

Hermite Distributed Approximating Functionals as Almost-Ideal Low-Pass Filters

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Abstract

The two-parameter family of Hermite Distributed Approximating Functionals (HDAFs) is shown to possess all properties that are essential requirements in filter design. When properly scaled, HDAFs provide an arbitrarily sharp high-frequency cut-off while retaining their smoothness. More precisely, bounds on the Fourier transform of the HDAF integral kernel show that it converges almost-uniformly to the ideal window, and that the pass and transition bands can be tuned independently to any width while preserving Gaussian decay in *both* time and frequency domains. The effective length of the HDAF filter in both domains is controlled by an estimate of the Heisenberg uncertainty product. In addition, a new asymptotic relationship between the HDAF and a windowed sinc function is obtained. In all calculations, we have aimed at precise error estimates that may assist numerical implementations.

Key words and phrases : Filter design, ideal low-pass filter, Gaussian decay in time and frequency domains, uncertainty product.

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1 Introduction

Optimizing the design of band-pass filters constitutes a major field of study in signal processing [1–3]. Depending on the specifics of an application, such as the numerical solution of linear and non-linear ordinary and partial differential equations [4], data compression [5], imaging [6], or pattern recognition [7], the goals of filter design may vary from plain frequency selectivity to localization in the time domain or to other, more specific requirements. Compared to filters in the analog domain, where causality imposes a severe restriction, there is more freedom in digital signal processing that can be exploited to attain these specific goals. Here, we restrict ourselves to discussing the context of the basic low-pass filter defined on the entire real line, because once this is constructed, a wide variety of related filters and computational tools can be obtained. In the Conclusion we will give an idea how such a basic low-pass filter can be used for purposes of digital signal processing.

The contemporary view of filter design emphasizes the role of the time-frequency domain [1]. From this perspective, it is intuitively clear what general conditions must be satisfied in order to produce a robust low-pass filter. *These conditions are: 1) the filter must be able to approach the so-called “ideal window” in a practicable and efficient way; 2) the filter must show a strong type of decay in both the time and frequency domains or equivalently, a high degree of regularity including smoothness in both domains; and 3) the Heisenberg uncertainty product of the filter should grow slowly as the frequency selectivity increases.* These conditions are at the moment rather vague, but we will make them more precise later on. For now, we mention that the first condition guarantees arbitrarily accurate separation of the desired and undesired frequency bands. Band-pass and band-stop filters are easily designed starting with the basic low-pass filter. The second condition prevents that the filter itself introduces uncontrolled aliasing and Gibbs’ oscillations. This is important for the stability of numerical implementation of various operations. The third condition ensures that the effective length of the filter is small in both the time *and* frequency domains, in the spirit of filter design in the time-frequency domain. As far as we are aware, up to now, no filter has been shown to satisfy *all* three conditions in a mathematically rigorous fashion.

The purpose of this paper is to show that such a class of filters has, in fact, been created, by demonstrating that there is a precise sense in which all three conditions can be satisfied. This class of filters is called the Distributed Approximating Functionals (DAFs) [8], and the particular DAF which is shown to satisfy all the conditions is known as the Hermite DAF (HDAF). For the sake of space, we will simply posit the HDAF, but detailed derivations from a variety of viewpoints have been given elsewhere [8]. This paper is organized as follows: In Section 2, we introduce the notion of filters. We show how to understand

convolution with the HDAF integral kernels as the implementation of low-pass filters that approximate the identity operator. In Section 3, we obtain an integral expression for the HDAF in time and frequency domains, which is needed to characterize its behavior. We analyze the detailed behavior of the HDAF in order to prove that, when properly scaled, it provides a smooth, arbitrarily close approximation to the ideal window. We also discuss the Gaussian decay of HDAFs in time and frequency domains and estimate the Heisenberg uncertainty product for the HDAFs. Finally, we summarize our results and point out further developments in Section 4.

2 Definitions and Notation

Definition 2.1. In this work, we understand a filter M as a bounded operator defined on the space $L^2(\mathbb{R})$ of square-integrable functions on \mathbb{R} such that M commutes with the family of unitary shift operators $\{T_a\}_{a \in \mathbb{R}}$ given by $T_a f(x) = f(x - a)$ for all $f \in L^2(\mathbb{R})$. Consequently, M acts by multiplication with an essentially bounded function \widehat{M} in the frequency domain. When speaking about a filter M , we often identify it and its properties with those of the associated function \widehat{M} . For the Fourier transform, we adopt the convention

$$\hat{f}(k) := \int_{-\infty}^{\infty} f(x) e^{-ikx} dx \quad (1)$$

when $f \in L^1(\mathbb{R}) \cap L^2(\mathbb{R})$, and as usual extend $f \mapsto \hat{f}$ to a bounded map (unitary up to a normalization factor $\frac{1}{2\pi}$) on all of $L^2(\mathbb{R})$.

Example 2.2. The ideal low-pass filter X is associated with the characteristic function of an interval centered at $k = 0$, so $\widehat{X}(k) = \chi_{[-\kappa/2, \kappa/2]}(k)$. The length $\kappa > 0$ of the interval is called the pass-bandwidth of X . A more general low-pass filter M is given by a complex-valued, bounded function \widehat{M} that has the limits $\lim_{k \rightarrow 0} \widehat{M}(k) = 1$ and $\lim_{k \rightarrow \pm\infty} \widehat{M}(k) = 0$.

Definition 2.3. The family of Hermite Distributed Approximating Functional (HDAF) filters in one dimension is given by operators $\{D_{n,\sigma}\}$ indexed by order and length-scale parameters $n \in \mathbb{N}_0$ and $\sigma > 0$. Each $D_{n,\sigma}$ acts on $f \in L^2(\mathbb{R})$ by convolution with an integral kernel, $D_{n,\sigma} f(x) = \int_{-\infty}^{\infty} D_n(x, x'; \sigma) f(x') dx'$. The HDAF integral kernel [8] is defined as

$$D_n(x, x'; \sigma) := \frac{1}{\sqrt{2\pi\sigma}} e^{-\frac{1}{2\sigma^2}(x-x')^2} \sum_{j=0}^n \frac{(-1)^j}{2^{2j} j!} H_{2j} \left(\frac{x-x'}{\sqrt{2\sigma}} \right), \quad (2)$$

where $x, x' \in \mathbb{R}$ and the m -th Hermite polynomial is denoted as $H_m(x) = (-1)^m e^{x^2} \frac{d^m}{dx^m} e^{-x^2}$ for any $m \in \mathbb{N}_0$.

Remarks 2.4. The quantity $D_n(x, x'; \sigma)$ has the property that it approaches the Dirac distribution $\delta(x - x')$ in either of the limits $\sigma \rightarrow 0$ or $n \rightarrow \infty$, see the following paragraph. In a joint limit where $n, \sigma \rightarrow \infty$ and n/σ^2 is kept constant, it has been claimed [9], but not proved, that convolving with HDAF kernels implements an ideal filter to arbitrary, controllable accuracy.

