Continuous window functions for NFFT

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Dedicated to Jürgen Prestin on the occasion of his 60th birthday

In this paper, we study the error behavior of the nonequispaced fast Fourier transform (NFFT). This approximate algorithm is mainly based on the convenient choice of a compactly supported window function. Here we consider the continuous Kaiser–Bessel, continuous exp-type, sinh-type, and continuous cosh-type window functions with the same support and same shape parameter. We present novel explicit error estimates for NFFT with such a window function and derive rules for the optimal choice of the parameters involved in NFFT. The error constant of a window function depends mainly on the oversampling factor and the truncation parameter. For the considered continuous window functions, the error constants have an exponential decay with respect to the truncation parameter.

Key words: nonequispaced fast Fourier transform, NFFT, error estimate, oversampling factor, truncation parameter, continuous window function with compact support, Kaiser–Bessel window function.

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1 Introduction

The nonequispaced fast Fourier transform (NFFT), see [7, 6, 21, 17] and [16, Chapter 7] is an important generalization of the fast Fourier transform (FFT). The window-based approximation leads to the most efficient algorithms under different approaches [8, 19]. Recently a new class of window functions were suggested in [3] and asymptotic error estimates are given in [4]. After [21], the similarities of the window-based algorithms for NFFT became clear. Recently we have analyzed the window-based NFFT used so far and presented the related error estimates in [18]. Now we continue this investigation and present new error estimates for the some other window functions. More precisely,

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we consider the continuous Kaiser–Bessel window function and two close relatives of the sinh-type window function, namely the continuous exp-type and cosh-type window functions. All these window functions have the same support and the same shape parameter. We show that these window functions are very useful for NFFT, since they produce very small errors.

In this paper, we present novel explicit error estimates (2.3) with so-called error constants (2.1). The error constants of NFFT are defined by values of the Fourier transform of the window function. We show that an upper bound of (2.1) depends only on the oversampling factor $\sigma > 1$ and the truncation parameter $m \ge 2$ and decreases with exponential rate with respect to m. In numerous applications of NFFT, one uses quite often an oversampling factor $\sigma \in \left[\frac{5}{4}, 2\right]$ and a truncation parameter $m \in \{2, 3, \ldots, 6\}$. Therefore we will assume that $\sigma \ge \frac{5}{4}$.

The outline of the paper is as follows. In Section 2 we introduce the set Φ_{m,N_1} of continuous, even window functions with support $\left[-\frac{m}{N_1}, \frac{m}{N_1}\right]$, where $m \in \mathbb{N} \setminus \{1\}$ and $N_1 = \sigma N \in 2\mathbb{N}$ (with $N \in 2\mathbb{N}$ and $2m \ll N_1$) are fixed. We emphasize that a continuous window function $\varphi \in \Phi_{m,N_1}$ tends to zero at the endpoints $\pm \frac{m}{N_1}$ of its compact support $\left[-\frac{m}{N_1}, \frac{m}{N_1}\right] \subset \left[-\frac{1}{2}, \frac{1}{2}\right]$. In Section 3 we show that the simple rectangular window function (3.1) is not convenient for NFFT.

The main results of this paper are contained in Sections 4-7. For the first time, we present explicit estimates of the error constants (2.1) for fixed truncation parameter m and oversampling factor σ . In Section 4, we derive explicit error estimates for the continuous Kaiser-Bessel window function (4.1). In comparison, we show that the popular standard Kaiser-Bessel window function (4.13) has a similar error behavior as (4.1). A very useful continuous window function is the sinh-type window function (5.1) which is handled in Section 5.

The main drawback for the numerical analysis of the exp-type and cosh-type window function is the fact that an explicit Fourier transform of this window function is unknown. In Sections 6 and 7, we develop a new technique. We split the continuous exp-type/coshtype window function into a sum $\psi + \rho$, where the Fourier transform of the compactly supported function ψ is explicitly known and where the compactly supported function ρ has small magnitude. Here we use the fact that both window functions (6.1) and (7.1) are close relatives of the sinh-type window function (5.1) which was introduced by the authors in [18]. The Fourier transform of ρ is explicitly estimated for small as well as large frequencies, where σ and m are fixed. We present many numerical results so that the error constants of the different window functions can be easily compared. After this investigation, we favor the use of a continuous window function with small error constant, which can be very fast computed, such as the sinh-type, standard/continuous exp-type, or continuous cosh-type window function.

2 Continuous window functions for NFFT

Let $\sigma > 1$ be an oversampling factor. Assume that $N \in 2\mathbb{N}$ and $N_1 := \sigma N \in 2\mathbb{N}$ are given. For fixed truncation parameter $m \in \mathbb{N} \setminus \{1\}$ with $2m \ll N_1$, we introduce the

open interval $I := \left(-\frac{m}{N_1}, \frac{m}{N_1}\right)$ and the set Φ_{m,N_1} of all continuous window functions $\varphi : \mathbb{R} \to [0, 1]$ with following properties:

- Each window function φ is even, has the support \overline{I} , and is continuous on \mathbb{R} .
- Each restricted window function $\varphi|_{[0, m/N_1]}$ is decreasing with $\varphi(0) = 1$.
- For each window function φ , the Fourier transform

$$\hat{\varphi}(v) := \int_{I} \varphi(t) \,\mathrm{e}^{-2\pi\mathrm{i}\,vt} \,\mathrm{d}t = 2 \,\int_{0}^{m/N_{1}} \varphi(t)\,\cos(2\pi\,vt) \,\mathrm{d}t$$

is positive for all $v \in [-N/2, N/2]$.

Examples of continuous window functions of Φ_{m,N_1} are the (modified) B-spline window functions, algebraic window functions, Bessel window functions, and sinh-type window functions (see [18] and [16, Chapter 7]). More examples are presented in Sections 4 – 7.

In the following, we denote the torus \mathbb{R}/\mathbb{Z} by \mathbb{T} and the Banach space of continuous, 1-periodic functions by $C(\mathbb{T})$. Let $I_N := \{-N/2, \ldots, N/2 - 1\}$ be the index set for $N \in 2\mathbb{N}$.

We say that a continuous window function $\varphi \in \Phi_{m,N_1}$ is convenient for NFFT, if the $C(\mathbb{T})$ -error constant

$$e_{\sigma}(\varphi) := \sup_{N \in 2\mathbb{N}} e_{\sigma,N}(\varphi) \tag{2.1}$$

with

$$e_{\sigma,N}(\varphi) := \max_{n \in I_N} \Big\| \sum_{r \in \mathbb{Z} \setminus \{0\}} \frac{\hat{\varphi}(n+rN_1)}{\hat{\varphi}(n)} e^{2\pi i rN_1 \cdot} \Big\|_{C(\mathbb{T})}$$
(2.2)

fulfills the condition $e_{\sigma}(\varphi) \ll 1$ for conveniently chosen oversampling factor $\sigma > 1$. Now we show that the error of the nonequispaced fast Fourier transform (NFFT) with a window function $\varphi \in \Phi_{m,N_1}$ can be estimated by the error constant (2.1). The NFFT (with nonequispaced spatial data and equispaced frequencies) is an approximate, fast algorithm which computes approximately the values $p(x_j), j = 1, \ldots, M$, of any 1periodic trigonometric polynomial

$$p(x) := \sum_{k \in I_N} c_k \, \mathrm{e}^{2\pi \mathrm{i} \, kx}$$

at finitely many nonequispaced nodes $x_j \in \left[-\frac{1}{2}, \frac{1}{2}\right), j = 1, \ldots, M$, where $c_k \in \mathbb{C}$, $k \in I_N$, are given coefficients. By the properties of the window function $\varphi \in \Phi_{m,N_1}$, the 1-periodic function

$$\tilde{\varphi}(x) := \sum_{k \in \mathbb{Z}} \varphi(x+k) \,, \quad x \in \mathbb{T} \,,$$

is continuous on \mathbb{T} and of bounded variation over $\left[-\frac{1}{2}, \frac{1}{2}\right]$. Then from the Convergence Theorem of Dirichlet–Jordan (see [23, Vol. 1, pp. 57–58]), it follows that $\tilde{\varphi}$ possesses the uniformly convergent Fourier expansion

$$\tilde{\varphi}(x) = \sum_{k \in \mathbb{Z}} \hat{\varphi}(k) e^{2\pi i kx}$$

with the Fourier coefficients

$$\hat{\varphi}(k) = \int_{\mathbb{R}} \varphi(x) e^{-2\pi i kx} dx = \int_0^1 \tilde{\varphi}(x) e^{-2\pi i kx} dx.$$

We approximate the trigonometric polynomial p by the 1-periodic function

$$s(x) := \sum_{\ell \in I_{N_1}} g_\ell \, \tilde{\varphi} \left(x - \frac{\ell}{N_1} \right) \in C(\mathbb{T})$$

with the coefficients

$$g_{\ell} := \frac{1}{N_1} \sum_{k \in I_N} \frac{c_k}{\hat{\varphi}(k)} e^{2\pi i \, k \ell / N_1}, \quad \ell \in I_{N_1}.$$

The computation of the values $s(x_j)$, j = 1, ..., M, is very easy, since φ is compactly supported. The computational cost of the algorithm is $\mathcal{O}(N_1 \log N_1 + (2m+1)M)$, see [16, Algorithm 7.1] and [11] for details.

We interpret s - p as the *error function of the* NFFT which we measure in the norm of $C(\mathbb{T})$.

Theorem 2.1 Let $\sigma > 1$, $N \in 2\mathbb{N}$, and $N_1 = \sigma N \in 2\mathbb{N}$ be given. Further let $m \in \mathbb{N} \setminus \{1\}$ with $2m \ll N_1$.

Then the error function of the NFFT can be estimated by

$$\|s - p\|_{C(\mathbb{T})} \le e_{\sigma}(\varphi) \sum_{n \in I_N} |c_n|.$$
(2.3)

The proof of Theorem 2.1 is based on the equality

$$s(x) - p(x) = \sum_{n \in I_N} \sum_{r \in \mathbb{Z} \setminus \{0\}} c_n \frac{\hat{\varphi}(n+rN_1)}{\hat{\varphi}(n)} e^{2\pi i (n+rN_1)x}, \quad x \in \mathbb{T}.$$

For details of the proof see [18, Lemma 2.3] and [13, Lemma 2.1]. In order to describe the behavior of $C(\mathbb{T})$ -error constant (2.1), we have to study the Fourier transform of a window function φ with the support \overline{I} .

3 Rectangular window function

In this Section, we present a simple discontinuous window function which is not convenient for NFFT. Later we will use this discontinuous window function in Remarks 4.6 and 6.8, where we estimate the $C(\mathbb{T})$ -error constants for the standard Kaiser-Bessel window function and the original exp-type window function, respectively.

The simplest window function is the rectangular window function

$$\varphi_{\text{rect}}(x) := \begin{cases} 1 & x \in I, \\ \frac{1}{2} & x = \pm \frac{m}{N_1}, \\ 0 & x \in \mathbb{R} \setminus \bar{I}, \end{cases}$$
(3.1)

Now we show that the rectangular window function (3.1) is not convenient for NFFT. The Fourier transform of (3.1) has the form

$$\hat{\varphi}_{\text{rect}}(v) = \frac{2m}{N_1} \operatorname{sinc} \frac{2\pi m v}{N_1}, \quad v \in \mathbb{R}.$$
(3.2)

The discontinuous window function φ_{rect} doesn't belong to Φ_{m,N_1} .

Lemma 3.1 For each $n \in I_N \setminus \{0\}$, the Fourier series

$$\frac{n}{N_1} \sum_{r \in \mathbb{Z}} \frac{1}{r + \frac{n}{N_1}} e^{2\pi i r N_1 x}$$

converges pointwise to the $\frac{1}{N_1}$ -periodic function of the form

$$\frac{\pi n \mathbf{i}}{N_1} \left(1 - \mathrm{e}^{-2\pi \mathrm{i}\, n/N_1} \right)^{-1} \mathrm{e}^{-2\pi \mathrm{i}\, n\, x} \,, \quad x \in (0, \, \frac{1}{N_1}) \,.$$

Proof. For fixed $n \in I_N \setminus \{0\}$, we consider the Fourier series of the special $\frac{1}{N_1}$ -periodic function $g_n(x) := e^{-2\pi i nx}$ for $x \in (0, \frac{1}{N_1})$ with

$$g_n(0) = g_n(\frac{1}{N_1}) := \frac{1}{2} \left(1 + e^{-2\pi i n/N_1} \right).$$

For n = 0 we have $g_0(x) = 1$. Then the *r*th Fourier coefficient of g_n reads as follows

$$N_1 \int_0^{1/N_1} g_n(t) e^{-2\pi i r N_1 t} dt = \frac{1}{\pi i} \left(1 - e^{-2\pi i n/N_1} \right) \frac{1}{r + \frac{n}{N_1}}$$

for $r \in \mathbb{Z}$. By the Convergence Theorem of Dirichlet–Jordan, the Fourier series of g_n is pointwise convergent such that for each $x \in \mathbb{R}$ and $n \in I_N \setminus \{0\}$

$$g_n(x) = \frac{1}{\pi i} \left(1 - e^{-2\pi i n/N_1} \right) \sum_{r \in \mathbb{Z}} \frac{1}{r + \frac{n}{N_1}} e^{2\pi i r N_1 x}.$$

This completes the proof. \blacksquare

Lemma 3.2 The $C(\mathbb{T})$ -error constant of the rectangular window function (3.1) can be estimated by

$$0.18 < \frac{1}{2} - \frac{1}{\pi} \le e_{\sigma}(\varphi_{\text{rect}}) \le \frac{1}{2} + \frac{\pi}{4} < 1.3, \qquad (3.3)$$

i.e., the rectangular window function (3.1) is not convenient for NFFT.

