

# **Eigenvalues and Biorthogonal Eigensystems of Scaling Operators**

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# Organization of Talk

1. Introduction
2. Appell Systems of Random Variables
3. Biorthogonal Systems of Eigenfunctions
4. Gaussian,  $B$ -splines, Hermite and Bernoulli Polynomials

# 1 Introduction

## 1.1 Appell Sequences

- Take a compactly supported distribution  $\phi$ , normalized so that

$$\widehat{\phi}(0) = 1.$$

- Generate an Appell sequence

$$\frac{e^{xz}}{\widehat{\phi}(iz)} = \sum_{m=0}^{\infty} Q_m(x) z^m. \quad (1)$$

- Action of  $\phi^{(k)}$  on (1)

$$\frac{\langle \phi^{(k)}(x), e^{xz} \rangle}{\widehat{\phi}(iz)} = \sum_{m=0}^{\infty} \langle \phi^{(k)}, Q_m \rangle z^m$$

$\Downarrow$

$$(-1)^k z^k = \sum_{m=0}^{\infty} \langle \phi^{(k)}, Q_m \rangle z^m$$

$\Downarrow$

$$\langle (-1)^k \phi^{(k)}, Q_m \rangle = \delta_{k,m}.$$

## 1.2 Scaling Functions

- A scaling function satisfies a *scaling equation*,

$$\phi(x) = 2 \sum_j h(j) \phi(2x - j), \quad x \in \mathbb{R}, \quad (2)$$

where  $h \in \ell_0(\mathbb{Z})$  with sum 2.

- In the Fourier domain

$$\widehat{\phi}(u) = \widehat{h}(u/2) \widehat{\phi}(u/2), \quad u \in \mathbb{R}, \quad (3)$$

where  $\widehat{h}(u) = \sum_j h(j) e^{-iju}$ .

- **Fact 1.** (2) has a unique solution  $\phi$ , which is a compactly supported distribution whose FT is

$$\widehat{\phi}(u) = \prod_{j=1}^{\infty} \widehat{h}(u/2^j), \quad u \in \mathbb{R}$$

- **Fact 2.** The cascade sequence  $\prod_{j=1}^n \widehat{h}(u/2^j)$  converges locally uniformly to  $\widehat{\phi}(u)$ .
- **Fact 3.**

$$\phi^{(k)}(x) = 2^{k+1} \sum_j h(j) \phi^{(k)}(2x - j), \quad x \in \mathbb{R},$$

showing that  $\phi^{(k)}$  is an eigenfunction of the *scaling operator*

$$T_h f = 2 \sum_j h(j) f(2 \cdot -j), \quad f \in \ell(\mathbb{Z}),$$

with eigenvalues  $2^{-k}$ ,  $k = 1, 2, \dots$ .

### 1.3 Appell Sequences of Scaling Functions

Take a scaling function that satisfies (2) and generate an Appell sequence by (1).

- **Example 1.** Take  $\phi = \chi_{(0,1]}$ . Then  $\widehat{\phi}(u) = \left(\frac{1-e^{iu}}{iu}\right)$ , and (1) becomes

$$\frac{ze^{xz}}{e^z - 1} = \sum_{m=0}^{\infty} B_m(x)z^m,$$

which generates the classical Bernoulli polynomials.

- **Example 2.** Take  $\phi = \chi_{(0,1]} * \cdots * \chi_{(0,1]}$ ,  $N$  times. Then  $\phi$  is the  $B$ -spline of order  $N$ ,  $\widehat{\phi}(u) = \left(\frac{1-e^{iu}}{iu}\right)^N$ , and (1) becomes

$$\frac{z^N e^{xz}}{(e^z - 1)^N} = \sum_{m=0}^{\infty} B_{N,m}(x)z^m,$$

which generates the classical Bernoulli polynomials,  $B_{N,m}(x)$ , of order  $N$ .

## 1.4 Aim of Talk

- To explore some connections between scaling functions and the corresponding Appell sequences.
- Formulate the problem using simple probability.

## 2 Appell Systems of Random Variables

### 2.1 Stochastic Multiscale Operators

Consider the conditional expectation,

$$E\{f(Y + Z)|Y\}$$

of functions of two independent random variables  $Y$  and  $Z$ .

#### Examples

$$E\{\alpha f(\alpha X - T)|X\} = \alpha \int_{\mathbb{R}} f(\alpha X - t) d\psi_T(t) =: W_{T,\alpha} f(X) \quad (4)$$

is a *scaling operator*, and

$$E\left\{f\left(\frac{X + T}{\alpha}\right) | X\right\} = \int_{\mathbb{R}} f\left(\frac{X + t}{\alpha}\right) d\psi_T(t) =: W_{T,\alpha}^* f(X) \quad (5)$$

is its adjoint, where  $\alpha > 1$  is the dilation.

