

LEAST SQUARES APPROXIMATION WITH CONSTRAINT: GENERALIZED GEGENBAUER CASE

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Abstract. This paper considers the least squares approximation of function $f \in L^2[-1, 1]$, $f(-1)=f(1)=0$. Using generalized Gegenbauer weight function $p(x)=|x|^\mu(1-x^2)^\alpha$ ($\mu, \alpha > -1$), some of the results from [8] are generalized. This approximation is compared with the least square approximation without constraint. The approximation is illustrated on two numerical examples.

1. Introduction

In [11] Wrigge and Fransén considered two families of functions and showed how these functions can be approximated on $[0, 1]$ by the polynomials of the form $\sum_{n=1}^k c_{n,k}(x(1-x))^n$ and $(1-2x) \sum_{n=1}^k C_{n,k}(x(1-x))^n$. They used the L^2 -norm with respect to the weight function $p(x)=(x(1-x))^q$, where $q \in \{0, 1, \dots\}$. In [8] Milovanović and Wrigge presented a better and more natural way of approximation using Gegenbauer polynomials $C_{k,\lambda}(x)$ orthogonal with respect to the weight function $p(x)=(1-x^2)^{\lambda-1/2}$, $x \in [-1, 1]$, $\lambda > -1/2$. In this way they generalized the results from [11] and also avoided complicated manipulations with matrices.

Further generalization of these results can be obtained by using generalized Gegenbauer monic polynomials $W_k^{(\alpha,\beta)}(x)$, orthogonal on $[-1, 1]$ with respect to the weight function $p(x)=|x|^\mu(1-x^2)^\alpha$, $\mu, \alpha > -1$, $\beta=(\mu-1)/2$, which was introduced by Lascenov in [7] (see, also, [2, pp. 155—156]). It is interesting to say that these polynomials have been again „discovered” by J. Radecki ([9]).

2. Preliminaries

The relation between generalized Gegenbauer (monic) polynomials $W_k^{(\alpha,\beta)}(x)$ and Jacobi polynomials $P_k^{(\alpha,\beta)}(x)$ is given by

$$(2.1) \quad W_{2k}^{(\alpha,\beta)}(x) = \frac{k!}{(k+\alpha+\beta+1)_k} P_k^{(\alpha,\beta)}(2x^2-1),$$

$$(2.2) \quad W_{2k+1}^{(\alpha,\beta)}(x) = \frac{k!}{(k+\alpha+\beta+2)_k} x P_k^{(\alpha,\beta+1)}(2x^2-1).$$

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Notice that

$$(2.3) \quad W_{2k+1}^{(\alpha, \beta)}(x) = xW_{2k}^{(\alpha, \beta+1)}(x).$$

Using relations (2.1), (2.2) and Jacobi polynomials theory, it is possible to obtain a set of relations for generalized Gegenbauer polynomials. For example, the recurrence relation is

$$(2.4) \quad \begin{aligned} W_{k+1}^{(\alpha, \beta)}(x) &= xW_k^{(\alpha, \beta)}(x) - \Lambda_k W_{k-1}^{(\alpha, \beta)}(x), \quad k = 0, 1, \dots, \\ W_{-1}^{(\alpha, \beta)}(x) &= 0, \quad W_0^{(\alpha, \beta)}(x) = 1, \end{aligned}$$

where

$$\Lambda_{2k} = \frac{k(k+\alpha)}{(2k+\alpha+\beta)(2k+\alpha+\beta+1)}, \quad \Lambda_{2k-1} = \frac{(k+\beta)(k+\alpha+\beta)}{(2k+\alpha+\beta-1)(2k+\alpha+\beta)},$$

for $k=1, 2, \dots$, except when $\alpha+\beta=-1$; then $\Lambda_1=(\beta+1)/(\alpha+\beta+2)$.

