# MUS421 Lecture 3A FFT Windows

Julius O. Smith III (jos@ccrma.stanford.edu)
Center for Computer Research in Music and Acoustics (CCRMA)
Department of Music, Stanford University
Stanford, California 94305

June 27, 2020

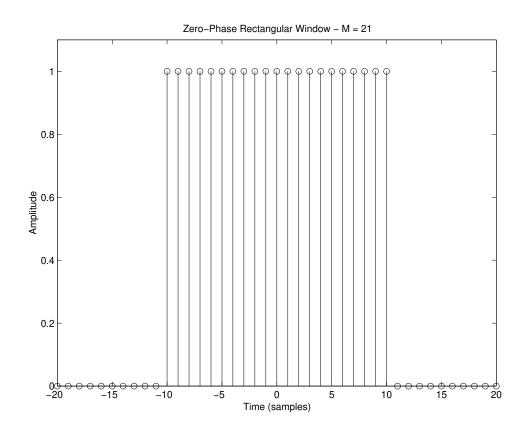
#### Outline

- Rectangular, Hann, Hamming
- MLT Sine
- Blackman-Harris Window Family
- Bartlett
- Poisson
- Slepian and Kaiser
- Dolph-Chebyshev
- Gaussian
- Optimal Windows

# The Rectangular Window

Previously, we looked at the rectangular window:

$$w_R(n) \stackrel{\Delta}{=} \left\{ egin{array}{l} 1, & |n| \leq rac{M-1}{2} \\ 0, & ext{otherwise} \end{array} 
ight.$$



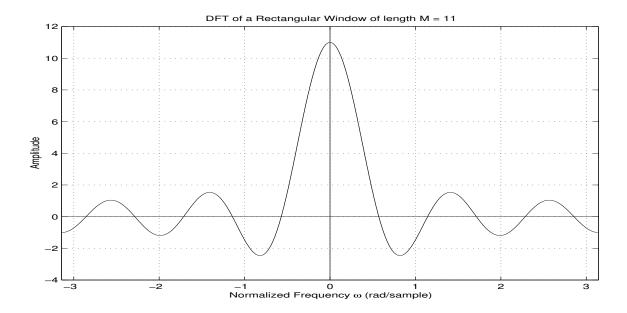
The window transform (DTFT) was found to be

$$W_R(\omega) = \frac{\sin\left(M\frac{\omega}{2}\right)}{\sin\left(\frac{\omega}{2}\right)} \stackrel{\Delta}{=} M \cdot \mathsf{asinc}_M(\omega) \tag{1}$$

where  $\operatorname{asinc}_M(\omega)$  denotes the aliased sinc function.

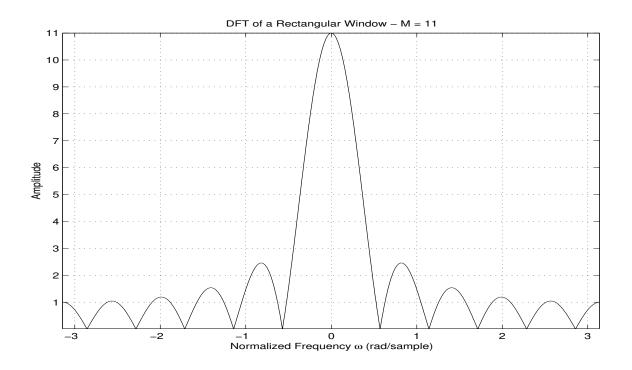
$$\mathsf{asinc}_M(\omega) \stackrel{\Delta}{=} \frac{\sin(M\omega/2)}{M \cdot \sin(\omega/2)}$$

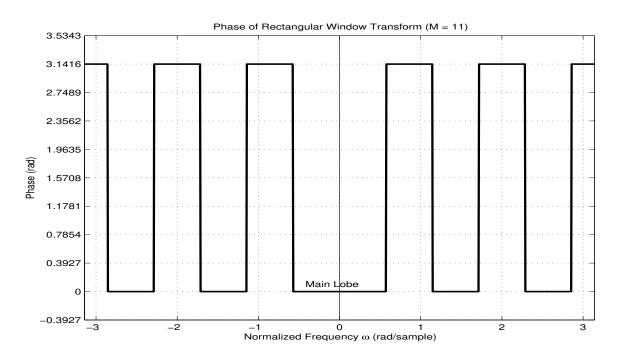
This result is plotted below:



