

Generalized Jacobi Chebyshev Wavelet Approximation: Perkiraan Wavelet Jacobi Chebyshev Umum

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General Background: Wavelet approximations are fundamental in numerical analysis and signal processing, with classical orthogonal polynomials like Jacobi and Chebyshev serving as key tools due to their strong approximation properties. **Specific Background:** The use of Chebyshev wavelets has been extended through generalized polynomial frameworks, such as Koornwinder's generalization of Jacobi polynomials, offering more flexibility for function approximation on finite intervals. **Knowledge Gap:** Despite existing wavelet frameworks, the integration of generalized Jacobi and Chebyshev structures into a unified wavelet approximation scheme remains underexplored. **Aims:** This study introduces the Generalized Jacobi Chebyshev Wavelet (GJCW) approximation, establishing its theoretical foundations and demonstrating convergence and approximation capabilities. **Results:** It is shown that for a uniformly bounded function expanded in the GJCW basis, the partial sums yield both convergent and best uniform polynomial approximations. **Novelty:** The formulation of a new wavelet approximation based on a hybrid of generalized Jacobi and Chebyshev polynomials constitutes a novel contribution, supported by rigorous recurrence relations and multiresolution analysis. **Implications:** This work enhances the theoretical landscape of wavelet-based function approximation, with potential applications in computational mathematics, signal analysis, and numerical solutions of differential equations.

Highlight :

- Wavelet Construction: The paper defines and constructs *generalized Jacobi Chebyshev wavelets* using orthogonal polynomials.
- Approximation Theory: It proves that if the wavelet series converges, then a *uniform best polynomial approximation* exists.
- Multiresolution Framework: The approach is grounded in *Mallat's multiresolution analysis*, enabling efficient function approximation.

Keywords : Jacobi Polynomials, Chebyshev Wavelets, Multiresolution Analysis, Polynomial Approximation, Orthonormal Basis

INTRODUCTION

Jacobi polynomials $P_n^{(\alpha, \beta)}(x)$ constitute a category of classical orthogonal polynomials. They are orthogonal about the weight $(1-x)^\alpha (1+x)^\beta$ on the interval $[-1, 1]$. We have $P_n^{((\alpha, \beta))}(x)$ Chebyshev polynomials with $|x| \leq 1$.

n	Jacobi Polynomial
0	1
1	$(\alpha + \beta) + \frac{1}{2}(\alpha + \beta + 2)x$
2	$-4 + \alpha^2 - \beta + \beta^2 - \alpha(1 + 2\beta) + 2(3\alpha + \alpha^2 - \beta(3 + \beta))x$ $+ \frac{1}{8}(3 + \alpha + \beta)(4 + \alpha + \beta)x^2$
3	$(\alpha + \beta)(-16 + \alpha^2 + (-3 + \beta)\beta - \alpha(3 + 2\beta)) + 3(\alpha + \beta + 4)$ $(\alpha^2 - (2\beta + 1)\alpha + \beta^2 - \beta - 6)x + 3(\alpha + \beta(\alpha + \beta + 4)(\alpha + \beta + 5)x^2$ $+ \frac{1}{48}((\alpha + \beta + 4)(\alpha + \beta + 5)(\alpha + \beta + 6)x^3$

Figure 1.

also for $n=4,5,\dots$

$$P_n^{(\alpha,\beta)}(x) = \frac{(-1)^n}{2^n n!} (1-x)^{-\alpha} (1+x)^{-\beta} \frac{d^n}{dx^n} \{ (1-x)^\alpha (1+x)^\beta (1-x^2)^n \}.$$

Figure 2.

The Jacobi polynomials are generated by the three-term recurrence relation, for scalars $a_n^{(\alpha,\beta)}, b_n^{(\alpha,\beta)}$ and $c_n^{(\alpha,\beta)}$,

$$P_{n+1}^{(\alpha,\beta)}(x) = (a_n^{(\alpha,\beta)} x - b_n^{(\alpha,\beta)}) P_n^{(\alpha,\beta)}(x) - c_n^{(\alpha,\beta)} P_{n-1}^{(\alpha,\beta)}(x) \quad n \geq 1,$$

Figure 3.