Starting from the relation ([1, p. 782])

$$(2k+\alpha+\beta+3)P_{k+1}^{(\alpha, \beta)}(x) = (k+\alpha+\beta+2)P_{k+1}^{(\alpha+1, \beta)}(x) - (k+\beta+1)P_k^{(\alpha+1, \beta)}(x)$$

and using (2.1), we obtain

$$(2.5) \quad W_{2k+2}^{(\alpha+1, \beta)}(x) = W_{2k+2}^{(\alpha, \beta)}(x) + \frac{(k+1)(k+\beta+1)}{(2k+\alpha+\beta+3)(2k+\alpha+\beta+2)} W_{2k}^{(\alpha+1, \beta)}(x).$$

In the sequel, the following formulas will be necessary

$$(2.6) \quad \begin{aligned} \|W_{2k}^{(\alpha, \beta)}\|^2 &= \int_{-1}^1 W_{2k}^{(\alpha, \beta)}(x)^2 p(x) dx = \frac{k!}{(k+\alpha+\beta+1)_k} B(k+\alpha+1, k+\beta+1), \\ \|W_{2k+1}^{(\alpha, \beta)}\|^2 &= \|W_{2k}^{(\alpha, \beta+1)}\|^2 = \frac{k!}{(k+\alpha+\beta+2)_k} B(k+\alpha+1, k+\beta+2), \\ W_{2k}^{(\alpha, \beta)}(1) &= \frac{(\alpha+1)_k}{(k+\alpha+\beta+1)_k}, \quad W_{2k+1}^{(\alpha, \beta)}(1) = W_{2k}^{(\alpha, \beta+1)}(1) = \frac{(\alpha+1)_k}{(k+\alpha+\beta+2)_k} \end{aligned}$$

and, also

$$(2.7) \quad \begin{aligned} W_{2k}^{(\alpha, \beta)}(x) &= W_{2k}^{(\alpha, \beta)}(1) {}_2F_1(-k, k+\alpha+\beta+1; \alpha+1; 1-x^2), \\ W_{2k+1}^{(\alpha, \beta)}(x) &= xW_{2k+1}^{(\alpha, \beta)}(1) {}_2F_1(-k, k+\alpha+\beta+2; \alpha+1; 1-x^2), \end{aligned}$$

where ${}_2F_1$ is the hypergeometric function.

Lemma 2.1. Let $h_k = \|W_k^{(\alpha, \beta)}\|^2$. Then the identities

$$(2.8) \quad S_0^{(n)}(x) = \sum_{k=0}^n \frac{W_{2k}^{(\alpha, \beta)}(x) W_{2k}^{(\alpha, \beta)}(1)}{h_{2k}} = A_n^{(\alpha, \beta)} W_{2n}^{(\alpha+1, \beta)}(x)$$

and

$$(2.9) \quad S_1^{(n)}(x) = \sum_{k=0}^n \frac{W_{2k+1}^{(\alpha, \beta)}(x) W_{2k+1}^{(\alpha, \beta)}(1)}{h_{2k+1}} = A_n^{(\alpha, \beta+1)} W_{2n+1}^{(\alpha+1, \beta)}(x)$$

hold, where

$$(2.10) \quad A_n^{(\alpha, \beta)} = \frac{W_{2n}^{(\alpha, \beta)}(1)}{h_{2n}} = \frac{\Gamma(2n + \alpha + \beta + 2)}{n! \Gamma(\alpha + 1) \Gamma(n + \beta + 1)}$$

Proof. Our proof is based on induction. Note that (2.8) is correct for $n=0$. Now, let us suppose that (2.8) holds true for some n . Then we get

$$S_0^{(n+1)}(x) = A_n^{(\alpha, \beta)} W_{2n}^{(\alpha+1, \beta)}(x) + \frac{W_{2n+2}^{(\alpha, \beta)}(x) W_{2n+2}^{(\alpha, \beta)}(1)}{h_{2n+2}}$$

On the basis of (2.10), we obtain

$$\begin{aligned} S_0^{(n+1)}(x) &= A_{n+1}^{(\alpha, \beta)} \left(\frac{A_n^{(\alpha, \beta)}}{A_{n+1}^{(\alpha, \beta)}} W_{2n}^{(\alpha+1, \beta)}(x) + W_{2n+2}^{(\alpha, \beta)}(x) \right) \\ &= A_{n+1}^{(\alpha, \beta)} \left(\frac{(n+1)(n+\beta+1)}{(2n+\alpha+\beta+3)(2n+\alpha+\beta+2)} W_{2n}^{(\alpha+1, \beta)}(x) + W_{2n+2}^{(\alpha, \beta)}(x) \right). \end{aligned}$$

Finally, according to (2.5), we find $S_0^{(n+1)}(x) = A_{n+1}^{(\alpha, \beta)} W_{2n+2}^{(\alpha+1, \beta)}(x)$, showing that identity (2.8) holds true also for $n := n+1$. The identity (2.9) is simply proved if we multiply identity (2.8), for $\beta := \beta+1$, by x and using relations (2.3) and (2.6).

