

Lecture 15: G-S process and Orthogonal complement

G-S process : produce orthonormal basis

Goal: Let $V = P_n(\mathbb{R})$. Can we get orthonormal basis for $P_n(\mathbb{R})$ with respect to some (weighted) inner product?

Important applications in physics!

e.g. Legendre polynomial, Chebyshev polynomial, Laguerre ...

Definition 1: Let $V = P_n(\mathbb{R})$. Let $w: \mathbb{R} \rightarrow \mathbb{R}$ such that $w(x) > 0$.

Then the weighted inner product is defined as :

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

Using G-S process, we can obtain orthonormal basis with respect to different weight. e.g. $w(x) = 1 \leftrightarrow$ Legendre; $w(x) = \frac{1}{\sqrt{1-x^2}} \leftrightarrow$ Chebyshev

Example 1: (Legendre polynomials) Let $V = P_n(\mathbb{R})$. Define:

$$\langle f(x), g(x) \rangle = \int_{-1}^1 f(t) g(t) dt \quad (w(x) = 1)$$

Clearly, $\beta = \{1, x, x^2, \dots, x^n\}$ = basis for V . We can obtain orthogonal polynomials from β using G-S process.

G-S process: $p_0(x) = 1,$

$$p_1(x) = x - \frac{\langle x, p_0 \rangle}{\langle p_0, p_0 \rangle} p_0(x)$$

$$p_2(x) = x^2 - \frac{\langle x^2, p_0 \rangle}{\langle p_0, p_0 \rangle} p_0(x) - \frac{\langle x^2, p_1 \rangle}{\langle p_1, p_1 \rangle} p_1(x)$$

$$p_n(x) = x^n - \frac{\langle x^n, p_0 \rangle}{\langle p_0, p_0 \rangle} p_0(x) - \dots - \frac{\langle x^n, p_{n-1} \rangle}{\langle p_{n-1}, p_{n-1} \rangle} p_{n-1}(x)$$

} Orthogonal
polynomials

Note that : $\langle x^n, x^m \rangle = 0$ if $n+m$ is odd

$$\therefore p_0(x) = 1, p_1(x) = x - 0 = x; p_2(x) = x^2 - \frac{\langle x^2, 1 \rangle}{\langle 1, 1 \rangle} 1 - 0 = x^2 - \frac{1}{3}$$

$$p_3(x) = x^3 - 0 - \frac{\langle x^3, x \rangle}{\langle x, x \rangle} x - 0 = x^3 - \frac{3}{5}x$$

$$p_4(x) = \frac{1}{8}(35x^4 - 30x^2 + 3)$$

$\{p_0, p_1, \dots, p_n\}$ are called Legendre's polynomials.

It becomes orthonormal polynomial by normalization:

$$\left\{ \frac{p_0}{\|p_0\|}, \dots, \frac{p_n}{\|p_n\|} \right\} \text{ (exercise)}$$

Example 2: (Chebyshev polynomial) Let $V = P_n(\mathbb{R})$. Define:

$$\langle f, g \rangle = \int_{-1}^1 \frac{1}{\sqrt{1-t^2}} f(t) g(t) dt. \quad (w(x) = \frac{1}{\sqrt{1-x^2}})$$

Again, we G-S process to obtain orthogonal basis from $\{1, x, \dots, x^n\}$.

$$p_0(x) = 1, p_1(x) = x - \frac{\langle x, p_0 \rangle}{\langle p_0, p_0 \rangle} p_0(x) = x - \frac{\int_{-1}^1 \frac{1}{\sqrt{1-t^2}} t dt}{\int_{-1}^1 \frac{1}{\sqrt{1-t^2}} dt} 1 = x$$

$$p_2(x) = x^2 - \frac{\langle x^2, p_0 \rangle}{\langle p_0, p_0 \rangle} p_0(x) - \frac{\langle x^2, p_1 \rangle}{\langle p_1, p_1 \rangle} p_1(x) = x^2 - \frac{1}{2}$$

$$p_n(x) = \dots$$

In fact, we can prove that: $p_{n+1}(x) = 2x p_n(x) - p_{n-1}(x)$.

Remark: Same procedure can be applied to different weight function to get different orthogonal polynomials.

e.g. $w(x) = 1$ (Legendre); $w(x) = \frac{1}{\sqrt{1-x^2}}$ (Chebyshev, 1st kind)

$w(x) = \sqrt{1-x^2}$ (Chebyshev, 2nd kind); $w(x) = e^{-x}$ (Laguerre)

$w(x) = (1+x)^\alpha (1-x)^\beta$ (Jacobi) etc

Definition 2: Let β = orthonormal subset (can be infinite) of V = inner product space. Let $x \in V$. Define the Fourier coefficients of x relative to β to be scalars $\langle x, y \rangle$ where $y \in \beta$.