By $D_n(x, x'; \sigma) = D_n(x - x', 0; \sigma) = D_n(x' - x, 0; \sigma)$, one may verify that each HDAF filter $D_{n,\sigma}$ is self-adjoint and indeed commutes with translations. Consequently,

$$\int_{-\infty}^{\infty} e^{-ikx} D_{n,\sigma} f(x) dx = \widehat{D}_n(k; \sigma) \int_{-\infty}^{\infty} e^{-ikx} f(x) dx \quad (3)$$

for all $f \in L^2(\mathbb{R}) \cap L^1(\mathbb{R})$ and $k \in \mathbb{R}$ with a real-valued function that is seen to be

$$\widehat{D}_n(k; \sigma) = \exp\left(-\frac{k^2 \sigma^2}{2}\right) e_n\left(\frac{k^2 \sigma^2}{2}\right), \quad (4)$$

containing the exponential, locally compensated by the truncated exponential series

$$e_n(y) := \sum_{j=0}^n \frac{y^j}{j!}. \quad (5)$$

For each fixed pair $n \in \mathbb{N}_0$ and $\sigma > 0$ the value $\widehat{D}_n(k; \sigma)$ is non-negative, bounded above by $\widehat{D}_n(k; \sigma) \leq \widehat{D}_n(0; \sigma) = 1$ and decays faster than any polynomial in k . Consequently, $D_{n,\sigma}$ is a low-pass filter. As either $n \rightarrow \infty$ or $\sigma \rightarrow 0$, we have $\widehat{D}_n(k; \sigma) \nearrow 1$ monotonically and thus the operator $D_{n,\sigma}$ approximates the identity on $L^2(\mathbb{R})$ from below.

3 Bounds on HDAFs with Relevance for Filtering

In this section we investigate certain properties of HDAFs that demonstrate their usefulness for separating desired and undesired frequency bands in a signal, their regularity properties, and the behavior of their Heisenberg uncertainty products. In addition, we derive an asymptotic approximation formula for the HDAF integral kernel that shows its relation to a windowed sinc function, also known as Gaussian sinc-DAF [10].

3.1 An Integral Representation for HDAFs

The following lemma provides integral representations for the Fourier transform $\widehat{D}_n(k; \sigma)$ and for the HDAF integral kernel itself. In view of $D_n(x, x'; \sigma) = D_n(x - x', 0; \sigma)$, we can assume $x' = 0$ and simplify notation. The integral representations will be instrumental in the derivation of various estimates that control the properties of HDAF filters.

Lemma 3.1. For any $n \in \mathbb{N}_0$, $k \in \mathbb{R}$ and $x \in \mathbb{R}$,

$$\widehat{D}_n(k; \sigma) = \frac{\Gamma(n+1, \frac{k^2\sigma^2}{2})}{n!} \equiv \frac{1}{n!} \int_{\frac{k^2\sigma^2}{2}}^{\infty} \lambda^n e^{-\lambda} d\lambda \quad (6)$$

and

$$D_n(x, 0; \sigma) = \frac{2}{\pi n! x} \int_0^{\infty} \sin\left(\frac{\sqrt{2}x}{\sigma} t\right) t^{2n+1} e^{-t^2} dt. \quad (7)$$

Proof. The representation for $\widehat{D}_n(k; \sigma)$ is seen to be true by taking the derivative

$$\frac{d}{dk} \widehat{D}_n(k; \sigma) = -\frac{k\sigma^2}{n!} \left(\frac{k^2\sigma^2}{2}\right)^n \exp\left(-\frac{k^2\sigma^2}{2}\right) \quad (8)$$

of Eq. (4) and integrating it again, while adjusting the constant of integration to obtain the value $\widehat{D}_n(0; \sigma) = 1$ at $k = 0$. Eq. (8) follows from Eq. (4) by the telescoping derivatives $e'_n(y) = e_{n-1}(y)$ for $n \in \mathbb{N}$.

We recover the integral representation of the HDAF integral kernel by the inverse Fourier transform of (6),

$$D_n(x, 0; \sigma) = \frac{1}{2\pi n!} \int_{-\infty}^{\infty} e^{ikx} \int_{\frac{k^2\sigma^2}{2}}^{\infty} \xi^n e^{-\xi} d\xi dk. \quad (9)$$

Writing this result as an integral over positive values of k and switching the order of integration, we obtain

$$\begin{aligned} D_n(x, 0; \sigma) &= \frac{1}{\pi n!} \int_0^{\infty} \int_0^{\frac{\sqrt{2\xi}}{\sigma}} \cos(kx) \xi^n e^{-\xi} dk d\xi \\ &= \frac{1}{\pi n! x} \int_0^{\infty} \sin\left(\frac{\sqrt{2\xi}}{\sigma} x\right) \xi^n e^{-\xi} d\xi \\ &= \frac{2}{\pi n! x} I_n(x; \sigma), \end{aligned} \quad (10)$$

where $I_n(x; \sigma)$ is defined by

$$I_n(x; \sigma) := \int_0^{\infty} \sin\left(\frac{\sqrt{2}x}{\sigma} t\right) t^{2n+1} e^{-t^2} dt. \quad (11)$$

□

Remarks 3.2. It is possible to bypass the Fourier transform in an alternative derivation of the integral representation for $D_n(x, 0; \sigma)$. The summation of Eq. (2) with the Christoffel-Darboux formula [14, p. 307] is the first step in this derivation, yielding

$$D_n(x, 0; \sigma) = \frac{(-1)^n}{2^{2n+1} n! \sqrt{\pi}} \frac{H_{2n+1}\left(\frac{x}{\sqrt{2}\sigma}\right)}{x} e^{-\frac{1}{2\sigma^2} x^2}. \quad (12)$$

for $n \in \mathbb{N}_0$. Inserting the known integral representation [11, 22.10.15] for Hermite polynomials,

$$H_m(x) = e^{x^2} \frac{2^{m+1}}{\sqrt{\pi}} \int_0^\infty \cos(2tx - \frac{m\pi}{2}) t^m e^{-t^2} dt \quad (13)$$

valid for all $m \in \mathbb{N}_0$, results in Eq. (10). The virtue of the approach via the frequency domain is that it painlessly generalizes to non-integral values of $n \geq 0$, whereas Eq. (12) requires an additional effort to interpolate $(-1)^n$.

Using integration by parts the integrals $I_n(x; \sigma)$ of Eq. (11) are seen to obey a recursion relation that is equivalent to the standard one between even order Hermite polynomials. Specifically,

$$I_n(x; \sigma) = \frac{1}{4} \left[\left(8n - \frac{2x^2}{\sigma^2} - 2 \right) I_{n-1}(x; \sigma) - (2n-1)(2n-2) I_{n-2}(x; \sigma) \right], \quad (14)$$

which simplifies the calculation of the HDAF kernel, since then one only needs $I_n(x; \sigma)$ for $n \in \{0, 1\}$ to obtain its value for any $n \in \mathbb{N}_0$.

3.2 Bounds on the Width of the Transition Bands

While the ideal low-pass filter is characterized solely by its bandwidth, our HDAF approximation is controlled by two parameters n and σ . Next to the width of the pass band, the second scale parameter implicit in our approximation is the width of the transition bands. In this subsection we illustrate how tuning n and σ controls these bandwidths, enabling a high-frequency cut-off that is arbitrarily sharp and tuned to any given frequency.

Definition 3.3. The pass band associated with a low-pass filter M and an error $\epsilon_1 > 0$ is the set $\{k \in \mathbb{R} : |\widehat{M}(k) - 1| \leq \epsilon_1\}$. The stop or attenuation band associated with M and an error $\epsilon_2 > 0$ is the set $\{k \in \mathbb{R} : |\widehat{M}(k)| \leq \epsilon_2\}$. Given M and errors $\epsilon_{1,2}$, the so-called transition band is what is contained in neither pass nor stop band. Typically, these bands are given by intervals or by finite unions thereof. If the transition band is given by two intervals, we speak of them separately as of two transition bands.

Now we investigate the properties implicit in the integral representation for the Fourier transform of the HDAF kernel. In Eq. (6), $\widehat{D}_n(k; \sigma)$ is a manifestly even function, thus we can restrict ourselves to investigating the behavior of $\widehat{D}_n(k; \sigma)$ for positive values of k . We begin with a heuristic consideration.

