

Orthogonal Polynomials

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Interpolation works for a
"black-box" function.

Equi-spaced nodes not a good
choice in general

$$P_n(x_k) = f(x_k)$$

\nwarrow nodes \nearrow

$$\text{? } \epsilon(x) = |P_n(x) - f(x)|$$

$$x \neq x_k$$

Piece-wise polynomial interp
works for any nodes, but $\epsilon(x)$ is
not very small

Another way to approximate

$$f(x) \approx p_n(x) \in \mathcal{P}_n \text{ on } [a, b]$$

$$p_n^* = \arg \min_{p_n \in \mathcal{P}_n} \| f(x) - p_n(x) \|$$

L₂-approximation

$$p_n^* = \arg \min_{p_n \in \mathcal{P}_n} \| f(x) - p_n(x) \|_2$$

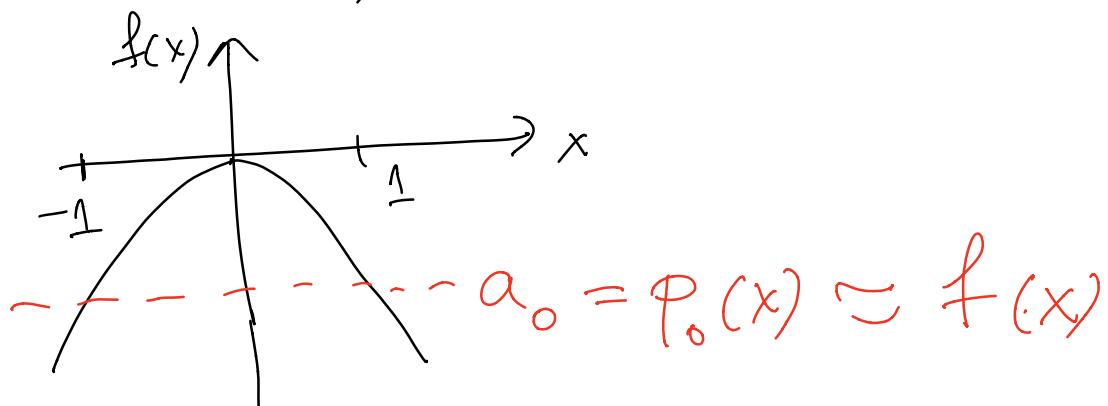


Least-squares
"fitting"

$$\left(\int |f(x) - p_n(x)|^2 dx \right)^{1/2}$$

E.g. Approximate $f(x) = -2x^2$
 in $P_{n=0}$ (space of constants)

on $[-1, 1]$



Idea: Average of function

$$a_0 = \frac{\int_{-1}^1 f(x) dx}{2} \quad (?)$$

$$= \frac{\int_{-1}^1 -2x^2 dx}{2} = -\frac{x^3}{3} \Big|_{-1}^1 = -\frac{2}{3}$$

$$\underline{L_2} : \quad \|f - a_0\|^2 = \int_{-1}^1 (-2x - a_0)^2 dx$$

$$a_0^* = \arg \min_{a_0} 2a_0^2 + \frac{8a_0}{3} + \frac{8}{5}$$

$$\text{Derivative} = 0 \Rightarrow a_0^* = -\frac{2}{3}$$

$$P_0^* \approx a_0^* = -\frac{2}{3}$$

More generally:

$$a_0^* = \arg \min_{a_0} \int_{-1}^1 (f(x) - a_0)^2 dx$$

$$= \arg \min_{a_0} \left[\int_{-1}^1 f^2(x) dx - 2a_0 \int_{-1}^1 f(x) dx + a_0^2 \int_{-1}^1 dx \right]$$

$$\frac{\partial}{\partial a_0} \left[\cancel{\int f^2 dx} - 2a_0 \int f dx + a_0^2 \int dx \right]$$

$$-\cancel{\int f dx} + a_0 \cancel{\int dx} = 0$$

$$\Rightarrow a_0 = \frac{\int f dx}{\int dx} = \frac{\int f x dx}{a}$$

average of
function

Now make even more general

$$P_n^* \in \mathcal{P}_n$$

Step 1 :

Basis = $\{ P_0(x), P_1(x), \dots, P_{n+1}(x) \} \subset \mathcal{P}_n$

and linearly independent

$$P_n^*(x) = \sum_{j=0}^n a_j P_j(x)$$

Best L₂ approx

$$F(\vec{a}) = \int_a^b \left(f(x) - \sum_{j=0}^n a_j P_j(x) \right)^2 dx$$

given
unknown

$$\frac{\partial F}{\partial a_k} = 0, \quad k=0, \dots, n$$

"normal equations"