Remark: More specifically, Fourier studied:

$$a_n = \int_0^{2\pi} f(t) \sin nt dt; \quad b_n = \int_0^{2\pi} f(t) \cos nt dt \quad \text{or more generally,}$$

$$c_n = \frac{1}{2\pi} \int_0^{2\pi} f(t) e^{-int} dt$$

Recall: $S = \{e^{int}\}_{n \in \mathbb{Z}}$ is an orthonormal basis w.r.t.

$\langle f, g \rangle = \frac{1}{2\pi} \int_0^{2\pi} f \bar{g} dt$. $\therefore c_n$ is the n -th Fourier coefficient of f relative to S ,

$$\text{since } \langle f, e^{int} \rangle = \frac{1}{2\pi} \int_0^{2\pi} f(t) e^{-int} dt$$

Example 3: Let $S = \{e^{int} : n \in \mathbb{Z}\}$ = orthonormal set in $V = C([0, 2\pi])$

Compute the n -th Fourier coefficient of t relative to S :

$$c_n = \frac{1}{2\pi} \int_0^{2\pi} t e^{-int} dt = -\frac{1}{in} \quad (\text{by integration by part}) \quad (n \neq 0)$$

$$\text{For } n=0, \quad c_0 = \frac{1}{2\pi} \int_0^{2\pi} t dt = \pi$$

We can prove that: $\|x\|^2 \geq \sum_{i=1}^n |\langle x, v_i \rangle|^2$ where = $\{v_1, v_2, \dots, v_n\}$ = orthonormal subset.

$$\begin{aligned} \therefore \|f\|^2 &\geq \sum_{n=-R}^{-1} |\langle f, f_n \rangle|^2 + |\langle f, 1 \rangle|^2 + \sum_{n=1}^R |\langle f, f_n \rangle|^2 \\ &= \sum_{n=-R}^{-1} \frac{1}{n^2} + \pi^2 + \sum_{n=1}^R \frac{1}{n^2} = 2 \sum_{n=1}^R \frac{1}{n^2} + \pi^2 \text{ for } \forall k. \end{aligned}$$

Now, $\|f\|^2 = \frac{1}{2\pi} \int_0^{2\pi} t^2 dt = \frac{4}{3}\pi^2$, we get:

$$\frac{4}{3}\pi^2 \geq 2 \sum_{n=1}^k \frac{1}{n^2} + \pi^2 \Rightarrow \sum_{n=1}^k \frac{1}{n^2} \leq \frac{\pi^2}{6} \Rightarrow \sum_{n=1}^{\infty} \frac{1}{n^2} \leq \frac{\pi^2}{6}.$$

(for $\forall k$)

Orthogonal complement

Definition 3: Let S = non-empty subset in V (inner product space). Define S^\perp (S "perp") = set of all $\vec{v} \in V$ that are orthogonal to $\forall \vec{w} \in S$. (i.e. $S^\perp = \{\vec{x} \in V : \langle \vec{x}, \vec{y} \rangle = 0 \text{ for } \forall \vec{y} \in S\}$) S^\perp is called orthogonal complement of S .

Remark: (Trivial examples)

- $\{\vec{0}\}^\perp = V$
- $V^\perp = \{\vec{0}\}$
- $V = \mathbb{R}^3$, $\{e_1\}^\perp = y-z \text{ plane}$
 $= \{(0, y, z) : y, z \in \mathbb{R}\}$

Application: Shortest distance from P to a plane



Theorem 1: Let W = finite-dim subspace of inner product space V . Let $\vec{y} \in V$. Then there exists an unique $\vec{u} \in W$ and an unique $\vec{z} \in W^\perp$ such that $\vec{y} = \vec{z} + \vec{u}$.

Also, if $\{\vec{v}_1, \vec{v}_2, \dots, \vec{v}_k\}$ is an orthonormal basis for W , then

$$\vec{u} = \sum_{i=1}^k \langle \vec{y}, \vec{v}_i \rangle \vec{v}_i \quad (\text{Orthogonal projection of } \vec{y} \text{ on } W)$$

Proof: Let $\{\vec{v}_1, \dots, \vec{v}_k\}$ = orthonormal basis for W .

