

Last time: Instead of solving a square matrix system for polynomial interpolation, what if we used $m > n+1$ points so that $p(x_j) = f(x_j)$ $j=1, \dots, m$

$$\text{Suppose } p(x) = \sum_{j=0}^n c_j \varphi_j(x) \quad \underbrace{\varphi_j}_{V_{ij} = \varphi_j(x_i)}$$

\Rightarrow the interpolation system $[V][c] = [y]$ (*)
is over-determined

We can solve (*) using least squares: $\min_c \|Vc - y\|_2^2$
 $\Rightarrow c$ solves the normal equations $V^T V c = V^T y$

As $m \rightarrow \infty$, $\frac{1}{m} V^T V c = \frac{1}{m} V^T y$ looks like
it converges to a system involving integrals.

$$\left. \begin{aligned} \frac{1}{m} (V^T V)_{ij} &\rightarrow \frac{1}{n} \sum_{k=1}^n \varphi_i(x_k) \varphi_j(x_k) \\ \frac{1}{m} (V^T y)_i &\rightarrow \dots \end{aligned} \right\} \begin{aligned} &\int_a^b \varphi_i(x) \varphi_j(x) \\ &\int_a^b f(x) \varphi_i(x) \end{aligned} \quad \begin{aligned} &\text{theory} \\ &\text{for this} \\ &\text{system?} \end{aligned}$$

\Rightarrow Minimizing polynomial with respect to the continuous L^2 -norm

Ex: Given $f(x)$, find the linear polynomial $p(x)$ which minimizes

$$\frac{1}{2} \left\| p(x) - f(x) \right\|_{L^2([a,b])}^2 = \frac{1}{2} \int_a^b (p - f)^2$$

$p_1(x) = c_0 + c_1 x$, so we just need to determine $c_0 + c_1$.

$$J(c_0, c_1) = \frac{1}{2} \int_a^b (c_0 + c_1 x - f(x))^2$$



Find c_0, c_1 st.

$$\frac{\partial J}{\partial c_0} = 0, \quad \frac{\partial J}{\partial c_1} = 0$$

$$\textcircled{1} \quad \frac{\partial J}{\partial c_0} = \int_a^b (c_0 + c_1 x - f(x)) = 0$$

$$\textcircled{2} \quad \frac{\partial J}{\partial c_1} = \int_a^b (c_0 + c_1 x - f(x)) x = 0$$

$$\textcircled{1} + \textcircled{2} \quad c_0 \int_a^b 1 + c_1 \int_a^b x = \int_a^b f(x)$$

$$c_0 \int_a^b x + c_1 \int_a^b x^2 = \int_a^b f(x) x$$

$$\Rightarrow \begin{bmatrix} \int_a^b 1 & \int_a^b x \\ \int_a^b x & \int_a^b x^2 \end{bmatrix} \begin{bmatrix} c_0 \\ c_1 \end{bmatrix} = \begin{bmatrix} \int_a^b f(x) \\ \int_a^b f(x)x \end{bmatrix}$$

$$p(x) = \sum_{j=1}^n c_j \varphi_j(x)$$

$\varphi_1(x) = 1$
 $\varphi_2(x) = x$

$$\Rightarrow \begin{bmatrix} \int_a^b \varphi_0(x) \varphi_0(x) & \int_a^b \varphi_0(x) \varphi_1(x) \\ \int_a^b \varphi_1(x) \varphi_0(x) & \int_a^b \varphi_1(x) \varphi_1(x) \end{bmatrix} \begin{pmatrix} c_0 \\ c_1 \end{pmatrix} = \begin{pmatrix} \int_a^b f \varphi_0 \\ \int_a^b f \varphi_1 \end{pmatrix}$$

Ex: $p(x) = \sum_{j=0}^n c_j \varphi_j(x)$

$$\min_c \|f - p\|_{L^2([a,b])}^2 = \min_{c_0, \dots, c_n} \underbrace{\|f - p\|_{L^2([a,b])}^2}_{J(c_0, \dots, c_n)}$$

Find c_0, \dots, c_n st. $\frac{\partial J}{\partial c_i} = 0$

$$\Rightarrow \frac{\partial}{\partial c_i} \int_a^b \left(f(x) - \sum_{j=0}^n c_j \varphi_j(x) \right)^2$$

$$\Rightarrow \int_a^b \left(f(x) - \sum_{j=0}^n c_j \varphi_j(x) \right) \varphi_i(x) = 0$$

for $i = 0, \dots, n$.