## 2.2 Appell Systems

- Let  $Y_n$ ,  $n = 1, 2, 3, \dots$ , be i.i.d. random variables with c.f.  $\widehat{\psi}$ , analytic at 0,

$$S_n := \frac{Y_1}{\alpha} + \frac{Y_2}{\alpha^2} + \dots + \frac{Y_n}{\alpha^n},$$

where  $\alpha > 1$ .

- The c.f. of  $S_n$  (*cascade sequence*),

$$\widehat{\phi}_n(u) = \prod_{j=1}^n \widehat{\psi}(u/\alpha^j),$$

generates two families of Appell sequences

$$\widehat{\phi}_n(iz)e^{-xz} = \sum_{m=0}^{\infty} P_m(x; n)z^m$$

and

$$\frac{e^{xz}}{\widehat{\phi}_n(iz)} = \sum_{m=0}^{\infty} Q_m(x; n)z^m,$$

which we shall call the *Appell system of  $S_n$* .



- $\widehat{\phi}_n(iz) \rightarrow \widehat{\phi}(iz)$  locally uniformly as  $n \rightarrow \infty$ ,  
where

$$\widehat{\phi}(iz) = \prod_{j=1}^{\infty} \widehat{\psi}(iz/\alpha^j).$$

- $P_m(x; n) \rightarrow P_m(x)$  and  $Q_m(x; n) \rightarrow Q_m(x)$ ,  
 $x \in \mathbb{R}$ , as  $n \rightarrow \infty$ , where  $(P_m)$  and  $(Q_m)$  are  
generated by

$$\widehat{\phi}(iz)e^{-xz} = \sum_{m=0}^{\infty} P_m(x)z^m$$

and

$$\frac{e^{xz}}{\widehat{\phi}(iz)} = \sum_{m=0}^{\infty} Q_m(x)z^m .$$

- $(P_m)$  and  $(Q_m)$  are an Appell system of

$$S = \sum_{j=0}^{\infty} Y_j/\alpha^j .$$

### Proposition 3.1

Let  $(P_m(\cdot; n))_m, (Q(\cdot; n))_m, n = 1, 2, \dots,$  be an Appell system of  $S_n$ . For any two independent random variables  $X$  and  $T$ , where  $T$  has finite moments of all order, and for any  $n \geq 1$ ,

$$\sum_{m=0}^{\infty} E \{P_m(X + T; n)|X\} \frac{z^m}{m!} = \prod_{j=1}^n \widehat{\psi}(iz/\alpha^j) e^{-Xz} E\{e^{-Tz}\}$$

and

$$\sum_{m=0}^{\infty} E \{Q_m(X + T; n)|X\} \frac{z^m}{m!} = \frac{e^{Xz} E\{e^{Tz}\}}{\prod_{j=1}^n \widehat{\psi}(iz/\alpha^j)}$$

in a neighborhood of the origin.

## Proof of Proposition 3.1

We shall prove only the first relation.

$$\begin{aligned} \sum_{m=0}^{\infty} E \{ P_m(X + T; n) | X \} \frac{z^m}{m!} &= E \left\{ \sum_{m=0}^{\infty} P_m(X + T; n) \frac{z^m}{m!} | X \right\} \\ &= E \left\{ \prod_{j=1}^n \psi(iz/\alpha^j) e^{-(X+T)z} | X \right\} \\ &= \prod_{j=1}^n \psi(iz/\alpha^j) e^{-Xz} E \{ e^{-Tz} | X \}, \end{aligned}$$

which leads to the first relation, since  $T$  and  $X$  are independent.

QED

## Theorem 3.2

Let  $(P_m(x; n))_m, (Q(x; n))_m, n = 1, 2, \dots$ , be an Appell system of  $S_n$ . If  $X$  is independent of  $Y_j$ , then for any  $m \geq 0$  and  $n \geq 1$ , we have the following relations,

$$E \{P_m(\alpha X - Y_j; n) | X\} = \alpha^m P_m(X; n + 1),$$

$$E \{P_m(X - Y_j/\alpha^{n+1}; n) | X\} = P_m(X; n + 1),$$

$$E \left\{ Q_m \left( \frac{X + Y_j}{\alpha}; n \right) | X \right\} = \alpha^{-m} Q_m(X; n - 1),$$

and

$$E \{Q_m(X + Y_j/\alpha^n; n) | X\} = Q_m(X; n - 1).$$

### Corollary 3.3

Let  $(P_m(x; n))_m, (Q_m(x; n))_m, n = 1, 2, \dots,$  be an Appell system of  $S_n = \sum_{j=1}^n Y_j/\alpha^j$ . Then for any  $m \geq 0$  and  $n \geq 1$ ,