where

$$P_0^{(\alpha,\beta)}(x) = 1, \quad P_1^{(\alpha,\beta)}(x) = \frac{1}{2}(\alpha + \beta + 2)x + \frac{1}{2}(\alpha - \beta),$$

Figure 4.

also

$$\begin{aligned}a_n^{(\alpha,\beta)} &= \frac{(2n+\alpha+\beta+1)(2n+\alpha+\beta)}{2(n+1)(n+\alpha+\beta+1)}, \\b_n^{(\alpha,\beta)} &= \frac{(\beta^2-\alpha^2)(2n+\alpha+\beta+1)}{2(n+1)(n+\alpha+\beta+1)(n+\alpha+\beta)}, \\c_n^{(\alpha,\beta)} &= \frac{(2n+\alpha+\beta+2)(n+\alpha)(n+\beta)}{(n+1)(n+\alpha+\beta+1)(2n+\alpha+\beta)},\end{aligned}$$

Figure 5.

The Jacobi polynomials $y=P_n^{(\alpha,\beta)}(x)$ solve the linear second-order differential equation

$$(1-x^2)y'' + [\beta - \alpha - (\alpha + \beta + 2)x]y' + n(n + \alpha + \beta + 1)y = 0.$$

Figure 6.

A. Jacobi polynomials

Theorem 1.1 *There are scalars $a_0^{(\alpha,\beta)}, a_1^{(\alpha,\beta)}, \dots, a_n^{(\alpha,\beta)} \in \mathbb{R}$ such that*

$$P_n^{(\alpha,\beta)}(x) = \sum_{m=0}^n a_{n,m}^{(\alpha,\beta)} x^m.$$

Proof. We use mathematical induction.

For this

$$P_0^{(\alpha,\beta)}(x) = 1,$$

$$P_1^{(\alpha,\beta)}(x) = \frac{1}{2}(\alpha + \beta + 2)x + \frac{1}{2}(\alpha - \beta).$$

The hypothesis of mathematical induction:

Suppose for $0 \leq k < n$, we have

$$P_k^{(\alpha,\beta)}(x) = \sum_{m=0}^k c_{k,m}^{(\alpha,\beta)} x^m,$$

Therefore

$$P_{n-1}^{(\alpha,\beta)}(x) = \sum_{m=0}^{n-1} c_{n-1,m}^{(\alpha,\beta)} x^m$$

and

$$P_{n-2}^{(\alpha,\beta)}(x) = \sum_{m=0}^{n-2} c_{n-2,m}^{(\alpha,\beta)} x^m,$$

Figure 7.

The rule of mathematical induction:

For scalars $a_n^{(\alpha, \beta)}$, $b_n^{(\alpha, \beta)}$ and $c_n^{(\alpha, \beta)}$

$$\begin{aligned}
 & | \\
 & P_n^{(\alpha, \beta)}(x) = (d_n^{(\alpha, \beta)} x - e_n^{(\alpha, \beta)}) P_{n-1}^{(\alpha, \beta)}(x) + f_n^{(\alpha, \beta)} P_{n-2}^{(\alpha, \beta)}(x) \\
 & = (c_0^{(\alpha, \beta)} c_{n,0}^{(\alpha, \beta)} - f_0^{(\alpha, \beta)} c_{n,0}^{(\alpha, \beta)}) + \sum_{m=1}^{n-2} [c_{n,m-1}^{(\alpha, \beta)} - c_n^{(\alpha, \beta)} c_{n,m}^{(\alpha, \beta)} - \\
 & f_n^{(\alpha, \beta)} c_{n,m}^{(\alpha, \beta)}] x^m \\
 & + (c_{n-2}^{(\alpha, \beta)} c_{n,n-2}^{(\alpha, \beta)} - f_{n-2}^{(\alpha, \beta)} c_{n,n-2}^{(\alpha, \beta)}) x^{n-2} + c_{n-1}^{(\alpha, \beta)} c_{n,n-1}^{(\alpha, \beta)} x^{n-1} + c_{n,n-1}^{(\alpha, \beta)} d_n^{(\alpha, \beta)} x^n \\
 & = \sum_{m=0}^n h_{n,m}^{(\alpha, \beta)} x^m.
 \end{aligned}$$

Figure 8.