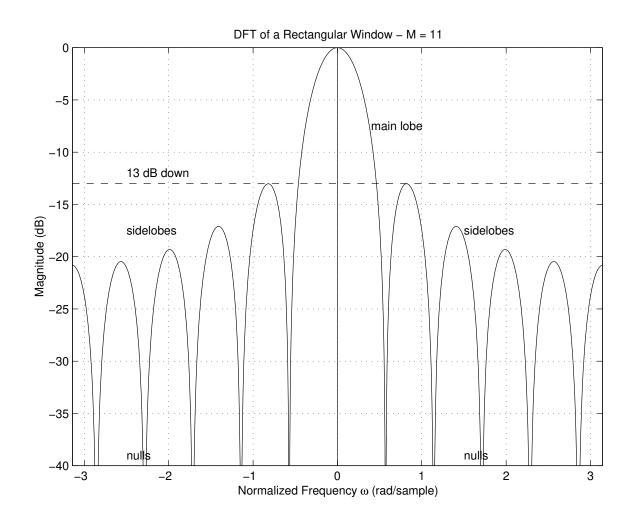
Note that this is the complete window transform, not just its magnitude. We obtain real window transforms like this only for symmetric, zero-centered windows.

More generally, we may plot both the *magnitude* and *phase* of the window versus frequency:

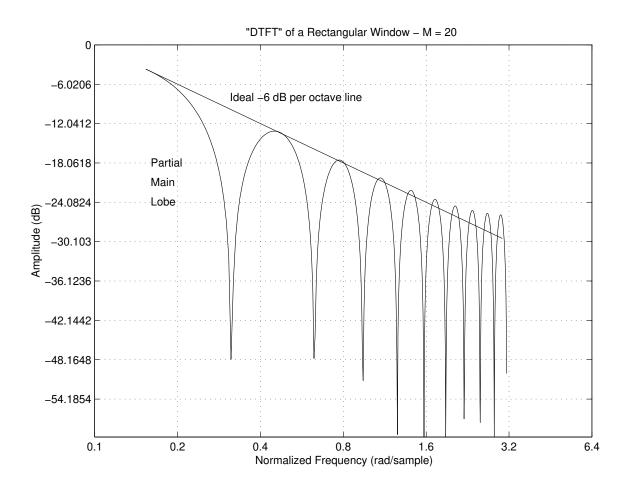




# In audio work, we more typically plot the window transform magnitude on a *decibel* (*dB*) *scale*:



Since the DTFT of the rectangular window approximates the sinc function, it should "roll off" at approximately 6 dB per octave, as verified in the log-log plot below:



As the sampling rate approaches infinity, the rectangular window transform converges exactly to the sinc function. Therefore, the departure of the roll-off from that of the sinc function can be ascribed to *aliasing* in the frequency domain, due to sampling in the time domain.

#### Sidelobe Roll-Off Rate

In general, when only the first n terms exist in the power-series expansion of a continuous function w(t) (i.e., each term is finite), then the Fourier Transform magnitude  $|W(\omega)|$  is asymptotically proportional to

$$|W(\omega)| \to \frac{1}{\omega^n} \quad (\text{as } \omega \to \infty)$$

**Proof:** Papoulis, **Signal Analysis**, McGraw-Hill, 1977 Thus, we have the following rule-of-thumb:

$$n \text{ terms} \leftrightarrow -6n \text{ dB per octave roll-off rate}$$

(since 
$$-20 \log_{10}(2) = 6.0205999...$$
).

This is also -20n dB per *decade*.

To apply this result, we normally only need to look at the window's *endpoints*. The interior of the window is usually differentiable of all orders.

## Example Roll-Off Rates:

- Amplitude discontinuity  $(n=1) \leftrightarrow -6 \text{ dB/octave}$
- Slope discontinuity  $(n=2) \leftrightarrow -12 \text{ dB/octave}$
- Curvature discontinuity  $(n=3) \leftrightarrow -18 \text{ dB/octave}$

For discrete-time windows, the roll-off rate slows down at high frequencies due to aliasing.

In summary, the DTFT of the M-sample rectangular window is proportional to the 'aliased sinc function':

