

Cues
Lecture 2.
 Orthogonal polynomials:
 Spectral theorem, and
 Christoffel-Darboux
 kernel

Notes
 1. Three-term recurrence and Jacobi operator
 In this part, we show that orthogonality implies
 a 3-term recurrence relation.

• This proposition means:
 orthogonality \Rightarrow 3-term
 recurrence!

Prop. 2.1 (3-term recurrence): The orthogonal
 polynomials $\{P_n(x)\}_{n \geq 0}$ satisfy a 3-term recurrence
 relation $xP_n(x) = a_{n+1}P_{n+1}(x) + b_nP_n(x) + a_nP_{n-1}(x)$
 where $a_0 = 0$ and

$$a_n = \int_{\mathbb{R}} x P_n(x) P_{n-1}(x) d\mu(x), \quad b_n = \int_{\mathbb{R}} x P_n^2(x) d\mu(x).$$

E.x. 1: Verify that $P_n(x) = P_{n+1}(x)$
 $+ b_n P_n(x) + a_n^2 P_{n-1}(x)$ by using
 orthogonality.

Proof: Since $\{P_n(x)\}_{n \geq 0}$ is a basis of $L^2(\mu)$, we have
 $xP_n(x) = \sum_{k=0}^n C_{n,k} P_k(x)$, where, by orthogonality,
 $C_{n,k} = \int_{\mathbb{R}} x P_n(x) P_k(x) d\mu(x)$. Moreover, from
 $\deg(xP_k) = k+1$, we have $C_{n,k} = 0$ if $k+1 < n$, so that
 only three terms remain in the recurrence. Obviously,

E.x. 2: Show that $b_n = \tilde{D}_{n+1} / D_n$
 where

$$\tilde{D}_{n+1} = \det \begin{pmatrix} m_0 & m_1 & \dots & m_n \\ m_1 & m_2 & \dots & m_{n+1} \\ \vdots & \vdots & & \vdots \\ m_{n+1} & m_n & \dots & m_{2n+1} \\ m_{n+1} & m_{n+2} & \dots & m_{2n+1} \end{pmatrix}.$$

$$C_{n,n-1} = \int_{\mathbb{R}} x P_n(x) P_{n-1}(x) d\mu(x), \quad C_{n,n} = \int_{\mathbb{R}} x P_n^2(x) d\mu(x),$$

$$C_{n,n+1} = \int_{\mathbb{R}} x P_n(x) P_{n+1}(x) d\mu(x),$$

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which depends on a single index n .

Remark 2.2: We have an explicit determinant formula for a_n . By comparing
 the coefficient of the highest order, we have $\gamma_n = a_{n+1} \gamma_{n+1}$, and thus $a_{n+1} = \gamma_n \gamma_{n+1}^{-1}$
 $= (D_{n+2} D_n / D_{n+1}^2)^{1/2}$, where $D_n = \det (m_{i+j})_{i,j=0}^{n-1}$ is the Hankel determinant.

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- Another expression for 3-term recurrence:

$$\begin{pmatrix} P_{n+1}(x) \\ P_n(x) \end{pmatrix} = L_n \begin{pmatrix} P_n(x) \\ P_{n+1}(x) \end{pmatrix},$$

where

$$L_n = \begin{pmatrix} a_{n+1}^+(x-b_n) & a_{n+1}^+ a_n \\ 1 & 0 \end{pmatrix}.$$

This formula is useful in the construction of Lax pair in classic integrable systems.

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Alternative expression for 3-term recurrence.

$$x \begin{pmatrix} P_0(x) \\ P_1(x) \\ P_2(x) \\ \vdots \end{pmatrix} = \begin{pmatrix} b_0 & a_1 & 0 & \dots \\ a_1 & b_1 & a_2 & \dots \\ 0 & a_2 & b_2 & \dots \\ \vdots & \vdots & \vdots & \ddots \end{pmatrix} \begin{pmatrix} P_0(x) \\ P_1(x) \\ P_2(x) \\ \vdots \end{pmatrix} := J \begin{pmatrix} P_0(x) \\ P_1(x) \\ P_2(x) \\ \vdots \end{pmatrix}$$

where J is called as an infinite tridiagonal Jacobi matrix. Denote the upper left corner $(n \times n)$ of J by J_n , then we have

$$J_n \begin{pmatrix} P_0(x) \\ P_1(x) \\ \vdots \\ P_{n-2}(x) \\ P_{n-1}(x) \end{pmatrix} = x \begin{pmatrix} P_0(x) \\ P_1(x) \\ \vdots \\ P_{n-2}(x) \\ P_{n-1}(x) \end{pmatrix} - \begin{pmatrix} 0 \\ 0 \\ \vdots \\ 0 \\ a_n P_n(x) \end{pmatrix}.$$

Recall that $P_n(x)$ has n simple, real roots. Therefore, every zero $x_{n,j}$ of $P_n(x)$ is an eigenvalue of J_n , with corresponding eigenvector $(P_0(x_{n,j}), \dots, P_{n-1}(x_{n,j}))^T$.