Proof. By (3.2) we obtain for n = 0 and $r \in \mathbb{Z}$ that

$$\frac{\hat{\varphi}_{\text{rect}}(rN_1)}{\hat{\varphi}_{\text{rect}}(0)} = \begin{cases} 1 & r = 0, \\ 0 & r \in \mathbb{Z} \setminus \{0\} \end{cases}$$

and hence

$$\sum_{r \in \mathbb{Z} \setminus \{0\}} \frac{\hat{\varphi}_{\operatorname{rect}}(rN_1)}{\hat{\varphi}_{\operatorname{rect}}(0)} e^{2\pi \mathrm{i} rN_1 x} = 0.$$

By (3.2) we have for $n \in I_N \setminus \{0\}$ and $r \in \mathbb{Z}$, $\frac{\hat{\varphi}_{\text{rect}}(n+rN_1)}{\hat{\varphi}_{\text{rect}}(n)} = \frac{n}{n+rN_1}$. Thus we obtain by Lemma 3.1 that for $x \in (0, \frac{1}{N_1})$ and $n \in I_N \setminus \{0\}$

$$\sum_{r \in \mathbb{Z} \setminus \{0\}} \frac{\hat{\varphi}_{\text{rect}}(n+rN_1)}{\hat{\varphi}_{\text{rect}}(n)} e^{2\pi i rN_1 x} = \frac{n}{N_1} \sum_{r \in \mathbb{Z} \setminus \{0\}} \frac{1}{r + \frac{n}{N_1}} e^{2\pi i rN_1 x}$$
$$= \frac{\pi n i}{N_1} \left(1 - e^{-2\pi i n/N_1}\right)^{-1} e^{-2\pi i nx} - \frac{n}{N_1}.$$

Using $\left|1 - e^{-2\pi i n/N_1}\right| = 2 \left|\sin \frac{\pi n}{N_1}\right|$, it follows that

$$\left|\frac{\pi n \mathbf{i}}{N_{1}} \left(1 - \mathrm{e}^{-2\pi \mathbf{i} n/N_{1}}\right)^{-1} \mathrm{e}^{-2\pi \mathbf{i} nx} - \frac{n}{N_{1}}\right| \ge \frac{|n|}{N_{1}} \left(\frac{\pi}{2} |\sin \frac{\pi n}{N_{1}}|^{-1} - 1\right)$$
$$= \frac{|n|}{2N_{1}} \frac{\pi - 2 |\sin \frac{\pi n}{N_{1}}|}{|\sin \frac{\pi n}{N_{1}}|} \ge \frac{1}{2\pi} \left(\pi - 2 |\sin \frac{\pi n}{\sigma N}|\right) \ge \frac{1}{2\pi} \left(\pi - 2\right).$$

Analogously, we estimate

$$\begin{aligned} \left|\frac{\pi n \mathbf{i}}{N_1} \left(1 - \mathrm{e}^{-2\pi \mathbf{i} \, n/N_1}\right)^{-1} \mathrm{e}^{-2\pi \mathbf{i} \, nx} - \frac{n}{N_1}\right| &\leq \frac{|n|}{N_1} \left(\frac{\pi}{2} \,|\sin\frac{\pi n}{N_1}|^{-1} + 1\right) \\ &= \frac{|n|}{2N_1} \,\frac{\pi + 2 \,|\sin\frac{\pi n}{N_1}|}{|\sin\frac{\pi n}{N_1}|} &\leq \frac{1}{4} \left(\pi + 2 \,|\sin\frac{\pi n}{N_1}|\right) \leq \frac{1}{4} \left(\pi + 2\right). \end{aligned}$$

Consequently the rectangular window function is not convenient for NFFT, since the corresponding $C(\mathbb{T})$ -error constant $e_{\sigma}(\varphi_{\text{rect}})$ can be estimated by (3.3).

Lemma 3.3 The Fourier transform of rectangular window function (3.1) has the following property

$$\left(1-\frac{2}{\pi}\right)\frac{m}{N_1}\left|\operatorname{sinc}\frac{\pi m}{\sigma}\right| \le \left\|\sum_{r\in\mathbb{Z}\setminus\{0\}}\hat{\varphi}_{\operatorname{rect}}(n+rN_1)\,\mathrm{e}^{2\pi\mathrm{i}rN_1\cdot}\right\|_{C(\mathbb{T})} \le \frac{3m}{N_1}\,.\tag{3.4}$$

Proof. By (2.1) - (2.2), we obtain

$$\max_{n \in I_N} \left\| \sum_{r \in \mathbb{Z} \setminus \{0\}} \hat{\varphi}_{\operatorname{rect}}(n+rN_1) e^{2\pi \operatorname{i} rN_1 \cdot} \right\|_{C(\mathbb{T})} \ge e_{\sigma}(\varphi_{\operatorname{rect}}) \min_{n \in I_N} \left| \hat{\varphi}_{\operatorname{rect}}(n) \right|.$$

Then from Lemma 3.2 and (3.2) it follows immediately the lower estimate in (3.4). Now we show the upper estimate in (3.4). For each $n \in I_N \setminus \{0\}$, the $\frac{1}{N_1}$ -periodic Fourier series

$$f_n(x) := \sum_{r \in \mathbb{Z}} \hat{\varphi}_{\text{rect}}(n+rN_1) e^{2\pi i r N_1 x} = \frac{2m}{N_1} \sum_{r \in \mathbb{Z}} \operatorname{sinc} \frac{2\pi m (n+rN_1)}{N_1} e^{2\pi i r N_1 x}$$
$$= \frac{1}{\pi N_1} \sin \frac{2\pi m n}{N_1} \sum_{r \in \mathbb{Z}} \frac{1}{r + \frac{n}{N_1}} e^{2\pi i r N_1 x}$$

is pointwise convergent by Lemma 3.1 such that

$$f_n(x) = \begin{cases} \frac{i}{2m} \sin \frac{2\pi mn}{N_1} \left(1 - e^{-2\pi i n/N_1}\right)^{-1} e^{-2\pi i n x} & x \in \left(0, \frac{1}{N_1}\right), \\ \frac{1}{4m} \sin \frac{2\pi mn}{N_1} \cot \frac{\pi n}{N_1} & x \in \left\{0, \frac{1}{N_1}\right\}. \end{cases}$$

In the case n = 0 we have $f_0(x) := \sum_{r \in \mathbb{Z}} \operatorname{sinc}(2\pi m r) e^{2\pi i r N_1 x} = 1$. For arbitrary $x \in \mathbb{R}$ and $m \in \mathbb{N}$, it holds obviously $|\sin(2mx)| \leq 2m |\sin x|$ and so $|1 - e^{-2\pi i n/N_1}| = 2 |\sin \frac{\pi n}{N_1}|$. We obtain for $n \in I_N \setminus \{0\}$ that

$$\max_{x \in \mathbb{R}} |f_n(x)| \le \frac{1}{2N_1} \left| \sin \frac{2\pi mn}{N_1} \right| \left| \sin \frac{\pi n}{N_1} \right|^{-1} \le \frac{m}{N_1}$$

Thus for $n \in I_N \setminus \{0\}$ and $x \in [0, \frac{1}{N_1}]$, we receive

$$\sum_{\mathbf{r}\in\mathbb{Z}\setminus\{0\}}\hat{\varphi}_{\mathrm{rect}}(n+rN_1)\,\mathrm{e}^{2\pi\mathrm{i}rN_1x} = f_n(x) - \frac{2m}{N_1}\operatorname{sinc}\frac{2\pi mn}{N_1}$$

and hence

$$\left\|\sum_{r\in\mathbb{Z}\setminus\{0\}}\hat{\varphi}_{\mathrm{rect}}(n+rN_1)\,\mathrm{e}^{2\pi\mathrm{i}rN_1\cdot}\right\|_{C(\mathbb{T})}\leq\frac{3m}{N_1}$$

In the case n = 0, the above estimate is also true, since

$$\sum_{r \in \mathbb{Z} \setminus \{0\}} \hat{\varphi}_{\text{rect}}(rN_1) e^{2\pi i r N_1 x} = 0.$$

This completes the proof. \blacksquare

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4 Continuous Kaiser–Bessel window function

In the following, we consider the so-called continuous Kaiser-Bessel window function

$$\varphi_{cKB}(x) := \begin{cases} \frac{1}{I_0(\beta) - 1} \left(I_0 \left(\beta \sqrt{1 - \left(\frac{N_1 x}{m}\right)^2} \right) - 1 \right) & x \in I, \\ 0 & x \in \mathbb{R} \setminus I \end{cases}$$
(4.1)

with $b := 2\pi(1-\frac{1}{2\sigma}), \sigma > 1, \beta := bm, N \in 2\mathbb{N}$, and $N_1 \in 2\mathbb{N}$, see [13], where

$$I_0(x) := \sum_{k=0}^{\infty} \frac{1}{(k!)^2} \left(\frac{x}{2}\right)^{2k}, \quad x \in \mathbb{R},$$

denotes the modified Bessel function of first kind. Further we assume that $m \in \mathbb{N} \setminus \{1\}$ fulfills $2m \ll N_1$. We emphasize that the shape parameter β in (4.1) depends on m and σ . Obviously, $\varphi_{cKB} \in \Phi_{m,N_1}$ is a continuous window function. Note that a discontinuous version (4.13) of (4.1) (see Remark 4.6) was considered in [9], but the $C(\mathbb{T})$ -error constant of (4.13) was not determined. By [14, p. 95] and (3.2), the Fourier transform of the continuous Kaiser–Bessel window function (4.1) has the form

$$\hat{\varphi}_{cKB}(v) = \frac{2m}{(I_0(\beta) - 1) N_1} \left(\frac{\sinh\left(\beta \sqrt{1 - \left(\frac{2\pi v}{N_1 b}\right)^2}\right)}{\beta \sqrt{1 - \left(\frac{2\pi v}{N_1 b}\right)^2}} - \operatorname{sinc}\frac{2\pi m v}{N_1}\right)$$

for $|v| < \frac{N_1 b}{2\pi}$ and

$$\hat{\varphi}_{cKB}(v) = \frac{2m}{(I_0(\beta) - 1)N_1} \left(\operatorname{sinc} \left(\beta \sqrt{\left(\frac{2\pi v}{N_1 b}\right)^2 - 1} \right) - \operatorname{sinc} \frac{2\pi m v}{N_1} \right)$$

for $|v| \ge \frac{N_1 b}{2\pi}$. One can show that $\hat{\varphi}_{cKB}|_{[0, \frac{N_1 b}{2\pi})}$ is positive and decreasing such that

$$\min_{n \in I_N} \hat{\varphi}_{cKB}(n) = \hat{\varphi}_{cKB}\left(\frac{N}{2}\right) \ge \frac{2m}{\left(I_0(\beta) - 1\right)N_1} \left[\frac{\sinh\left(2\pi m\sqrt{1 - 1/\sigma}\right)}{2m\pi\sqrt{1 - 1/\sigma}} - \frac{\sigma}{\pi m}\right].$$
(4.2)

Using the scaled frequency $w = 2\pi m v / N_1$, we obtain

$$\hat{\varphi}_{\text{cKB}}\left(\frac{N_1w}{2\pi m}\right) = \frac{2m}{\left(I_0(\beta) - 1\right)N_1} \cdot \begin{cases} \left(\frac{\sinh(\sqrt{\beta^2 - w^2})}{\sqrt{\beta^2 - w^2}} - \sin c w\right) & |w| < \beta, \\ \left(\sin c(\sqrt{w^2 - \beta^2}) - \sin c w\right) & |w| \ge \beta. \end{cases}$$
(4.3)

Note that we have $\frac{N_1b}{2\pi} = N_1 - \frac{N}{2} > \frac{N}{2}$ by the special choice of b.

Lemma 4.1 For $|w| \ge \beta$ we have

$$\left|\operatorname{sinc}(\sqrt{w^2 - \beta^2}) - \operatorname{sinc} w\right| \le \frac{2\beta^2}{w^2}$$

Proof. Since the sinc-function is even, we consider only the case $w \ge \beta$. For $w = \beta$, the above inequality is true, since $|\operatorname{sinc} 0 - \operatorname{sinc} \beta| \le 1 + |\operatorname{sinc} \beta| < 2$. For $w > \beta$ we obtain

$$\begin{aligned} \left| \operatorname{sinc}(\sqrt{w^2 - \beta^2}) - \operatorname{sinc} w \right| &= \left| \frac{\sin \sqrt{w^2 - \beta^2}}{\sqrt{w^2 - \beta^2}} - \frac{\sin w}{w} \right| \\ &\leq \frac{1}{w} \left| \sin \sqrt{w^2 - \beta^2} - \sin w \right| + \left| \sin \sqrt{w^2 - \beta^2} \right| \left(\frac{1}{\sqrt{w^2 - \beta^2}} - \frac{1}{w} \right) \\ &\leq \frac{2}{w} \left| \sin \frac{1}{2} \left(\sqrt{w^2 - \beta^2} - w \right) \right| + \left| \sin \sqrt{w^2 - \beta^2} \right| \left(\frac{1}{\sqrt{w^2 - \beta^2}} - \frac{1}{w} \right). \end{aligned}$$

From

$$\sqrt{w^2 - \beta^2} - w = w \left(1 - \sqrt{1 - \frac{\beta^2}{w^2}} \right) = w \frac{1 - \left(1 - \frac{\beta^2}{w^2} \right)}{1 + \sqrt{1 - \frac{\beta^2}{w^2}}} \le \frac{\beta^2}{w}$$
(4.4)

it follows that

$$\left|\sin\frac{1}{2}\left(\sqrt{w^2-\beta^2}-w\right)\right| \le \frac{1}{2}\left(\sqrt{w^2-\beta^2}-w\right) \le \frac{\beta^2}{2w}$$

Further we receive by (4.4) that

$$\left| \sin \sqrt{w^2 - \beta^2} \right| \left(\frac{1}{\sqrt{w^2 - \beta^2}} - \frac{1}{w} \right) = \left| \sin \sqrt{w^2 - \beta^2} \right| \frac{w - \sqrt{w^2 - \beta^2}}{w \sqrt{w^2 - \beta^2}}$$

$$\leq \left| \sin \sqrt{w^2 - \beta^2} \right| \frac{\beta^2}{w^2} \leq \frac{\beta^2}{w^2} .$$

This completes the proof. \blacksquare

In our study we use the following

Lemma 4.2 Let $f : (0, \infty) \to (0, \infty)$ be a decreasing function which is integrable on each interval $[1 - |u|, \infty)$ with arbitrary $u \in (-1, 1)$. Further we extend f by f(-x) := f(x) for all x > 0.

Then for each $u \in (-1, 1)$, it follows

$$\sum_{r \in \mathbb{Z} \setminus \{0, \pm 1\}} f(u+r) \leq 2 \int_{1-|u|}^{\infty} f(t) dt,$$
$$\sum_{r \in \mathbb{Z} \setminus \{0\}} f(u+r) \leq 2 f(1-|u|) + 2 \int_{1-|u|}^{\infty} f(t) dt.$$

Proof. For arbitrary $u \in (-1, 1)$ and $r \in \mathbb{N}$, we have

$$f(u+r) \le f(r-|u|).$$
 (4.5)

Using (4.5), the following series can be estimated by

$$\sum_{r=2}^{\infty} f(u+r) \leq \sum_{r=2}^{\infty} f(r-|u|),$$
$$\sum_{r=2}^{\infty} f(u-r) \leq \sum_{r=2}^{\infty} f(r-|u|).$$

Hence it follows by the integral test for convergence of series that

$$\sum_{r \in \mathbb{Z} \setminus \{0, \pm 1\}} f(u+r) = \sum_{r=2}^{\infty} f(u+r) + \sum_{r=2}^{\infty} f(u-r) \le 2 \sum_{r=2}^{\infty} f(r-|u|)$$
$$\le 2 \int_{1}^{\infty} f(x-|u|) \, \mathrm{d}x = 2 \int_{1-|u|}^{\infty} f(t) \, \mathrm{d}t \, .$$

Hence we conclude that

r

$$\sum_{\in \mathbb{Z} \setminus \{0\}} f(u+r) \le 2 f(1-|u|) + 2 \int_{1-|u|}^{\infty} f(t) \, \mathrm{d}t \,. \quad \blacksquare$$

We illustrate Lemma 4.2 for some special functions f, which we need later.