$$E \left\{ Q_m \left( \frac{S_{n-1} + Y_n}{\alpha}; n \right) | S_{n-1} \right\} = \alpha^{-m} Q_m(S_{n-1}; n-1),$$

and

$$E \{ Q_m(S_n; n) | S_{n-1} \} = Q_m(S_{n-1}; n-1).$$

## Remark

- Since

$$E \{ E \{ Q_m(S_n; n) | S_{n-1} \} | S_{n-2} \} = E \{ Q_m(S_n; n) | S_{n-2} \},$$

a repeated application of the last relation in Corollary 3.3 gives the martingale equality

$$E \{ Q_m(S_n; n) | S_k \} = Q_m(S_k; k), \quad k = 1, 2, \dots, n-1,$$

which shows that for each  $m \geq 0$ ,  $(Q_m(S_n; n))_{n \geq 1}$  is a martingale sequence.

- This martingale relation has been observed by Wim Schoutens (1998) in a different context.

### Corollary 3.4

Let  $(P_m)$  and  $(Q_m)$  be an Appell system of  $S = \sum_{j=1}^{\infty} Y_j/\alpha^j$ . If  $X$  is independent of  $Y_j$ , then for any  $m \geq 0$ , we have the following equalities,

$$E \{P_m(\alpha X - Y_j)|X\} = \alpha^m P_m(X),$$

$$E \left\{ Q_m \left( \frac{X + Y_j}{\alpha} \right) | X \right\} = \alpha^{-m} Q_m(X).$$

### 3.5 Biorthogonal Systems of Eigenfunctions

Let  $\psi$  be a probability measure whose c.f. is analytic in a neighborhood of the origin.

- The first relation in Corollary 3.4 implies that the Appell polynomials  $P_m$  are eigenfunctions with eigenvalues  $\alpha^{m+1}$ ,  $m = 0, 1, \dots$ , of the scaling operator

$$W_{\psi, \alpha} f(x) := \alpha \int_{\mathbb{R}} f(\alpha x - y) d\psi(y). \quad (6)$$

- The second implies that  $Q_m$  are eigenfunctions of its adjoint  $W_{\psi, \alpha}^*$  with eigenvalues  $\alpha^{-m}$ ,  $m = 0, 1, \dots$ .
- There is a unique probability measure  $\phi$ , which is the invariant distribution of  $W_{\psi, \alpha}$ , and

$$\widehat{\phi}(u) = \prod_{j=1}^{\infty} \widehat{\psi}(u/\alpha^j).$$

- (6) shows that the distributional derivatives  $\phi^{(m)}$  are eigenfunctions of  $W_{\psi, \alpha}$  with eigenvalues  $\alpha^{-m}$ ,  $m = 0, 1, \dots$ .



### Theorem 3.5

Let  $\psi$  be a probability measure whose F.T. is analytic in a neighbourhood of the origin and  $\widehat{\phi}(u) = \prod_{j=1}^{\infty} \widehat{\psi}(u/\alpha^j)$ . Then  $(\phi^{(m)})_{m \geq 0}$  and  $(Q_m)_{m \geq 0}$  form a biorthogonal eigensystem, i.e.,

$$\langle (-1)^m \phi^{(m)}, Q_n \rangle = \delta_{m,n}, \quad m, n = 0, 1, \dots$$

where  $\langle \cdot, \cdot \rangle$  denotes the action of a distribution on a test function.

## 4 Gaussian, $B$ -splines, Hermite and Bernoulli Polynomials

### 4.1 Gaussian Function

$$G(x) = \frac{1}{\sqrt{2\pi}} e^{-x^2/2}.$$

- Optimal in time-frequency localization, i.e. it attains the bound in the Heissenberg uncertainty product.
- Invariant density of  $W_{\psi,\alpha}$ , i.e.

$$W_{\psi,\alpha}G = G,$$

where, in this case,  $\psi$  is the Gaussian density function

$$\psi(x) = \frac{1}{\sqrt{2\pi(\alpha^2 - 1)}} e^{-x^2/2(\alpha^2-1)}.$$

## 4.2 Gaussian and Hermite Polynomials

- Take i.i.d. Gaussian random variables  $Y_j$  with density

$$\psi(x) = \frac{1}{\sqrt{2\pi(\alpha^2 - 1)}} e^{-x^2/2(\alpha^2 - 1)} .$$

- Then  $S = \sum_{j=1}^{\infty} Y_j/\alpha^j$  has density

$$\phi(x) = G(x),$$

which is the invariant density of  $W_{\psi, \alpha}$ , and

$$\widehat{\phi}(u) = e^{-u^2/2} .$$

- The same method applies giving the generating function

$$e^{xz - z^2/2} = \sum_{m=0}^{\infty} H_m(x) z^m ,$$

which generates the Hermite polynomials.