Theorem 1.2

Let $P_n^{(\alpha, \beta)}(x) = \sum_{m=0}^n a_{n,m}^{(\alpha, \beta)} x^m$ be a representation of Jacobi polynomials. Then The following conditions are satisfying:

- i) $a_{n,m}^{(\alpha, \beta)} = 0$ for all $m \geq n$,
- ii) $(m-1)(m+2)a_{n,m+2}^{(\alpha, \beta)} - (\beta - \alpha)(m+1)a_{n,m+1}^{(\alpha, \beta)} + [-m(m-1) - m(\alpha + \beta + 2) + n(n + \alpha + \beta + 1)]a_{n,m}^{(\alpha, \beta)} = 0$ for all $0 \leq m \leq n$.

Proof. Suppose $y = P_n^{(\alpha, \beta)}(x) = \sum_{m=0}^n a_{n,m}^{(\alpha, \beta)} x^m$, then

$$y' = \sum_{m=1}^n m a_{n,m}^{(\alpha, \beta)} x^{m-1} = \sum_{m=0}^{n-1} (m+1) a_{n,m+1}^{(\alpha, \beta)} x^m$$

$$y'' = \sum_{m=2}^n m(m-1) a_{n,m}^{(\alpha, \beta)} x^{m-2} = \sum_{m=0}^{n-2} (m+1)(m+2) a_{n,m+2}^{(\alpha, \beta)} x^m.$$

We put in above equation we have

$$\begin{aligned}
 & \sum_{m=0}^{n-2} (m+1)(m+2) a_{n,m+2}^{(\alpha, \beta)} x^m - \sum_{m=2}^n m(m-1) a_{n,m}^{(\alpha, \beta)} x^m + (\beta - \alpha) \sum_{m=0}^{n-1} (m \\
 & + 1) a_{n,m+1}^{(\alpha, \beta)} x^m - ((\alpha + \beta + 2) \sum_{m=0}^n m a_{n,m}^{(\alpha, \beta)} x^m + n(n + \alpha + \beta \\
 & + 1) \sum_{m=0}^n a_{n,m}^{(\alpha, \beta)} x^m = 0,
 \end{aligned}$$

it follows that

$$\begin{aligned}
 & \sum_{m=0}^n - (m-1)(m+2) a_{n,m+2}^{(\alpha, \beta)} - (\beta - \alpha)(m+1) a_{n,m+1}^{(\alpha, \beta)} \\
 & + [-m(m-1) - m(\alpha + \beta + 2) + n(n + \alpha + \beta + 1)] a_{n,m}^{(\alpha, \beta)} x^m = 0.
 \end{aligned}$$

Figure 9.

B. Generalized Jacobi Chebyshev Wavelets

Let $\alpha > -1, \beta > -1, M \geq 0$ and $N \geq 0$. Koornwinder in [12] it is demonstrated that the generalised Jacobi polynomials $\{P_n^{(\alpha, \beta, M, N)}(x)\}_{n=0}^{\infty}$ can be written as

$$P_n^{(\alpha, \beta, M, N)}(x) = P_n^{(\alpha, \beta)}(x) + M Q_n^{(\alpha, \beta)}(x) + N R_n^{(\alpha, \beta)}(x) + M N S_n^{(\alpha, \beta)}(x), \quad n = 0, 1, 2, \dots,$$

where

$$Q_0^{(\alpha, \beta)}(x) = R_0^{(\alpha, \beta)}(x) = S_0^{(\alpha, \beta)}(x) = 0,$$

and for $n = 1, 2, 3, \dots$,

$$Q_n^{(\alpha, \beta)}(x) = \frac{(\beta+2)_{n-1}(\alpha+\beta+2)_{n-1}}{(\alpha+1)_n n!} \times [n(n+\alpha+\beta+1)P_n^{(\alpha, \beta)}(x) - (\beta+1)(x-1)DP_n^{(\alpha, \beta)}(x)],$$

$$R_n^{(\alpha, \beta)}(x) = \frac{(\alpha+2)_{n-1}(\alpha+\beta+2)_{n-1}}{(\beta+1)_n n!} \times [n(n+\alpha+\beta+1)P_n^{(\alpha, \beta)}(x) - (\alpha+1)(x+1)DP_n^{(\alpha, \beta)}(x)],$$

and

$$S_n^{(\alpha, \beta)} = \frac{(\alpha+\beta+2)_n(\alpha+\beta+2)_{n-1}}{(\alpha+1)(\beta+1)n!(n-1)!} \times [n(n+\alpha+\beta+1)P_n^{(\alpha, \beta)}(x) - \{(\beta+1)(x-1) + (\alpha+1)(x+1)\}DP_n^{(\alpha, \beta)}(x)].$$

