

April 14, 2020

Numerical Analysis

Last time: minimax function approximation

Goal: Find $p_n \in P_n$ such that

$$\|f - p_n\|_\infty = \min_{q \in P_n} \max_{x \in [a,b]} |f(x) - q(x)|$$

Thm Let $n \geq 0$, $f(x) = x^{n+1}$, then $\|f - p_n\|_\infty$ is minimized

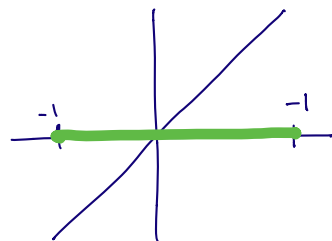
when $p_n(x) = x^{n+1} - \frac{1}{2^n} \cos((n+1) \arccos x)$.

T_{n+1} = Chebyshev polynomial.

Ex: $n=0 \Rightarrow f(x) = x$

$$p_n(x) = x - \frac{1}{2^0} \cdot \cos(\arccos x)$$

$$= x - 1 \cdot x = 0$$



Chebyshev polynomials

$$T_n(x) = \cos(n \arccos x) \quad n = 0, 1, 2, \dots$$

$$T_0(x) = 1$$

$$T_1(x) = x$$

$$T_2(x) = \cos(2 \arccos x)$$

\vdots

$$T_{n+1}(x) = 2x \cdot T_n(x) - T_{n-1}(x)$$

← polynomial of degree $n+1$

Usually only concerned with $T_n(x)$ for $x \in [-1, 1]$.

can prove using trig identities applied to this definition

Trivially, the zeros of T_n can be computed as:

$$\cos(n \arccos x) = 0$$

$$\Rightarrow n \arccos x = \frac{\pi}{2} (2m+1) \quad \text{for } m = \dots, -2, -1, 0, 1, 2, \dots$$

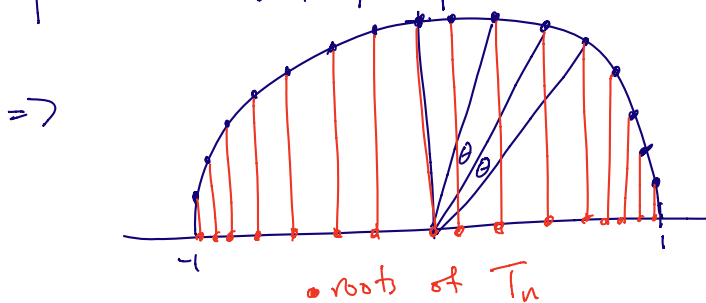
$$\arccos x = \frac{\pi}{2n} (2m+1)$$

$$x = \cos \left(\frac{(2m+1)\pi}{2n} \right) \quad m = 0, 1, \dots, n-1 \quad \left(\begin{array}{l} \text{roots repeat} \\ \text{for } m \geq n \end{array} \right).$$

The roots on $[-1, 1]$ can be ordered from $[-1, 1]$ as:

$$x_j = -\cos \left(\frac{(2j-1)\pi}{2n} \right) \quad j = 1, \dots, n \quad (n \text{ roots}).$$

this is an angle, as $j = 1 \dots n$, we get equispaced points in $(0, \pi)$.



Claim: Interpolation of a function f on $[a, b]$ with a degree n polynomial p_n at the roots of T_n - i.e. Chebyshev nodes - yields a near minimax polynomial approximant.

Idea: The approximation error of the interpolation is

$$f(x) - p_n(x) = \frac{f^{(n+1)}(\xi)}{(n+1)!} \prod_{j=0}^n (x - x_j)$$

If x_j are chosen to be the roots of T_n (properly scaled to this interval), then $\prod_{j=0}^n (x - x_j) = \frac{1}{2^n} T_n(x)$
 Factorization of $\frac{1}{2^n} T_n(x)$.

What is special about $\frac{1}{2^n} T_n(x)$?

It can be shown that $\frac{1}{2^n} T_n$ is the minimum norm monic polynomial. A monic polynomial of degree n is one whose coefficient on the x^n term is 1.

$$\|f - p_n\|_\infty \leq \frac{M_{n+1}}{(n+1)!} \underbrace{\left\| \prod_{j=0}^n (x - x_j) \right\|_\infty}_{\text{just minimize this by choosing the "best" } x_j\text{'s.}}$$

Approximation in the 2-norm

The 2-norm of a function, with some general continuous weight function $w > 0$, on (a, b) is:

$$\|f\|_2^2 = \int_a^b (f(x))^2 w(x) dx.$$

Goal: Find $p_n \in P_n$ such that $\|f - p_n\|_2$ is minimized. This is a least squares approximation to f - exactly analogous to solving least squares problems in finite dimensions (ie, Linear Alg.)

Therefore the solution is obtained by computing the orthogonal projection of f onto the space P_n , under the inner product:

$$(f, g)_w = \text{inner product of } f \text{ with } g$$

$$= \int_a^b f(x) g(x) w(x) dx.$$

There is no reason why the best p_n for the 2-norm error is the same p_n for the ∞ -norm error. In more detail...

