

$$= \frac{W(e^{i\theta})}{R} \sum_{n=-\infty}^{\infty} x_m R(n), \quad (2b)$$

the short-time Fourier transform is evaluated, R

$$(n) = x(n) \left[\frac{W(e^{j\omega_0})}{N} \sum_{m=-\infty}^{\infty} w(mR-n) \right] \quad (2a)$$

The short-time Fourier transform of a signal $x(u)$, at time mR , is defined as

$$X^m(e^{j\omega_0 u}) = \sum_{l=0}^{mR-L+1} x(l)w(mR-l) e^{-j\omega_0 l} \quad (1a)$$

[1] that any signal $x(u)$ can be recovered within small window [1–3], by overlap adding the windowed signals $X^m(e^{j\omega_0 u})$.

[2] The short-time Fourier transform is the short-time transformation of the signal $w(u)$ is a causal window function of length L and [3]. It denotes the discrete Fourier transform of adjacent samples of the short-time transformation of the signal $w(u)$.

[3] The short-time Fourier transform of a signal $x(u)$ is defined as

$$F[x](\tau)(MR-u) = \sum_{l=0}^{mR-L+1} x(l)w(mR-l) \quad (1b)$$

[4] that any short-time transform from the short-time transform is specified at or above the Nyquist estimation rate of the window [1–3] by overlap adding the windows.

Introduction

The short-time Fourier transform of a signal

Keywords. Short-time analysis, aliasing errors, window functions, digital filters, sum.

Le terme est écrit de la même manière que l'anglais *multiple*. Les résultats de simulations sont présentés pour les cas des émetteurs de Hamming, Dölpf-Schöbyscheff, Kaiser et rectangle uniforme.

Derzeit besteht die Fakultät für Betriebswirtschaftslehre aus drei Institute mit insgesamt 12 Professoren und ca. 100 weiteren Lehr- und Forschungspersonal. Die Fakultät ist in sechs Institute gegliedert: Institut für Betriebswirtschaftslehre, Institut für Betriebswirtschaftslehre der Betriebswirtschaft, Institut für Betriebswirtschaftslehre der Betriebswirtschaft, Institut für Betriebswirtschaftslehre der Betriebswirtschaft, Institut für Betriebswirtschaftslehre der Betriebswirtschaft und Institut für Betriebswirtschaftslehre der Betriebswirtschaft.

Computer simulations using rectangular, Hamming, Dohc-Chelyshev, and Kaiser windows are compared to results of the digital Poisson motion formula. The results are discussed in terms of error specification.

Received 31 May 1980
Revised 5 January 1981

Acoustics Research Department, Bell Laboratories, Murray Hill, NJ 07974, USA

School of Electrical Engineering, Oklahoma State University, Stillwater, OK 74078, USA

R. YARLAGADDA

VALIDATING ERRORS IN SHORT-TIME ANALYSIS

VALIDATING ERRORS IN SHORT-TIME ANALYSIS

SHORT COMMUNICATION

North-Holland Publishing Company
Signal Processing 4 (1982) 79-84

where $W(e^{j0})$ corresponds to the zero frequency value of the window.

The derivation of eq. (2) makes use of the approximate relation

$$\frac{R}{W(e^{j0})} \sum_{m=-\infty}^{\infty} w(mR - n) \approx 1. \quad (3)$$

This relation is exact if $w(n)$ is bandlimited to a frequency of $(L/2RT)$, where L is the length of the window in samples and T is the length of the time interval for which $w(t)$ is nonzero [1]. Since $w(n)$ is time limited, its spectrum cannot be bandlimited. Hence in order to obtain an equality, eq. (3) must be modified to include an error term

$$\frac{R}{W(e^{j0})} \sum_{m=-\infty}^{\infty} w(mR - n) = 1 + e_R(n), \quad (4)$$

where $e_R(n)$ takes into consideration the aliasing errors due to the non-bandlimited window function. Note that the subscript R is introduced to denote that the error signal $e_R(n)$ depends on R , the decimation period.

The purpose of this paper is to quantify the magnitude of the error term $e_R(n)$. This may be done in terms of a discrete version of the Poisson summation formula. We apply this result in order to find explicit formulae for $\max_n(|e_R(n)|)$ and $\text{RMS}(e_R(n))$. These results are then illustrated using rectangular, Hamming, Dolph-Chebyshev, and Kaiser windows as a function of R .