Example 4.3 For the function $f(x) = x^{-\mu}$, x > 0, with $\mu > 1$, Lemma 4.2 provides that for each $u \in (-1, 1)$ it holds

$$\sum_{r \in \mathbb{Z} \setminus \{0\}} |u+r|^{-\mu} \le 2\left(1-|u|\right)^{-\mu} + \frac{2\left(1-|u|\right)^{1-\mu}}{\mu-1} \,. \tag{4.6}$$

Especially for $\mu = 2$, it follows that

$$\sum_{r \in \mathbb{Z} \setminus \{0\}} |u+r|^{-2} \le \frac{4-2|u|}{(1-|u|)^2} \le \frac{4}{(1-|u|)^2} \,. \tag{4.7}$$

For the function $f(x) = e^{-ax}$, x > 0, with a > 0, we obtain by Lemma 4.2 that for each $u \in (-1, 1)$,

$$\sum_{r \in \mathbb{Z} \setminus \{0\}} e^{-a |u+r|} \le \left(2 + \frac{2}{a}\right) e^{a |u| - a}.$$
(4.8)

Choosing the function $f(x) = \frac{1}{ax} e^{-\sqrt{ax}}$, x > 0, with a > 0, Lemma 4.2 implies that for each $u \in (-1, 1)$,

$$\sum_{r \in \mathbb{Z} \setminus \{0\}} \frac{1}{a |u+r|} e^{-\sqrt{a |u+r|}} \le \frac{2}{a (1-|u|)} e^{-\sqrt{a-a |u|}} + \frac{4}{a} E_1\left(\sqrt{a (1-|u|)}\right), \quad (4.9)$$

where E_1 denotes the exponential integral $E_1(x) := \int_x^\infty \frac{1}{t} e^{-t} dt, x > 0.$

Lemma 4.4 For all $n \in I_N$ we estimate

$$\sum_{r \in \mathbb{Z} \setminus \{0\}} \left| \hat{\varphi}_{cKB}(n+rN_1) \right| \le \frac{8m}{\left(I_0(\beta) - 1 \right) N_1} \,.$$

Proof. By (4.3) and Lemma 4.1 we obtain for all frequencies $|v| \ge \frac{N_1 b}{2\pi} = N_1 - \frac{N}{2}$ that

$$\left| \hat{\varphi}_{\text{cKB}}(v) \right| \le \frac{mb^2 N_1}{2 \left(I_0(bm) - 1 \right) \pi^2 v^2} \,.$$

Thus we have for each $n \in I_N$

$$\sum_{r \in \mathbb{Z} \setminus \{0\}} \left| \hat{\varphi}_{cKB}(n+rN_1) \right| \le \frac{b\beta}{2 \left(I_0(\beta) - 1 \right) N_1 \pi^2} \sum_{r \in \mathbb{Z} \setminus \{0\}} \left(r + \frac{n}{N_1} \right)^{-2} \\ \le \frac{2b\beta}{\left(I_0(\beta) - 1 \right) N_1 \pi^2} \left(1 - \frac{|n|}{N_1} \right)^{-2} \le \frac{2b\beta}{\left(I_0(\beta) - 1 \right) N_1 \pi^2} \left(1 - \frac{1}{2\sigma} \right)^{-2},$$

since by (4.7) it holds for all $n \in I_N$,

$$\sum_{r \in \mathbb{Z} \setminus \{0\}} \left(r + \frac{n}{N_1} \right)^{-2} \le 4 \left(1 - \frac{|n|}{N_1} \right)^{-2} \le 4 \left(1 - \frac{1}{2\sigma} \right)^{-2}.$$

By the special choice of $b = 2\pi \left(1 - \frac{1}{2\sigma}\right)$, we obtain the above inequality.

From Lemma 4.4 it follows immediately that for all $n \in I_N$,

$$\left\|\sum_{r\in\mathbb{Z}\setminus\{0\}}\hat{\varphi}_{cKB}(n+rN_1)e^{2\pi i rN_1}\right\|_{C(\mathbb{T})} \leq \sum_{r\in\mathbb{Z}\setminus\{0\}}\left|\hat{\varphi}_{cKB}(n+rN_1)\right| \leq \frac{8m}{(I_0(\beta)-1)N_1}.$$
(4.10)

Theorem 4.5 Let $b = 2\pi (1 - \frac{1}{2\sigma}), \sigma > 1, \beta = bm, N \in 2\mathbb{N}, and N_1 = \sigma N \in 2\mathbb{N}.$ Further $m \in \mathbb{N} \setminus \{1\}$ with $2m \ll N_1$ is given.

Then the $C(\mathbb{T})$ -error constant of the continuous Kaiser-Bessel window function (4.1) can be estimated by

$$e_{\sigma}(\varphi_{\rm cKB}) \le 16 \, m\pi \sqrt{1 - 1/\sigma} \left[e^{2\pi m \sqrt{1 - 1/\sigma}} - e^{-2\pi m \sqrt{1 - 1/\sigma}} - 4\sqrt{\sigma^2 - \sigma} \right]^{-1}.$$
 (4.11)

Note that for $\sigma \geq \frac{5}{4}$, it holds $2\pi \sqrt{1 - 1/\sigma} \geq 2\pi/\sqrt{5}$ and hence

$$e^{-2\pi m \sqrt{1-1/\sigma}} \le e^{-2\pi m/\sqrt{5}} < 0.06^m$$
. (4.12)

Proof. By the definition (2.1) of the $C(\mathbb{T})$ -error constant, it holds

$$e_{\sigma}(\varphi_{cKB}) = \sup_{N \in 2\mathbb{N}} e_{\sigma,N}(\varphi_{cKB})$$

with

$$e_{\sigma,N}(\varphi_{cKB}) = \max_{n \in I_N} \left\| \sum_{r \in \mathbb{Z} \setminus \{0\}} \frac{\hat{\varphi}_{cKB}(n+rN_1)}{\hat{\varphi}_{cKB}(n)} e^{2\pi i rN_1 \cdot} \right\|_{C(\mathbb{T})}$$

$$\leq \frac{1}{\min_{n \in I_N} \hat{\varphi}_{cKB}(n)} \max_{n \in I_N} \left\| \sum_{r \in \mathbb{Z} \setminus \{0\}} \hat{\varphi}_{cKB}(n+rN_1) e^{2\pi i rN_1 \cdot} \right\|_{C(\mathbb{T})},$$

where it holds (4.2), i.e.,

$$\begin{aligned} \hat{\varphi}_{cKB}\Big(\frac{N}{2}\Big) &\geq \frac{1}{(I_0(\beta)-1)N_1\pi\sqrt{1-1/\sigma}} \left[\sinh\left(2\pi m\sqrt{1-1/\sigma}\right) - 2\sqrt{\sigma^2 - \sigma}\right] \\ &= \frac{1}{2(I_0(\beta)-1)N_1\pi\sqrt{1-1/\sigma}} \left[e^{2\pi m\sqrt{1-1/\sigma}} - e^{-2\pi m\sqrt{1-1/\sigma}} - 4\sqrt{\sigma^2 - \sigma}\right]. \end{aligned}$$

Thus from (4.10) it follows the assertion (4.11). \blacksquare

Remark 4.6 As in [9] and [4], we consider also the standard Kaiser–Bessel window function

$$\varphi_{\mathrm{KB}}(x) := \begin{cases} \frac{1}{I_0(\beta)} I_0\left(\beta \sqrt{1 - \left(\frac{N_1 x}{m}\right)^2}\right) & x \in I, \\ \frac{1}{2I_0(\beta)} & x = \pm \frac{m}{N_1}, \\ 0 & x \in \mathbb{R} \setminus \bar{I} \end{cases}$$
(4.13)

with the shape parameter $\beta = mb = 2\pi m (1 - \frac{1}{2\sigma}), \sigma > 1, N \in 2\mathbb{N}$, and $N_1 \in 2\mathbb{N}$. Further we assume that $m \in \mathbb{N} \setminus \{1\}$ fulfills $2m \ll N_1$. This window function possesses jump discontinuities at $x = \pm \frac{m}{N_1}$ with very small jump height $I_0(\beta)^{-1}$, such that (4.13) is "almost continuous". The Fourier transform of (4.13) is even and reads by [14, p. 95] as follows

$$\hat{\varphi}_{\mathrm{KB}}(v) = \begin{cases} \frac{2m}{I_0(\beta)N_1} \frac{\sinh\left(\beta\sqrt{1-\left(\frac{2\pi v}{N_1 b}\right)^2}\right)}{\beta\sqrt{1-\left(\frac{2\pi v}{N_1 b}\right)^2}} & |v| < \frac{N_1 b}{2\pi},\\ \frac{2m}{I_0(\beta)N_1} \operatorname{sinc}\left(\beta\sqrt{\left(\frac{2\pi v}{N_1 b}\right)^2 - 1}\right) & |v| \ge \frac{N_1 b}{2\pi}. \end{cases}$$

Thus $\hat{\varphi}_{\text{KB}}|_{[0, \frac{N_1 b}{2\pi})}$ is positive and decreasing such that

$$\min_{n \in I_N} \hat{\varphi}_{\mathrm{KB}}(n) = \hat{\varphi}_{\mathrm{KB}}\left(\frac{N}{2}\right) \ge \frac{1}{I_0(\beta) N_1 \pi} \frac{\sinh\left(2\pi m \sqrt{1 - 1/\sigma}\right)}{\sqrt{1 - 1/\sigma}}$$

Splitting $\varphi_{\rm KB}$ in the form

$$\varphi_{\mathrm{KB}}(x) = \left(1 - \frac{1}{I_0(\beta)}\right) \varphi_{\mathrm{cKB}}(x) + \frac{1}{I_0(\beta)} \varphi_{\mathrm{rect}}(x), \quad x \in \mathbb{R},$$

we can estimate for all $n \in I_N$,

$$\begin{split} \|\sum_{r\in\mathbb{Z}\setminus\{0\}}\hat{\varphi}_{\mathrm{KB}}(n+rN_{1})\,\mathrm{e}^{2\pi\mathrm{i}rN_{1}\,\cdot}\|_{C(\mathbb{T})} &\leq \left(1-\frac{1}{I_{0}(\beta)}\right)\sum_{r\in\mathbb{Z}\setminus\{0\}}\left|\hat{\varphi}_{\mathrm{cKB}}(n+rN_{1})\right| \\ &+\frac{1}{I_{0}(\beta)}\,\|\sum_{r\in\mathbb{Z}\setminus\{0\}}\hat{\varphi}_{\mathrm{rect}}(n+rN_{1})\,\mathrm{e}^{2\pi\mathrm{i}rN_{1}\,\cdot}\|_{C(\mathbb{T})}\,. \end{split}$$

Then from Lemma 4.4 and (3.4) it follows that

$$\max_{n \in I_N} \left\| \sum_{r \in \mathbb{Z} \setminus \{0\}} \hat{\varphi}_{\mathrm{KB}}(n+rN_1) \,\mathrm{e}^{2\pi \mathrm{i} rN_1 \cdot} \right\|_{C(\mathbb{T})} \leq \frac{11m}{I_0(\beta) \, N_1} \,.$$

Consequently, we obtain the following estimate of the $C(\mathbb{T})$ -error constant of (4.13)

$$e_{\sigma}(\varphi_{\text{KB}}) \leq 11 \, m\pi \sqrt{1 - \frac{1}{\sigma}} \left[\sinh(2\pi m \sqrt{1 - 1/\sigma}) \right]^{-1}$$
$$= 22\pi m \sqrt{1 - \frac{1}{\sigma}} \left[e^{2\pi m \sqrt{1 - 1/\sigma}} - e^{-2\pi m \sqrt{1 - 1/\sigma}} \right]^{-1}.$$

For $\sigma \geq \frac{5}{4}$, we sustain by (4.12) that

$$e_{\sigma}(\varphi_{\text{KB}}) \le 22\pi m \sqrt{1 - \frac{1}{\sigma}} \left[e^{2\pi m \sqrt{1 - 1/\sigma}} - 0.06^m \right]^{-1}.$$
 (4.14)

5 sinh-type window function

For fixed shape parameter $\beta = bm$ with $m \in \mathbb{N} \setminus \{1\}$, $b = 2\pi \left(1 - \frac{1}{2\sigma}\right)$, and oversampling factor $\sigma \geq \frac{5}{4}$, we consider the sinh-type window function

$$\varphi_{\sinh}(x) := \begin{cases} \frac{1}{\sinh\beta} \sinh\left(\beta \sqrt{1 - (N_1 x/m)^2}\right) & x \in I, \\ 0 & x \in \mathbb{R} \setminus I. \end{cases}$$
(5.1)

Obviously, $\varphi_{\sinh} \in \Phi_{m,N_1}$ is a continuous window function which was introduced by the authors in [18] (with another shape parameter $\beta = 4m$). Let $N \ge 8$ be an even integer. For a sampling factor $\sigma \ge \frac{5}{4}$ and $b = 2\pi (1 - \frac{1}{2\sigma})$, we form $N_1 = \sigma N \in 2\mathbb{N}$ and the shape parameter $\beta = bm$, where $m \in \mathbb{N} \setminus \{1\}$ with $2m \ll N_1$.