Remark 3.4. Intuitively, the region of rapid transition in $k \geq 0$ can be understood as the interval between the points, denoted as k_t^\pm , at which $\frac{d^2}{dk^2} \widehat{D}_n(k; \sigma)$ is an extremum. Taking further derivatives of Eq. (8), we obtain

$$\frac{d^2}{dk^2} \widehat{D}_n(k; \sigma) = -\frac{k\sigma^2}{n!} \left(\frac{k^2\sigma^2}{2}\right)^n \exp\left(-\frac{k^2\sigma^2}{2}\right) \left[\frac{2n+1}{k} - k\sigma^2\right]$$

and

$$\frac{d^3}{dk^3}\widehat{D}_n(k; \sigma) = \frac{k\sigma^2}{n!} \left(\frac{k^2\sigma^2}{2}\right)^n \exp\left(-\frac{k^2\sigma^2}{2}\right) \left\{ \frac{2n+1}{k^2} + \sigma^2 - \left[\frac{2n+1}{k} - k\sigma^2\right]^2 \right\}.$$

By setting $\frac{d^3}{dk^3}\widehat{D}_n(k; \sigma)$ to zero we find

$$k_t^\pm = \frac{1}{\sigma} \sqrt{\frac{(4n+3) \pm \sqrt{16n+9}}{2}}.$$

Moreover, setting Eq. (15) equal to zero, we find that the only inflection point in \mathbb{R}^+ , (in our case this is, in fact, the point where the slope $\frac{d}{dk}\widehat{D}_n(k; \sigma)$ has a global minimum) is given by

$$k_{sl} = \sqrt{2n+1}/\sigma \quad (15)$$

Hence, this inflection point lies within the heuristically defined transition region. If we fix the position of the inflection point by scaling σ with $\sqrt{2n+1}$ then we expect that the width of the transition region decreases to leading order as $k_t^+ - k_t^- \sim n^{-1/2}\sigma^{-1}$. The qualitative characteristics of $\widehat{D}_n(k; \sqrt{2n+1}\sigma)$ are such that for small values of k , it has a value close to one; and over a controllably short frequency range in the vicinity of $k = \frac{1}{\sigma}$, it falls rapidly and monotonically to zero. This behavior makes it appropriate for a low-pass filter that approximates the ideal filter. In Fig. 1 we show $\widehat{D}_n(k; \sqrt{2n+1}\sigma)$ as a function of k and n , to illustrate that we can obtain a function that is arbitrarily close to the ideal window.

The remaining part of this section is a rigorous verification of the above-described intuitive claims. As before, let $k > 0$. We begin by rewriting Eq. (6) as

$$\widehat{D}_n(k; \sigma) = \frac{2(n+\frac{1}{2})^{n+1}}{n!} \int_{k\sigma/\sqrt{2n+1}}^{\infty} t^{2n+1} e^{-(n+\frac{1}{2})t^2} dt \quad (16)$$

such that the maximum $e^{-(n+\frac{1}{2})}$ of the integrand occurs at $t = 1$, independently of n . By scaling σ with $\sqrt{2n+1}$ we also eliminate the dependence of the lower limit of integration on n . In the following, we want to compare the integrand with Gaussians to control the $n \rightarrow \infty$ asymptotics.

Lemma 3.5. Given $0 < x \leq 1 \leq y \leq \infty$, we claim the upper bound

$$t^{2n+1} e^{-(n+\frac{1}{2})t^2} \leq e^{-(n+\frac{1}{2})} e^{-(n+\frac{1}{2})(1+\frac{1}{y^2})(t-1)^2} \quad (17)$$

for all $0 \leq t \leq y$ and the lower bound

$$e^{-(n+\frac{1}{2})} e^{-(n+\frac{1}{2})(1+\frac{1}{x^2})(t-1)^2} \leq t^{2n+1} e^{-(n+\frac{1}{2})t^2} \quad (18)$$

Figure 1: HDAF low-pass filter as function of k and n with $k_{sl} = 1$ fixed according to Eq. (15).

for all $t \geq x$, with equality in either case only when $t = 1$. In addition, we have a complementary upper bound

$$t^{2n+1} e^{-(n+\frac{1}{2})t^2} \leq y^{2n+1} e^{(n+\frac{1}{2})(\frac{1}{y^2}-2)} e^{-(n+\frac{1}{2})(t-\frac{1}{y})^2} \quad (19)$$

for $t \geq y$ and correspondingly

$$t^{2n+1} e^{-(n+\frac{1}{2})t^2} \leq x^{2n+1} e^{-(n+\frac{1}{2})[(1+\frac{1}{x^2})(t-\frac{2x}{1+x^2})^2 - \frac{4}{1+x^2} + 3]} \quad (20)$$

when $0 \leq t \leq x$.

Proof. To show the various estimates, we define $g(t) := (2n+1) \ln t - (n+\frac{1}{2})t^2$ for all $t > 0$ and use that in an interval (x, y) with $0 < x \leq 1 \leq y \leq \infty$, $-(2n+1)(1+\frac{1}{x^2}) \leq g''(t) \leq -(2n+1)(1+\frac{1}{y^2})$. Integrating this inequality twice starting from $t = 1$ and using the monotonicity of the exponential function yields the first pair of upper and lower bounds. In addition, integrating the inequality for g'' twice starting at $y \geq 1$ gives the improved upper bound for $t \geq y$ and similarly, starting at $x \leq 1$, for $0 \leq t \leq x$. \square

Once the integrand is bounded by suitable Gaussians, we use an estimate of

the complementary error function

$$\frac{e^{-x^2}}{x + \sqrt{x^2 + 2}} < \int_x^\infty e^{-t^2} dt \leq \frac{e^{-x^2}}{x + \sqrt{x^2 + \frac{4}{\pi}}}, \quad (21)$$

valid whenever $x \geq 0$ [15, 16]. The remaining ingredient in the derivation of upper and lower bounds is the Stirling-Robbins [17] estimate for the factorial,

$$\sqrt{2\pi n} n^{n+\frac{1}{2}} e^{-n+\frac{1}{12n+1}} < n! < \sqrt{2\pi n} n^{n+\frac{1}{2}} e^{-n+\frac{1}{12n}}. \quad (22)$$

Now we derive an upper bound on $\widehat{D}(k; \sigma)$ in case $k^2\sigma^2 \geq 1$ and a lower bound in case $k^2\sigma^2 \leq 1$. For simplicity, we again consider only $k > 0$.

Lemma 3.6. Let $n \in \mathbb{N}_0$ and $k, \sigma > 0$. If $k\sigma \geq 1$ then

$$\widehat{D}_n(k; \sqrt{2n+1}\sigma) < \sqrt{\frac{2}{\pi n}} e^{n+\frac{1}{2}} \frac{(k\sigma)^{2n+1} e^{-(n+\frac{1}{2})k^2\sigma^2}}{k\sigma - \frac{1}{k\sigma} + \sqrt{(k\sigma - \frac{1}{k\sigma})^2 + \frac{4}{\pi(n+\frac{1}{2})}}}. \quad (23)$$

and if $0 < k\sigma \leq 1$ then

$$\begin{aligned} \widehat{D}_n(k; \sqrt{2n+1}\sigma) &> 1 - \sqrt{\frac{2}{\pi n}} e^{n+\frac{1}{2}} \frac{1}{k\sigma + \frac{1}{k\sigma}} \\ &\quad \times \frac{(k\sigma)^{2n+1} e^{-(n+\frac{1}{2})k^2\sigma^2}}{\frac{2}{1+k^2\sigma^2} - 1 + \sqrt{\left(\frac{2}{1+k^2\sigma^2} - 1\right)^2 + \frac{4}{\pi(n+\frac{1}{2})(1+k^2\sigma^2)}}}. \end{aligned} \quad (24)$$