1. Numerical examples

For notational purposes, Fig. 1 identifies the basic definitions of the characteristic time T and the frequency length F for a Dolph-Chebyshev window. T denotes the length of the time interval for which $w(t)$ is nonzero, $W(f)$ denotes the Fourier transform of $w(t)$, and F denotes the smallest frequency for which $|W(f)| < \delta$ for $|f| > \frac{1}{2}F$. F is approximately equal to the main lobe width. It is clear that T and F can be defined for other windows similarly. Using $(1/F)$ as the sampling period (i.e., the Nyquist period), then $T/(1/F) = TF$ gives

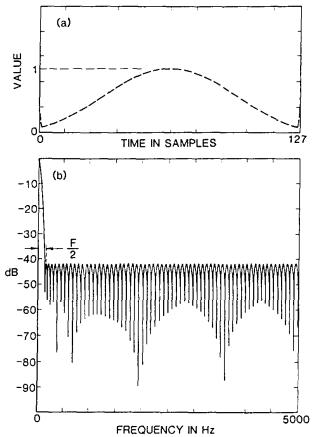


Fig. 1. Dolph-Chebyshev window with a -42 dB side-lobe level and its spectrum.

the number of sample periods in time T for the short-time analysis output. Therefore the ratio (L/TF) gives the Nyquist upper bound for the decimation period R identified earlier. The product TF (the time-bandwidth product of the window) will be useful, and is identified here by Q ,

$$Q = TF. \quad (5)$$

Using this notation, $R < L/Q$. For a Hamming window which has a -42 dB side-lobe level, Q is approximately 4.0 [1]. For a Dolph-Chebyshev window, Q is approximately [4, eq. (17)]

$$Q \cong \frac{2}{\pi} \ln \left[\frac{2}{\delta} \right]. \quad (6)$$

For example, for a Chebyshev window with a -42 dB ($\delta = 0.0079$) side-lobe level, Q is about 3.52. For the sake of comparison, windows with the same Q were used here. Thus, a Chebyshev window with $Q = 4$, which has $\delta = 0.00374$ (-48.5 dB) side-lobe level was chosen. For a

the spectrum of a rectangular window drop off at 12 dB/oct, the error curves for a rectangular window do not compare favorably with the error curves of the other windows for arbitrary values of R , as will be shown analytically.

2. Digital Poisson summation formula

If $w(t)$ is any arbitrary function, with $W(f)$ its Fourier transform, then according to the classical Poisson formula [8]

$$\sum_{m=-\infty}^{\infty} w(t+mT) = \frac{1}{T} \sum_{m=-\infty}^{\infty} W\left(\frac{m}{T}\right) e^{j2\pi(m/T)t}$$

The corresponding digital equivalent formula is given as follows. Let $w(n)$ be any sequence with

$$W(z) = \sum_{l=-\infty}^{\infty} w(l)z^{-l}, \quad (8)$$

its z transform. Then the digital Poisson formula is [9, eqs. (3-15), (3-16)],

$$\sum_{m=-\infty}^{\infty} w(m+R) = \frac{1}{R} \sum_{p=0}^{R-1} W(e^{j(2\pi/R)p}) e^{j(2\pi/R)pn}. \quad (9)$$

Eq. (9) differs from the classical "analogue" Poisson formula in that the righthand sum is finite and is over the z transform of $w(n)$ evaluated on the unit circle at R equispaced points.

3. Analysis of overlap-add errors

In this section, the error, $e_R(n)$, in eq. (4) is quantified using the results of the last section. From eq. (9) and eq. (4), and letting $\omega_p' = 2\pi p/R$, we have

$$\begin{aligned} e_R(n) &= \frac{1}{W(e^{j0})} \sum_{p=0}^{R-1} W(e^{j\omega_p'}) e^{jn\omega_p'} - 1 \\ &= \sum_{p=1}^{R-1} \left[\frac{W(e^{j\omega_p'})}{W(e^{j0})} \right] e^{jn\omega_p'}, \end{aligned} \quad (10)$$

since the $p=0$ term is identically 1. This implies that the error can be computed from the samples of the z transform of the window sequence. It is clear that when $R=1$ the error is exactly zero. For $R>1$, eq. (10) gives a simple procedure for the computation of $e_R(n)$ using an inverse DFT. This can be seen by noting that the DFT coefficients of $e_R(n)$ can be determined from eq. (10) and are given by

$$E_R(p) = \text{DFT}(e_R(n)) = \begin{cases} 0, & p=0 \\ \frac{R}{W(e^{j0})} W(e^{j\omega_p}), & 1 \leq p \leq R-1. \end{cases} \quad (11)$$

Two useful bounds on $e_R(n)$ can be obtained. First, using the triangle inequality on eq. (10), we have

$$|e_R(n)| \leq \frac{1}{W(e^{j0})} \sum_{p=1}^{R-1} |W(e^{j\omega_p})|. \quad (12)$$

We define \hat{e}_{\max} as the right side of eq. (12). Second, the root mean square error (RMS(e)) per sample can be obtained using Parseval's theorem [9] and eq. (11), where the RMS error per sample is defined as

$$\begin{aligned} e_{\text{RMS}} &= \left(\frac{1}{R} \sum_{n=0}^{R-1} e_R^2(n) \right)^{1/2} \\ &= \frac{1}{W(e^{j0})} \left(\sum_{p=1}^{R-1} |W(e^{j\omega_p})|^2 \right)^{1/2}. \end{aligned} \quad (13)$$