Substituting $t = N_1 x/m$, we determine the even Fourier transform

$$\hat{\varphi}_{\sinh}(v) = \int_{\mathbb{R}} \varphi_{\sinh}(x) \,\mathrm{e}^{-2\pi\mathrm{i}\,vx} \,\mathrm{d}x = \frac{2m}{N_1 \,\sinh\beta} \,\int_0^1 \sinh\left(\beta \sqrt{1-t^2}\right) \,\cos\frac{2\pi \,mvt}{N_1} \,\mathrm{d}t \,.$$

Using the scaled frequency $w := 2\pi m v / N_1$, the Fourier transform of (5.1) reads by [14, p. 38] as follows

$$\hat{\varphi}_{\sinh}\left(\frac{N_1w}{2\pi m}\right) = \frac{\pi m\beta}{N_1 \sinh\beta} \begin{cases} (\beta^2 - w^2)^{-1/2} I_1\left(\sqrt{\beta^2 - w^2}\right) & w \in (-\beta, \beta), \\ 1/4 & w = \pm\beta, \\ (w^2 - \beta^2)^{-1/2} J_1\left(\sqrt{w^2 - \beta^2}\right) & w \in \mathbb{R} \setminus [-\beta, \beta]. \end{cases}$$
(5.2)

By the power series expansion of the modified Bessel function I_1 , we obtain for $w \in (-\beta, \beta)$,

$$(\beta^2 - w^2)^{-1/2} I_1(\sqrt{\beta^2 - w^2}) = \frac{1}{2} \sum_{k=0}^{\infty} \frac{1}{4^k \, k! \, (k+1)!} \, (\beta^2 - w^2)^k$$

such that $\hat{\varphi}_{\sinh}(v)$ is decreasing for $v \in [0, N_1(1-\frac{1}{2\sigma})]$. The frequency $v = \frac{N}{2}$ corresponds to the scaled frequency $w = \frac{\pi m}{\sigma} < \beta = 2\pi m (1-\frac{1}{2\sigma})$. Hence we have

$$\begin{split} \min_{n \in I_N} \hat{\varphi}_{\sinh}(n) &= \hat{\varphi}_{\sinh}\left(\frac{N}{2}\right) \\ &= \frac{m\pi\beta}{N_1 \sinh\beta} \left(\beta^2 - \frac{\pi^2 m^2}{\sigma^2}\right)^{-1/2} I_1\left(\sqrt{\beta^2 - \frac{\pi^2 m^2}{\sigma^2}}\right) \\ &= \frac{\beta}{2N_1 \sinh\beta} \left(1 - \frac{1}{\sigma}\right)^{-1/2} I_1\left(2\pi m \sqrt{1 - \frac{1}{\sigma}}\right). \end{split}$$

By $m \ge 2$ and $\sigma \ge \frac{5}{4}$, we maintain

$$2\pi m \sqrt{1-\frac{1}{\sigma}} \ge 4\pi \sqrt{1-\frac{1}{\sigma}} \ge x_0 := \frac{4\pi}{\sqrt{5}}.$$

By the inequality (see [1] or [18, Lemma 3.3])

$$\sqrt{2\pi x_0} e^{-x_0} I_1(x_0) \le \sqrt{2\pi x} e^{-x} I_1(x), \quad x \ge x_0,$$

we preserve for all $x \ge x_0$,

$$I_1(x) \ge \sqrt{x_0} e^{-x_0} I_1(x_0) x^{-1/2} e^x > \frac{2}{5} x^{-1/2} e^x$$

and hence

$$\hat{\varphi}_{\sinh}\left(\frac{N}{2}\right) \ge \frac{\beta}{5 N_1 \sqrt{2\pi m} \sinh\beta} \left(1 - \frac{1}{\sigma}\right)^{-3/4} e^{2\pi m \sqrt{1 - 1/\sigma}}.$$
(5.3)

Now we estimate $\hat{\varphi}_{\sinh}\left(\frac{N_1w}{2\pi m}\right)$ for $|w| > \beta = bm$. For $|w| > \beta$, $\sigma \ge \frac{5}{4}$, and $N \ge 8$, it holds

$$\frac{N_1|w|}{2\pi m} > N_1 \left(1 - \frac{1}{2\sigma}\right) = N\left(\sigma - \frac{1}{2}\right) \ge \frac{3N}{4} \ge 6$$

By an inequality of the Bessel function J_1 (see [12] or [18, Lemma 3.2]), we have for all $x \ge 6$,

$$|J_1(x)| \le \frac{1}{\sqrt{x}} \,.$$

Using (5.2), it follows that for $|w| > \beta$,

$$\begin{aligned} \left| \hat{\varphi}_{\sinh} \left(\frac{N_1 w}{2\pi m} \right) \right| &= \frac{\beta \pi m}{N_1 \sinh \beta} \left(w^2 - \beta^2 \right)^{-1/2} \left| J_1 \left(\sqrt{w^2 - \beta^2} \right) \right| \\ &\leq \frac{\beta \pi m}{N_1 \sinh \beta} \left(w^2 - \beta^2 \right)^{-3/4} \end{aligned}$$

i.e., for $|v| > N_1 (1 - \frac{1}{2\sigma})$,

$$|\hat{\varphi}_{\sinh}(v)| \le \frac{\sqrt{\pi m}}{\sqrt{2} N_1 \sinh \beta} \left(\frac{v^2}{N_1^2} - \left(1 - \frac{1}{2\sigma}\right)^2\right)^{-3/4} \left(1 - \frac{1}{2\sigma}\right).$$
(5.4)

For each $n \in I_N$ with $|n \pm N_1| > N_1 \left(1 - \frac{1}{2\sigma}\right)$, we obtain by (5.2) and $|J_1(x)| \le \frac{1}{2} |x|$ that

$$\left|\hat{\varphi}_{\sinh}(n\pm N_1)\right| \le \frac{\pi m\beta}{2N_1 \sinh\beta} \,. \tag{5.5}$$

In the case $-\frac{N}{2} + N_1 = N_1 (1 - \frac{1}{2\sigma})$, we get by (5.2) that

$$\hat{\varphi}_{\sinh}\left(-\frac{N}{2}+N_{1}\right)=\frac{\pi m\beta}{4N_{1}\sinh\beta},$$

i.e., we can use (5.5) for all $n \in I_N$. For all $n \in I_N$ and $r \in \mathbb{Z} \setminus \{0, \pm 1\}$, we have by $\sigma \geq \frac{5}{4}$ that

$$\left|\frac{n}{N_1} + r\right| \ge 2 - \frac{1}{2\sigma} \ge 2 - \frac{2}{5} = \frac{8}{5} \,.$$

Since the decreasing function $h: \left[\frac{8}{5}, \infty\right) \to \mathbb{R}$,

$$h(x) := x^{3/2} \left(x^2 - \left(1 - \frac{1}{2\sigma}\right)^2 \right)^{-3/4} = \left(1 - \left(1 - \frac{1}{2\sigma}\right)^2 x^{-2}\right)^{-3/4},$$

is bounded by

$$h\left(\frac{8}{5}\right) = \left(1 - \frac{25}{64}\left(1 - \frac{1}{2\sigma}\right)^2\right)^{-3/4} < \left(1 - \frac{25}{64}\right)^{-3/4} < \frac{3}{2},$$

we receive for all $n \in I_N$ and $r \in \mathbb{Z} \setminus \{0, \pm 1\}$,

$$\begin{aligned} \left| \hat{\varphi}_{\sinh}(n+rN_{1}) \right| &\leq \frac{\sqrt{\pi m}}{\sqrt{2} N_{1} \sinh \beta} \left(\left(r + \frac{n}{N_{1}} \right)^{2} - \left(1 - \frac{1}{2\sigma} \right)^{2} \right)^{-3/4} \\ &\leq \frac{3\sqrt{\pi m}}{2\sqrt{2} N_{1} \sinh \beta} \left| r + \frac{n}{N_{1}} \right|^{-3/2}. \end{aligned}$$

Therefore we get by (4.6) that for all $n \in I_N$,

$$\sum_{\substack{r \in \mathbb{Z} \setminus \{0, \pm 1\} \\ \leq \frac{3\sqrt{2\pi m}}{N_1 \sinh \beta}} \left| \hat{\varphi}_{\text{Bessel}}(n+rN_1) \right| \leq \frac{3\sqrt{\pi m}}{2\sqrt{2}N_1 \sinh \beta} \sum_{\substack{r \in \mathbb{Z} \setminus \{0, \pm 1\} \\ r \in \mathbb{Z} \setminus \{0, \pm 1\}}} \left| r + \frac{n}{N_1} \right|^{-3/2}$$

and hence by (5.5),

$$\sum_{r\in\mathbb{Z}\setminus\{0\}} \left|\hat{\varphi}_{\text{Bessel}}(n+rN_1)\right| \le \frac{1}{N_1\sinh\beta} \left[\pi m\beta + 3\sqrt{2\pi m} \left(1 - \frac{1}{2\sigma}\right)^{-1/2}\right].$$
 (5.6)

Then from (5.2) and (5.5) it follows that

$$\begin{split} e_{\sigma,N}(\varphi_{\sinh}) &\leq \frac{1}{\hat{\varphi}_{\sinh}(N/2)} \max_{n \in I_N} \sum_{r \in \mathbb{Z} \setminus \{0\}} \left| \hat{\varphi}_{\text{Bessel}}(n+rN_1) \right| \\ &\leq \left[5\pi m \sqrt{2\pi m} + 3 \left(1 - \frac{1}{2\sigma}\right)^{-3/2} \right] \left(1 - \frac{1}{\sigma}\right)^{3/4} e^{-2\pi m \sqrt{1 - 1/\sigma}} \,. \end{split}$$

Using $5\pi\sqrt{2\pi} < 40$, we summarize:

Theorem 5.1 Let $N \in 2\mathbb{N}$, $N \geq 8$, and $\sigma \geq \frac{5}{4}$ be given, where $N_1 = \sigma N \in 2\mathbb{N}$. Further let $m \in \mathbb{N} \setminus \{1\}$ with $2m \ll N_1$ and $\beta = 2\pi m \left(1 - \frac{1}{2\sigma}\right)$.

Then the $C(\mathbb{T})$ -error constant of the sinh-type window function (5.1) can be estimated by

$$e_{\sigma}(\varphi_{\sinh}) \leq \left[40 \, m^{3/2} + 3\left(1 - \frac{1}{2\sigma}\right)^{-3/2}\right] \left(1 - \frac{1}{\sigma}\right)^{3/4} e^{-2\pi m \sqrt{1 - 1/\sigma}},$$

i.e., the \sinh -type window function (5.1) is convenient for NFFT.

For a fixed oversampling factor $\sigma \geq \frac{5}{4}$, the $C(\mathbb{T})$ -error constant of (5.1) decays exponentially with the truncation parameter $m \geq 2$. On the other hand, the computational cost of NFFT increases with respect to m (see [16, pp. 380–381]) such that m should be not too large. For $\sigma = 2$ and m = 4, we obtain $e_{\sigma}(\varphi_{\sinh}) \leq 3.7 \cdot 10^{-6}$.

6 Continuous exp-type window function

For fixed shape parameter $\beta = bm$ with $m \in \mathbb{N} \setminus \{1\}$, $b = 2\pi \left(1 - \frac{1}{2\sigma}\right)$, and oversampling factor $\sigma \geq \frac{5}{4}$, we consider the *continuous* exp-type window function

$$\varphi_{\text{cexp}}(x) := \begin{cases} \frac{1}{\mathrm{e}^{\beta} - 1} \left(\mathrm{e}^{\beta \sqrt{1 - (N_1 x/m)^2}} - 1 \right) & x \in I, \\ 0 & x \in \mathbb{R} \setminus I. \end{cases}$$
(6.1)

Obviously, $\varphi_{\text{cexp}} \in \Phi_{m,N_1}$ is a continuous window function. Note that a discontinuous version of this window function was suggested in [3, 4]. A corresponding error estimate for the NFFT was proved in [4], where an asymptotic value of its Fourier transform was determined for $\beta \to \infty$ by saddle point integration. We present new explicit error estimates for fixed shape parameter β of moderate size.

In the following, we present a new approach to an error estimate for the NFFT with the continuous exp-type window function (6.1). Unfortunately, the Fourier transform of (6.1) is unknown analytically. Therefore we represent (6.1) as sum

$$\varphi_{\text{cexp}}(x) = \psi(x) + \rho(x), \qquad (6.2)$$

where the Fourier transform of ψ is known and where the correction term ρ has small magnitude $|\rho|$. We choose

$$\psi(x) := \begin{cases} \frac{2}{\mathrm{e}^{\beta} - 1} \sinh\left(\beta \sqrt{1 - (N_1 x)^2 / m^2}\right) & x \in I, \\ 0 & x \in \mathbb{R} \setminus I \end{cases}$$

and

$$\rho(x) := \begin{cases} \frac{1}{\mathrm{e}^{\beta} - 1} \left(\mathrm{e}^{-\beta \sqrt{1 - (N_1 x)^2 / m^2}} - 1 \right) & x \in I, \\ 0 & x \in \mathbb{R} \setminus I \end{cases}$$

m	$-\rho(0) = e^{-\beta}$
2	$8.06 \cdot 10^{-5}$
3	$7.24 \cdot 10^{-7}$
4	$6.51 \cdot 10^{-9}$
5	$5.85 \cdot 10^{-11}$
6	$5.25 \cdot 10^{-13}$

Table 6.1: Upper bounds of the function $-\rho$ for $m = 2, 3, \ldots, 6$ and $\sigma = 2$.

The Fourier transform of ψ reads as follows (see [14, p. 38])

$$\hat{\psi}(v) = \frac{2\pi m\beta}{\left(e^{\beta} - 1\right)N_{1}} \begin{cases} \frac{1}{2\pi m} \left(\left(1 - \frac{1}{2\sigma}\right)^{2} - \frac{v^{2}}{N_{1}^{2}}\right)^{-1/2} I_{1}\left(2\pi m\sqrt{\left(1 - \frac{1}{2\sigma}\right)^{2} - \frac{v^{2}}{N_{1}^{2}}}\right) & |v| < N_{1}\left(1 - \frac{1}{2\sigma}\right), \\ \frac{1}{4} & v = \pm N_{1}\left(1 - \frac{1}{2\sigma}\right), \\ \frac{1}{2\pi m}\left(\frac{v^{2}}{N_{1}^{2}} - \left(1 - \frac{1}{2\sigma}\right)^{2}\right)^{-1/2} J_{1}\left(2\pi m\sqrt{\frac{v^{2}}{N_{1}^{2}} - \left(1 - \frac{1}{2\sigma}\right)^{2}}\right) & |v| > N_{1}\left(1 - \frac{1}{2\sigma}\right). \end{cases}$$

$$(6.3)$$

Since ρ is even and $\rho|_{[0, \frac{m}{N_1}]}$ is increasing, we have $0 \ge \rho(x) \ge \rho(0) = \frac{1}{e^{\beta} - 1} (e^{-\beta} - 1) = -e^{-\beta}$.