Since $P_n^{(\alpha, \beta, M, N)}(x), R_n^{(\alpha, \beta)}, Q_n^{(\alpha, \beta)}, S_n^{(\alpha, \beta)}$ are linear combination of $P_n^{(\alpha, \beta)}, DP_n^{(\alpha, \beta)}$, we have

Figure 10.

Proposition 2.1 The generalized Jacobi polynomials $P_n^{(\alpha, \beta, M, N)}(x)$ are generated by the three-term recurrence relation, for scalars $d_n^{(\alpha, \beta, M, N)}, e_n^{(\alpha, \beta, M, N)}$ and $f_n^{(\alpha, \beta, M, N)}$, for $n \geq 1$

$$P_{n+1}^{(\alpha, \beta, M, N)}(x) = (d_n^{(\alpha, \beta, M, N)}x - e_n^{(\alpha, \beta, M, N)})P_n^{(\alpha, \beta, M, N)}(x) - f_n^{(\alpha, \beta, M, N)}P_{n-1}^{(\alpha, \beta, M, N)}(x).$$

In the following table, we define generalized Chebyshev wavelets. (see [2, 4, 7, 8, 9, 10, 11, 15])

Suppose $k \in \mathbb{N}$ (degree of multiresolution), $m \geq 0, n = 1, 2, \dots, 2^k$

$$\Psi_{n,m}^{(\alpha, \beta, M, N)}(t) = \begin{cases} \sqrt{\frac{2^{k+1}}{n}} P_m^{(\alpha, \beta, M, N)}(2^k t - 2n + 1) & t \in [\frac{n-1}{2^{k-1}}, \frac{n}{2^{k-1}}], \\ 0 & \text{otherwise} \end{cases}$$

A function $f \in L^2[-1, 1)$ is expanded by generalized Chebyshev wavelets series as

$$f(t) = \sum_{n=0}^{2^k} \sum_{m=0}^{\infty} c_{n,m} \Psi_{n,m}^{(\alpha, \beta, M, N)}(t),$$

where

$$c_{n,m} = \int_{-1}^1 f(t) \Psi_{n,m}^{(\alpha, \beta, M, N)}(t) \omega_{n,m}^{(\alpha, \beta, M, N)}(t) dt,$$

and $\omega_{n,m}^{(\alpha, \beta, M, N)}$ is the weight function of (α, β, M, N) generalized Chebyshev polynomials. Also

$$\int_{-1}^1 |\Psi_{n,m}^{(\alpha, \beta, M, N)}(t)|^2 \omega_{n,m}^{(\alpha, \beta, M, N)}(t) dt = L_{n,m}^{(\alpha, \beta, M, N)},$$

Figure 11.

It is necessary to study multiresolution analysis and Mallat's Theorem for wavelet approximation.

Definition 2.1 Multiresolution Analysis: An MRA with scaling function ϕ constitutes a collection of closed subspaces

$\{V_j\}_{j \in \mathbb{Z}}$ of $L^2(\mathbb{R})$, such that

(i) $V_j \subset V_{j+1}$;

(ii) $f(x) \in V_j \Leftrightarrow f(2x) \in V_{j+1}$;

(iii) $\overline{\bigcup V_j} = L^2(\mathbb{R})$,

(iv) $V_j \cap V_k = \{0\}$;

(v) There exists a function $\phi \in V_0$ such that the collection $\{\phi(x-k) : k \in \mathbb{Z}\}$ is a Riesz basis of V_0 .

Figure 12.

The series of wavelet subspaces W_j of $L^2(\mathbb{R})$ is such that $V_j \perp W_j$, for all j and $V_{j+1} = V_j \oplus W_j$. Closure of $\bigoplus W_j$ is dense in $L^2(\mathbb{R})$ for L^2 norm.

Figure 13.