3. Approximations with constraint

Following Milovanović and Wrigge [8], we introduce two families of real functions, viz.

$$F_e = \{f : f(-x) = f(x), f(1) = 0, f \in L^2[-1, 1]\}$$

and

$$F_o = \{f : f(-x) = -f(x), f(1) = 0, f \in L^2[-1, 1]\},$$

where $L^2[-1, 1] = L_p^2[-1, 1]$, $p(x) = |x|^\mu (1-x^2)^\alpha$, $\alpha, \mu > -1$, and

$$(3.1) \quad (f, g) = \int_{-1}^1 f(x) g(x) p(x) dx \quad (f, g \in L^2[-1, 1]).$$

Let further \mathcal{P}_m be the set of all real polynomials of degree at most m and such that the polynomials belong to the set F_e if m is even and to the set F_o if m is odd.

In this section, we will give the least squares approximation Φ_{2n} (or Φ_{2n+1}) for the function $f \in F_e$ (or F_o) in the class \mathcal{P}_{2n} (or \mathcal{P}_{2n+1}), with respect to the norm $\|f\| = ((f, f))^{1/2}$, where the inner product (\cdot, \cdot) is defined by (3.1). For this approximation we have

$$(3.2) \quad \min_{\Phi \in \mathcal{P}_{2n}} \|f - \Phi\| = \|f - \Phi_{2n}\| \quad \text{when } f \in F_e,$$

or

$$(3.3) \quad \min_{\Phi \in \mathcal{P}_{2n+1}} \|f - \Phi\| = \|f - \Phi_{2n+1}\| \quad \text{when } f \in F_o.$$

Applying the same method as in the paper [8] and using the relations (2.6) — (2.10), we can prove two following theorems:

Theorem 3.1. *If $f \in F_e$ then the least squares approximation in the class \mathcal{P}_{2n} is given by*

$$(3.4) \quad \Phi_{2n}(x) = \sum_{i=1}^n d_{n,i} (1-x^2)^i,$$

where

$$d_{n,i} = \frac{(-1)^i}{\Gamma(\alpha+i+2)} \sum_{k=0}^n (f, W_{2k}^{(\alpha,\beta)}) \frac{\Gamma(2k+\alpha+\beta+2)}{k! \Gamma(k+\beta+1)} S_{k,i}^{(\alpha,\beta)}$$

and

$$S_{k,i}^{(\alpha,\beta)} = \begin{cases} -\binom{n}{i} (\alpha+1) (n+\alpha+\beta+2)_i, & k < i, \\ \binom{k}{i} (\alpha+i+1) (k+\alpha+\beta+1)_i - \binom{n}{i} (\alpha+1) (n+\alpha+\beta+2)_i, & k \geq i. \end{cases}$$

Theorem 3.2. *If $f \in F_o$, then the least squares approximation in the class \mathcal{P}_{2n+1} is given by*

$$(3.5) \quad \Phi_{2n+1}(x) = x \sum_{i=1}^n b_{n,i} (1-x^2)^i,$$

where

$$b_{n,i} = \frac{(-1)^i}{\Gamma(\alpha+i+2)} \sum_{k=0}^n (f, W_{2k+1}^{(\alpha,\beta)}) \frac{\Gamma(2k+\alpha+\beta+3)}{k! \Gamma(k+\beta+2)} g_{k,i}^{(\alpha,\beta)}$$

and

$$g_{k,i}^{(\alpha,\beta)} = \begin{cases} -\binom{n}{i} (n+\alpha+\beta+3)_i (\alpha+1), & k < i, \\ \binom{k}{i} (k+\alpha+\beta+2)_i (\alpha+i+1) - \binom{n}{i} (n+\alpha+\beta+3)_i (\alpha+1), & k \geq i. \end{cases}$$