Proof. We use first the upper bound Ineq. (19) for the integrand of Eq. (16) and then Eqs. (21)-(22) to obtain

$$\begin{aligned} \widehat{D}_n(k; \sqrt{2n+1}\sigma) &< \frac{2(n+\frac{1}{2})^{n+1}}{\sqrt{2\pi n} n^{n+\frac{1}{2}}} e^{n-\frac{1}{12n+1}} e^{(n+\frac{1}{2})(\frac{1}{k^2\sigma^2}-2)} \\ &\quad \times (k\sigma)^{2n+1} \int_{k\sigma}^\infty e^{-(n+\frac{1}{2})(t-\frac{1}{k\sigma})^2} dt \\ &< \sqrt{\frac{2}{\pi n}} \left(1 + \frac{1}{2n}\right)^n e^{-\frac{1}{2}-\frac{1}{12n+1}} e^{(n+\frac{1}{2})(\frac{1}{k^2\sigma^2}-1)} \\ &\quad \times (k\sigma)^{2n+1} \frac{e^{-(n+\frac{1}{2})(k\sigma-\frac{1}{k\sigma})^2}}{k\sigma - \frac{1}{k\sigma} + \sqrt{(k\sigma - \frac{1}{k\sigma})^2 + \frac{4}{\pi(n+\frac{1}{2})}}} \end{aligned} \quad (25)$$

To obtain Ineq. (23), the simple inequality $(1 + \frac{y}{n})^n < e^y$ for all $y, n > 0$ is applied and the marginal correction term $e^{-\frac{1}{12n+1}}$ is dropped.

If $0 < k\sigma \leq 1$, we use the normalization $\widehat{D}(0, \sigma) = 1$ and the upper bound Ineq. (20) for the integrand Eq. (16) in the interval $[0, k\sigma]$ and then extend the domain of integration to $(-\infty, k\sigma]$, combined with the estimate Ineq. (21),

$$\widehat{D}_n(k; \sqrt{2n+1}\sigma) > 1 - \frac{2(n+\frac{1}{2})^{n+1}}{n!} \int_0^{k\sigma} t^{2n+1} e^{-(n+\frac{1}{2})t^2} dt \quad (26)$$

$$\begin{aligned} &> 1 - \frac{2(n+\frac{1}{2})^{n+1}}{n!} (k\sigma)^{2n+1} e^{(n+\frac{1}{2})[\frac{4}{1+k^2\sigma^2}-3]} \\ &\quad \times \int_{-\infty}^{k\sigma} e^{-(n+\frac{1}{2})(1+\frac{1}{k^2\sigma^2})(t-\frac{2k\sigma}{1+k^2\sigma^2})^2} dt \end{aligned} \quad (27)$$

$$\begin{aligned} &> 1 - \frac{2(n+\frac{1}{2})^{n+1}}{n!} (k\sigma)^{2n+1} e^{(n+\frac{1}{2})[\frac{4}{1+k^2\sigma^2}-3]} \frac{1}{(n+\frac{1}{2})(1+\frac{1}{k^2\sigma^2})} \\ &\quad \times \frac{e^{-(n+\frac{1}{2})(1+k^2\sigma^2)(1-\frac{2}{1+k^2\sigma^2})^2}}{\frac{2k\sigma}{1+k^2\sigma^2} - k\sigma + \sqrt{(\frac{2k\sigma}{1+k^2\sigma^2} - k\sigma)^2 + \frac{4}{\pi(n+\frac{1}{2})(1+\frac{1}{k^2\sigma^2})}}} \end{aligned} \quad (28)$$

which simplifies to Ineq. (24) after applying Ineq. (22), the elementary estimate $(1 + \frac{y}{n})^n < e^y$ for $y, n > 0$, and some cancellations in the exponent. \square

The upper and lower bounds on the Fourier transform of the HDAF integral kernel are illustrated in Fig. 2 and Fig. 3 for various values of n .

The properties of the locally compensated exponential series appearing in Eq. (4) have been discussed in [12]. From these properties one can deduce the uniform convergence of $\widehat{D}_n(k; \sqrt{2n+1}\sigma)$ to the ideal filter on compact sets in \mathbb{R} that do not include the points $k = \pm\frac{1}{\sigma}$. A result from [13] extends this result to certain subsets of the complex plane. Here, we restrict ourselves to real arguments, but derive a uniform convergence result that allows to shrink the excluded neighborhoods of $k = \pm\frac{1}{\sigma}$ as n increases.

Theorem 3.7. In the limit $n \rightarrow \infty$, the scaled HDAF low-pass filter approaches the ideal filter with band-width $2/\sigma$

$$\lim_{n \rightarrow \infty} \widehat{D}_n(k; \sqrt{2n+1}\sigma) = \begin{cases} 1, & k \in (-\frac{1}{\sigma}, \frac{1}{\sigma}) \\ \frac{1}{2}, & k = \pm\frac{1}{\sigma} \\ 0, & \text{else} \end{cases}. \quad (29)$$

The convergence is uniform at an asymptotic rate of $n^{-1/2}$ outside of two transition bands centered at $k = \pm\frac{1}{\sigma}$, the widths of which decay asymptotically at least as $\frac{1}{\sigma}\sqrt{\ln n/n}$. More precisely, we have

$$\lim_{n \rightarrow \infty} \max \left\{ \sqrt{n}(1 - \widehat{D}_n(k; \sqrt{2n+1}\sigma)) : 0 \leq k \leq \frac{1}{\sigma} - \frac{1}{2\sigma} \sqrt{\frac{\ln n}{n + \frac{1}{2}}} \right\} = 0 \quad (30)$$

Figure 2: Upper bounds on the HDAF low-pass filter as function of k and n with $k_{sl} = 1$ fixed according to Eq. (15).

and

$$\lim_{n \rightarrow \infty} \max \left\{ \sqrt{n} \widehat{D}_n(k; \sqrt{2n+1}\sigma) : k \geq \frac{1}{\sigma} + \frac{1}{2\sigma} \sqrt{\frac{\ln n}{n + \frac{1}{2}}} \right\} = 0. \quad (31)$$

Proof. From Ineq. (24) we obtain

$$\liminf_{n \rightarrow \infty} \widehat{D}_n \left(\frac{1}{\sigma} - \frac{c}{\sqrt{n + \frac{1}{2}\sigma}}; \sqrt{2n+1}\sigma \right) \geq 1 - \frac{1}{\sqrt{2\pi}} \frac{e^{-2c^2}}{c + \sqrt{c^2 + \frac{2}{\pi}}} \quad (32)$$

for any constant $c > 0$. Replacing this constant with any sequence $\{c_n\}_{n \geq 0}$ that satisfies $c_n \rightarrow \infty$ gives a limit inferior of one. Together with the simple bound $\widehat{D}_n(k; \sigma) \leq 1$ this establishes the limit

$$\lim_{n \rightarrow \infty} \min \left\{ \widehat{D}_n(k; \sqrt{2n+1}\sigma) : 0 \leq k \leq \frac{1}{\sigma} - \frac{c_n}{\sqrt{n + \frac{1}{2}\sigma}} \right\} = 1. \quad (33)$$

The growth of $\{c_n\}$ can be arbitrarily slow, but of course it determines how quickly the limit is approached. Using l'Hôpital's rule Ineq. (24) shows that to obtain the convergence rate of $n^{-1/2}$ as described in Eq. (30), it suffices to choose $c_n = \sqrt{\frac{1}{4} \ln n}$.

Figure 3: Lower bounds on the HDAF low-pass filter as function of k and n with $k_{sl} = 1$ fixed.