Since ρ has small absolute values in the small support \overline{I} , the Fourier transform $\hat{\rho}$ is small too and it holds

$$|\hat{\rho}(v)| = \left| \int_{I} \rho(x) e^{-2\pi i v x} dx \right| \le \frac{2m}{N_1} e^{-\beta}.$$

Substituting $t = N_1 x/m$, we determine the Fourier transform

$$\begin{split} \hat{\varphi}_{\text{cexp}}(v) &= \int_{I} \varphi_{\text{cexp}}(x) \, \mathrm{e}^{-2\pi \mathrm{i} \, vx} \, \mathrm{d}x = \int_{I} \psi(x) \, \mathrm{e}^{-2\pi \mathrm{i} \, vx} \, \mathrm{d}x + \int_{I} \rho(x) \, \mathrm{e}^{-2\pi \mathrm{i} \, vx} \, \mathrm{d}x \\ &= \frac{2m}{(\mathrm{e}^{\beta} - 1) \, N_{1}} \Big[2 \, \int_{0}^{1} \sinh\left(\beta \sqrt{1 - t^{2}}\right) \, \cos\frac{2\pi m v t}{N_{1}} \, \mathrm{d}t \\ &+ \int_{0}^{1} \left(\mathrm{e}^{-\beta \sqrt{1 - t^{2}}} - 1 \right) \, \cos\frac{2\pi m v t}{N_{1}} \, \mathrm{d}t \Big] \, . \end{split}$$

For simplicity, we introduce the scaled frequency $w := 2\pi m v / N_1$ such that

$$\hat{\varphi}_{cexp}\left(\frac{N_1 w}{2\pi m}\right) = \frac{2m}{(e^{\beta} - 1) N_1} \left[2 \int_0^1 \sinh\left(\beta \sqrt{1 - t^2}\right) \cos(wt) dt + \int_0^1 \left(e^{-\beta \sqrt{1 - t^2}} - 1\right) \cos(wt) dt \right].$$
(6.4)

From [14, p. 38] it follows that

$$\int_{0}^{1} \sinh\left(\beta\sqrt{1-t^{2}}\right)\,\cos(wt)\,\mathrm{d}t = \frac{\pi\beta}{2} \begin{cases} (\beta^{2}-w^{2})^{-1/2}\,I_{1}\left(\sqrt{\beta^{2}-w^{2}}\right) & w\in\left(-\beta,\,\beta\right),\\ 1/4 & w=\pm\beta,\\ (w^{2}-\beta^{2})^{-1/2}\,J_{1}\left(\sqrt{w^{2}-\beta^{2}}\right) & w\in\mathbb{R}\setminus\left[-\beta,\,\beta\right], \end{cases}$$

where I_1 denotes the modified Bessel function and J_1 the Bessel function of first order. Therefore we consider

$$\hat{\rho}\left(\frac{N_1w}{2\pi m}\right) = \frac{2m}{(e^{\beta}-1)N_1} \int_0^1 \left(e^{-\beta\sqrt{1-t^2}}-1\right) \cos(wt) dt$$
$$= \frac{m}{(e^{\beta}-1)N_1} \int_{-1}^1 \left(e^{-\beta\sqrt{1-t^2}}-1\right) e^{iwt} dt$$
(6.5)

as correction term of (6.4). Now we estimate the integral

$$\mathcal{I}(w) := \int_{-1}^{1} \left(e^{-\beta \sqrt{1-t^2}} - 1 \right) e^{i w t} dt$$
(6.6)

by complex contour integrals.

Lemma 6.1 For each $w \in \mathbb{R}$ with $|w| \ge \beta = bm, m \in \mathbb{N} \setminus \{1\}$, we have

$$|\mathcal{I}(w)| \le \left(2 + \frac{2}{e}\right) \beta \sqrt[4]{5} |w|^{-5/4} + 4 |w|^{-1} e^{-\sqrt{|w|}} + \left(2 e^{-\sqrt{2}\beta} + 1\right) e^{-|w|}.$$

Proof. Here we use the same technique as in [4, Lemma 10], where the integral

$$\int_{-1}^{1} \left(e^{\beta \sqrt{1-t^2}} - 1 \right) e^{i \, wt} \, dt$$

for $|w| \ge \beta^4$ was estimated. Since $\mathcal{I}(w) = \mathcal{I}(-w)$, we consider only the case $w \ge \beta$. Let C_1 be the line segment from -1 to -1 + i. Further, C_2 is the line segment from -1 + i to 1 + i, C_3 is the line segment from 1 + i to 1, and C_4 is the line segment from 1 to -1. Since the principal square root function is holomorphic in \mathbb{C} except the nonpositive real axis, the complex function

$$g(z) := \left(\mathrm{e}^{-\beta\sqrt{1-z^2}} - 1\right)\mathrm{e}^{\mathrm{i}\,wz}$$

is holomorphic in \mathbb{C} except the set $(-\infty, -1] \cup [1, \infty)$. Hence g is holomorphic on the interior of the closed curve $C := C_1 \cup C_2 \cup C_3 \cup C_4$ and continuous in $D \cup C$, since simple calculation shows that

$$\lim_{\substack{z \to -1 \\ z \in D}} g(z) = \lim_{\substack{z \to 1 \\ z \in D}} g(z) = 0 \,.$$

Then the stronger form of Cauchy's Integral Theorem (see [5]) provides

$$\mathcal{I}(w) = \mathcal{I}_1(w) + \mathcal{I}_2(w) + \mathcal{I}_3(w) \tag{6.7}$$

with the contour integrals

$$\mathcal{I}_k(w) = \int_{C_k} \left(e^{-\beta \sqrt{1-z^2}} - 1 \right) e^{i \, wz} \, dz \,, \quad k = 1, \, 2, \, 3 \,.$$

Note that $\mathcal{I}_3(w)$ is the complex conjugate of $\mathcal{I}_1(w)$ such that $|\mathcal{I}_3(w)| = |\mathcal{I}_1(w)|$.

The line segment C_2 can be parametrized by $z = t + i, t \in [-1, 1]$ such that

$$\mathcal{I}_{2}(w) = e^{-w} \int_{-1}^{1} \left(e^{-\beta\sqrt{2-t^{2}-2it}} - 1 \right) e^{iwt} dt$$

and hence

$$|\mathcal{I}_2(w)| \le e^{-w} \int_{-1}^1 |e^{-\beta\sqrt{2-t^2-2it}}| dt + 2e^{-w}$$

Then we have $\left|e^{-\beta\sqrt{2-t^2-2it}}\right| = e^{-\beta\operatorname{Re}\sqrt{2-t^2-2it}}$. Since for $t \in [-1, 1]$,

$$|2 - t^2 - 2it| = \sqrt{t^2 + 4} \in [2, \sqrt{5}],$$

we obtain the estimate $\sqrt{2} \leq \text{Re}\sqrt{2-t^2-2it} \leq \sqrt[4]{5}$ for $t \in [-1, 1]$. Note that $\text{Re}\sqrt{2-t^2-2it} > 0$ for all $t \in [-1, 1]$. Thus we have

$$|\mathcal{I}_2(w)| \le 2 e^{-w} \left(e^{-\sqrt{2}\beta} + 1 \right).$$

A parametrization of the line segment C_1 is $z = -1 + it, t \in [0, 1]$, such that

$$\mathcal{I}_{1}(w) = i e^{-iw} \int_{0}^{1} \left(e^{-\beta \sqrt{2it+t^{2}}} - 1 \right) e^{-wt} dt$$

For $w \ge \beta > 0$, we split $\mathcal{I}_1(w)$ into the sum of two integrals

$$\begin{aligned} \mathcal{I}_{1,0}(w) &:= \mathrm{i} \, \mathrm{e}^{-\mathrm{i} w} \, \int_0^{w^{-1/2}} \left(\mathrm{e}^{-\beta \sqrt{2\mathrm{i} t + t^2}} - 1 \right) \mathrm{e}^{-wt} \, \mathrm{d} t \,, \\ \mathcal{I}_{1,1}(w) &:= \mathrm{i} \, \mathrm{e}^{-\mathrm{i} w} \, \int_{w^{-1/2}}^1 \left(\mathrm{e}^{-\beta \sqrt{2\mathrm{i} t + t^2}} - 1 \right) \mathrm{e}^{-wt} \, \mathrm{d} t \,. \end{aligned}$$

Since $2 \le |\sqrt{2i+t}| \le \sqrt[4]{5}$, $t \in [0, 1]$, the integral $\mathcal{I}_{1,0}(w)$ is bounded in magnitude by

$$|\mathcal{I}_{1,0}(w)| \le \max_{t \in [0, w^{-1/2}]} \left(1 - e^{-\beta\sqrt{t} \sqrt[4]{5}}\right) \int_0^{w^{-1/2}} e^{-wt} \, \mathrm{d}t \, .$$

From

$$\int_0^{w^{-1/2}} e^{-wt} dt = \frac{1 - e^{-\sqrt{w}}}{w} \le \frac{1}{w}$$

and

$$\max_{t \in [0, w^{-1/2}]} \left(1 - e^{-\beta \sqrt{t} \sqrt[4]{5}} \right) = 1 - e^{-\beta \sqrt[4]{5/w}} \le \beta \sqrt[4]{\frac{5}{w}}$$

it follows that

$$|\mathcal{I}_{1,0}(w)| \le (1 + e^{-1}) \beta \sqrt[4]{5} w^{-5/4}$$

Above we have used the simple inequality $1 - e^{-x} \le x$ for $x \ge 0$.

Finally we estimate the integral $\mathcal{I}_{1,1}(w)$ as follows

$$|\mathcal{I}_{1,1}(w)| \le \int_{w^{-1/2}}^{1} e^{-\beta\sqrt{t}\operatorname{Re}\sqrt{2i+t}-wt} dt + \int_{w^{-1/2}}^{1} e^{-wt} dt.$$

From $\operatorname{Re}\sqrt{2i+t} > 0$ for all $t \in [0, 1]$ and

$$\int_{w^{-1/2}}^{1} e^{-wt} dt = w^{-1} \left(e^{-\sqrt{w}} - e^{-w} \right) \le w^{-1} e^{-\sqrt{w}}$$

it follows that

$$|\mathcal{I}_{1,1}(w)| \le \int_{w^{-1/2}}^{1} e^{-wt} dt + w^{-1} e^{-\sqrt{w}} \le 2 w^{-1} e^{-\sqrt{w}}.$$

Thus we receive for $w \ge \beta$,

$$\begin{aligned} |\mathcal{I}(w)| &\leq |\mathcal{I}_{1}(w)| + |\mathcal{I}_{2}(w)| + |\mathcal{I}_{3}(w)| &= 2 |\mathcal{I}_{1}(w)| + |\mathcal{I}_{2}(w)| \\ &\leq 2 |\mathcal{I}_{1,0}(w)| + 2 |\mathcal{I}_{1,1}(w)| + |\mathcal{I}_{2}(w)| \\ &\leq \left(2 + \frac{2}{e}\right) \beta \sqrt[4]{5} w^{-5/4} + 4 w^{-1} e^{-\sqrt{w}} + \left(2 e^{-\sqrt{2}\beta} + 1\right) e^{-w}. \end{aligned}$$

This completes the proof. \blacksquare

Remark 6.2 In the proof of Lemma 6.1, it was shown that for real w with $|w| \ge \beta$, the contour integral $\mathcal{I}_1(w)$ can be estimated by

$$|\mathcal{I}_1(w)| \le \left(1 + \frac{1}{e}\right) \beta \sqrt[4]{5} |w|^{-5/4} + 2 |w|^{-1} e^{-\sqrt{|w|}}.$$

Splitting the integral $\mathcal{I}_1(w)$ in the form

$$\mathcal{I}_{1}(w) = i e^{-i w} \left(\int_{0}^{(\log \sqrt{w})/w} + \int_{(\log \sqrt{w})/w}^{1} \right) \left(e^{-\beta \sqrt{2it + t^{2}}} - 1 \right) e^{-wt} dt,$$

this yields the better estimate for $|w| \ge \beta$,

$$|\mathcal{I}_1(w)| \le \left(\frac{1}{\sqrt{2}}\beta\sqrt[4]{5}\sqrt{\log|w|}+2\right)|w|^{-3/2}.$$

m	$\gamma(m,2)$
2	$1.17 \cdot 10^{-2}$
3	$5.08\cdot10^{-3}$
4	$2.84\cdot10^{-3}$
5	$1.81 \cdot 10^{-3}$
6	$1.25 \cdot 10^{-3}$

Table 6.2: Maximum error $\gamma(m, \sigma)$ between $\int_0^1 \left(e^{-\beta \sqrt{1-t^2}} - 1 \right) \cos(wt) dt$ and $-\sin w$ for $m = 2, 3, \ldots, 6$ and $\sigma = 2$.

Lemma 6.3 Let $\beta = bm$ be given. Then for each $w \in \mathbb{R}$, it holds the estimate

$$-\operatorname{sinc} w - \gamma(m,\sigma) \le \int_0^1 \left(e^{-\beta\sqrt{1-t^2}} - 1 \right) \, \cos(wt) \, \mathrm{d}t \le -\operatorname{sinc} w + \gamma(m,\sigma)$$

with the small positive constant

$$\gamma(m,\sigma) := \int_0^1 \mathrm{e}^{-\beta\sqrt{1-t^2}} \,\mathrm{d}t \,.$$

Proof. For each $w \in \mathbb{R}$, we have

$$\begin{aligned} \left| \int_0^1 \left(\mathrm{e}^{-\beta\sqrt{1-t^2}} - 1 \right) \, \cos(wt) \, \mathrm{d}t + \operatorname{sinc} w \right| &= \left| \int_0^1 \mathrm{e}^{-\beta\sqrt{1-t^2}} \, \cos(wt) \, \mathrm{d}t \right| \\ &\leq \int_0^1 \mathrm{e}^{-\beta\sqrt{1-t^2}} \, \mathrm{d}t = \gamma(m,\sigma) \end{aligned}$$

such that

$$\max_{w \in \mathbb{R}} \left| \int_0^1 \left(e^{-\beta \sqrt{1-t^2}} - 1 \right) \, \cos(wt) \, \mathrm{d}t + \operatorname{sinc} w \right| \le \gamma(m, \sigma) \, \mathrm{d}t$$

see Table 6.2. \blacksquare

Using (6.4) and Lemma 6.3, we receive for $w \in (-\beta, \beta)$,

$$\hat{\varphi}_{cexp}\left(\frac{N_1 w}{2\pi m}\right) \geq \frac{2m}{(e^{\beta} - 1) N_1} \Big[\pi \beta \left(\beta^2 - w^2\right)^{-1/2} I_1\left(\sqrt{\beta^2 - w^2}\right)$$
(6.8)
$$-\sin w - \gamma(m, \sigma)\Big].$$
(6.9)

The function $h:\,[0,\beta]\to\mathbb{R}$ defined by

$$h(w) := \pi\beta \left(\beta^2 - w^2\right)^{-1/2} I_1\left(\sqrt{\beta^2 - w^2}\right) - \operatorname{sinc} w$$
$$= \frac{\pi\beta}{2} \sum_{s=0}^{\infty} \frac{(\beta^2 - w^2)^s}{2^{2s} s! (s+1)!} - \operatorname{sinc} w$$

has the derivative

$$h'(w) = -\pi\beta \sum_{s=1}^{\infty} \frac{(\beta^2 - w^2)^{s-1}}{2^{2s} (s-1)! (s+1)!} - \frac{\mathrm{d}}{\mathrm{d}w} \operatorname{sinc} w.$$
(6.10)