Let $\tilde{\Phi}_{2n+q}(x)$ be the least squares approximation without constraint given by (see, e. g., [10, pp. 50—51])

$$\tilde{\Phi}_{2n+q}(x) = \sum_{k=0}^n \frac{(f, W_{2k+q}^{(\alpha,\beta)})}{h_{2k+q}} W_{2k+q}^{(\alpha,\beta)}(x),$$

where $q=0$ or $q=1$. It is easy to see that the approximation with constraint $\Phi_{2n+q}(x)$ turns out to be the truncated expansion in generalized Gegenbauer polynomials with a multiple of $S_q^{(n)}(x) (= A_n^{(\alpha,\beta+q)} W_{2n+q}^{(\alpha+1,\beta)}(x))$ added to satisfy the constraint at $x=1$, i. e.,

$$(3.7) \quad \Phi_{2n+q}(x) = \tilde{\Phi}_{2n+q}(x) - \tilde{\Phi}_{2n+q}(1) \frac{S_q^{(n)}(x)}{S_q^{(n)}(1)},$$

where $q=0$ or $q=1$.

4. General case

In the general case, when f is neither an even nor an odd function, but $f \in L^2[-1, 1]$ and $f(-1) = f(1) = 0$, then the least squares approximation ψ_m (in the class of real polynomials of degree $\leq m$), which satisfies the conditions $\psi_m(-1) = \psi_m(1) = 0$, is given by

$$(4.1) \quad \psi_m(x) = \Phi_{2n}(x) + \Phi_{2n+1}(x) \quad \text{when } m = 2n + 1$$

and

$$\psi_m(x) = \Phi_{2n}(x) + \Phi_{2n-1}(x) \quad \text{when } m = 2n,$$

where Φ_{2n} and Φ_{2n+1} are the solutions of (3.2) and (3.3). This can be seen by writing

$$(4.2) \quad f(x) = f^{(e)}(x) + f^{(o)}(x) = \frac{1}{2}(f(x) + f(-x)) + \frac{1}{2}(f(x) - f(-x)).$$

It is of some interest to compare approximations without constraint with our approximations with constraint.

In the set of all real polynomials Π_{2n+1} of degree at most $m = 2n + 1$, the least squares approximation without constraint for the function f , given by (4.2), can be represented in the form

$$(4.3) \quad \tilde{\psi}_{2n+1}(x) = \tilde{\Phi}_{2n}(x) + \tilde{\Phi}_{2n+1}(x),$$

where the even and odd parts, $\tilde{\Phi}_{2n}$ and $\tilde{\Phi}_{2n+1}$, are given by (3.6) for $q = 0$ and $q = 1$ respectively.

According to (4.1), (3.7) and (4.3) we have that the corresponding least squares approximation with constraint is given by

$$(4.4) \quad \psi_{2n+1}(x) = \tilde{\psi}_{2n+1}(x) - h(x),$$

where

$$h(x) = \tilde{\Phi}_{2n}(1) \frac{S_0^{(n)}(x)}{S_0^{(n)}(1)} + \tilde{\Phi}_{2n+1}(1) \frac{S_1^{(n)}(x)}{S_1^{(n)}(1)}.$$

Assuming $f \in L^2[-1, 1]$ and $f(-1) = f(1) = 0$, define

$$(4.5) \quad D^* = \min_{\psi \in \mathcal{P}_{2n} \cup \mathcal{P}_{2n+1}} \|f - \psi\|^2 = \|f - \psi_{2n+1}\|^2$$

and

$$(4.6) \quad \tilde{D}^* = \min_{\psi \in \Pi_{2n+1}} \|f - \psi\|^2 = \|f - \tilde{\psi}_{2n+1}\|^2.$$

We note that $\mathcal{P}_{2n} \cup \mathcal{P}_{2n+1} \subset \Pi_{2n+1}$.

