

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

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

L_2 -approximation

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

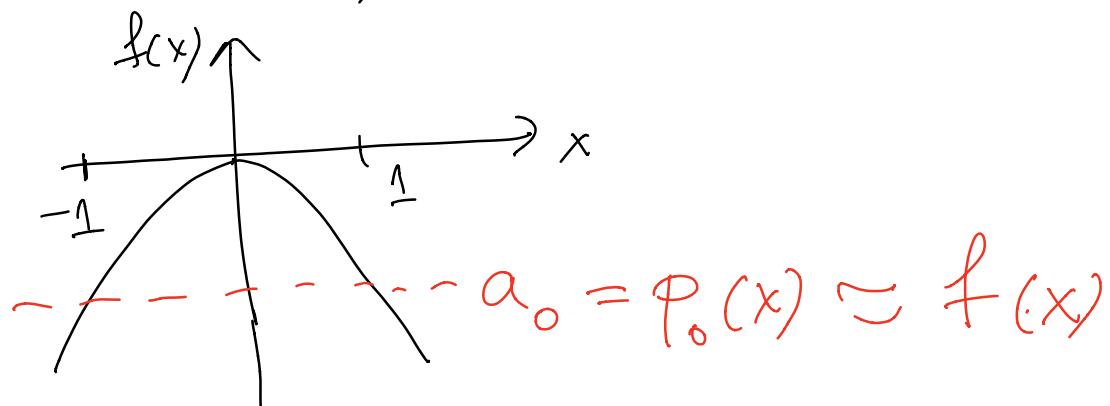


Least-squares
"fitting"

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

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:

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

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

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

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

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

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"

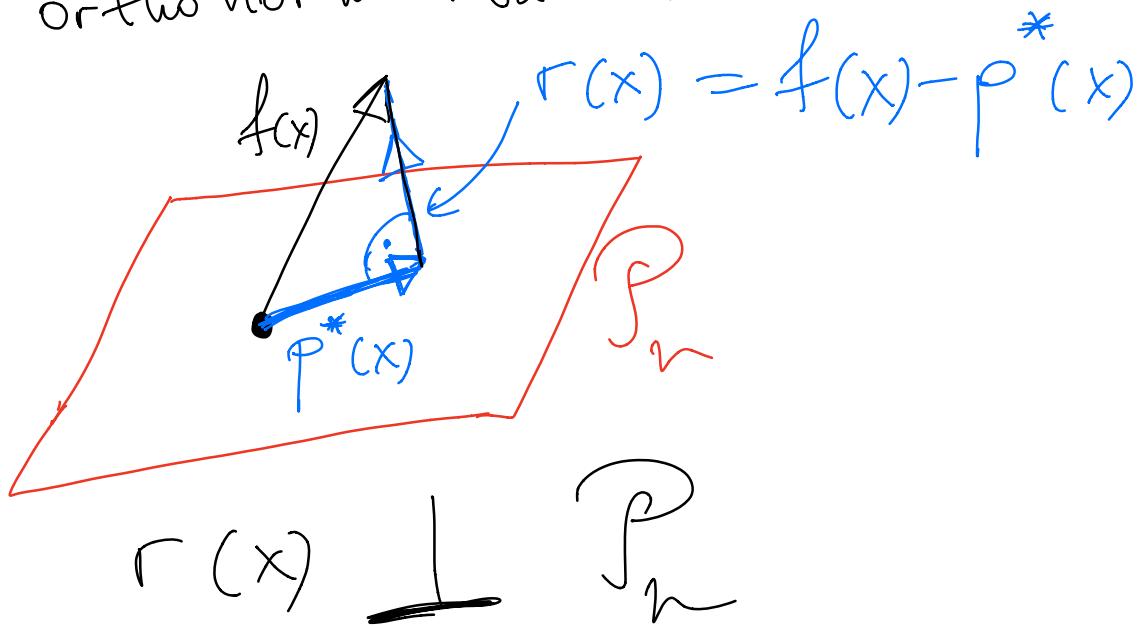
Linear System of $(n+1)$ equations for $(n+1)$ unknowns \vec{a}

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

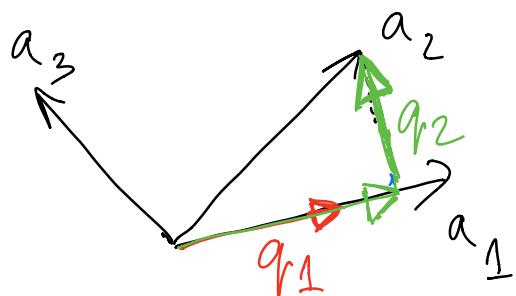
"columns of A " $\equiv \{1, x, x^2, \dots, x^n\}$

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

\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

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

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

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

$$\rightarrow \underbrace{\text{Proj}_{\{a_j\}} a_k}_{\{a_j\}} = ?$$