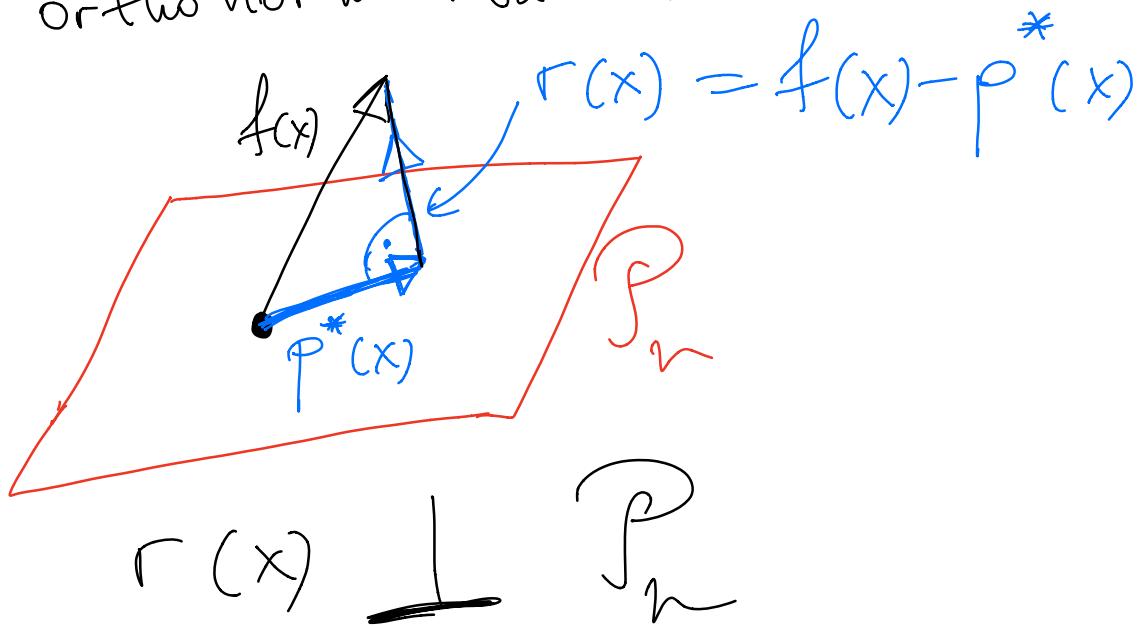
Linear System of $(n+1)$ equations for $(n+1)$ unknowns \vec{a}

Reminder: $Ax = b$, over-determined

$$x^* = \arg \min_x \|Ax - b\|^2$$

① Normal egs: $(A^T A)x = A^T b$

② QR factorization, Gram-Schmidt
orthonormalization.



Conclusion: $r(x)$ must be
orthogonal to all polynomials
of degree $n \iff$

$$r(x) \perp P_k(x), k=0, \dots, n$$

$$(r(x), P_k(x)) = 0 \quad \forall k$$

|

$$\left(\underset{\downarrow}{f} - p^*, p_k \right)_2 = 0 \quad \forall k$$

$$\left(f - \sum_{j=0}^n a_j p_j, p_k \right) = 0$$

$$(f, p_k) = \sum_{j=0}^n a_j (p_j, p_k)$$

$$\sum_{j=0}^n (p_j, p_k) a_j = (f, p_k)$$

$k = 0, \dots, n$

$(n+1)$ equations for $(n+1)$ coeffs.

$$\underbrace{V}_{\swarrow \rightarrow} \vec{a} = \vec{f}$$

Vandermonde matrix

$$V_{jk} = (P_j, P_k) =$$

Can pre-compute
exactly

$$= \int_a^b P_j(x) P_k(x) dx$$

$$f_k = (f, P_k) = \int_a^b f(x) P_k(x) dx$$

cannot compute
exactly for any $f(x)$

E.g.