We now present Mallat's theorem, which ensures that in the context of an orthogonal multiresolution analysis (MRA), an orthonormal basis exists for $L^2(\mathbb{R})$. These basis functions are essential in wavelet theory, facilitating the development of sophisticated computational algorithms.

Lemma 2.1 (Mallat's Theorem) In the context of an orthogonal multiresolution analysis (MRA) characterised by a scaling function ϕ , a corresponding wavelet exists $\psi \in L^2(\mathbb{R})$ such that for each $j \in \mathbb{Z}$, the family $\{\psi_{j,k}\}_{k \in \mathbb{Z}}$ is an orthonormal basis for W_j . Hence the family $\{\psi_{j,k}\}_{j,k \in \mathbb{Z}}$ is an orthonormal basis for $L^2(\mathbb{R})$.

Definition 2.2 (i) Let family be an orthonormal basis for $L^2(\mathbb{R})$ and $P_n(f)$ the orthogonal projection of $L^2([-1,1])$ onto V_n . Then

$$P_n(f) = \sum_{k=-\infty}^{\infty} \langle f, \psi_{n,k} \rangle \psi_{n,k}, n=1,2,3,\dots$$

(ii) The wavelet approximation of the Chebyshev polynomial is defined by

$$E_n(f) = \|f - P_n(f)\|_2 = \int_{-1}^1 |f(t) - P_n(f)(t)|^2 dt =$$

$o(\phi(n))$, $o(\phi(n))$ is a small function.

Theorem 2.1 Let $f \in L^2([-1, 1])$ be a uniformly bounded function and

$f(t) = \sum_{n=0}^{2^k} \sum_{m=0}^{\infty} t_{n,m} \Psi_{n,m}^{(\alpha, \beta, M, N)}(t)$ be expanded in terms of generalized Chebyshev wavelets and the series $\sum_{n=0}^{2^k} \sum_{m=0}^{\infty} |t_{n,m}|^2 L_{n,m}^{(\alpha, \beta, M, N)}$ be convergent. Then generalized Chebyshev wavelet approximation f , for every M is the partial sums

$$u_{2^k, M-1}^{(\alpha, \beta, M, N)}(t) = \sum_{n=0}^{2^k} \sum_{m=0}^{M-1} t_{n,m} \Psi_{n,m}^{(\alpha, \beta, M, N)}(t),$$

and

$$E_{2^k, M-1}(f) = o\left(\left(\sum_{n=0}^{2^k} \sum_{m=M}^{\infty} L_{n,m}^{(\alpha, \beta, M, N)} |t_{n,m}|^2\right)^{\frac{1}{2}}\right).$$

Figure 14.

Proof. We have

$$\begin{aligned} & \|f - u_{2^k, M-1}^{(\alpha, \beta, M, N)}\|_2^2 \\ &= \int_{-1}^1 \left| \sum_{n=0}^{2^k} \sum_{m=0}^{\infty} t_{n,m} \Psi_{n,m}^{(\alpha, \beta, M, N)}(t) \right. \\ &\quad \left. - \sum_{n=0}^{2^k} \sum_{m=0}^{M-1} t_{n,m} \Psi_{n,m}^{(\alpha, \beta, M, N)}(t) \right|^2 \omega_{n,m}^{(\alpha, \beta, M, N)}(t) dt \\ &= \int_{-1}^1 \left| \sum_{n=0}^{2^k} \sum_{m=M}^{\infty} t_{n,m} \Psi_{n,m}^{(\alpha, \beta, M, N)}(t) \right|^2 \omega_{n,m}^{(\alpha, \beta, M, N)}(t) dt \\ &\leq \sum_{n=0}^{2^k} \sum_{m=M}^{\infty} |t_{n,m}|^2 \int_{-1}^1 |\Psi_{n,m}^{(\alpha, \beta, M, N)}(t)|^2 \omega_{n,m}^{(\alpha, \beta, M, N)}(t) dt \\ &= \sum_{n=0}^{2^k} \sum_{m=M}^{\infty} |t_{n,m}|^2 L_{n,m}^{(\alpha, \beta, M, N)} \end{aligned}$$

Therefore $\|f - s_{M-1}\|_2 \leq \left(\sum_{m=M}^{\infty} |t_{n,m}|^2 L_{n,m}^{(\alpha, \beta, M, N)}\right)^{\frac{1}{2}}$. That is

$$E_{2^k, M-1}(f) = o\left(\left(\sum_{n=0}^{2^k} \sum_{m=M}^{\infty} |t_{n,m}|^2 L_{n,m}^{(\alpha, \beta, M, N)}\right)^{\frac{1}{2}}\right),$$

Figure 15.

