

Best Approx. in Hilbert Space

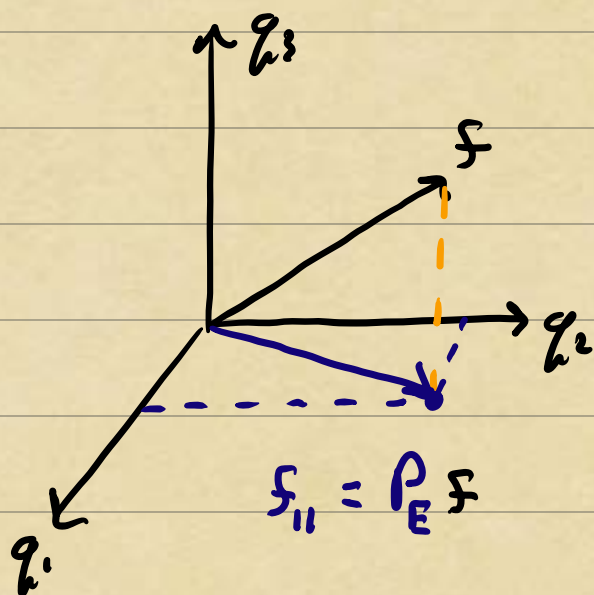
Goal: Minimize $\|f - E\epsilon\|$ in Hilbert norm.

$E: \mathbb{R}^N \rightarrow H$ (symmetric)

Two step procedure:

\Rightarrow Project $f_{||} = P_E f$

\Rightarrow Solve $E\epsilon = f_{||}$



Here, $P_E: H \rightarrow \text{col}(E)$ is the orthogonal projection of H onto $\text{col}(E)$, characterized by

- i) $f \rightarrow P_E f$ is linear
- ii) $P_E f = f$ when $f \in \text{col}(E)$
- iii) $P_E f = 0$ when $f \in \text{col}(E)^\perp$
- iv) $\|P_E f\| \leq \|f\|$ for all $f \in H$.

Objective $\|f - E\epsilon\|$ is minimized when $E\epsilon = P_E f$.

Today: Orthonormal Bases in Hilbert Space

Orthonormal Bases (Finite-Dimension)

An orthonormal basis for a subspace $V_N \subset H$ with dimension $N < \infty$ is a basis $\{q_1, \dots, q_N\} \subset V_N$ such that $\langle q_i, q_j \rangle = \begin{cases} 1 & i=j \\ 0 & i \neq j \end{cases}$.

Claim: The orthogonal projection onto V_N can be computed explicitly via

$$P_N f = \sum_{k=1}^N \langle f, q_k \rangle q_k.$$

pf For any $f \in H$, we have that

$$f = f_N + f_N^\perp, \text{ where } f_N \in V_N, f_N^\perp \in V_N^\perp,$$

and $P_N f = f_N$ by definition. Since $\{q_1, \dots, q_N\}$ is a basis for V_N , there are unique scalars $\alpha_1, \dots, \alpha_N$ s.t.

$$f_N = \alpha_1 q_1 + \dots + \alpha_N q_N.$$

Using the orthonormality of $\{q_1, \dots, q_N\}$ and the linearity of $\langle \cdot, \cdot \rangle$, we have

$$\begin{aligned}\langle f_N, q_j \rangle &= \alpha_1 \langle q_1, q_j \rangle + \dots + \alpha_j \langle q_j, q_j \rangle + \dots + \alpha_N \langle q_N, q_j \rangle \\ &= \alpha_j\end{aligned}$$

Therefore, $f_N = \sum_{k=1}^N \langle f, q_k \rangle q_k$ as claimed.

If q_1, \dots, q_N are orthogonal, but not normalized ($\|q_k\| \neq 1$), then normalize $\tilde{q}_k = \frac{q_k}{\|q_k\|}$.

$$P_N f = \sum_{k=1}^N \frac{\langle f, q_k \rangle}{\langle q_k, q_k \rangle} q_k.$$

Example: Fourier Series (Best Approx. in $L^2([-1, 1])$).