Together with the non-negativity of $\widehat{D}_n(k; \sigma)$, the upper bound Ineq. (23) results analogously in

$$\limsup_{n \rightarrow \infty} \widehat{D}_n\left(\frac{1}{\sigma} + \frac{c}{\sqrt{n + \frac{1}{2}\sigma}}; \sqrt{2n + 1}\sigma\right) \leq \sqrt{\frac{2}{\pi}} \frac{e^{-2c^2}}{2(c + \sqrt{c^2 + \frac{1}{\pi}})} \quad (34)$$

and

$$\lim_{n \rightarrow \infty} \max\left\{\widehat{D}_n(k; \sqrt{2n + 1}\sigma) : k \geq \frac{1}{\sigma} + \frac{c_n}{\sqrt{n + \frac{1}{2}\sigma}}\right\} = 0 \quad (35)$$

for any sequence $\{c_n\}$, $c_n \rightarrow \infty$. Inspecting Ineq. (23) and using l'Hôpital's rule shows we can choose again $c_n = \sqrt{\frac{1}{4} \ln n}$ to ensure a convergence rate of $n^{-1/2}$ as in Eq. (31).

Finally, we verify the convergence $\widehat{D}_n(\frac{1}{\sigma}; \sqrt{2n + 1}\sigma) = \frac{1}{2}$ by the limit of the lower bound Ineq. (24) for $k\sigma = 1$ as $n \rightarrow \infty$ and a corresponding sharp upper

bound. This is achieved via

$$\begin{aligned} \widehat{D}_n\left(\frac{1}{\sigma}; \sqrt{2n+1}\sigma\right) &= \frac{2(n+\frac{1}{2})^{n+1}}{n!} \int_1^{1+\frac{cn}{\sqrt{n+\frac{1}{2}}}} t^{2n+1} e^{-(n+\frac{1}{2})t^2} dt \\ &\quad + \widehat{D}_n\left(\frac{1}{\sigma} + \frac{1}{\sigma} \frac{cn}{\sqrt{n+\frac{1}{2}}}; \sqrt{2n+1}\sigma\right) \end{aligned} \quad (36)$$

with any sequence $\{c_n\}_{n \in \mathbb{N}}$ satisfying $c_n \rightarrow \infty$. Using the previously established convergence of $\widehat{D}_n\left(\frac{1}{\sigma} + \frac{1}{\sigma} \frac{cn}{\sqrt{n+\frac{1}{2}}}; \sqrt{2n+1}\sigma\right) \rightarrow 0$ and the upper bound Ineq. (17) for the integrand between the limits $x = 1$ and $y_n := 1 + \frac{cn}{\sqrt{n+\frac{1}{2}}}$ results in

$$\begin{aligned} \limsup_{n \rightarrow \infty} D\left(\frac{1}{\sigma}; \sqrt{2n+1}\sigma\right) &\leq \frac{2(n+\frac{1}{2})^{n+1}}{n!} e^{-(n+\frac{1}{2})} \int_1^{y_n} e^{-(n+\frac{1}{2})(1+\frac{1}{y_n^2})(t-1)^2} dt \\ &= \frac{2(n+\frac{1}{2})^{n+\frac{1}{2}}}{n!} e^{-(n+\frac{1}{2})} \int_0^{cn} e^{-(1+\frac{1}{y_n^2})t^2} dt. \end{aligned} \quad (37)$$

Now dominated convergence applies and again with Ineq. (22)

$$\limsup_{n \rightarrow \infty} \widehat{D}_n\left(\frac{1}{\sigma}; \sqrt{2n+1}\sigma\right) \leq \frac{1}{2} \quad (38)$$

follows, which finishes the proof. \square

3.3 Gaussian decay of the HDAF kernel and of its Fourier transform

The upper bound Ineq. (23) on $\widehat{D}_n(k; \sqrt{2n+1}\sigma)$ serves two purposes: The first one is in the separation of frequency bands as described above; the second one concerns regularity properties of the HDAF kernel. Ineq. (23) shows Gaussian decay of its Fourier transform. This has computational relevance because it implies that when a function is convolved with the HDAF kernel, the result is smooth, and we may then even apply a large class of pseudo-differential operators without spoiling this property.

Theorem 3.8. The scaled HDAF integral kernel of order $n \in \mathbb{N}$ is for any $x \in \mathbb{R}$ bounded by

$$|D_n(x, 0; \sqrt{2n+1}\sigma)| \leq \frac{(1 + \frac{1}{2n})^{1/2}}{\pi\sigma}. \quad (39)$$

For $|x| > (2n + 1)\sigma$, it exhibits Gaussian decay,

$$|D_n(x, 0; \sqrt{2n+1}\sigma)| \leq \frac{1}{\sqrt{n + \frac{1}{2}}\pi n!} \left[\left(\frac{x^2}{(4n+2)\sigma^2} \right)^{n+\frac{1}{2}} e^{-\frac{x^2}{(4n+2)\sigma^2}} + \frac{\frac{1}{2} \left(\frac{x^2}{(2n+1)\sigma^2} \right)^{n+\frac{1}{2}} e^{-\frac{x^2}{(2n+1)\sigma^2}}}{\frac{|x|^2}{(2n+1)\sigma} - (2n+1)\sigma + \sqrt{\left(\frac{|x|^2}{(2n+1)\sigma} - (2n+1)\sigma \right)^2 + \frac{4|x|^2}{\pi(n+\frac{1}{2})}}} \right]. \quad (40)$$

Proof. For the bounds on $D_n(x, 0; \sqrt{2n+1}\sigma)$ we distinguish several cases, depending on the choice of n and x . In the simplest one, we have $|x| \leq \sigma$ and apply $|\sin(s)| \leq |s|$ to the integrand of Eq. (11) resulting in

$$|I_n(x; \sqrt{2n+1}\sigma)| \leq \frac{|x|}{\sqrt{n + \frac{1}{2}}\sigma} \int_0^\infty t^{2n+2} e^{-t^2} dt. \quad (41)$$

Inserted in Eq. (10), this shows that $D(x, 0; \sqrt{2n+1}\sigma)$ is bounded by

$$|D_n(x, 0; \sqrt{2n+1}\sigma)| \leq \frac{\Gamma(n + \frac{3}{2})}{\pi n! \sqrt{n + \frac{1}{2}}\sigma} \leq \frac{(1 + \frac{1}{2n})^{1/2}}{\pi\sigma} \quad (42)$$

with the gamma function $\Gamma(m+1)$ that extends the factorial $m!$ to real arguments. The last estimate is again obtained from Ineq. (22) and only serves to illustrate the asymptotic behavior for large n .

The next case we consider is $\sigma < |x| \leq (2n+1)\sigma$. This time, we estimate $|\sin(s)| \leq 1$ in Eq. (11) and obtain algebraic decay,

$$|D_n(x, 0; \sqrt{2n+1}\sigma)| \leq \frac{1}{\pi|x|}. \quad (43)$$

The last case is the main part and establishes Gaussian decay of the HDAF kernel. It is in the domain given by $|x| > (2n+1)\sigma$. The estimate starts from the expression in Eqs. (10)-(11), written as

$$D_n(x, 0; \sqrt{2n+1}\sigma) = -i \frac{(n + \frac{1}{2})^{n+1}}{\pi n! x} \int_{-\infty}^\infty t^{2n+1} e^{ixt/\sigma - (n+\frac{1}{2})t^2} dt. \quad (44)$$

As a first step, we complete the square in the exponent. Since the integrand extends to a holomorphic function that exhibits Gaussian decay along the direction of the real axis, we can shift the line of integration from \mathbb{R} to $\mathbb{R} - i\frac{x}{(2n+1)\sigma}$ and obtain

$$\begin{aligned} D_n(x, 0; \sqrt{2n+1}\sigma) &= -i \frac{(n + \frac{1}{2})^{n+1}}{\pi n! x} e^{-\frac{x^2}{(4n+2)\sigma^2}} \int_{-\infty}^\infty t^{2n+1} e^{-(n+\frac{1}{2})(t - i\frac{x}{(2n+1)\sigma})^2} dt \\ &= -i \frac{(n + \frac{1}{2})^{n+1}}{\pi n! x} e^{-\frac{x^2}{(4n+2)\sigma^2}} \int_{-\infty}^\infty \left(t + i\frac{x}{(2n+1)\sigma} \right)^{2n+1} e^{-(n+\frac{1}{2})t^2} dt. \end{aligned} \quad (45)$$