$$\Rightarrow G = \begin{bmatrix} \int_a^b \varphi_0 \varphi_0 & \int_a^b \varphi_0 \varphi_1 & \cdots & \int_a^b \varphi_0 \varphi_n \\ \int_a^b \varphi_1 \varphi_0 & \ddots & & \\ \vdots & & \ddots & \\ \int_a^b \varphi_n \varphi_0 & \cdots & \int_a^b \varphi_n \varphi_n \end{bmatrix} \rightarrow \text{i}^{th} \text{ entry is } \int_a^b \varphi_i \varphi_j$$

$$Gc = b \quad \text{w/} \quad b = \begin{pmatrix} \int_a^b f(x) \varphi_0(x) \\ \vdots \\ \int_a^b f(x) \varphi_n(x) \end{pmatrix}, \quad b_i = \int_a^b f(x) \varphi_i(x)$$

$\Rightarrow L^2$ norm minimization is just one of a huge class of continuous least squares approximations

L^2 norm = one instance of an inner product norm.

Def: An inner product (IP) is a map from $V \times V \rightarrow \mathbb{R}$ st.

① Bilinearity: Let $\alpha, \beta \in \mathbb{R}$. Then, if $f, g, h \in V$

$$(\alpha f + \beta g, h) = \alpha(f, h) + \beta(g, h)$$

② Symmetry: $(f, g) = (g, f)$

③ Positivity: $(f, f) \geq 0$ \quad $(f, f) = 0$ if $f = 0$

Ex: L^2 inner product: $(f, g) = \int_a^b f(x)g(x)$

$$(f, f) = \int_a^b f^2 \geq 0$$

= inner product over (Lebesgue) integrable functions on $[a, b] \rightarrow L^2([a, b])$

Define $\|f\|_{L^2([a, b])} = \sqrt{(f, f)}$

Def: Inner prod. norm $\|f\| = \sqrt{(\bar{f}, f)}$

Recall: $\|\cdot\|: V \rightarrow \mathbb{R}$ is a norm if

- (1) $\|f\| \geq 0$ & $\|f\| = 0$ iff $f = 0$
- (2) $\alpha \in \mathbb{R}$, $\|\alpha f\| = |\alpha| \|f\|$
- (3) $\|f + g\| \leq \|f\| + \|g\|$

All IPs satisfy the Cauchy-Schwarz inequality
(CS)

$$\Rightarrow (f, g) \leq \|f\| \|g\|$$

Pf: let $\alpha \in \mathbb{R}$. $0 \leq \|f - \alpha g\|^2 = (f - \alpha g, f - \alpha g)$

$$= \|f\|^2 - 2\alpha (f, g) + \alpha^2 \|g\|^2$$

Specify $\alpha = \frac{(f, g)}{\|g\|^2}$

$$\Rightarrow 0 \leq \|f\|^2 - 2 \frac{(f, g)^2}{\|g\|^2} + \frac{(f, g)^2}{\|g\|^2}$$

$$0 \leq \|f\|^2 - \frac{(f, g)^2}{\|g\|^2}$$

$$(f, g)^2 \leq (\|f\| \|g\|)^2 \Rightarrow (f, g) \leq \|f\| \|g\|$$

Triangle Ineq: use CS

$$\begin{aligned}\Rightarrow \|f + g\|^2 &\leq \|f\|^2 + 2(f, g) + \|g\|^2 \\ &\leq \|f\|^2 + 2\|f\| \|g\| + \|g\|^2\end{aligned}$$

$$\Rightarrow \|f + g\| \leq \|f\| + \|g\|$$

Ex: L^2 inner prod on $[a, b]$

Ex: weighted L^2 inner products on $[a, b]$

$$(f, g) = \int_a^b f(x)g(x) \omega(x) \quad \text{w/ } \omega(x) > 0$$

Ex: Sobolev inner product

$$(f, g)_{H^1([a, b])} = (f, g)_{L^2([a, b])} + (f', g')_{L^2([a, b])}$$

Differences b/w Cont. Lsg. & Interpolation

- interp. still depends on point distribution
- cont. lsg. is indep. of points, just depends on IP & basis functions $\varphi_j(x)$
- optimal in IP norm.
- cont. lsg. more involved.

Similarities: monomials = bad.

$$G_{ij} = \int_a^b \varphi_j(x) \varphi_i(x) . \quad \text{if } \varphi_j(x) = x^j$$

$$\Rightarrow G_{ij} = \int_0^1 x^i x^j = \int_0^1 x^{i+j} = \frac{x^{i+j+1}}{i+j+1} \Big|_0^1$$

Hilbert matrix = $\frac{1}{i+j+1}$

\downarrow	$\begin{bmatrix} 1 & 2 & 3 & 4 & 5 \\ 2 & x_3 & y_4 & y_5 & \frac{y_6}{16} \\ y_3 & \frac{y_8}{8} & \ddots & y_7 & y_9 \\ \vdots & & & y_7 & y_9 \\ y_6 & & & y_9 & y_{11} \end{bmatrix}$	\rightarrow
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