According to (4.5), (4.4) and (4.6) we have

$$D^* = (f - \psi_{2n+1}, f - \psi_{2n+1}) = \tilde{D}^* + 2(f - \tilde{\psi}_{2n+1}, h) + (h, h).$$

It is easy to show that $(f - \tilde{\Psi}_{2n+1}, h) = 0$ and

$$(h, h) = \frac{(\tilde{\Phi}_{2n}(1))^2}{S_0^{(n)}(1)} + \frac{(\tilde{\Phi}_{2n+1}(1))^2}{S_1^{(n)}(1)}.$$

So, we obtain

$$D^* - \tilde{D}^* = \frac{(\tilde{\Phi}_{2n}(1))^2}{A_n^{(\alpha, \beta)} W_{2n}^{(\alpha+1, \beta)}(1)} + \frac{(\tilde{\Phi}_{2n+1}(1))^2}{A_n^{(\alpha, \beta+1)} W_{2n+1}^{(\alpha+1, \beta)}(1)}.$$

Thus we see that the difference $D^* - \tilde{D}^*$ is a linear combination of the squares of errors $\tilde{\Phi}_{2n}(1) - f^{(e)}(1) = \tilde{\Phi}_{2n}(1)$ and $\tilde{\Phi}_{2n+1}(1) - f^{(o)}(1) = \tilde{\Phi}_{2n+1}(1)$, where $f^{(e)}(1) = f^{(o)}(1) = f(1) = 0$.

A similar result can be obtained if we consider our approximations in the case $m = 2n$.

5. Examples

As may be seen from Theorems 3.1 and 3.2, a main difficulty when calculating the least squares approximations is to achieve high — precision values of the inner products $(f, W_k^{(\alpha, \beta)})$.

An appropriate numerical method for the determination of these inner products is the application of Gauss — Christoffel quadrature with the generalized Gegenbauer weight. The parameters of these quadratures can be calculated from corresponding Jacobi matrix by using QR — algorithm ([4], [5]). The elements of Jacobi matrix are determined by three — term recurrence relation (2.4).

A better approach is use of Gauss — Lobatto quadratures with the same weight ([6]). Namely, we can achieve higher accuracy than with the above one, using the same number of knots because of the conditions $f(-1) = f(1) = 0$.

Example 5.1. $f(x) = \cos(\pi x/2)$, $x \in [-1, 1]$.

In this case the approximating polynomial is given by (3.4), where the coefficients $d_{n,i}$ ($n = 1, 2, 3, 4$) are displayed in Table 5.1. The corresponding absolute errors

$$e_n = \max_{-1 \leq x \leq 1} |f(x) - \Phi_{2n}(x)| \quad (n = 1, \dots, 4)$$

are given, too. Numbers in parenthesis indicate decimal exponents.

Table 5.1

n	i	$\mu = 0$	$\alpha = -0.5$	$\mu = -0.5$	$\alpha = -0.5$
		$d_{n,i}$	e_n	$d_{n,i}$	e_n
1	1	0.962270459871	3.84 (—2)	0.979346973677	4.60 (—2)
2	1	0.777230028062	7.47 (—4)	0.776199638179	9.00 (—4)
	2	0.222048518171		0.223462069048	
3	1	0.785557128489	8.05 (—6)	0.785579574340	9.69 (—6)
	2	0.195401796805		0.195322260565	
	3	0.019033372405		0.019094870042	
4	1	0.785396470018	5.46 (—8)	0.785396215336	6.56 (—8)
	2	0.196365747628		0.196367406887	
	3	0.017380885279		0.017377843941	
	4	0.000856845176		0.000858513050	

We can notice that for $\beta = -0.5$ ($\mu = 0$) and $\alpha = \lambda - 1/2$, the problem is reduced to the Gegenbauer's case which is considered in [8].