This integral can be estimated by replacing the integrand with its absolute value and by splitting the domain of integration into lower and higher frequency parts,

$$\begin{aligned} & \int_{-\infty}^{\infty} \left(t^2 + \frac{x^2}{(2n+1)^2\sigma^2} \right)^{n+\frac{1}{2}} e^{-(n+\frac{1}{2})t^2} dt \\ & \leq 2 \int_0^{\frac{|x|}{(2n+1)\sigma}} \left(t^2 + \frac{x^2}{(2n+1)^2\sigma^2} \right)^{n+\frac{1}{2}} e^{-(n+\frac{1}{2})t^2} dt \\ & \quad + 2 \int_{\frac{|x|}{(2n+1)\sigma}}^{\infty} t^{2n+1} \left(1 + \frac{x^2}{(2n+1)^2\sigma^2 t^2} \right)^{n+\frac{1}{2}} e^{-(n+\frac{1}{2})t^2} dt. \end{aligned} \quad (46)$$

The first integral in Eq. (46) is bounded by replacing the exponential in the integrand with a larger, algebraic expression,

$$\begin{aligned} \left(t^2 + \frac{x^2}{(2n+1)^2\sigma^2} \right)^{n+\frac{1}{2}} e^{-(n+\frac{1}{2})t^2} & \leq \left(t^2 + \frac{x^2}{(2n+1)^2\sigma^2} \right)^{n+\frac{1}{2}} (1+t^2)^{-(n+\frac{1}{2})} \\ & \leq \left(1 + \frac{\frac{x^2}{(2n+1)^2\sigma^2} - 1}{1+t^2} \right)^{n+\frac{1}{2}} \leq \left(\frac{x^2}{(2n+1)^2\sigma^2} \right)^{n+\frac{1}{2}}. \end{aligned} \quad (47)$$

To bound the second integral in Eq. (46), we use $\frac{x^2}{(2n+1)^2\sigma^2 t^2} \leq 1$ and again Ineq. (19), analogous to the derivation of Eq. (23). Together, the two integrals are thus estimated by

$$\begin{aligned} & \int_{-\infty}^{\infty} \left(t^2 + \frac{x^2}{(2n+1)^2\sigma^2} \right)^{n+\frac{1}{2}} e^{-(n+\frac{1}{2})t^2} dt \\ & \leq \left(\frac{x^2}{(2n+1)^2\sigma^2} \right)^{n+\frac{1}{2}} \frac{|x|}{(n+\frac{1}{2})\sigma} \\ & \quad + \frac{\frac{|x|^{2n+1}}{2^{n-\frac{1}{2}}(n+\frac{1}{2})^{2n+2}\sigma^{2n+1}} e^{-\frac{x^2}{(4n+2)\sigma^2}}}{\frac{|x|}{(2n+1)\sigma} - \frac{(2n+1)\sigma}{|x|} + \sqrt{\left(\frac{|x|}{(2n+1)\sigma} - \frac{(2n+1)\sigma}{|x|} \right)^2 + \frac{4}{\pi(n+\frac{1}{2})\sigma}}}. \end{aligned} \quad (48)$$

Collecting all terms, we have

$$\begin{aligned} |D_n(x, 0; \sqrt{2n+1}\sigma)| & \leq \frac{(n+\frac{1}{2})^{n+1}}{\pi n! |x|} e^{-\frac{x^2}{(4n+2)\sigma^2}} \left[\left(\frac{x^2}{(2n+1)^2\sigma^2} \right)^{n+\frac{1}{2}} \frac{|x|}{(n+\frac{1}{2})\sigma} \right. \\ & \quad \left. + \frac{\frac{|x|^{2n+1}}{2^{n-\frac{1}{2}}(n+\frac{1}{2})^{2n+2}\sigma^{2n+1}} e^{-\frac{x^2}{(4n+2)\sigma^2}}}{\frac{|x|}{(2n+1)\sigma} - \frac{(2n+1)\sigma}{|x|} + \sqrt{\left(\frac{|x|}{(2n+1)\sigma} - \frac{(2n+1)\sigma}{|x|} \right)^2 + \frac{4}{\pi(n+\frac{1}{2})\sigma}} \right] \end{aligned} \quad (49)$$

which simplifies to the claimed Ineq. (40). \square

Remark 3.9. We are grateful for Mark Arnold's suggestion to consider non-integral values of n . It is interesting to note that for such n , the Gaussian bounds in the frequency domain generalize painlessly, whereas in the time domain there is only algebraic decay. The reason is that $\widehat{D}_n(k; \sigma)$ is for non-integral n only finitely often continuously differentiable, see Eq. (8).

3.4 Heisenberg Uncertainty Product for HDAFs

In this section we study the behavior of the Heisenberg uncertainty product for the HDAFs and show that it grows, up to logarithmic corrections, inverse proportionally to the square-root of the transition bandwidth.

Definition 3.10. The Heisenberg uncertainty product $\Delta x \Delta k$ of a function $f \in L^2(\mathbb{R})$ is a non-negative, possibly infinite quantity given by

$$\Delta x \Delta k := \inf_{a, b \in \mathbb{R}} \frac{\left(\int_{-\infty}^{\infty} (x-a)^2 |f(x)|^2 dx \int_{-\infty}^{\infty} (k-b)^2 |\widehat{f}(k)|^2 \frac{dk}{2\pi} \right)^{1/2}}{\int_{-\infty}^{\infty} |f(x)|^2 dx}. \quad (50)$$

Theorem 3.11. The uncertainty product of the HDAF integral kernel $D_n(x, 0; \sigma)$ is independent of the choice of $\sigma > 0$ and will be denoted by $(\Delta x)_n (\Delta k)_n$. Asymptotically, its growth is bounded by $n^{1/4}$. More explicitly, given any $\epsilon > 0$, there is $N \in \mathbb{N}_0$ such that $(\Delta x)_n (\Delta k)_n$ is for all $n \geq N$ bounded by

$$\begin{aligned} (\Delta x)_n (\Delta k)_n &= \frac{\left(\int_{-\infty}^{\infty} x^2 |D_n(x, 0; \sigma)|^2 dx \int_{-\infty}^{\infty} k^2 |\widehat{D}_n(k; \sigma)|^2 \frac{dk}{2\pi} \right)^{1/2}}{\int_{-\infty}^{\infty} |D_n(x, 0; \sigma)|^2 dx} \\ &\leq (1 + \epsilon) \frac{1}{\sqrt{3\pi}^{1/4}} (n + \frac{1}{2})^{1/4}. \end{aligned} \quad (51)$$

Proof. In Eq. (51), we can omit the infimization over the shift parameters a and b in Eq. (50), because the HDAF kernel and its Fourier transform are even functions. From the functional form of the HDAF integral kernel given in Eq. (2) or Eq. (12) and its Fourier transform Eq. (4), we conclude that the uncertainty product is independent of the length parameter σ . Therefore, we can set $\sigma = \sqrt{2n+1}$ and use that $\widehat{D}_n(k; \sqrt{2n+1})$ is a sequence of uniformly bounded functions that converges almost uniformly to the ideal window with bandwidth 2. Consequently, by the unitarity of the Fourier transform and by dominated convergence using the bound given in Eq. (23), we obtain

$$\lim_{n \rightarrow \infty} \int_{-\infty}^{\infty} |D_n(x, 0; \sqrt{2n+1})|^2 dx = \lim_{n \rightarrow \infty} \int_{-\infty}^{\infty} |\widehat{D}_n(k; \sqrt{2n+1})|^2 \frac{dk}{2\pi} = \frac{1}{\pi} \quad (52)$$

and

$$\lim_{n \rightarrow \infty} \int_{-\infty}^{\infty} k^2 |\widehat{D}_n(k; \sqrt{2n+1})|^2 \frac{dk}{2\pi} = \frac{1}{3\pi}. \quad (53)$$