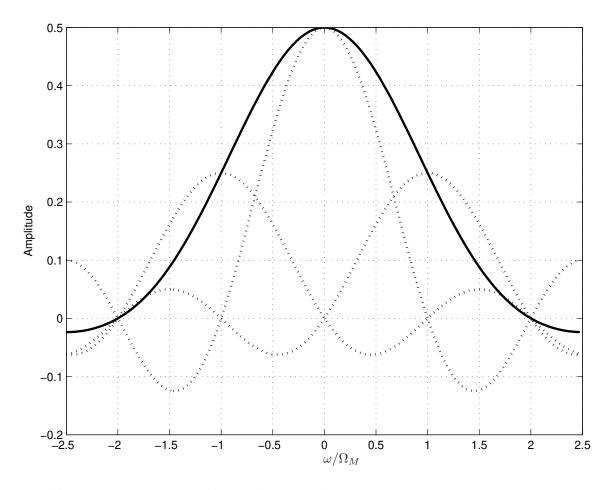
$$\begin{aligned} \operatorname{asinc}_M(\omega T) \; & \stackrel{\triangle}{=} \; \frac{\sin(\omega MT/2)}{M \cdot \sin(\omega T/2)} \\ & \approx \; \frac{\sin(\pi fMT)}{M\pi fT} \stackrel{\triangle}{=} \operatorname{sinc}(fMT) \end{aligned}$$

#### Points to note:

- Zero crossings at integer multiples of  $\Omega_M \stackrel{\Delta}{=} \frac{2\pi}{M}$  where  $\Omega_M \stackrel{\Delta}{=} \frac{2\pi}{M} =$  frequency sampling interval for a length M DFT
- Main lobe width is  $2\Omega_M = \frac{4\pi}{M}$
- ullet As M gets bigger, the mainlobe narrows (better frequency resolution)
- *M* has no effect on the height of the side lobes (Same as the "Gibbs phenomenon" for Fourier series)
- First sidelobe only 13 dB down from main-lobe peak
- Side lobes roll off at approximately 6 dB per octave
- A phase term arises when we shift the window to make it causal, while the window transform is real in the zero-centered case (i.e., centered about time 0)

# **Generalized Hamming Window Family**

Consider the following picture in the frequency domain:



https://ccrma.stanford.edu/~jos/Windows/Generalized\_Hamming\_Window\_Family.html

We have added 2 extra aliased sinc functions (shifted), which results in the following behavior:

- There is some cancellation of the side lobes
- The width of the main lobe is doubled

In terms of the rectangular window transform  $W_R(\omega)=M\cdot {\sf asinc}_M(\omega)$  (zero-centered, unit-amplitude case), this can be written as:

$$W_H(\omega) \stackrel{\Delta}{=} \alpha W_R(\omega) + \beta W_R(\omega - \Omega_M) + \beta W_R(\omega + \Omega_M)$$

Using the Shift Theorem dual, we can take the inverse transform of the above equation:

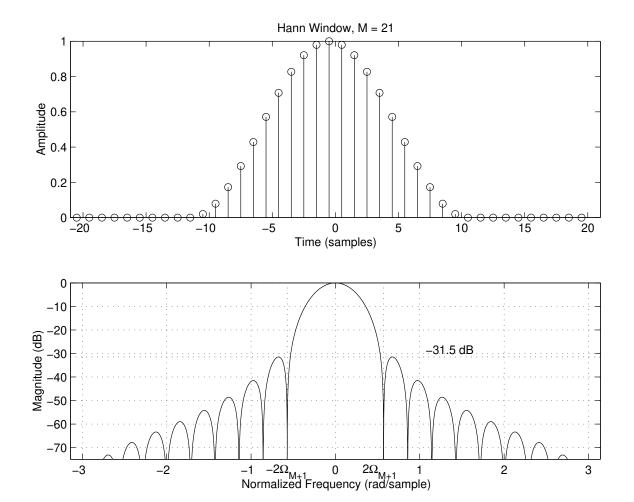
$$w_H = \alpha w_R(n) + \beta e^{-j\Omega_M n} w_R(n) + \beta e^{j\Omega_M n} w_R(n)$$
$$= w_R(n) \left[ \alpha + 2\beta \cos \left( \frac{2\pi n}{M} \right) \right]$$

Choosing various parameters for  $\alpha$  and  $\beta$  result in different windows in the generalized Hamming family, some of which have names.

## Hann or Hanning or Raised Cosine

The Hann window is defined by the settings  $\alpha=1/2$  and  $\beta=1/4$ :

$$w_H(n) = w_R(n) \left[ \frac{1}{2} + \frac{1}{2} \cos(\Omega_M n) \right] = w_R(n) \cos^2 \left( \frac{\Omega_M}{2} n \right)$$



## Hann window properties:

- ullet Main lobe is  $4\Omega_M$  wide
- $\bullet$  First side lobe is at  $-31~\mathrm{dB}$
- ullet Side-lobes roll off at  $pprox 18~\mathrm{dB}$  / octave

# Compare to the Rectangular window:

- $\bullet$  Main lobe is  $2\Omega_M$  wide
- First side lobe at -13 dB
- ullet Side-lobes roll off at  $pprox 6~\mathrm{dB}$  / octave