Example 5.2. The error function

$$\operatorname{erf}(x) = \frac{2}{\sqrt{\pi}} \int_0^x e^{-t^2} dt$$

is an odd one. In order to approximate this function by means of Theorem 3.2, let us define a new function f by

$$f(x) = \operatorname{erf}(ax) - \operatorname{erf}(a)x \quad (|x| \leq 1),$$

where a is a positive constant. For the function $f \in F_0$, according to Theorem 3.2, the approximating polynomial is given by (3.5). Taking into account the above, we obtain the approximation of the form

$$(5.1) \quad \operatorname{erf}(ax) \cong x \left(\operatorname{erf}(a) + \sum_{i=1}^n b_{n,i} (1-x^2)^i \right).$$

The coefficients $b_{n,i}$ ($i = 1, \dots, n$) and $b_{n,0} = \operatorname{erf}(a)$, for $a = 0.5$ and $\alpha = \beta = -0.5$ and 0.5 , are given in Table 5.2.

Table 5.2

		$\alpha = \beta = -0.5$	$\alpha = \beta = 0.5$
n	i	$d_{n,i}$	$d_{n,i}$
6	0	0.52049987781305	0.52049987781305
	1	0.04055429417069	0.04055429417017
	2	0.00295376508012	0.00295376508471
	3	0.00017297426446	0.00017297425369
	4	0.00000832270120	0.00000832270347
	5	0.00000033642194	0.00000033643767
	6	0.00000001309590	0.00000001308425

Table 5.3 contains the maximum values of the absolute error, i.e.

$$\max_{|x| \leq 1} |f(x) - \Phi_{2n+1}(x)|,$$

for $a = 0.5$ (0.5) 2, when $\alpha = \beta = -0.5$ and $\alpha = \beta = 0.5$.

It can be seen that the increase of the parameter a produces an increase of the error.

The obtained approximation for the error function given by (5.1) is very efficient for usage because it requires a small number of arithmetic operations, i.e. n additions, one subtraction, and $n+2$ multiplications, assuming that the Horner's scheme is used.

Table 5.3

$a=0.5$			$a=1.0$	
n	$\alpha=\beta=-0.5$	$\alpha=\beta=0.5$	$\alpha=\beta=-0.5$	$\alpha=\beta=0.5$
1	2.6 (—4)	2.9 (—4)	6.1 (—3)	7.1 (—3)
2	3.6 (—6)	4.7 (—6)	3.4 (—4)	4.5 (—4)
3	4.3 (—8)	6.4 (—8)	1.6 (—5)	2.4 (—5)
4	4.3 (—10)	7.2 (—10)	6.1 (—7)	1.1 (—6)
5	3.7 (—12)	7.0 (—12)	2.1 (—8)	4.2 (—8)
6	2.9 (—14)	5.9 (—14)	6.4 (—10)	1.4 (—9)
$a=1.5$			$a=2.0$	
n	$\alpha=\beta=-0.5$	$\alpha=\beta=0.5$	$\alpha=\beta=-0.5$	$\alpha=\beta=0.5$
1	3.1 (—2)	3.6 (—2)	7.7 (—2)	8.9 (—2)
2	3.6 (—3)	4.9 (—3)	1.6 (—2)	2.1 (—2)
3	3.7 (—4)	5.7 (—4)	2.7 (—3)	4.1 (—3)
4	3.2 (—5)	5.7 (—5)	4.0 (—4)	6.9 (—4)
5	2.5 (—6)	4.8 (—6)	5.2 (—5)	1.1 (—4)
6	1.7 (—7)	3.6 (—7)	6.1 (—6)	1.4 (—5)

All computations, which include the error function, have been performed using rational approximation to this function given in [3].

All calculations were performed in double precision arithmetic on a HONEYWELL DPS 6/92 computer.

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**SREDNJE-KVADRATNA APROKSIMACIJA SA OGRANIČENJEM:
GENERALISANI GEGENBAUEROV SLUČAJ****Gradimir V. Milovanović i Milan A. Kovačević**

U radu se razmatra srednje-kvadratna aproksimacija funkcije $f \in L^2[-1, 1]$ pod uslovom da je $f(-1) = f(1) = 0$. Korisćenjem generalisane Gegenbauerove težinske funkcije dobijena su uopštenja nekih rezultata Milovanovića i Wriggea [8]. Rezultati su upoređeni sa odgovarajućom srednje-kvadratnom aproksimacijom bez ograničenja. U poslednjem odeljku rada data su dva numerička primera.