With the scaling of σ used here, the only factor in the uncertainty product that contributes to its asymptotic growth with n is

$$\int_{-\infty}^{\infty} x^2 |D_n(x, 0; \sqrt{2n+1})|^2 = \int_{-\infty}^{\infty} \left| \frac{d}{dk} \widehat{D}_n(k; \sqrt{2n+1}) \right|^2 \frac{dk}{2\pi} \quad (54)$$

$$= \frac{2(n + \frac{1}{2})^{2n+2}}{\pi(n!)^2 (2n+1)^{2n+\frac{3}{2}}} \Gamma(2n + \frac{3}{2}) \quad (55)$$

$$\leq \frac{1}{\pi^{3/2} n^{2n+1}} (n + \frac{1}{2})^{2n+\frac{3}{2}} e^{-\frac{1}{2}} e^{\frac{1}{24n+6}} \leq \frac{1}{\pi^{3/2}} \sqrt{n + \frac{1}{2}} e^{\frac{1}{24n+6}}. \quad (56)$$

As usual, we have used simple inequality $(1 + \frac{y}{n})^n < e^y$ for $y > 1$ and Eq. (22) in the last two steps. The last inequality, together with the convergence of Eq. (52) and Eq. (53) as well as that of the marginal correction term $e^{\frac{1}{24n+6}} \rightarrow 1$ prove Ineq. (51). \square

3.5 Large-Order Behavior of the Scaled HDAF Kernel

To estimate the scaled HDAF kernel $D_n(x, 0; \sqrt{2n+1}\sigma)$ for large n , we follow ideas of asymptotic analysis in the evaluation of the integral in Eq. (11).

Theorem 3.12. The $n \rightarrow \infty$ asymptotics of the HDAF kernel are given by

$$D_n(x, 0; \sqrt{2n+1}\sigma) = \sqrt{\frac{2}{\pi}} \frac{(n + \frac{1}{2})^{n+\frac{1}{2}}}{n!} e^{-(n+\frac{1}{2})} \frac{\sin(x/\sigma)}{x} e^{-\frac{x^2}{(8n+4)\sigma^2}} + F_n(x; \sigma), \quad (57)$$

with an error term F_n that is bounded such that for every $\epsilon > 0$ there is an $N \in N_0$ satisfying

$$|F_n(x; \sigma)| \leq \frac{1 + \epsilon}{\pi|x|} \min\left\{\frac{|x|}{\sigma}, 1\right\} \frac{\sqrt{\ln n}}{\sqrt{n + \frac{1}{2}}} \quad (58)$$

for every $n \geq N$.

Proof. After a change of variables, we have

$$I_n(x; \sqrt{2n+1}\sigma) = (n + \frac{1}{2})^{n+1} \int_0^{\infty} \sin\left(\frac{tx}{\sigma}\right) t^{2n+1} e^{-(n+\frac{1}{2})t^2} dt \quad (59)$$

$$= (n + \frac{1}{2})^{n+1} e^{-(n+\frac{1}{2})} \int_{-\infty}^{\infty} \sin\left(\frac{tx}{\sigma}\right) e^{-(2n+1)(t-1)^2} dt + E_n(x; \sigma), \quad (60)$$

with an error term that is bounded by the sum of three contributions

$$|E_n(x; \sigma)| \leq (n + \frac{1}{2})^{n+1} \left[\Delta_1(x; \sigma) + \Delta_2(x; \sigma) + \Delta_3(x; \sigma) \right] \quad (61)$$

that arise by splitting the correction in integrals over three separate domains,

$$\Delta_1(x; \sigma) = \int_{-\infty}^{1-\tau} e^{-(n+\frac{1}{2})} e^{-(2n+1)(t-1)^2} dt \quad (62)$$

$$\Delta_2(x; \sigma) = \int_{1-\tau}^{1+\tau} \left| \sin\left(\frac{tx}{\sigma}\right) \right| \left| t^{2n+1} e^{-(n+\frac{1}{2})t^2} - e^{-(n+\frac{1}{2})} e^{-(2n+1)(t-1)^2} \right| dt \quad (63)$$

$$\Delta_3(x; \sigma) = \int_{1+\tau}^{\infty} t^{2n+1} e^{-(n+\frac{1}{2})t^2} dt \quad (64)$$

with an as yet unspecified split point at distance $0 < \tau < 1$ from the center of the Gaussian in the integrand. The first error term can be estimated directly with Ineq. (21) and gives

$$\Delta_1(x; \sigma) = \frac{e^{-(n+\frac{1}{2})} e^{-(2n+1)\tau^2}}{(2n+1) \left(\tau + \sqrt{\tau^2 + \frac{4}{\pi(2n+1)}} \right)}. \quad (65)$$

The second one involves Ineq. (17) and Ineq. (18) with $x = 1 - \tau$ and $y = 1 + \tau$, followed by extending the domain of integration to the entire real line,

$$\begin{aligned} \Delta_2(x; \sigma) &\leq \max_{1-\tau \leq t \leq 1+\tau} \left\{ \left| \sin\left(\frac{tx}{\sigma}\right) \right| \right\} \int_{1-\tau}^{1+\tau} e^{-(n+\frac{1}{2})} \left(e^{-(n+\frac{1}{2})(1+\frac{1}{(1+\tau)^2})(t-1)^2} \right. \\ &\quad \left. - e^{-(n+\frac{1}{2})(1+\frac{1}{(1-\tau)^2})(t-1)^2} \right) dt \quad (66) \\ &\leq \max_{1-\tau \leq t \leq 1+\tau} \left\{ \left| \sin\left(\frac{tx}{\sigma}\right) \right| \right\} \sqrt{\frac{\pi}{n+\frac{1}{2}}} e^{-(n+\frac{1}{2})} \left(\frac{1}{\sqrt{1+\frac{1}{(1+\tau)^2}}} - \frac{1}{\sqrt{1+\frac{1}{(1-\tau)^2}}} \right). \end{aligned}$$

To bound the third error term, we use again Ineq. (19) and Ineq. (21), resulting in

$$\begin{aligned} \Delta_3(x; \sigma) &\leq (1+\tau)^{2n+1} e^{(n+\frac{1}{2}) \left(\frac{1}{(1+\tau)^2} - 2 \right)} \int_{1+\tau}^{\infty} e^{-(n+\frac{1}{2})(t-\frac{1}{1+\tau})^2} dt \\ &\leq \frac{(1+\tau)^{2n+1}}{n+\frac{1}{2}} e^{(n+\frac{1}{2}) \left(\frac{1}{(1+\tau)^2} - 2 \right)} \frac{e^{-(n+\frac{1}{2})(1+\tau-\frac{1}{1+\tau})^2}}{1+\tau - \frac{1}{1+\tau} + \sqrt{(1+\tau - \frac{1}{1+\tau})^2 + \frac{4}{\pi(n+\frac{1}{2})}}}. \quad (67) \end{aligned}$$