Take monomials as basis for

$$P_n : \{1, x, x^2, \dots, x^n\}$$

$$\{P_0(x), P_1(x), \dots, P_n(x)\}$$

$$a=0, b=1$$

$$\begin{aligned}
 V_{ij} &= \int_0^1 P_i(x) P_j(x) dx = \\
 &= \int_0^1 x^i x^j dx = \int_0^1 x^{i+j} dx \\
 &= \left. \frac{x^{i+j+1}}{i+j+1} \right|_0^1 = \frac{1}{i+j+1}
 \end{aligned}$$

$$V_{ij} = \frac{1}{i+j+1} \quad \leftarrow \text{Hilbert matrix}$$

Very ill conditioned for $n \geq 10$
(Worksheets)

Using monomials as a basis
is a bad idea (ill-conditioned)

$$\nabla \vec{a} = \vec{f}$$

What's the simplest ∇ ?

$$\nabla = I \Rightarrow \vec{a} = \vec{f}$$

$$V_{ij} = \delta_{ij} = \begin{cases} 0 & \text{if } i \neq j \\ 1 & \text{if } i = j \end{cases}$$

$$\left\{ \begin{array}{l} \int p_i(x) p_j(x) dx = 0 \quad \text{if } i \neq j \\ \int p_i^2(x) dx = 1 \end{array} \right.$$

← skip this
in practice

orthonormal polynomials

$$a_k = f_k = (f, p_k) = \int f(x) p_k(x) dx$$

$$P_n^* = \sum_{k=0}^n a_k p_k(x)$$

aside: Often polynomials are orthogonal but not normalized

$$a_k = \frac{(f, P_k)}{(P_k, P_k)}$$

Orthogonal polynomials

[How to find an orthogonal basis for P_n ?]

Recall:

How to find an orthogonal basis for $\text{col } A$?

for upper triangular

$$A = Q R$$

\nearrow orthonormal \nwarrow upper triangular

Compute by Gram-Schmidt process

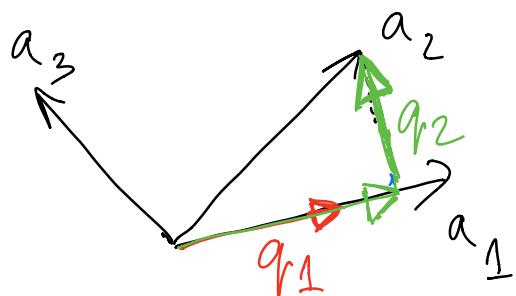
"columns of A " $\equiv \{1, x, x^2, \dots, x^n\}$

$\left\{ \begin{array}{l} \text{col}(A) \\ \downarrow \text{GS} \\ \textcircled{O} \end{array} \right\} \equiv P_n$

\equiv orthonormal basis for P_n

$\left\{ \begin{array}{l} \text{Start with monomials, then} \\ \text{do GS on them one by} \\ \text{one to get orthogonal polynomials} \end{array} \right\}$

Review:



$$q_1, q_2, q_3$$

$$q_1 \perp q_2 \perp q_3$$

$$q_1 = a_1$$

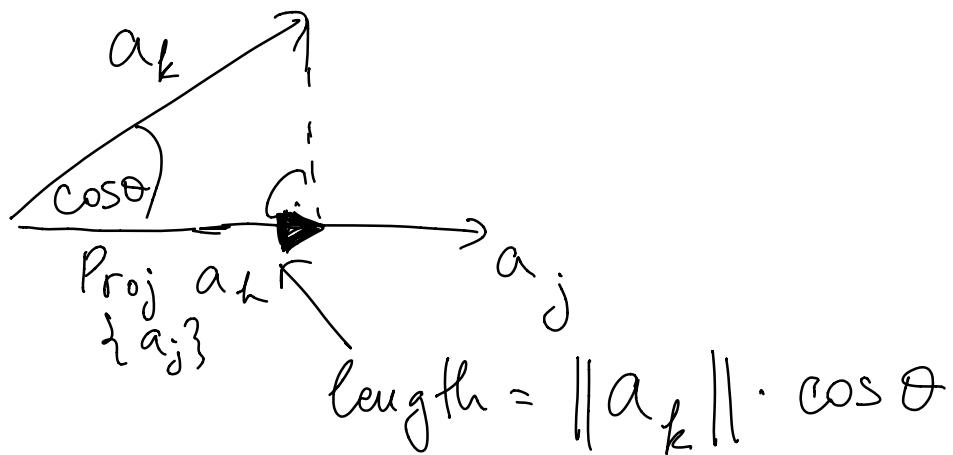
↑
skip normalization

$$q_2 = a_2 - \text{Proj}_{\{a_1\}} a_2$$

$$q_3 = a_3 - \text{Proj}_{\{a_1, a_2\}} a_3$$

$$= \vec{a}_3 - \underbrace{\text{Proj}_{\{\vec{a}_1\}} \vec{a}_3}_{\text{Proj}_{\{\vec{a}_1\}} \vec{a}_3} - \underbrace{\text{Proj}_{\{\vec{a}_2\}} \vec{a}_3}_{\text{Proj}_{\{\vec{a}_2\}} \vec{a}_3}$$