Obviously, it holds h'(0) = 0, and h'(w) < 0 for $w \in (0, \pi]$. For $w \in [\pi, \beta]$ and $m \ge 2$, we obtain

$$\left|\frac{\mathrm{d}}{\mathrm{d}w}\operatorname{sinc}w\right| \le \frac{1}{\pi^2} + \frac{1}{\pi}$$

and hence by (6.10),

$$h'(w) < -\frac{\pi\beta}{8} + \frac{1}{\pi^2} + \frac{1}{\pi} < 0.$$

Thus h(w) is decreasing with respect to $w \in [0, \beta]$. For $\beta = bm$ and $v = \frac{N_1 w}{2\pi m} \in \left(-N_1\left(1 - \frac{1}{2\sigma}\right), N_1\left(1 - \frac{1}{2\sigma}\right)\right)$, the inequality (6.8) implies

$$\hat{\varphi}_{\text{cexp}}(v) \geq \frac{2m}{(e^{\beta}-1)N_{1}} \Big[2\left(\left(1-\frac{1}{2\sigma}\right)^{2}-\frac{v^{2}}{N_{1}^{2}}\right)^{-1/2} I_{1}\left(2\pi m \sqrt{\left(1-\frac{1}{2\sigma}\right)^{2}-\frac{v^{2}}{N_{1}^{2}}}\right) \\ -\operatorname{sinc} \frac{2\pi m v}{N_{1}} - \gamma(m,\sigma) \Big] .$$

Consequently, the Fourier transform $\hat{\varphi}_{cexp}(v)$ is positive and decreasing for

$$v \in \left[0, N_1\left(1 - \frac{1}{2\sigma}\right)\right).$$

Hence we obtain

$$\begin{split} \min_{n \in I_N} \hat{\varphi}_{\text{cexp}}(n) &= \hat{\varphi}_{\text{cexp}} \Big(\frac{N}{2} \Big) \\ &\geq \frac{2m}{(\mathrm{e}^{\beta} - 1) N_1} \Big[\frac{b}{2} \left(1 - \frac{1}{\sigma} \right)^{-1/2} I_1 \Big(2\pi m \sqrt{1 - \frac{1}{\sigma}} \Big) - \operatorname{sinc} \frac{\pi m}{\sigma} - \gamma(m, \sigma) \Big] \\ &\geq \frac{2m}{(\mathrm{e}^{\beta} - 1) N_1} \Big[\frac{b}{2} \left(1 - \frac{1}{\sigma} \right)^{-1/2} I_1 \Big(2\pi m \sqrt{1 - \frac{1}{\sigma}} \Big) - 1 - \gamma(m, \sigma) \Big] \,. \end{split}$$

From $m \ge 2$ and $\sigma \ge \frac{5}{4}$ it follows that

$$2\pi m \sqrt{1 - \frac{1}{\sigma}} \ge 4\pi \sqrt{1 - \frac{1}{\sigma}} \ge x_0 := \frac{4\pi}{\sqrt{5}}$$

Hence by the inequality (see [1] or [18, Lemma 3.3])

$$\sqrt{2\pi x_0} e^{-x_0} I_1(x_0) \le \sqrt{2\pi x} e^{-x} I_1(x), \quad x \ge x_0,$$

we sustain that

$$I_1(x) \ge \sqrt{x_0} e^{-x_0} I_1(x_0) x^{-1/2} e^x, \quad x \ge x_0.$$

Thus for $x = 2\pi m \sqrt{1 - 1/\sigma}$, we get the estimate

$$I_1(2\pi m \sqrt{1-\frac{1}{\sigma}}) \ge \frac{2}{5\sqrt{2\pi m}} \left(1-\frac{1}{\sigma}\right)^{-1/4} e^{2\pi m \sqrt{1-1/\sigma}}$$

Thus we obtain the following

Lemma 6.4 Let $N \in 2\mathbb{N}$ and $\sigma \geq \frac{5}{4}$ be given, where $N_1 = \sigma N \in 2\mathbb{N}$. Further let $m \in \mathbb{N}$ with $2m \ll N_1$, $\beta = bm$, and $b = 2\pi \left(1 - \frac{1}{2\sigma}\right)$. Then we have

$$\hat{\varphi}_{\text{cexp}}\left(\frac{N}{2}\right) \geq \frac{2m}{\left(e^{\beta}-1\right)N_{1}} \left[\frac{b}{5\sqrt{2\pi m}} \left(1-\frac{1}{\sigma}\right)^{-3/4} e^{2\pi m \sqrt{1-1/\sigma}} - 1 - \gamma(m,\sigma)\right].$$

By (2.2), the constant $e_{\sigma,N}(\hat{\varphi}_{cexp})$ can be estimated as follows

$$e_{\sigma,N}(\hat{\varphi}_{\text{cexp}}) \leq \frac{1}{\hat{\varphi}_{\text{cexp}}(N/2)} \max_{n \in I_N} \Big\| \sum_{r \in \mathbb{Z} \setminus \{0\}} \hat{\varphi}_{\text{cexp}}(n+rN_1) e^{2\pi i rN_1} \Big\|_{C(\mathbb{T})},$$

where it holds by (6.2),

$$\left\|\sum_{r\in\mathbb{Z}\setminus\{0\}}\hat{\varphi}_{\operatorname{cexp}}(n+rN_1)\,\mathrm{e}^{2\pi\mathrm{i}\,rN_1\cdot}\right\|_{C(\mathbb{T})}\leq\sum_{r\in\mathbb{Z}\setminus\{0\}}\left|\hat{\psi}(n+rN_1)\right|+\sum_{r\in\mathbb{Z}\setminus\{0\}}\left|\hat{\rho}(n+rN_1)\right|.$$

Lemma 6.5 Let $N \in 2\mathbb{N}$ and $\sigma \geq \frac{5}{4}$ be given, where $N_1 = \sigma N \in 2\mathbb{N}$. Further let $m \in \mathbb{N}$ with $2m \ll N_1$, $\beta = bm$, and $b = 2\pi \left(1 - \frac{1}{2\sigma}\right)$. Then it holds for all $n \in I_N$,

$$\sum_{r \in \mathbb{Z} \setminus \{0\}} |\hat{\psi}(n+rN_1)| \le \frac{\beta}{(e^{\beta}-1)N_1} \Big[2\pi m + \frac{10}{\sqrt{2\pi m}} \Big(1 - \frac{1}{2\sigma}\Big)^{-1/2} \Big].$$

Proof. For all $n \in I_N$ and $r \in \mathbb{Z} \setminus \{0\}$ with $(n, r) \neq \left(-\frac{N}{2}, 1\right)$, it follows that

$$|n + rN_1| > N_1 - \frac{N}{2} = N_1 \left(1 - \frac{1}{2\sigma}\right)$$

and hence by $\sigma \geq \frac{5}{4}$,

$$\left|\frac{n}{N_1} + r\right| > 1 - \frac{1}{2\sigma} \ge \frac{3}{5}.$$

Thus by (6.3), we receive

$$\hat{\psi}(n+rN_1) = \frac{\beta}{(e^{\beta}-1)N_1} \left(\left(\frac{n}{N_1}+r\right)^2 - \left(1-\frac{1}{2\sigma}\right)^2 \right)^{-1/2} J_1 \left(2\pi m \sqrt{\left(\frac{n}{N_1}+r\right)^2 - \left(1-\frac{1}{2\sigma}\right)^2}\right).$$

In the case $(n,r) = \left(-\frac{N}{2},1\right)$, from (6.3) it follows that

$$\hat{\psi}\left(-\frac{N}{2}+N_{1}\right)=\hat{\psi}\left(N_{1}\left(1-\frac{1}{2\sigma}\right)\right)=\frac{\pi m\beta}{2(e^{\beta}-1)N_{1}}$$

Now we estimate the sum

$$\sum_{r \in \mathbb{Z} \setminus \{0\}} |\hat{\psi}(n+rN_1)| = |\hat{\psi}(n+N_1)| + |\hat{\psi}(n-N_1)| + \sum_{r \in \mathbb{Z} \setminus \{0,\pm 1\}} |\hat{\psi}(n+rN_1)|.$$

For $n \in I_N$ and $r \in \mathbb{Z} \setminus \{0, \pm 1\}$ we have by $\sigma \geq \frac{5}{4}$,

$$\left|\frac{n}{N_1} + r\right| \ge 2 - \frac{1}{2\sigma} \ge \frac{7}{5}$$

and therefore by $m \geq 2$,

$$2\pi m \sqrt{\left(\frac{n}{N_1} + r\right)^2 - \left(1 - \frac{1}{2\sigma}\right)^2} \geq 4\pi \sqrt{\left(2 - \frac{1}{2\sigma}\right)^2 - \left(1 - \frac{1}{2\sigma}\right)^2} \\ = 4\pi \sqrt{3 - \frac{1}{\sigma}} \geq 4\pi \sqrt{\frac{11}{5}} > 6.$$

By an inequality for the Bessel function $J_1(x)$ (see [12] or [18, Lemma 3.2]), it holds for all $x \ge 6$,

$$|J_1(x)| \le \frac{1}{\sqrt{x}}$$

such that for $x = 2\pi m \sqrt{(r + n/N_1)^2 - (1 - 1/(2\sigma))^2}$,

$$|J_1 \left(2\pi m \sqrt{\left(\frac{n}{N_1} + r\right)^2 - \left(1 - \frac{1}{2\sigma}\right)^2} \right)| \le \frac{1}{\sqrt{2\pi m}} \left(\left(\frac{n}{N_1} + r\right)^2 - \left(1 - \frac{1}{2\sigma}\right)^2 \right)^{-1/4}.$$

The decreasing function $h: \left[\frac{5}{4}, \infty\right) \to \mathbb{R}$ defined by

$$h(x) := x^{3/2} \left(x^2 - \left(1 - \frac{1}{2\sigma}\right)^2 \right)^{-3/4} = \left(1 - \left(1 - \frac{1}{2\sigma}\right)^2 x^{-2} \right)^{-3/4},$$

is bounded by

$$h\left(\frac{5}{4}\right) = \left(1 - \frac{16}{25}\left(1 - \frac{1}{2\sigma}\right)^2\right)^{-3/4} \le \left(1 - \frac{16}{25}\right)^{-3/4} = \left(\frac{5}{3}\right)^{3/2} < \frac{5}{2}.$$

Thus we receive for all $n \in I_N$ and $r \in \mathbb{Z} \setminus \{0, \pm 1\}$,

$$\left(\left(\frac{n}{N_1}+r\right)^2 - \left(1-\frac{1}{2\sigma}\right)^2\right)^{-1/2} \left|J_1\left(2\pi m\sqrt{\left(\frac{n}{N_1}+r\right)^2 - \left(1-\frac{1}{2\sigma}\right)^2}\right)\right| \\ \leq \frac{1}{\sqrt{2\pi m}} \left(\left(\frac{n}{N_1}+r\right)^2 - \left(1-\frac{1}{2\sigma}\right)^2\right)^{-3/4} \leq \frac{5}{2\sqrt{2\pi m}} \left|\frac{n}{N_1}+r\right|^{-3/2}.$$

Hence for all $n \in I_N$, we obtain by (4.6) that

$$\sum_{r \in \mathbb{Z} \setminus \{0, \pm 1\}} |\hat{\psi}(n+rN_1)| \leq \frac{5\beta}{2\sqrt{2\pi m} (e^{\beta}-1) N_1} \sum_{r \in \mathbb{Z} \setminus \{0, \pm 1\}} \left|\frac{n}{N_1} + r\right|^{-3/2} \\ \leq \frac{10\beta}{\sqrt{2\pi m} (e^{\beta}-1) N_1} \left(1 - \frac{1}{2\sigma}\right)^{-1/2}.$$

Now we estimate $\hat{\psi}(v)$ for $v = n \pm N_1$, $n \in I_N$. For $v = -\frac{N}{2} + N_1 = N_1 \left(1 - \frac{1}{2\sigma}\right)$, it holds by (6.3),

$$\hat{\psi}\left(-\frac{N}{2}+N_{1}\right) = \hat{\psi}\left(N_{1}\left(1-\frac{1}{2\sigma}\right)\right) = \frac{\pi m\beta}{2\left(e^{\beta}-1\right)N_{1}}.$$
(6.11)

For all the other $v = n \pm N_1 \neq -\frac{N}{2} + N_1$, $n \in I_N$, we have

$$|n \pm N_1| > \left(1 - \frac{1}{2\sigma}\right) N_1$$

such that by (6.3), $\hat{\psi}(n \pm N_1)$ reads as follows

$$\frac{\beta}{\left(e^{\beta}-1\right)N_{1}}\left(\left(1\pm\frac{n}{N_{1}}\right)^{2}-\left(1-\frac{1}{2\sigma}\right)^{2}\right)^{-1/2}J_{1}\left(2\pi m\sqrt{\left(1\pm\frac{n}{N_{1}}\right)^{2}-\left(1-\frac{1}{2\sigma}\right)^{2}}\right).$$

Since

$$2\pi m \sqrt{\left(1 \pm \frac{n}{N_1}\right)^2 - \left(1 - \frac{1}{2\sigma}\right)^2} > 0$$

can be small for $n \in I_N$, we estimate the Bessel function $J_1(x)$, $x \ge 0$, by Poisson's integral (see [22, p. 47])

$$|J_1(x)| = \frac{x}{\pi} \left| \int_0^\pi \cos(x \, \cos t) \, (\sin t)^2 \, \mathrm{d}t \right| \le \frac{x}{\pi} \, \int_0^\pi (\sin t)^2 \, \mathrm{d}t = \frac{x}{2} \, \mathrm{d}t$$

such that

$$\hat{\psi}(n\pm N_1)| \leq \frac{\pi m\beta}{\left(\mathrm{e}^\beta - 1\right)N_1}\,.$$

By (6.11), this estimate of $|\hat{\psi}(n \pm N_1)|$ is valid for all $n \in I_N$. This completes the proof.