Let $\vec{u} = \sum_{i=1}^k \langle \vec{y}, \vec{v}_i \rangle \vec{v}_i$.

We need to show $\vec{z} = \vec{y} - \vec{u} \in W^\perp$. Only need to show $\langle \vec{z}, \vec{v}_j \rangle = 0$

Now, $\langle \vec{z}, \vec{v}_j \rangle = \langle (\vec{y} - \sum_{i=1}^k \langle \vec{y}, \vec{v}_i \rangle \vec{v}_i), \vec{v}_j \rangle = \langle \vec{y}, \vec{v}_j \rangle - \sum_{i=1}^k \langle \vec{y}, \vec{v}_i \rangle \langle \vec{v}_i, \vec{v}_j \rangle$
 $= \langle \vec{y}, \vec{v}_j \rangle - \langle \vec{y}, \vec{v}_j \rangle \langle \vec{v}_j, \vec{v}_j \rangle$

For uniqueness of \vec{u} and \vec{z} , suppose $\vec{y} = \vec{u} + \vec{z} = \vec{u}' + \vec{z}'$ $\in W \cup W^\perp$

Then: $\vec{u} - \vec{u}' = \vec{z}' - \vec{z} \in W \cap W^\perp = \{\vec{0}\}$ (because if $\vec{w} \in W$ and $\vec{w} \in W^\perp$, then $\langle \vec{w}, \vec{w} \rangle = 0 \Rightarrow \vec{w} = \vec{0}$)
 $\therefore \vec{u} = \vec{u}'$ and $\vec{z} = \vec{z}'$.

Corollary 1: " \vec{u} is the unique vector in W closest to $\vec{y} \in V$ "

That is, for $\forall \vec{x} \in W$, $\|\vec{y} - \vec{x}\| \geq \|\vec{y} - \vec{u}\|$.

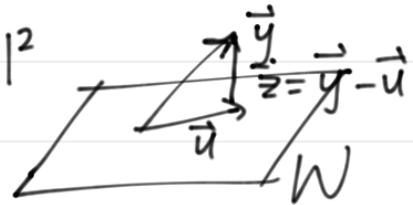
Also, equality holds $\Leftrightarrow \vec{x} = \vec{u}$.

Proof: Let $\vec{x} \in W$. Then $\vec{u} - \vec{x} \in W$ is orthogonal to $\vec{z} \in W^\perp$.

Note that $\|\vec{x} + \vec{y}\|^2 = \|\vec{x}\|^2 + \|\vec{y}\|^2$; if \vec{x}, \vec{y} are orthogonal
So, $\|\vec{y} - \vec{x}\|^2 = \|\vec{u} + \vec{z} - \vec{x}\|^2 = \|\vec{u} - \vec{x}\| + \|\vec{z}\|^2 = \|\vec{u} - \vec{x}\| + \|\vec{z}\|^2$ (exercise)
 $= \|\vec{u} - \vec{x}\|^2 + \|\vec{z}\|^2 \geq \|\vec{z}\|^2 = \|\vec{y} - \vec{u}\|^2$

If $\|\vec{y} - \vec{x}\| = \|\vec{y} - \vec{u}\|$; then: $\|\vec{u} - \vec{x}\|^2 + \|\vec{z}\|^2 = \|\vec{z}\|^2$

$\Rightarrow \|\vec{u} - \vec{x}\|^2 = 0 \Rightarrow \vec{x} = \vec{u}$.



Conversely, if $\vec{x} = \vec{u}$, then equality holds obviously.

Example 4: Let $V = \mathbb{R}^4$. Let $\beta = \left\{ \frac{1}{\sqrt{2}}(1, 1, 0, 0), \frac{1}{\sqrt{3}}(1, -1, 1, 0), \frac{1}{\sqrt{6}}(-1, 1, 2, 0) \right\}$

Let $W = \text{span}(\beta)$. β = orthonormal basis of W .

The orthogonal projection of $(2, 3, 4, 5)$ to W is:

$$\begin{aligned}\vec{w}_0 &= \langle \vec{w}, \vec{v}_1 \rangle \vec{v}_1 + \langle \vec{w}, \vec{v}_2 \rangle \vec{v}_2 + \langle \vec{w}, \vec{v}_3 \rangle \vec{v}_3 \\ &= \frac{5}{\sqrt{2}} \vec{v}_1 + \frac{3}{\sqrt{3}} \vec{v}_2 + \frac{9}{\sqrt{6}} \vec{v}_3.\end{aligned}$$