Prop. 2.2 (Relation between OPs and truncated Jacobi matrix): We have $P_n(x) = \det(xI_n - J_n)$.

proof: Firstly, $\det(xI_n - J_n)$ is a polynomial of degree n , with zeros $x_{n,j}, j=1, \dots, n$. Since $P_n(x) = \prod_{j=1}^n (x - x_{n,j})$, it is known that $P_n(x)$ has the same zeros with $P_n(x)$, i.e. $x_{n,j}, j=1, \dots, n$. Moreover, $P_n(x)$ and $\det(xI_n - J_n)$ are monic polynomials. Therefore, they are identical. #

Summary

In the first part, we have shown that orthogonality means 3-term recurrence relation. An important question is:

Whether could we get orthogonality from 3-term recurrence relation?

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Favard's theorem tells:
3-term recurrence \implies
orthogonality!

Let
Recall: If $A = A^T$, and
 λ_1, λ_2 are two distinct
eigenvalues of A . If v_1, v_2
are corresponding eigen-
vectors, then $v_1^T v_2 = 0$.

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2. Spectral theorem

Thm. 2.3 (Favard's theorem):

Suppose that $\{P_n\}_{n=0,1,2,\dots}$ is a sequence of polynomials that satisfy a 3-term recurrence relation

$$x P_n = a_{n+1} P_{n+1} + b_n P_n + a_n P_{n-1}, \quad n \geq 0$$

with $a_{n+1} > 0$ and $b_n \in \mathbb{R}$ for every $n \geq 0$, and initial conditions $P_{-1}(x) = 0$ and $P_0(x) = 1$. Then there is a probability measure μ on the real line, s.t.

$$\int_{\mathbb{R}} P_n(x) P_m(x) d\mu(x) = \delta_{nm}.$$

proof:

(Step 1). We first show that for polynomials whose degrees are less than or equal to n , we can construct a discrete measure μ_n , s.t. $\int_{\mathbb{R}} P_i(x) P_j(x) d\mu_n(x) = \delta_{ij}$, $0 \leq i, j \leq n-1$.

Consider the truncated Jacobi matrix J_n . It is known that zeros of $P_n(x)$ (i.e. $\{x_{n,j}\}_{j=1,\dots,n}$) coincide with eigenvalues of J_n , with eigenvector $v_j = (P_0(x_{n,j}), \dots, P_{n-1}(x_{n,j}))^T$.

Therefore, we have $J_n(v_1, \dots, v_n) = (v_1, \dots, v_n)D$, where $D = \text{diag}(x_{n,1}, \dots, x_{n,n})$. Since J_n is symmetric, we know that $\{v_j\}_{j=1,\dots,n}$ form an orthogonal system. By normalization $\tilde{v}_j = v_j / \|v_j\|_2$, we can construct $\tilde{V} = (\tilde{v}_1, \dots, \tilde{v}_n)$, which

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is an orthogonal matrix, s.t. $\tilde{V}^T \tilde{V} = I$.

Moreover, we have

$$\sum_{k=1}^n \frac{1}{\|v_k\|_2^2} P_i(x_{n,k}) P_j(x_{n,k}) = \delta_{ij}.$$

If we define a discrete measure

$$\mu_n = \sum_{k=1}^n \frac{1}{\|v_k\|_2^2} \delta_{x_{n,k}},$$

then this is a probability measure supported on the zeros of $P_n(x)$, for which $\int_{\mathbb{R}} P_i(x) P_j(x) d\mu_n(x) = \delta_{ij}$.

The first part of the proof is complete.

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(step 2) We show that $\{\mu_n\}$ has a subsequence that converges weakly to a probability measure μ , s.t.

$$\lim_{k \rightarrow \infty} \int_{\mathbb{R}} p_i(x) p_j(x) d\mu_{n_k}(x) = \int_{\mathbb{R}} p_i(x) p_j(x) d\mu(x).$$

① Show the sequence of probability measure $\{\mu_n\}_{n=0,1,2,\dots}$ has a subsequence that converges weakly to a measure μ (by Helly's selection principle);

② Show that μ is also a probability measure.

(Equivalent to show the sequence $\{\mu_n\}_{n=0,1,\dots}$ is tight. i.e. $\forall \varepsilon > 0, \exists$ a compact set $K \subseteq \mathbb{R}$, s.t. $\mu_n(K) \geq 1 - \varepsilon$ for every $n=0,1,2,\dots$) #

This part is left to be reading materials. \Leftarrow

Christoffel-Darboux kernel could be regarded as an application of 3-term recurrence relation, but it plays an important role in random matrix theory.