Now for arbitrary $n \in I_N$, we have to estimate the series

$$\sum_{r\in\mathbb{Z}\backslash\{0\}}\left|\hat{\rho}(n+rN_1)\right|.$$

By (6.5) and Lemma 6.1, we obtain for any $v \in \mathbb{R} \setminus \{0\}$,

$$\begin{aligned} |\hat{\rho}(v)| &\leq \frac{2m}{(e^{\beta}-1)N_{1}} \Big[\Big(1+\frac{1}{e}\Big) \frac{\beta}{2\pi m} \sqrt[4]{\frac{5}{2\pi m}} \Big(\frac{|v|}{N_{1}}\Big)^{-5/4} \\ &+ \frac{N_{1}}{\pi m |v|} e^{-\sqrt{2\pi m |v|/N_{1}}} + \Big(e^{-\sqrt{2}\beta} + \frac{1}{2}\Big) e^{-2\pi m |v|/N_{1}} \Big] \end{aligned}$$

Thus we obtain that

$$\sum_{r \in \mathbb{Z} \setminus \{0\}} |\hat{\rho}(n+rN_1)| \leq \frac{2m}{(e^{\beta}-1)N_1} \Big[\Big(1+\frac{1}{e}\Big) \frac{\beta}{2\pi m} \sqrt[4]{\frac{5}{2\pi m}} S_1(n) + 2S_2(n) + \Big(e^{-\sqrt{2}\beta} + \frac{1}{2}\Big) S_3(n) \Big]$$

with

$$S_1(n) := \sum_{r \in \mathbb{Z} \setminus \{0\}} \left| \frac{n}{N_1} + r \right|^{-5/4},$$
 (6.12)

$$S_2(n) := \sum_{r \in \mathbb{Z} \setminus \{0\}} \frac{1}{2\pi m |r + n/N_1|} e^{-\sqrt{2\pi m |r + n/N_1|}}, \qquad (6.13)$$

$$S_3(n) := \sum_{r \in \mathbb{Z} \setminus \{0\}} e^{-2\pi m |r+n/N_1|}.$$
 (6.14)

The inequalities (4.6), (4.8), and (4.9) imply that

$$S_1(n) \leq \left(\frac{4\sigma}{2\sigma - 1} + 8\right) \left(1 - \frac{1}{2\sigma}\right)^{-1/4},$$
 (6.15)

$$S_2(n) \leq \frac{2\sigma}{(2\sigma - 1)\pi m} e^{-\sqrt{2\pi m - \pi m/\sigma}} + \frac{2}{\pi m} E_1(\sqrt{2\pi m - \pi m/\sigma}), \quad (6.16)$$

$$S_3(n) \leq \left(2 + \frac{1}{\pi m}\right) e^{-2\pi m + \pi m/\sigma}$$
 (6.17)

Thus we obtain the following

Lemma 6.6 Let $N \in 2\mathbb{N}$ and $\sigma \geq \frac{5}{4}$ be given, where $N_1 = \sigma N \in 2\mathbb{N}$. Further let $m \in \mathbb{N} \setminus \{1\}$ with $2m \ll N_1$, $\beta = bm$, and $b = 2\pi (1 - \frac{1}{2\sigma})$. Then for each $n \in I_N$, it holds the estimate

$$\sum_{r \in \mathbb{Z} \setminus \{0\}} |\hat{\rho}(n+rN_1)| \leq \frac{1}{(e^{\beta}-1)N_1} \Big[\Big(1+\frac{1}{e}\Big) \frac{\beta}{\pi} \sqrt[4]{\frac{5}{2\pi m}} \Big(\frac{4\sigma}{2\sigma-1}+8\Big) \Big(1-\frac{1}{2\sigma}\Big)^{-1/4} \\ + \frac{8\sigma}{(2\sigma-1)\pi} e^{-\sqrt{2\pi m-\pi m/\sigma}} + \frac{8}{\pi} E_1\Big(\sqrt{2\pi m-\pi m/\sigma}\Big) \\ + \Big(e^{-\sqrt{2}\beta}+\frac{1}{2}\Big) \Big(4m+\frac{2}{\pi}\Big) e^{-2\pi m+\pi m/\sigma} \Big].$$

Hence from Lemmas 6.5 and 6.6 it follows that

$$\max_{n \in I_N} \left\| \sum_{r \in \mathbb{Z} \setminus \{0\}} \hat{\varphi}_{\text{cexp}}(n+rN_1) \, \mathrm{e}^{2\pi \mathrm{i} rN_1 \cdot} \right\|_{C(\mathbb{T})} \le \frac{\beta}{\left(\mathrm{e}^{\beta}-1\right) N_1} \, b(m,\sigma) \tag{6.18}$$

with the constant

$$b(m,\sigma) := 2\pi m + \frac{10}{\sqrt{2\pi m}} \left(1 - \frac{1}{2\sigma}\right)^{-1/2} + \left(1 + \frac{1}{e}\right) \frac{1}{\pi} \sqrt[4]{\frac{5}{2\pi m}} \left(\frac{4\sigma}{2\sigma - 1} + 8\right) \left(1 - \frac{1}{2\sigma}\right)^{-1/4} + \frac{8\sigma}{(2\sigma - 1)\pi\beta} e^{-\sqrt{2\pi m - \pi m/\sigma}} + \frac{8}{\pi\beta} E_1\left(\sqrt{2\pi m - \pi m/\sigma}\right) + \frac{4\pi m + 2}{\beta\pi} \left(e^{-\sqrt{2}\beta} + \frac{1}{2}\right) e^{-2\pi m + \pi m/\sigma}.$$
(6.19)

Using Lemma 6.4, we obtain by

$$e_{\sigma,N}(\varphi_{\text{cexp}}) \le \frac{1}{\hat{\varphi}_{\text{cexp}}(N/2)} \max_{n \in I_N} \left\| \sum_{r \in \mathbb{Z} \setminus \{0\}} \hat{\varphi}_{\text{cexp}}(n+rN_1) \, \mathrm{e}^{2\pi \mathrm{i} rN_1 \cdot} \right\|_{C(\mathbb{T})}$$

the following

Theorem 6.7 Let $N \in 2\mathbb{N}$ and $\sigma \geq \frac{5}{4}$ be given, where $N_1 = \sigma N \in 2\mathbb{N}$. Further let $m \in \mathbb{N}$ with $2m \ll N_1$, $\beta = bm$, and $b = 2\pi \left(1 - \frac{1}{2\sigma}\right)$.

Then the $C(\mathbb{T})$ -error constant of the continuous exp-type window function (6.1) can be estimated by

$$e_{\sigma}(\varphi_{\text{cexp}}) \le \frac{\beta b(m,\sigma)}{2m} \left[\frac{b}{5\sqrt{2\pi m}} \left(1 - \frac{1}{\sigma} \right)^{-3/4} e^{2\pi m \sqrt{1 - 1/\sigma}} - 1 - \gamma(m,\sigma) \right]^{-1}$$
(6.20)

In other words, the continuous exp-type window function (6.1) is convenient for NFFT. Note that for $\sigma \in \left[\frac{5}{4}, 2\right]$ and $m \ge 2$, it holds by (6.19),

$$\frac{\beta \, b(m,\sigma)}{2m} = \pi b(m,\sigma) \left(1 - \frac{1}{2\sigma}\right) \le \frac{3\pi^2}{2} \, m + b_0 < 15 \, m + b_0$$

with $b_0 < 17$.

In order to compute the Fourier transform $\hat{\varphi}$ of window function $\varphi \in \Phi_{m,N_1}$, we approximate this window function by numerical integration. In our next numerical examples we apply the following method. Since the window function $\varphi \in \Phi_{m,N_1}$ is even and supported in $\left[-\frac{m}{N_1}, \frac{m}{N_1}\right]$, we have

$$\begin{aligned} \hat{\varphi}(v) &= \int_{\mathbb{R}} \varphi(x) e^{-2\pi i v x} dx = \frac{m}{N_1} \int_{-1}^{1} \varphi\left(\frac{m}{N_1} t\right) e^{-2\pi i m v t/N_1} dt \\ &= \frac{2m}{N_1} \int_{0}^{1} \varphi\left(\frac{m}{N_1} t\right) \cos \frac{2\pi m v t}{N_1} dt \,. \end{aligned}$$

We evaluate the last integral using a global adaptive quadrature [20] for $\hat{\varphi}(k)$, $k = 0, \ldots, N$. In general, this values can be precomputed, see [10, 2].

Remark 6.8 As in [3, 4], we can consider also the original exp-type window function

$$\varphi_{\exp}(x) := \begin{cases} \exp\left(\beta\sqrt{1 - (N_1 x/m)^2} - \beta\right) & x \in I, \\ \frac{1}{2}e^{-\beta} & x = \pm \frac{m}{N_1}, \\ 0 & x \in \mathbb{R} \setminus \bar{I} \end{cases}$$
(6.21)

with the shape parameter $\beta = mb = 2\pi m \left(1 - \frac{1}{2\sigma}\right), \sigma \geq \frac{5}{4}, N \in 2\mathbb{N}$, and $N_1 = \sigma N \in 2\mathbb{N}$. Further we assume that $m \in \mathbb{N} \setminus \{1\}$ fulfills $2m \ll N_1$. This window function possesses jump discontinuities at $x = \pm \frac{m}{N_1}$ with very small jump height $e^{-\beta}$, such that (6.21) is "almost continuous". We split (6.21) in the form

$$\varphi_{\exp}(x) = (1 - e^{-\beta}) \varphi_{\exp}(x) + e^{-\beta} \varphi_{\operatorname{rect}}(x), \quad x \in \mathbb{R},$$

with the window functions (6.1) and (3.1). Then the Fourier transform of (6.21) reads as follows

$$\hat{\varphi}_{\exp}(v) = (1 - e^{-\beta}) \hat{\varphi}_{\exp}(v) + e^{-\beta} \hat{\varphi}_{\operatorname{rect}}(v), \quad v \in \mathbb{R},$$

By Lemma 6.4 it follows that

$$\begin{aligned} \hat{\varphi}_{\exp}\left(\frac{N}{2}\right) &= \min_{n \in I_N} \hat{\varphi}_{\exp}(n) \\ &= \min_{n \in I_N} \left[\left(1 - e^{-\beta}\right) \hat{\varphi}_{\exp}(n) + e^{-\beta} \hat{\varphi}_{\operatorname{rect}}(n) \right] \\ &\geq \frac{2m}{N_1 e^{\beta}} \left[\frac{b}{5\sqrt{2\pi m}} \left(1 - \frac{1}{\sigma}\right)^{-3/4} e^{2\pi m \sqrt{1 - 1/\sigma}} - \gamma(m, \sigma) \right]. \end{aligned}$$

Using (6.18) and (3.4), we estimate for all $n \in I_N$,

$$\begin{split} & \Big\| \sum_{r \in \mathbb{Z} \setminus \{0\}} \hat{\varphi}_{\exp}(n + rN_1) \, \mathrm{e}^{2\pi \mathrm{i} rN_1 \cdot} \Big\|_{C(\mathbb{T})} \leq \left(1 - \mathrm{e}^{-\beta}\right) \sum_{r \in \mathbb{Z} \setminus \{0\}} |\hat{\varphi}_{\exp}(n + rN_1)| \\ &+ \mathrm{e}^{-\beta} \, \Big\| \sum_{r \in \mathbb{Z} \setminus \{0\}} \hat{\varphi}_{\mathrm{rect}}(n + rN_1) \, \mathrm{e}^{2\pi \mathrm{i} rN_1 \cdot} \Big\|_{C(\mathbb{T})} \\ &\leq \frac{1}{N_1 \, \mathrm{e}^{\beta}} \left(\beta \, b(m, \sigma) + 3m\right). \end{split}$$

Thus we obtain

$$e_{\sigma}(\varphi_{\exp}) \le \left(\frac{\beta b(m,\sigma)}{2m} + \frac{3}{2}\right) \left[\frac{b}{5\sqrt{2\pi m}} \left(1 - \frac{1}{\sigma}\right)^{-3/4} e^{2\pi m \sqrt{1 - 1/\sigma}} - \gamma(m,\sigma)\right]^{-1}.$$

Thus the discontinuous window function (6.21) possesses a similar $C(\mathbb{T})$ -error constant as the continuous exp-type window function (6.1).

7 Continuous cosh-type window function

For fixed shape parameter $\beta = bm = 2\pi m \left(1 - \frac{1}{2\sigma}\right)$ and oversampling factor $\sigma \geq \frac{5}{4}$, we consider the *continuous* cosh-type window function

$$\varphi_{\cosh}(x) := \begin{cases} \frac{1}{\cosh\beta - 1} \left(\cosh\left(\beta\sqrt{1 - \left(\frac{N_1x}{m}\right)^2}\right) - 1 \right) & x \in I, \\ 0 & x \in \mathbb{R} \setminus I. \end{cases}$$
(7.1)

Obviously, $\varphi_{\cosh} \in \Phi_{m,N_1}$ is a continuous window function. Note that recently a discontinuous version of this window function was suggested in [3, Remark 13]. But up to now, a corresponding error estimate for the related NFFT was unknown. Now we show that the $C(\mathbb{T})$ -error constant $e_{\sigma}(\varphi_{\cosh})$ can be estimated by a similar upper bound as $e_{\sigma}(\varphi_{\text{cexp}})$ in Theorem 6.7. Thus the window functions (6.1) and (7.1) possess the same error behavior with respect to the NFFT.

In the following, we use the same technique as in Section 6. Since the Fourier transform of (7.1) is unknown analytically, we represent (7.1) as the sum

$$\varphi_{\cosh}(x) = \psi_1(x) + \rho_1(x)$$

where the Fourier transform of ψ_1 is known and where the correction term ρ_1 has small magnitude $|\rho_1|$. We choose

$$\psi_1(x) := \begin{cases} \frac{1}{\cosh \beta - 1} \sinh \left(\beta \sqrt{1 - (N_1 x)^2 / m^2} \right) & x \in I, \\ 0 & x \in \mathbb{R} \setminus I \end{cases}$$

and

$$\rho_1(x) := \begin{cases} \frac{1}{\cosh \beta - 1} \left(e^{-\beta \sqrt{1 - (N_1 x)^2 / m^2}} - 1 \right) & x \in I, \\ 0 & x \in \mathbb{R} \setminus I \end{cases}$$

Since ρ_1 is even and $\rho_1|_{[0, \frac{m}{N_1}]}$ is increasing, we have

$$0 \ge \rho_1(x) \ge \rho_1(0) = \frac{e^{-\beta} - 1}{\cosh \beta - 1} = -\frac{2}{e^{\beta} - 1}.$$

Since ρ_1 has a small values in the compact support \overline{I} , the Fourier transform $\hat{\rho}_1$ is small too and it holds

$$|\hat{\rho}_1(v)| = \left| \int_I \rho(x) e^{2\pi i \, vx} \, \mathrm{d}x \right| \le \frac{4m}{(e^\beta - 1) \, N_1}$$

Substituting $t = N_1 x/m$, we determine the Fourier transform

$$\begin{split} \hat{\varphi}_{\cosh}(v) &= \int_{I} \varphi_{\cosh}(x) \,\mathrm{e}^{-2\pi\mathrm{i}\,vx} \,\mathrm{d}x = \int_{I} \psi_{1}(x) \,\mathrm{e}^{-2\pi\mathrm{i}\,vx} \,\mathrm{d}x + \int_{I} \rho_{1}(x) \,\mathrm{e}^{-2\pi\mathrm{i}\,vx} \,\mathrm{d}x \\ &= \frac{2m}{(\cosh\beta - 1)\,N_{1}} \Big[\int_{0}^{1} \sinh\left(\beta\sqrt{1 - t^{2}}\right) \,\cos\frac{2\pi mvt}{N_{1}} \,\mathrm{d}t \\ &+ \int_{0}^{1} \left(\mathrm{e}^{-\beta\sqrt{1 - t^{2}}} - 1\right) \,\cos\frac{2\pi mvt}{N_{1}} \,\mathrm{d}t \Big] \,. \end{split}$$

For simplicity, we introduce the scaled frequency $w := 2\pi m v / N_1$ such that

$$\hat{\varphi}_{\exp}\left(\frac{N_1w}{2\pi m}\right) = \frac{2m}{\left(\cosh\beta - 1\right)N_1} \left[\int_0^1 \sinh\left(\beta\sqrt{1-t^2}\right)\,\cos(wt)\,\mathrm{d}t\right] + \int_0^1 \left(\mathrm{e}^{-\beta\sqrt{1-t^2}} - 1\right)\,\cos(wt)\,\mathrm{d}t\right].$$