Interestingly, if

$$\left| \frac{W(e^{j\omega_p})}{W(e^{j0})} \right| \leq \delta \quad \text{for } p=1, \dots, R-1, \quad (14)$$

then

$$e_{\text{RMS}} \leq \sqrt{R-1}\delta = \hat{e}_{\text{RMS}}. \quad (15)$$

Note that eq. (14) is realistic when R satisfies the Nyquist decimation criterion discussed earlier and δ bounds the out-of-band-peak spectral error. For larger values of R , eq. (14) and, therefore, eq. (15) are not valid. Eq. (15) is an excellent bound relating the RMS error per sample, window decimation period R , and the side-lobe attenuation of the window δ , when the Nyquist condition $R < L/Q$ has been satisfied.

For a rectangular window, $e_R(n)$ is exactly zero for some values of R . This can be seen by noting that $W(e^{j\omega_p})=0$ for $\omega_p = p2\pi/L$, $p \neq 0$, where L corresponds to length of the window [10]. When $\omega_p' = p2\pi/R$, $p=1, \dots, R-1$, coincide with these frequencies, the error is zero. This happens when L is divisible by R . An illustrated numerical example of this is given in the next section.

4. Computer simulations

In this section, computer simulations of some of the results in the last few sections are given. Let

$$\begin{aligned} e_{\max} &= \max |e_R(n)|, \\ \hat{e}_{\max} &= \frac{1}{W(e^{j0})} \sum_{p=1}^{R-1} |W(e^{j\omega_p})|. \end{aligned}$$

From our theory leading to eq. (12), we have

$$\hat{e}_{\max} \geq e_{\max}.$$

Fig. 4 gives the results for a 128 point Dolph-Chebyshev window having a 42 dB side-lobe attenuation with R varying from 4 to 128. Figs. 4a and 4b show respectively the theoretic \hat{e}_{\max} and measurement e_{\max} (eq. (12)) for the peak error. Figs. 4c and 4d respectively give the measured (true) and theoretic RMS error (right-hand side of eq. (13)). The spectral values required by the theory are computed using a 1024 point

fast Fourier transform (FFT). The deviations in the two curves Figs. 4c, d is due to the fact that the $|W(e^{j\omega_p})|$ cannot be computed exactly for all p and R values required from a single 1024 point FFT. Fig. 4e gives the plot of $(\sqrt{R}-1)\delta$, illustrating the utility of eq. (15) for the Chebyshev window.

Figs. 5a, b, c and d give respectively the RMS errors e_{RMS} (eq. (13)) for four different 128 point windows, a Dolph-Chebyshev, a Hamming window, a rectangular window, and a Kaiser window with R from 4 to 128. Fig. 5e is $(\sqrt{R}-1)\delta$, with

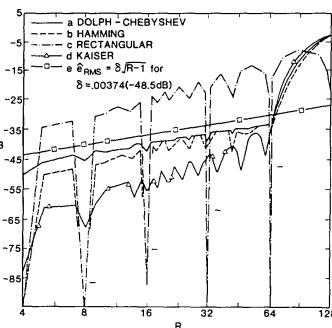


Fig. 5. Comparison of e_{RMS} for: a. Dolph-Chebyshev window, b. Hamming window, c. Rectangular window, d. Kaiser window, e. Plot of $\hat{e}_{\text{RMS}} = \sqrt{R}-1$, $\delta = 0.00374$ (-48.5 dB). These results are computed using the time domain method for each integer value of R , $4 \leq R \leq 128$. Squares and triangles are used on some curves to distinguish these curves from others.

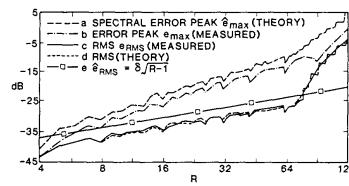


Fig. 4. For a Dolph-Chebyshev window with a -42 dB side-lobe level: a. Plot of \hat{e}_{\max} , b. Plot of e_{\max} , c. Plot of e_{RMS} from time values, d. Plot of e_{RMS} from spectral values, e. Plot of $\hat{e}_{\text{RMS}} = \sqrt{R}-1$, $\delta = 0.0079$. Squares are used on this curve to distinguish this curve from others.

$\delta = 0.00374$ (-48.5 dB). The rectangular window has zero RMS error for R a power of 2. The Hamming, Chebyshev, and Kaiser windows all have time-band width products Q of 4. For the case of $Q=4$, the Chebyshev window has a slightly larger RMS error (and much larger peak error, Figs. 2 and 3) when compared to the Hamming window. However, the Chebyshev window is more flexible as far as the range of side-lobe attenuation