Definition 2.3 Suppose $f \in L^2[a, b]$ and P_n is a set of all polynomials of degree n and smaller n . If there exists a function $q^* \in P_n$ such that $\lim_{n \rightarrow \infty} P_n(f) = 0$, where $P_n(f) = \inf_{p \in P_n} \|f - p\|_2$. Then f is called best uniform polynomial approximation to f on $[a, b]$.

Figure 16.

Theorem 2.2 Let $f(t) = \sum_{n=0}^{2^k} \sum_{m=0}^{\infty} t_{n,m} \Psi_{n,m}^{(\alpha,\beta,M,N)}(t)$ be expanded in terms of generalized Chebyshev wavelets. If $\sum_{n=0}^{2^k} \sum_{m=0}^{\infty} |t_{n,m}|^2 L_{n,m}^{(\alpha,\beta,M,N)}$ is converge, then (α, β, M, N) generalized Chebyshev wavelet approximation $E_{2^k,l}(f)$ of f is $f_1(t)$.

$$E_{2^k,l}(f) = o\left(\left(\sum_{n=0}^{2^k} \sum_{m=l+1}^{\infty} |t_{n,m}|^2 L_{n,m}^{(\alpha,\beta,M,N)}\right)^{\frac{1}{2}}\right).$$

Proof

$$\begin{aligned} \|f - f_1\|_2^2 &= \int_{-1}^1 \left| \sum_{n=0}^{2^k} \sum_{m=0}^{\infty} t_{n,m} \Psi_{n,m}^{(\alpha,\beta,M,N)}(t) - \sum_{n=0}^{2^k} \sum_{m=0}^l t_{n,m} \Psi_{n,m}^{(\alpha,\beta,M,N)}(t) \right|^2 \omega_{n,m}^{(\alpha,\beta,M,N)}(t) dt \\ &= \int_{-1}^1 \left| \sum_{n=0}^{2^k} \sum_{m=l+1}^{\infty} t_{n,m} \Psi_{n,m}^{(\alpha,\beta,M,N)}(t) \right|^2 \omega_{n,m}^{(\alpha,\beta,M,N)}(t) dt \\ &\leq \sum_{n=0}^{2^k} \sum_{m=l+1}^{\infty} |t_{n,m}|^2 \int_{-1}^1 |\Psi_{n,m}^{(\alpha,\beta,M,N)}(t)|^2 \omega_{n,m}^{(\alpha,\beta,M,N)}(t) dt \\ &\leq \sum_{n=0}^{2^k} \sum_{m=l+1}^{\infty} |t_{n,m}|^2 \int_{-1}^1 |\Psi_{n,m}^{(\alpha,\beta,M,N)}(t)|^2 \omega_{n,m}^{(\alpha,\beta,M,N)}(t) dt \\ &\leq L \sum_{n=0}^{2^k} \sum_{m=l+1}^{\infty} |t_{n,m}|^2 L_{n,m}^{(\alpha,\beta,M,N)}. \end{aligned}$$

Therefore

$$\|f - f_1(t)\|_2 \leq \left(\sum_{n=0}^{2^k} \sum_{m=l+1}^{\infty} |t_{n,m}|^2 L_{n,m}^{(\alpha,\beta,M,N)}\right)^{\frac{1}{2}},$$

$$E_{2^k,l}(f) = o\left(\left(\sum_{n=0}^{2^k} \sum_{m=l+1}^{\infty} |t_{n,m}|^2 L_{n,m}^{(\alpha,\beta,M,N)}\right)^{\frac{1}{2}}\right).$$

Figure 17.