3. Christoffel-Darboux kernel (C-D kernel)

Def. 2.4 (C-D kernel):

The Christoffel-Darboux kernel is defined by

$$\hat{K}_n(x, y) = \sum_{j=0}^{n-1} p_j(x) p_j(y), \quad (*)$$

where $\{p_j(x)\}_{j=0,1,2,\dots}$ is the set of orthonormal polynomials.

Prop. 2.5 (Alternative form for C-D kernel):

The kernel $(*)$ satisfies the C-D identity

$$\hat{K}_n(x, y) = \begin{cases} a_n \frac{1}{x-y} (p_n(x) p_{n+1}(y) - p_{n-1}(x) p_n(y)), & x \neq y \\ a_n (p_n'(x) p_{n-1}(x) - p_n(x) p_{n-1}'(x)), & x = y \end{cases}$$

where a_n is the coefficient of 3-term recurrence relation.

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E.x.: Regarding with monic orthogonal polynomials $\{P_n(x)\}$,

we have the C-D kernel

$$\widehat{K}_n(x, y) = \sum_{j=0}^{n-1} \frac{D_j}{D_{j+1}} P_j(x) P_j(y),$$

$$\text{where } \int_{\mathbb{R}} P_j^2(x) d\mu(x) = \frac{D_{j+1}}{D_j}.$$

Show that

$$(x-y) \widehat{K}_n(x, y) = \frac{D_{n-1}}{D_n} x$$

$$(P_n(x) P_{n-1}(y) - P_{n-1}(x) P_n(y)).$$

The proof of Prop. 2.6 is left to be an exercise.

In fact, if $q(x) \in \text{span}\{P_0(x), \dots,$

$P_{n-1}(x)\}$, then

$$\int_{\mathbb{R}} \widehat{K}_n(x, y) q(y) d\mu(y) = q(x).$$

Proof: $x \widehat{K}_n(x, y) = \sum_{j=0}^{n-1} x P_j(x) P_j(y)$
 $= \sum_{j=0}^{n-1} (a_{j+1} P_{j+1}(x) + b_j P_j(x) + a_j P_{j-1}(x)) P_j(y).$

To simplify this formula, we know

$$x \widehat{K}_n(x, y) = a_n P_n(x) P_{n-1}(y) + \sum_{j=0}^{n-1} a_j (P_j(x) P_{j-1}(y) + P_j(y) P_{j+1}(x))$$

$$+ \sum_{j=0}^{n-1} b_j P_j(x) P_j(y),$$

where the last two terms are symmetric in x & y . Noting that $y \widehat{K}_n(x, y) = x \widehat{K}_n(x, y) |_{x \leftrightarrow y}$, we know that $(x-y) \widehat{K}_n(x, y) = a_n (P_n(x) P_{n-1}(y) - P_n(y) P_{n-1}(x))$, and the result follows by the above formula. #

prop. 2.6 (properties of C-D kernel):

The C-D kernel $\widehat{K}_n(x, y)$ has the following properties:

$$\textcircled{1} \int_{\mathbb{R}} \widehat{K}_n(x, x) d\mu(x) = n;$$

$$\textcircled{2} \int_{\mathbb{R}} \widehat{K}_n(x, y) \widehat{K}_n(y, z) d\mu(y) = \widehat{K}_n(x, z); (**)$$

$$\textcircled{3} \int_{\mathbb{R}} \widehat{K}_n(x, y) P_k(y) d\mu(y) = P_k(x), \quad 0 \leq k \leq n-1.$$

We call $\widehat{K}_n(x, y)$ as reproducing kernel in $L^2(d\mu)$ due to the equation (**).

Summary

In general, $\widehat{K}_n(x, y)$ can induce an integral operator $\widehat{K}_n: L^2(d\mu) \rightarrow L^2(d\mu)$ s.t.

$$\widehat{K}_n f(x) = \int_{\mathbb{R}} \widehat{K}_n(x, y) f(y) d\mu(y).$$

Remark 2.7: If $d\mu(x) = w(x) dx$, then

$$K_n(x, y) = \sqrt{w(x)} \sqrt{w(y)} \widehat{K}_n(x, y) = \sqrt{w(x)} \sqrt{w(y)} \sum_{j=0}^{n-1} P_j(x) P_j(y)$$

which has the reproducing property in $L^2(\mathbb{R}, d\mu)$

$$\text{s.t. } \int_{-\infty}^{+\infty} K_n(x, y) K_n(y, z) dy = K_n(x, z).$$