the luxury of choosing the nodes $\{x_i\}$, (sometimes this is reasonable, like for numerical integration, though other times it's not reasonable, like if it's a measurement someone gives you).

then choose the nodes wisely. In particular, don't choose them equispaced, but rather they should concentrate near the boundaries (ex. **Chebyshev nodes** of the second kind)

$$x_i = -\cos\left(\frac{i \cdot \pi}{n}\right), i=0, 1, \dots, n$$

for the interval $[-1, 1]$

More on this later

(ex., ch. 4.7 Gaussian Quadrature → 2nd semester
ch. 8.3 Chebyshev polynomials)

Last bit of theory:

Weierstrass Approximation theorem

Thm: if $f \in C([a, b])$ then ($\forall \varepsilon > 0$) \exists a polynomial p
such that $|f(x) - p(x)| < \varepsilon$ ($\forall x \in [a, b]$)

proof via Bernstein polynomials.

Remark Can we find a polynomial p such that

$$|f(x) - p(x)| < \varepsilon \quad \forall x \in \mathbb{R} ? \text{ ie., change } [a, b] \text{ to } \mathbb{R} ?$$

No! All interesting polynomials (meaning all polynomials that are not identically constant) must go to $\pm \infty$ as $|x| \rightarrow \infty$.

So if $f(x) = \sin(x)$, we cannot hope to find a global (i.e., all of \mathbb{R}) polynomial approximation. Must settle for local approximation.

This is another reason for the piecewise / composite approach.