## **Hamming**

This window is determined by choosing  $\alpha$  to cancel the first side lobe and  $\beta$  to normalize peak amplitude to 1 in the time domain:

$$\alpha = \frac{25}{46} \approx 0.54$$

$$\beta = (1 - \alpha)/2$$

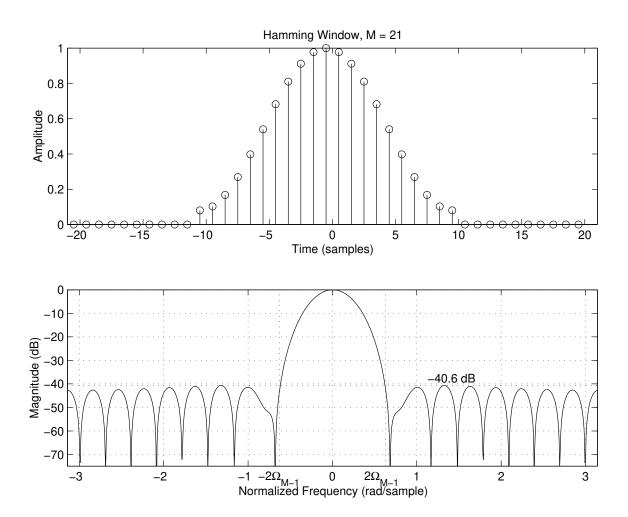
**Note:** The Hamming window is very close to the generalized Hamming window which *minimizes sidelobe level* within the family:

$$\alpha = 0.53836$$
 (minimum peak side-lobe magnitude)

Thus, the Hamming window is the "Chebyshev Generalized Hamming Window" rounded to two significant digits.

Chebyshev-type designs generally exhibit *equiripple* error behavior, since the worst-case error (sidelobe level in this case) is minimized (see Dolph-Chebyshev window below)

# **Hamming Window**

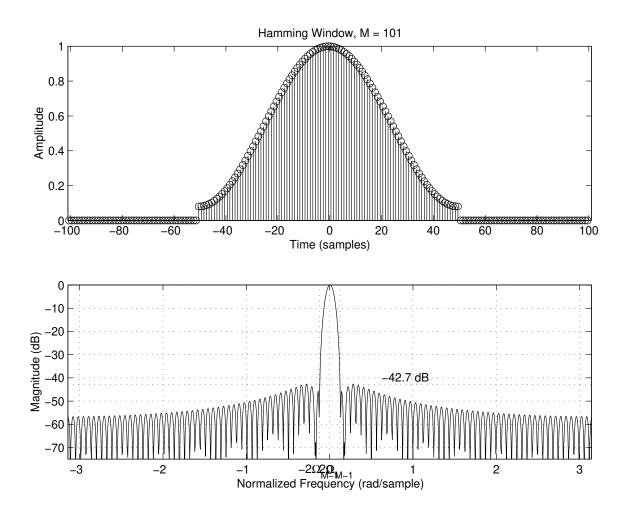


# **Hamming Window Properties**

- Discontinuous "slam to zero" at endpoints
- main lobe is  $4\Omega_M$  (like Hann)
- Roll off is approx. 6 dB/octave (but aliased)
- 1st side lobe is improved over Hann
- side lobes closer to "equal ripple"

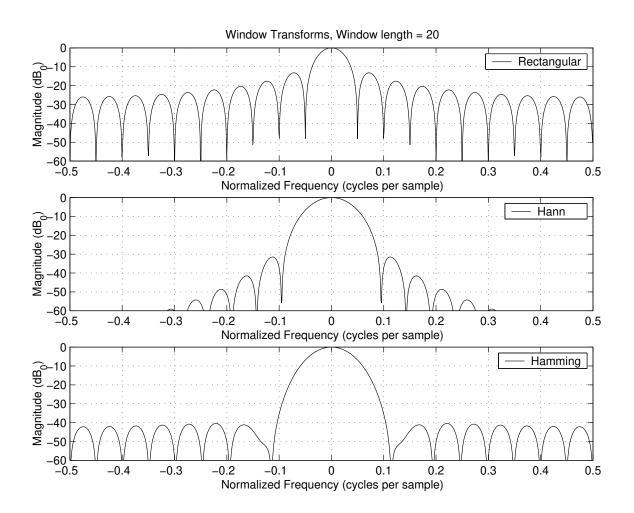
**Question:** How can side-lobes approximate "equal ripple" when they cannot roll-off slower than 6 dB / octave?