Easy to check that this is indeed an inner product

Our goal: For a function $f \in L^2_w(a, b)$ $\left(\|f\|_2 = \sqrt{\int_a^b (f(x))^2 w(x) dx} < \infty \right)$
 find $p_n \in \mathcal{P}_n$ such that:

$$\begin{aligned} \|f - p_n\|_2^2 &= \inf_{q \in \mathcal{P}_n} \|f - q\|_2^2 \\ &= \inf_{q \in \mathcal{P}_n} \int_a^b (f(x) - q(x))^2 w(x) dx. \end{aligned}$$

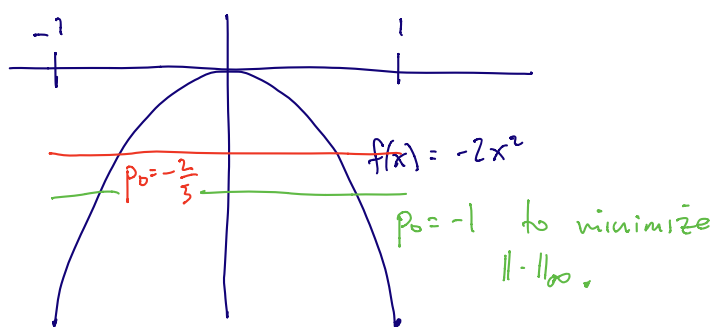
Ex: Let $n=0$, $f(x) = -2x^2$ on $[-1, 1]$, $w(x) = 1$.

Find $p_0 = c$ to minimize $\|f - p_0\|_2^2$.

$$\begin{aligned} \|f - p_0\|_2^2 &= \int_{-1}^1 (-2x^2 - c)^2 dx \\ &= 2c^2 + \frac{8c}{3} + \frac{8}{5} \end{aligned}$$

Find c to minimize this quantity.

$$\begin{aligned} \frac{d}{dc} \|f - p_0\|_2^2 &= 4c + \frac{8}{5} = 0 \\ \Rightarrow c &= -\frac{2}{3} \end{aligned}$$



What would the best approximation in $\| \cdot \|_\infty$ be?

Computing the best 2-norm approximation

Recall: The least squares solution to $A\vec{x} = \vec{b}$ is obtained

by solving $A\vec{x} = \underbrace{QQ^T}_{\text{projection of } \vec{b} \text{ onto } \text{col}A} \vec{b}$, Q is obtained by applying the Gram-Schmidt process to the columns of A .

To apply Gram-Schmidt, all that was needed was a vector space and an inner product.

We can do the same thing for polynomial least squares:

(1) Trivially, P_n is a vector space.

(2) We can define an inner product on P_n by:
 $(f, g) = \int_a^b f(x) g(x) w(x) dx.$

Two functions are orthogonal if $(f, g) = 0.$

Let $p_0, p_1, p_2, \dots, p_n$ be a basis for $P_n.$

The 2-norm approximation problem takes the form:

For $q(x) = \sum_{j=0}^n c_j p_j(x)$, $\|f - q\|_2^2$ is given by: (set $w=1$ for now)

$$\begin{aligned} A &= \int_a^b \left(f(x) - \sum_{j=0}^n c_j p_j(x) \right)^2 dx \\ &= \int_a^b (f(x))^2 dx - 2 \sum_{j=0}^n c_j \int_a^b f(x) p_j(x) dx + \sum_{j=0}^n \sum_{k=0}^n c_j c_k \int_a^b p_j(x) p_k(x) dx \\ &= (f, f) - 2 \sum_{j=0}^n c_j (f, p_j) + \sum_{j=0}^n \sum_{k=0}^n c_j c_k (p_j, p_k) \end{aligned}$$

Writing down

$\nabla A = \vec{0}$, we have that:

$$\frac{\partial A}{\partial c_0} = 0, \quad \frac{\partial A}{\partial c_1} = 0, \quad \dots, \quad \frac{\partial A}{\partial c_n} = 0$$

$$\Rightarrow \frac{\partial A}{\partial c_\ell} = -2 (f, p_\ell) + 2 \sum_{k=0}^n c_k (p_\ell, p_k) = 0$$

$$\Rightarrow \sum_{k=0}^n c_k (p_\ell, p_k) = (f, p_\ell) \quad \leftarrow \begin{array}{l} \text{linear} \\ \text{n+1 equations, in n+1 variables} \\ c_0, \dots, c_n. \end{array}$$

In matrix form:

$$\begin{pmatrix} (p_0, p_0) & \dots & (p_0, p_n) \\ (p_1, p_0) & \dots & (p_1, p_n) \\ \vdots & & \vdots \\ (p_n, p_0) & \dots & (p_n, p_n) \end{pmatrix} \begin{pmatrix} c_0 \\ c_1 \\ \vdots \\ c_n \end{pmatrix} = \begin{pmatrix} (f, p_0) \\ (f, p_1) \\ \vdots \\ (f, p_n) \end{pmatrix}$$

Therefore once this linear system has been solved, the best 2-norm approximation to f is:

$$q = c_0 p_0 + c_1 p_1 + \dots + c_n p_n.$$

If the p_k 's were orthonormal, i.e. $(p_k, p_m) = \delta_{km}$, then the above system is of the form:

$$\mathbf{I} \vec{c} = \text{RHS}.$$

$$s_0 = c_k = (f, p_k). \quad (\text{if } p_k \text{'s are orthonormal}).$$