Collecting the terms,

$$\begin{aligned}
|E_n(x; \sigma)| &\leq (n + \frac{1}{2})^{n+\frac{1}{2}} e^{-(n+\frac{1}{2})} \left(\frac{1}{\sqrt{4n+2}} \frac{e^{-(2n+1)\tau^2}}{\tau + \sqrt{\tau^2 + \frac{4}{\pi(2n+1)}}} \right. \\
&\quad + \sqrt{\pi}(1+\tau) \min\{\frac{|x|}{\sigma}, 1\} \left(\frac{1}{\sqrt{1 + \frac{1}{(1+\tau)^2}}} - \frac{1}{\sqrt{1 + \frac{1}{(1-\tau)^2}}} \right) \\
&\quad + \frac{1}{\sqrt{n + \frac{1}{2}}} (1+\tau)^{2n+1} e^{(n+\frac{1}{2})(\frac{1}{(1+\tau)^2}-1)} \\
&\quad \left. \times \frac{e^{-(n+\frac{1}{2})(1+\tau-\frac{1}{1+\tau})^2}}{1 + \tau - \frac{1}{1+\tau} + \sqrt{(1 + \tau - \frac{1}{1+\tau})^2 + \frac{4}{\pi(n+\frac{1}{2})}}} \right). \tag{68}
\end{aligned}$$

Choosing $\tau = \frac{c}{\sqrt{n+\frac{1}{2}}} \equiv \frac{c}{r_n}$ with some fixed $c > 0$ and the abbreviation $r_n := \sqrt{n + \frac{1}{2}}$, we can further estimate

$$\begin{aligned}
|E_n(x; \sigma)| &\leq (n + \frac{1}{2})^{n+\frac{1}{2}} e^{-(n+\frac{1}{2})} \left(\frac{e^{-2e^2}}{2(c + \sqrt{c^2 + \frac{2}{\pi}})} \right. \\
&\quad + \sqrt{\pi}(1 + \frac{c}{r_n}) \min\{\frac{|x|}{\sigma}, 1\} \frac{1}{4} \left(\frac{1}{(1 - \frac{c}{r_n})^2} - \frac{1}{(1 + \frac{c}{r_n})^2} \right) \\
&\quad \left. + \frac{(1 + \frac{c}{r_n})^{2n+1} e^{-2cr_n(1+\frac{c}{r_n})^{-3}} e^{-c^2(1+\frac{1}{(1+\frac{c}{r_n})^2})^2}}{c(1 + \frac{1}{(1+\frac{c}{r_n})^2}) + \sqrt{c^2(1 + \frac{1}{(1+\frac{c}{r_n})^2})^2 + \frac{4}{\pi}}} \right). \tag{69}
\end{aligned}$$

The middle term comes from the inequality

$$\sqrt{1+y} - \sqrt{1+x} \leq \frac{y-x}{2\sqrt{1+x}} \tag{70}$$

with $y = (1 - \frac{c}{r_n})^{-2}$ and $x = (1 + \frac{c}{r_n})^{-2}$ applied to the numerator and some elementary estimates in the denominator; the last term comes from

$$1 + \tau - \frac{1}{1+\tau} \geq \tau \left(1 + \frac{1}{(1+\tau)^2} \right). \tag{71}$$

To adjust the leading order of the three error contributions, we replace the constant c by the sequence $c_n = \sqrt{\frac{1}{2} \ln(\frac{n\sigma}{|x|})}$ for $\frac{|x|}{\sigma} < 1$ and otherwise $c_n = \sqrt{\frac{1}{2} \ln n}$.

Inserting this in Ineq. (69) shows that independently of x and σ , the first and third term decay faster than the middle term. Consequently, for every $\epsilon > 0$

there is an $N > 0$ such that for each $n \geq N$ we have

$$|E_n(x; \sigma)| \leq (1 + \epsilon) \min\left\{\frac{|x|}{\sigma}, 1\right\} \sqrt{\pi} \frac{\sqrt{\frac{1}{2} \ln n}}{\sqrt{n + \frac{1}{2}}} \left(n + \frac{1}{2}\right)^{n + \frac{1}{2}} e^{-(n + \frac{1}{2})}. \quad (72)$$

To obtain this bound from Ineq. (69), we have again estimated the middle term using $\frac{1}{(1-\tau)^2} - \frac{1}{(1+\tau)^2} \leq 4\frac{\tau}{(1-\tau)^3}$ and bounded the denominator by a constant for sufficiently large n .

Now we perform the Fourier integral in Eq. (60) and obtain

$$D_n(x, 0; \sqrt{2n+1}\sigma) = \sqrt{\frac{2}{\pi}} \frac{(n + \frac{1}{2})^{n + \frac{1}{2}}}{n!} e^{-(n + \frac{1}{2})} \frac{\sin(x/\sigma)}{x} e^{-\frac{x^2}{(8n+4)\sigma^2}} + F_n(x; \sigma),$$

with the final error bound

$$|F_n(x; \sigma)| = \frac{2}{\pi n! |x|} |E_n(x; \sigma)| \leq \frac{1 + \epsilon}{\pi |x|} \min\left\{\frac{|x|}{\sigma}, 1\right\} \frac{\sqrt{\ln n}}{\sqrt{n + \frac{1}{2}}} \quad (73)$$

that is valid for any given $\epsilon > 0$ after a sufficiently large order n is reached. \square

Remark 3.13. The windowed sinc function appearing in Eq. (57) differs from the so-called Gaussian-sinc-DAF [10] only by a constant factor that can be shown to be arbitrarily close to one using the Stirling-Robbins estimates Ineq. (22). In the limit $n \rightarrow \infty$, F_n vanishes and Eq. (57) converges to the well known Fourier transform of the ideal low-pass filter

$$\lim_{n \rightarrow \infty} D_n(x, 0; \sqrt{2n+1}\sigma) = \frac{\sin(x/\sigma)}{\pi x} \quad (74)$$

with a cut-off frequency of $1/\sigma$. If error bounds are not desired, we can derive this limit by applying a slightly different scaling of σ with $\sqrt{2n}$ to Eq. (12) in combination with the scaling limit [11, 22.15.4] of Hermite polynomials.

4 Conclusion

In this paper, we have analytically verified previous, numerical evidence that suggests the use of HDAFs as arbitrarily controllable band-pass filters. The properties we demonstrated were that 1) by proper scaling, HDAFs of order n approximate the so-called ‘‘ideal window’’, pointwise bounds show the convergence is uniform at a rate of $n^{-1/2}$, up to a transition region that asymptotically decays in width at least as $\frac{1}{\sigma} \sqrt{\ln n/n}$; 2) HDAFs exhibit a high degree of regularity including Gaussian decay in both time and frequency domains; and 3) the growth of the Heisenberg uncertainty product for HDAFs is asymptotically

bounded by a constant times $n^{1/4}$. Although there was substantial computational evidence to support the conjecture that the HDAFs did, indeed, satisfy these conditions, an analytical demonstration was lacking [9]. At the same time, we were lacking a simple prescription for choosing the parameters to control the behavior of HDAF band-pass filter. The strategy we followed in the present paper was to examine the HDAF in the frequency domain in order to obtain quantitative conditions governing the choice of the parameters n and σ .

In addition, we have obtained an asymptotic relation between the HDAF integral kernel and a windowed sinc function that has proved highly useful in numerical applications and is known as the Gaussian-sinc-DAF [10], a sinc function windowed by a Gaussian envelope. Such a relationship has long been suspected, and the present result explains how it comes about.

Finally, HDAFs can be used for converting a signal from analog to digital in a customary two-step process of filtering and sampling. Instead of applying a simple high-frequency cut-off to an analog signal, we convolve it with an HDAF integral kernel and thus *effectively* truncate it in the frequency domain. While there remains an unwanted, higher frequency component, we can suppress its contribution to achieve any desired level of attenuation. The benefit of this procedure is the numerical robustness guaranteed by the Gaussian decay of HDAFs in both domains. For such HDAF-filtered signals, the results obtained here can be combined with an approximate sampling theorem [19, 20], which is anticipated to have a wide range of applications, in particular in the numerical solution of linear and non-linear ordinary and partial differential equations.

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