- It is known that

$$\phi^{(n)}(x) = (-1)^n H_n(x) G(x) .$$

- Hence the biorthogonal relation becomes

$$\delta_{m,n} = \langle (-1)^n \phi^{(n)}, H_m \rangle = \int_{\mathbb{R}} H_n(x) H_m(x) \frac{1}{\sqrt{2\pi}} e^{-x^2/2} dx ,$$

which is the familiar orthogonal relations of Hermite polynomials.

### 4.3 $B$ -splines

$$B_N = \chi_{[0,1]} * \cdots * \chi_{[0,1]}, \quad (N \text{ times}),$$

is a scaling function that satisfies the scaling equation

$$\phi(x) = 2 \sum_{j=0}^N \frac{1}{2^N} \binom{N}{j} \phi(2x - j).$$

#### Properties

- Piecewise polynomials with knots at integers.
- $\text{Supp}(B_N) = [0, N]$ .
- $\widehat{B}_N(u) = \left(\frac{1-e^{-iu}}{iu}\right)^N$ .
- $B_N$  approximates the Gaussian function:

$$\widetilde{B}_N \rightarrow G \text{ in } L^p(\mathbb{R}), \quad 1 \leq p \leq \infty,$$

where

$$\widetilde{B}_N(x) = \frac{1}{2} \sqrt{\frac{N}{3}} B_N \left( \frac{1}{2} \sqrt{\frac{N}{3}} x + \frac{N}{2} \right).$$

- provides efficient algorithm for computation.

## 4.4 $B$ -splines and Bernoulli Polynomials

- Take  $\alpha = 2$ , and i.i.d. random variables  $Y_j$  whose distribution,  $\psi$ , is the measure concentrated on the integers with binomial weights

$$\frac{1}{2^N} \binom{N}{m}, \quad m = 0, 1, \dots, N, \quad \text{and } 0, \text{ otherwise.}$$

- The density of  $S = \sum_{j=1}^{\infty} Y_j / \alpha^j$ ,  $\phi = B_N$ , and

$$\widehat{\phi}(u) = \left( \frac{1 - e^{-iu}}{iu} \right)^N .$$

- The corresponding generating function is

$$\frac{z^N e^{xz}}{(e^z - 1)^N} = \sum_{m=0}^{\infty} B_{N,m}(x) z^m ,$$

which generates the Bernoulli polynomials  $B_{N,m}$  of order  $N$ .

- Further,

$$T_{\psi}^* B_{N,m} = 2^{-m} B_{N,m},$$

and

$$\langle (-1)^n B_N^{(n)}, B_{N,m} \rangle = \delta_{n,m}, \quad n, m = 0, 1, \dots .$$

## 4.5 Normalized $B$ -splines and Bernoulli Polynomials

- Take

$$\phi = \tilde{B}_N := \sigma_N B_N(\sigma_N \cdot + \mu_N),$$

the normalized  $B$ -splines.

- Then

$$\hat{\phi}(u) = e^{iu\mu_N/\sigma_N} \left( \frac{1 - e^{iu/\sigma_N}}{iu/\sigma_N} \right)^N .$$

- The corresponding generating function is

$$\frac{(z/\sigma_N)^N e^{(\sigma_N x + \mu_N)(z/\sigma_N)}}{(e^{z/\sigma_N} - 1)^N} = \sum_{m=0}^{\infty} \tilde{B}_{N,m}(x) z^m ,$$

which generates the normalized Bernoulli polynomials of order  $N$ ,

$$\tilde{B}_{N,m}(x) = \sigma_N^{-m} B_{N,m}(\sigma_N x + \mu_N) .$$

- The biorthogonal relation is

$$\langle (-1)^n \tilde{B}_N^{(n)}, \tilde{B}_{N,m} \rangle = \delta_{n,m}, \quad n, m = 0, 1, \dots .$$

$$\begin{array}{ccc}
\tilde{B}_N & \xrightarrow{L^p} & G \\
\Downarrow & & \Downarrow \\
\tilde{B}_{N,m} & \xrightarrow{?} & H_m
\end{array}$$

$$\begin{aligned}
\lim_{N \rightarrow \infty} \sum_{m=0}^{\infty} \tilde{B}_{N,m}(x) z^m &= \lim_{N \rightarrow \infty} \frac{(z/\sigma_N)^N e^{(\sigma_N x + \mu_N)(z/\sigma_N)}}{(e^{z/\sigma_N} - 1)^N} \\
&= \dots = \\
&= e^{xz - z^2/2} \\
&= \sum_{m=0}^{\infty} H_m(x) z^m,
\end{aligned}$$

which shows that for  $m = 0, 1, \dots$ ,

$$\lim_{N \rightarrow \infty} \tilde{B}_{N,m}(x) = H_m(x), \quad x \in \mathbb{R}.$$