Theorem 2.3 Let $f(t) = \sum_{n=0}^{2^k} \sum_{m=0}^{\infty} t_{n,m} \Psi_{m,n}^{(\alpha,\beta,\mathcal{M},N)}(t)$ be expanded in terms of generalized Chebyshev wavelets. If $\sum_{n=0}^{2^k} \sum_{m=0}^{\infty} |t_{n,m}|^2 L_{n,m}^{(\alpha,\beta,\mathcal{M},N)}$ is converge, then generalized Chebyshev wavelet approximation $E_{2^k,l}(t)$ of f is $-f_2(t)$ and

$$E_{2^k,l}(f) = o((\sum_{n=0}^{2^k} \sum_{m=l+1}^{\infty} |t_{n,m}|^2 L_{n,m}^{(\alpha,\beta,\mathcal{M},N)})^{\frac{1}{2}}).$$

Proof.

$$\begin{aligned} \|f - (-f_2)\|_2^2 &= \int_{-1}^1 |\sum_{n=0}^{2^k} \sum_{m=0}^l t_{n,m} \Psi_{m,n}^{(\alpha,\beta,\mathcal{M},N)-}(t)|^2 \omega_{n,m}^{(\alpha,\beta,\mathcal{M},N)}(t) dt \\ &+ \sum_{n=0}^{2^k} \sum_{m=0}^l |t_{n,m}|^2 \omega_{n,m}^{(\alpha,\beta,\mathcal{M},N)}(t) dt \\ &= \int_{-1}^1 |\sum_{n=0}^{2^k} \sum_{m=l+1}^{\infty} t_{n,m} \Psi_{m,n}^{(\alpha,\beta,\mathcal{M},N)-}(t)|^2 \omega_{n,m}^{(\alpha,\beta,\mathcal{M},N)}(t) dt \\ &\leq \sum_{n=0}^{2^k} \sum_{m=l+1}^{\infty} \int_{-1}^1 |t_{n,m}|^2 |\Psi_{m,n}^{(\alpha,\beta,\mathcal{M},N)-}(t)|^2 \omega_{n,m}^{(\alpha,\beta,\mathcal{M},N)}(t) dt \\ &\leq \sum_{n=0}^{2^k} \sum_{m=l+1}^{\infty} |t_{n,m}|^2 \int_{-1}^1 |\Psi_{m,n}^{(\alpha,\beta,\mathcal{M},N)}(t)|^2 \omega_{n,m}^{(\alpha,\beta,\mathcal{M},N)}(t) dt \\ &\leq \sum_{n=0}^{2^k} \sum_{m=l+1}^{\infty} |t_{n,m}|^2 L_{n,m}^{(\alpha,\beta,\mathcal{M},N)}. \end{aligned}$$

Therefore

$$\|f - (-f_2(t))\|_2 \leq (\sum_{n=0}^{2^k} \sum_{m=l+1}^{\infty} |t_{n,m}|^2 L_{n,m}^{(\alpha,\beta,\mathcal{M},N)})^{\frac{1}{2}},$$

$$E_{2^k,l}(f) = o((\sum_{n=0}^{2^k} \sum_{m=l+1}^{\infty} |t_{n,m}|^2 L_{n,m}^{(\alpha,\beta,\mathcal{M},N)})^{\frac{1}{2}}).$$

Figure 18.

Corollary 2.2 Let $f(t) = \sum_{n=0}^{2^k} \sum_{m=0}^{\infty} t_{n,m} \Psi_{m,n}^{(\alpha,\beta,\mathcal{M},N)+}(t)$ be expanded in terms of generalized Chebyshev wavelets. If $\sum_{n=0}^{2^k} \sum_{m=0}^{\infty} |t_{n,m}|^2 L_{n,m}^{(\alpha,\beta,\mathcal{M},N)}$ is converge, then uniform best polynomial approximation of f is $f_1(t)$.

Corollary 2.3 Let $f(t) = \sum_{n=0}^{2^k} \sum_{m=0}^{\infty} t_{n,m} \Psi_{m,n}^{(\alpha,\beta,\mathcal{M},N)-}(t)$ be expanded in terms of generalized Chebyshev wavelets. If $\sum_{n=0}^{2^k} \sum_{m=0}^{\infty} |t_{n,m}|^2 L_{n,m}^{(\alpha,\beta,\mathcal{M},N)}$ is converge, then uniform best polynomial approximation of f is $-f_2(t)$.

Figure 19.

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