$$\rightarrow \underbrace{\text{Proj}_{\{\vec{a}_j\}} \vec{a}_k}_{\text{Proj}_{\{\vec{a}_j\}} \vec{a}_k} = \underbrace{\lambda_{kj}}_{\text{coefficient}} \underbrace{\vec{a}_j}_{\vec{a}_j}$$



$$(\lambda_{kj} \vec{a}_j, \vec{a}_j) = (\vec{a}_k, \vec{a}_j)$$

$$\lambda_{kj} = \frac{\vec{a}_k \cdot \vec{a}_j}{\vec{a}_j \cdot \vec{a}_j} = \frac{(\vec{a}_k, \vec{a}_j)}{\|\vec{a}_j\|^2}$$

$$\underbrace{\text{Proj}_{\{\vec{a}_j\}} \vec{a}_k}_{\text{Proj}_{\{\vec{a}_j\}} \vec{a}_k} = \frac{(\vec{a}_k, \vec{a}_j)}{\|\vec{a}_j\|^2} \vec{a}_j$$

GS for polynomials

$$\left\{ \begin{array}{l} 1, x, x^2, x^3, \dots \\ P_0, P_1, P_2, P_3, \dots \end{array} \right\}$$

Orthogonal polynomials

$$\left\{ \varphi_1, \varphi_2, \dots \right\}$$

$$\varphi_1 = P_0 = 1 \quad (\text{Degree zero})$$

$$\varphi_2 = P_1 - \text{Proj}_{\{\varphi_1\}} P_1$$

$$= P_1 - \frac{(\varphi_0, P_1)}{(\varphi_0, \varphi_0)} \varphi_0$$

$$= P_1 - \frac{\int_{-1}^1 x \cdot dx}{\int_{-1}^1 dx} \cdot 1$$

$$= p_1 - \underset{\text{anti-symmetry}}{\underset{\text{zero by}}{\cancel{p_0}}} = p_1 = x$$

$$\varphi_1 = x$$

$$\varphi_2 = p_2 - \frac{(\varphi_0, p_2)}{(\varphi_0, \varphi_0)} \varphi_0 - \frac{(\varphi_1, p_2)}{(\varphi_1, \varphi_1)} \varphi_1$$

$$= x^2 - \frac{\int x^2 dx}{\int dx} - \frac{\cancel{\int x^3 dx}}{\cancel{\int x^2 dx}} \cdot x$$

zero by
antisymmetry

$$= x^2 - \frac{x^3}{3} \Big|_1^{-1} \Rightarrow$$

$$\varphi_2 = x^2 - \frac{1}{3}$$

$$\Psi_n(x) = x^n - \sum_{j=0}^{n-1} \frac{\left(\int_{-1}^1 \Psi_n(x) \Psi_j(x) dx \right)}{\left(\int_{-1}^1 \Psi_j^2(x) dx \right)} \Psi_j(x)$$

↑
 Orthogonal polynomials on $[-1, 1]$
 in the standard L_2 inner product

Legendre polynomials
 (table in Wiki)

For a general $[a, b]$

$$t = \frac{x+1}{2}(b-a) + a$$

$t \in [a, b]$ when

$$x \in [-1, 1]$$

$$0 = \int_{-1}^1 \Psi_j(x) \Psi_k(x) dx = \int_a^b \Psi_j(t) \Psi_k(t) dt$$

$x \leftrightarrow t$
 Legendre polys
 on $[-1, 1]$ \longleftrightarrow Orthogonal
 polys on
 $[a, b]$

Recall: Optimal L_2 approx:

$$f(x) \approx p_n^*(x) = \sum_{k=0}^n a_k \varphi_k(x)$$

Legendre
polynomials

$$a_k = \frac{(f, \varphi_k)}{(\varphi_k, \varphi_k)} = \frac{\int f(x) \varphi_k(x) dx}{\int \varphi_k^2(x) dx}$$