From [14, p. 38] it follows that

$$\int_{0}^{1} \sinh\left(\beta\sqrt{1-t^{2}}\right)\,\cos(wt)\,\mathrm{d}t = \frac{\pi\beta}{2} \begin{cases} (\beta^{2}-w^{2})^{-1/2}\,I_{1}\left(\sqrt{\beta^{2}-w^{2}}\right) & w\in\left(-\beta,\,\beta\right),\\ 1/4 & w=\pm\beta,\\ (w^{2}-\beta^{2})^{-1/2}\,J_{1}\left(\sqrt{w^{2}-\beta^{2}}\right) & w\in\mathbb{R}\setminus\left[-\beta,\,\beta\right], \end{cases}$$

where I_1 denotes the modified Bessel function and J_1 the Bessel function of first order. Using Lemma 6.3, we receive for $w \in (-\beta, \beta)$,

$$\hat{\varphi}_{\cosh}\left(\frac{N_1 w}{2\pi m}\right) \ge \frac{2m}{\left(\cosh\beta - 1\right) N_1} \left[\frac{\pi\beta}{2} \left(\beta^2 - w^2\right)^{-1/2} I_1\left(\sqrt{\beta^2 - w^2}\right) - \operatorname{sinc} w - \gamma(m, \sigma)\right].$$

This means for $\beta = bm$ and $v = \frac{N_1 w}{2\pi m} \in \left(-N_1 \left(1 - \frac{1}{2\sigma}\right), N_1 \left(1 - \frac{1}{2\sigma}\right)\right)$,

$$\hat{\varphi}_{\cosh}(v) \geq \frac{2m}{(\cosh\beta-1)N_1} \left[\frac{b}{4} \left(\left(1 - \frac{1}{2\sigma}\right)^2 - \frac{v^2}{N_1^2} \right)^{-1/2} I_1 \left(2\pi m \sqrt{\left(1 - \frac{1}{2\sigma}\right)^2 - \frac{v^2}{N_1^2}} \right) - \operatorname{sinc} \frac{2\pi m v}{N_1} - \gamma(m, \sigma) \right].$$

Since the function $h:\,[0,\beta)\to\mathbb{R}$ defined by

$$h(w) := \frac{\pi\beta}{2} \left(\beta^2 - w^2\right)^{-1/2} I_1\left(\sqrt{\beta^2 - w^2}\right) - \operatorname{sinc} w > 0$$

is decreasing and

$$\lim_{w \to \beta = 0} h(w) = \frac{\pi\beta}{4} - \operatorname{sinc} \beta \ge \pi^2 \left(1 - \frac{1}{2\sigma}\right) - 1 \ge \frac{3\pi^2}{5} - 1,$$

the Fourier transform $\hat{\varphi}_{\cosh}(v)$ is positive and decreasing for $v \in \left[0, N_1\left(1 - \frac{1}{2\sigma}\right)\right)$ too. Hence we obtain

$$\begin{aligned} \min_{n \in I_N} \hat{\varphi}_{\cosh}(n) &= \hat{\varphi}_{\cosh}\left(\frac{N}{2}\right) \\ &\geq \frac{2m}{\left(\cosh\beta - 1\right)N_1} \left[\frac{b}{4}\left(1 - \frac{1}{\sigma}\right)^{-1/2} I_1\left(2\pi m\sqrt{1 - \frac{1}{\sigma}}\right) - \operatorname{sinc}\frac{\pi m}{\sigma} - \gamma(m, \sigma)\right] \\ &\geq \frac{2m}{\left(\cosh\beta - 1\right)N_1} \left[\frac{b}{4}\left(1 - \frac{1}{\sigma}\right)^{-1/2} I_1\left(2\pi m\sqrt{1 - \frac{1}{\sigma}}\right) - 1 - \gamma(m, \sigma)\right]. \end{aligned}$$

From $m \ge 2$ and $\sigma \ge \frac{5}{4}$ it follows that

$$2\pi m \sqrt{1 - \frac{1}{\sigma}} \ge 4\pi \sqrt{1 - \frac{1}{\sigma}} \ge x_0 := \frac{4\pi}{\sqrt{5}}$$

By the inequality (see [1] or [18, Lemma 3.3])

$$\sqrt{2\pi x_0} e^{-x_0} I_1(x_0) \le \sqrt{2\pi x} e^{-x} I_1(x), \quad x \ge x_0,$$

we sustain that

$$I_1(x) \ge \sqrt{x_0} e^{-x_0} I_1(x_0) x^{-1/2} e^x > \frac{2}{5} x^{-1/2} e^x, \quad x \ge x_0$$

Hence for $x = 2\pi m \sqrt{1 - 1/\sigma}$ we get the estimate

$$I_1(2\pi m \sqrt{1-\frac{1}{\sigma}}) > \frac{\sqrt{2}}{5\sqrt{\pi m}} \left(1-\frac{1}{\sigma}\right)^{-1/4} e^{2\pi m \sqrt{1-1/\sigma}}.$$

Thus we obtain the following

Lemma 7.1 Let $N \in 2\mathbb{N}$ and $\sigma \geq \frac{5}{4}$ be given, where $N_1 = \sigma N \in 2\mathbb{N}$. Further let $m \in \mathbb{N}$ with $2m \ll N_1$, $\beta = bm$, and $b = 2\pi \left(1 - \frac{1}{2\sigma}\right)$. Then we have

$$\hat{\varphi}_{\cosh}\left(\frac{N}{2}\right) \le \frac{2m}{\left(\cosh\beta - 1\right)N_1} \left[\frac{\sqrt{\pi}}{5\sqrt{2m}} \left(1 - \frac{1}{2\sigma}\right) \left(1 - \frac{1}{\sigma}\right)^{-3/4} e^{2\pi m \sqrt{1 - 1/\sigma}} - 1 - \gamma(m, \sigma)\right].$$

By (2.2) the constant $e_{\sigma,N}(\hat{\varphi}_{\cosh})$ can be estimated as follows

$$e_{\sigma,N}(\hat{\varphi}_{\cosh}) \leq \frac{1}{\min_{n \in I_N} \hat{\varphi}_{\cosh}(n)} \max_{n \in I_N} \left\| \sum_{r \in \mathbb{Z} \setminus \{0\}} \hat{\varphi}_{\cosh}(n+rN_1) e^{2\pi i rN_1 \cdot} \right\|_{C(\mathbb{T})}$$
$$= \frac{1}{\hat{\varphi}_{\cosh}(N/2)} \max_{n \in I_N} \left\| \sum_{r \in \mathbb{Z} \setminus \{0\}} \hat{\varphi}_{\cosh}(n+rN_1) e^{2\pi i rN_1 \cdot} \right\|_{C(\mathbb{T})},$$

where it holds

$$\left\|\sum_{r\in\mathbb{Z}\setminus\{0\}}\hat{\varphi}_{\cosh}(n+rN_{1})e^{2\pi i rN_{1}\cdot}\right\|_{C(\mathbb{T})}$$

$$\leq \sum_{r\in\mathbb{Z}\setminus\{0\}}\left|\hat{\psi}_{1}(n+rN_{1})\right| + \sum_{r\in\mathbb{Z}\setminus\{0\}}\left|\hat{\rho}_{1}(n+rN_{1})\right|.$$
(7.2)

Analogously to Lemma 6.5, we get

Lemma 7.2 Let $N \in 2\mathbb{N}$ and $\sigma \geq \frac{5}{4}$ be given, where $N_1 = \sigma N \in 2\mathbb{N}$. Further let $m \in \mathbb{N}$ with $2m \ll N_1$, $\beta = bm$, and $b = 2\pi (1 - \frac{1}{2\pi})$. Then it holds for all $n \in I_N$,

$$\sum_{r \in \mathbb{Z} \setminus \{0\}} |\hat{\psi}_1(n+rN_1)| \le \frac{\beta}{(\cosh \beta - 1) N_1} \left[2\pi m + \frac{10}{\sqrt{2\pi m}} \left(1 - \frac{1}{2\sigma} \right)^{-1/2} \right].$$

Finally for any $n \in I_N$, we estimate the sum

$$\sum_{r\in\mathbb{Z}\setminus\{0\}}\left|\hat{\rho}_1(n+rN_1)\right|.$$

Analogously to Lemma 6.6, we obtain

Lemma 7.3 Let $N \in 2\mathbb{N}$ and $\sigma \geq \frac{5}{4}$ be given, where $N_1 = \sigma N \in 2\mathbb{N}$. Further let $m \in \mathbb{N}$ with $2m \ll N_1$, $\beta = bm$, and $b = 2\pi \left(1 - \frac{1}{2\sigma}\right)$. Then for all $n \in I_N$,

$$\sum_{r \in \mathbb{Z} \setminus \{0\}} |\hat{\rho}_1(n+rN_1)| \leq \frac{1}{\left(\cosh\beta - 1\right)N_1} \left[\left(1 + \frac{1}{e}\right) \frac{\beta}{\pi} \sqrt[4]{\frac{5}{2\pi m}} \left(\frac{4\sigma}{2\sigma - 1} + 8\right) \left(1 - \frac{1}{2\sigma}\right)^{-1/4} + \frac{8\sigma}{(2\sigma - 1)\pi} e^{-\sqrt{2\pi m - \pi m/\sigma}} + \frac{8}{\pi} E_1 \left(\sqrt{2\pi m - \pi m/\sigma}\right) + \left(e^{-\sqrt{2}\beta} + \frac{1}{2}\right) \left(4m + \frac{2}{\pi}\right) e^{-2\pi m + \pi m/\sigma} \right].$$

From Lemmas 7.2 and 7.3 it follows that by (7.2),

$$\left\|\sum_{r\in\mathbb{Z}\setminus\{0\}}\hat{\varphi}_{\cosh}(n+rN_1)\,\mathrm{e}^{2\pi\mathrm{i}\,rN_1\cdot}\right\|_{C(\mathbb{T})} \le \frac{\beta\,b(m,\sigma)}{\left(\cosh\beta-1\right)N_1} \tag{7.3}$$

with the constant (6.19). Using

$$e_{\sigma,N}(\varphi_{\cosh}) \le \frac{1}{\hat{\varphi}_{\cosh}(N/2)} \max_{n \in I_N} \left\| \sum_{r \in \mathbb{Z} \setminus \{0\}} \hat{\varphi}_{\cosh}(n+rN_1) e^{2\pi i rN_1 \cdot} \right\|_{C(\mathbb{T})},$$

it follows from (7.3) and Lemma 7.1:

Theorem 7.4 Let $N \in 2\mathbb{N}$ and $\sigma \geq \frac{5}{4}$ be given, where $N_1 = \sigma N \in 2\mathbb{N}$. Further let $m \in \mathbb{N}$ with $2m \ll N_1$, $\beta = bm$, and $b = 2\pi (1 - \frac{1}{2\sigma})$. Then the $C(\mathbb{T})$ error constant of the continuous cosh time window function (7.1) can be

Then the $C(\mathbb{T})$ -error constant of the continuous \cosh -type window function (7.1) can be estimated by

$$e_{\sigma}(\varphi_{\cosh}) \leq \frac{\beta \, b(m,\sigma)}{2m} \left[\frac{\sqrt{\pi}}{5\sqrt{2m}} \left(1 - \frac{1}{2\sigma} \right) \left(1 - \frac{1}{\sigma} \right)^{-3/4} e^{2\pi m \sqrt{1 - 1/\sigma}} - 1 - \gamma(m,\sigma) \right]^{-1},$$

i.e., the continuous \cosh -type window function (6.1) is convenient for NFFT.

Conclusion

In this paper, we prefer the use of continuous, compactly supported window functions for NFFT (with nonequispaced spatial data and equispaced frequencies). Such window functions simplify the algorithm for NFFT, since the truncation error of NFFT vanishes. Further, such window functions can produce very small errors of NFFT. Examples of such window functions are the continuous Kaiser-Bessel window function (4.1), continuous exp-type window function (6.1), sinh-type window function (5.1), and continuous coshtype window function (7.1) which possess the same support and shape parameter. For these window functions, we present novel explicit error estimates for NFFT and we derive rules for the convenient choice of the truncation parameter $m \geq 2$ and the oversampling parameter $\sigma \geq \frac{5}{4}$. The main tool of this approach is the decay of the Fourier transform $\hat{\varphi}(v)$ of $\varphi \in \Phi_{m,N_1}$ for $|v| \to \infty$. A rapid decay of $\hat{\varphi}$ is essential for small error constants. Unfortunately, the Fourier transform of certain window function φ , such as (6.1) and (7.1), is unknown analytically. Therefore we propose a new technique and split φ into a sum of two compactly supported functions ψ and ρ , where the Fourier transform ψ is explicitly known and where $|\rho|$ is sufficiently small. Further, it is shown that the standard Kaiser-Bessel window function and original exp-type window function which have jump discontinuities with very small jump heights at the endpoints of their support, possess a similar error behavior as the corresponding continuous window functions. In summary, the $C(\mathbb{T})$ -error constant of the continuous/standard Kaiser–Bessel window

function is of best order $\mathcal{O}(m e^{-2\pi m \sqrt{1-1/\sigma}})$. For the sinh-type, continuous/original

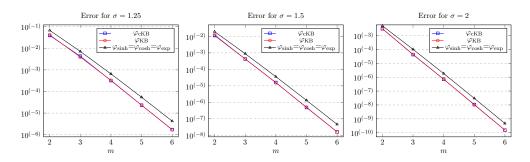


Figure 7.1: The constants $e_{\sigma,N}(\varphi)$ of the different window functions with shape parameter $\beta = \pi m(2 - 1/\sigma)$ for $\sigma \in \{1.25, 1.5, 2\}$ and $m \in \{2, 3, 4\}$.

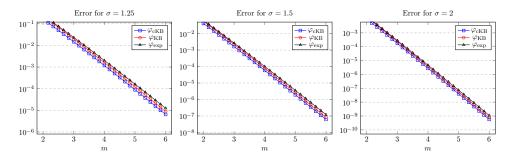


Figure 7.2: The constants $e_{\sigma}(\varphi)$ of the different window functions with shape parameter $\beta = \pi m(2 - 1/\sigma)$ for $\sigma \in \{1.25, 1.5, 2\}$ and $m \in \{2, 3, 4, 5, 6\}$ based on the error estimates of equation (4.11), (4.14) and (6.20).

exp-type, and continuous cosh-type window functions, the corresponding $C(\mathbb{T})$ -error constants are of order $\mathcal{O}(m^{3/2} e^{-2\pi m \sqrt{1-1/\sigma}})$. Nevertheless, our numerical results show that all window functions proposed here yield a very similar error, see Figure 7.1. The presented error estimates are very precise, see Figure 7.2. Thus we can recommend the use of all these window functions.

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