A continuous, periodic function f on $[-1, 1]$ has

$$f(x) = \frac{1}{\sqrt{2}} \sum_{k=-\infty}^{+\infty} \hat{f}_k e^{inkx}, \quad \hat{f}_k = \frac{1}{\sqrt{2}} \int_{-1}^{+1} f(x) e^{-inkx} dx.$$

Consider the best approximation of f in the

subspace $V_N = \{e^{-inNx}, \dots, e^{+inNx}\}$ of

$$L^2([-1,1]) = \left\{ f: [-1,1] \rightarrow \mathbb{R} \mid \int_{-1}^{+1} |f(x)|^2 dx < \infty \right\},$$

which is a Hilbert space w/inner product

$$\langle f, g \rangle = \int_{-1}^{+1} f(x) \overline{g(x)} dx.$$

Since $\{e^{ikx}\}_{k=-\infty}^{+\infty}$ are pairwise orthogonal,

$$f(x) = \underbrace{\frac{1}{\sqrt{2}} \sum_{k=-N}^{+N} \hat{f}_k e^{ikx}}_{f_N} + \underbrace{\frac{1}{\sqrt{2}} \sum_{|k| > N} \hat{f}_k e^{ikx}}_{f_N^\perp}$$

where $f_N \in V_N$ and $f_N^\perp \in V_N^\perp$. Therefore,
 $\langle f - f_N, g \rangle = 0$ for all $g \in V_N$ and f_N is
the best approximation of f in V_N . Moreover,
if P_N is the orthogonal projection onto V_N ,

$$P_N f = \frac{1}{\sqrt{2}} \sum_{k=-N}^N \hat{f}_k e^{ikx}, \quad \hat{f}_k = \frac{1}{\sqrt{2}} \int_{-1}^{+1} f(x) e^{-ikx} dx.$$

Fourier modes $\{e^{-inNx}, \dots, e^{+inNx}\}$ are an ONB for V_N .

Orthonormal Bases (Infinite Dimension)

Given a Hilbert Space H , an **orthonormal set** $\{q_n\}_{n=1}^{\infty}$ satisfies $\langle q_i, q_j \rangle = \begin{cases} 0 & i \neq j \\ 1 & i = j \end{cases}$.

A set $S \subset H$ is called **dense** in H if for any $f \in H$, for any $\varepsilon > 0$, there is a $\phi \in S$ s.t. $\|f - \phi\| < \varepsilon$. Equivalently, there is a sequence $\{\phi_n\}_{n=1}^{\infty}$ such that

$$\lim_{n \rightarrow \infty} \|f - \phi_n\| = 0.$$

An orthonormal set $\{q_n\}_{n=1}^{\infty} \subset H$ is an **orthonormal basis** for H if $\text{span}\{q_n\}_{n=1}^{\infty}$ is dense in H . In other words, each $f \in H$ is a limit of finite linear combos of $\{q_n\}$,

$$\lim_{N \rightarrow \infty} \|f - \sum_{k=1}^N c_k q_k\| = 0.$$

We write $f = \sum_{k=1}^{\infty} c_k q_k$ and the coefficients are again given by $c_k = \langle f, q_k \rangle$.

For any orthonormal set $\{q_k\}_{k=1}^{\infty}$,

Bessel's Inequality $\|f\|^2 \leq \sum_{k=1}^{\infty} |\langle f, q_k \rangle|^2.$

If $\{q_k\}_{k=1}^{\infty}$ is an ONB, then

Parseval's Identity $\|f\|^2 = \sum_{k=1}^{\infty} |\langle f, q_k \rangle|^2.$

Example: Every $f \in L^2([-1, 1])$ has

$$f = \frac{1}{\sqrt{2}} \sum_{k=-\infty}^{+\infty} c_k e^{inkx}, \text{ where } c_k = \frac{1}{\sqrt{2}} \int_{-\infty}^{+\infty} f(x) e^{-inkx} dx.$$

Caution: The series converges in the L^2 norm:

$$\lim_{N \rightarrow \infty} \int \left| f(x) - \underbrace{\frac{1}{\sqrt{2}} \sum_{k=-\infty}^{+\infty} c_k e^{inkx}}_{S_N} \right|^2 dx = 0.$$

This is sometimes called convergence "in mean", but it does not imply that $S_N(x) \rightarrow f(x)$ pointwise (for each $x \in [-1, 1]$). Convergence may fail at certain points unless f is smooth.

Example: What is the orthogonal projection of $f \in L^2([-1,1])$ onto the subspace of even functions $\{g \in L^2 / g(x) = g(-x)\}$?

$$f(x) = \underbrace{\frac{a_0}{2} + \sum_{k=1}^{\infty} a_k \cos(k\pi x)}_{P_N f} + \sum_{k=1}^{\infty} b_k \sin(k\pi x)$$

Quasimatrix Notation

We can write orthogonal projectors in a compact matrix form using quasimatrices.

$$P_N f = \sum_{k=1}^N \langle f, q_k \rangle q_k = Q Q^* f$$

Here, Q and $Q^* f$ are defined as

$$Q = \begin{bmatrix} 1 & & 1 \\ q_1 & \dots & q_N \\ 1 & & 1 \end{bmatrix}, \quad Q^* f = \begin{bmatrix} \langle f, q_1 \rangle \\ \vdots \\ \langle f, q_N \rangle \end{bmatrix}$$

Quasimatrix

Vector