Homework: $f(x) = \sin(x)$ on
 $[0, \pi]$

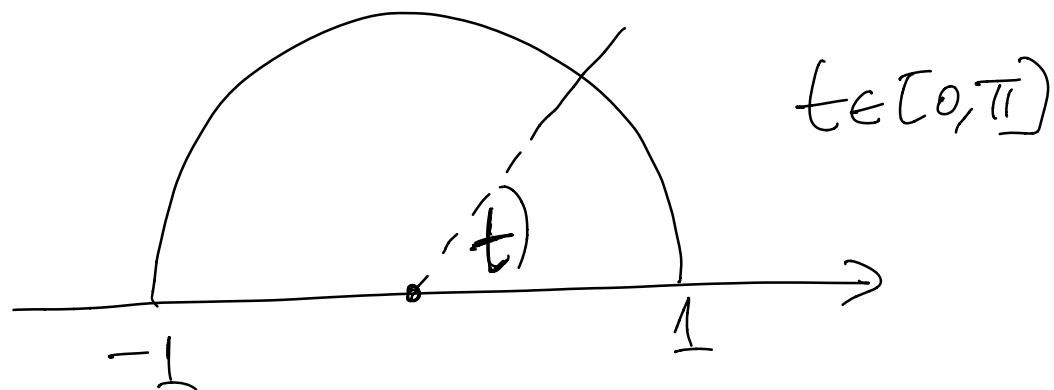
$$\text{E.g. } \varphi_2(t) = x^2 - \frac{1}{3} = \left(\frac{2t}{\pi} - 1\right)^2 - \frac{1}{3}$$

Chebyshev polynomials

$$\int_0^{\pi} \cos(nt) \cos(mt) dt = 0 \quad \text{if } m \neq n$$

\cos 's are orthogonal in L_2

on $(0, \pi)$ (trigonometric
orthogonal polynomials)



$$x = \cos t \in [-1, 1]$$

$$t = \arccos(x)$$

$$\int_{x=-1}^1 \cos(m \cdot a \cos x) \cos(n \cdot a \cos x) \frac{dx}{\sqrt{1-x^2}} = 0$$

$T_m(x)$ $T_n(x)$

$$T_k(x) = \cos(k \cdot a \cos x)$$

\uparrow $k=0, 1, 2, \dots$

A polynomial

$$\left\{ \begin{array}{l} T_0 = 1 \\ T_1 = x \\ T_2 = 2x^2 - 1 \quad \text{or} \quad x^2 - \frac{1}{2} \\ T_3 = 4x^3 - 3x \quad \text{or} \quad x^3 - \frac{3}{4}x \\ \dots \end{array} \right.$$

not Legendre

$$\int_{x=-1}^1 T_m(x) T_n(x) \frac{dx}{\sqrt{1-x^2}} = 0$$

$$(T_m, T_n)_{(1-x^2)^{-1/2}} = 0$$

$$(f, g) = \int_{x=-1}^1 \frac{f(x) g(x) dx}{\sqrt{1-x^2}}$$

Chebyshev polynomials are orthogonal w.r.t. the weighted L₂ inner product with weight function

$$w(x) = \frac{1}{\sqrt{1-x^2}} \geq 0 \quad \text{in } [-1, 1]$$

Orthogonal polys & computing

Theorem:

#1 The roots of an orthogonal polynomial $\varphi_{j \geq 1}$ are real & distinct and in (a, b)

[See th 9.4 in theory textbook]

#2 These roots are "good" nodes for polynomial interpolation

E.g. Legendre: $x^2 - \frac{1}{3} = 0 \Rightarrow x = \pm \sqrt{\frac{1}{3}}$

Chebyshev: $x^2 - \frac{1}{2} = 0 \Rightarrow x = \pm \sqrt{\frac{1}{2}}$

Roots of Chebyshev polys:

$$T_k(x) = \cos(k \cdot a \cos(x)) = 0$$

$$k \cdot a \cos(x) = (2n+1) \frac{\pi}{2}$$
$$n \in \mathbb{Z}$$

$$a \cos x = \frac{2n+1}{k} \frac{\pi}{2}$$

$$\left\{ \begin{array}{l} x_n = \cos \left(\frac{2n+1}{k} \frac{\pi}{2} \right) \\ \text{is a root of } T_k(x) \end{array} \right.$$

$$t \in (0, \pi)$$

$$x \in (-1, 1)$$

$$0 \leq \frac{2n+1}{k} \leq 2$$

$$\Rightarrow 0 \leq n \leq \frac{(2k-1)}{2}$$

Do not include $-1, 1$

$$x_n = \cos \left(\frac{(n+1/2)\pi}{k} \right)$$

$$n = 0, \dots, \frac{2k-1}{2}$$

or equally good

$$x_n = \cos \left(\frac{n\pi}{k+1} \right)$$