## **Longer Hamming Window**



- Since the side-lobes nearest the main lobe are most affected by the Hamming optimization, we now have a larger frequency region over which the spectral envelope looks like that of the asinc function (an "aliased -6 dB/octave roll-off").
- The side-lobe level (-42.7 dB) is also improved over that of the shorter window (-40.6 dB).

# **Window Transform Summary**



#### The MLT Sine Window

The Modulated Lapped Transform (MLT) uses the sine window:

$$w(n) = \sin\left[\left(n + \frac{1}{2}\right)\frac{\pi}{2M}\right], \quad n = 0, 1, 2, \dots, 2M - 1.$$

- Used in MPEG-1, Layer 3 (MP3 format), MPEG-2 AAC, MPEG-4
- Sidelobes 24 dB down
- Asymptotically optimal coding gain
- Zero-phase-window transform ("truncated cosine window") has smallest moment of inertia over all windows:

$$\int_{-\pi}^{\pi} \omega^2 W(\omega) d\omega = \min$$

# **Blackman-Harris Window Family**

- The Blackman-Harris family of windows is basically a generalization of the Hamming family.
- In the case of the Hamming family, we constructed a summation of 3 shifted sinc functions.
- The Blackman-Harris family is derived by considering a more general summation of shifted sinc functions:

$$w_B(n) = w_R(n) \sum_{l=0}^{L-1} \alpha_l \cos(l\Omega_M n)$$

where 
$$\Omega_M \stackrel{\Delta}{=} 2\pi/M$$
,  $n = -(M-1)/2, \ldots (M-1)/2$ , ( $M$  odd).

## **Special Cases:**

- $L=1 \Rightarrow \mathsf{Rectangular}$
- $L=2 \Rightarrow$  Generalized Hamming
- $L = 3 \Rightarrow \mathsf{Blackman} \; \mathsf{Family}$
- $L > 3 \Rightarrow \mathsf{Blackman}\mathsf{-Harris}$  Family

## **Frequency-Domain Implementation**

The Blackman-Harris window family can be very efficiently implemented in the frequency domain as a (2L-1)-point convolution with the spectrum of the unwindowed data. Example:

- 1. Hann Window = 3-Point  $DFT_M$  Smoother:
  - ullet Start with a length M rectangular window
  - ullet Take an M-point DFT
  - Convolve the DFT data with the 3-point smoother [1/4, 1/2, 1/4] to implement a Hann window
  - Note that the Hann window requires *no multiplies* in linear fixed-point data formats
- 2. Any Blackman window is a 5-point smoother for a Length M (critically sampled) DFT

#### Classic Blackman

The so-called "Blackman Window" is the specific case in which  $\alpha_0=0.42$   $\alpha_1=0.5$ , and  $\alpha_2=0.08$ 

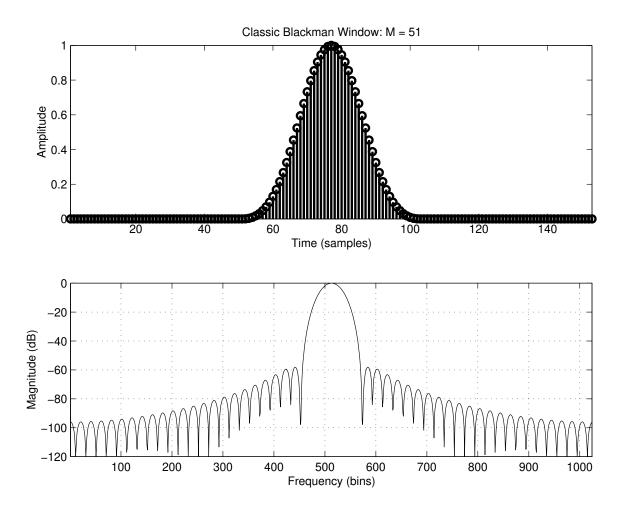
## **Properties:**

- ullet Sidelobes roll off about 18 dB per octave (as T o 0)
- -58 dB sidelobe level (worst case)
- One degree of freedom used to increase the roll-off rate from 6 dB/octave to 18 dB per octave
- One degree of freedom used to minimize sidelobes
- One degree of freedom used to scale the window

#### Matlab:

```
N = 101; L = 3; No2 = (N-1)/2; n=-No2:No2;
ws = zeros(L,3*N); z = zeros(1,N);
for l=0:L-1
   ws(l+1,:) = [z,cos(l*2*pi*n/N),z];
end
alpha = [0.42,0.5,0.08]; % Classic Blackman
w = alpha * ws;
```

## Classic Blackman Window and Transform



#### Three-Term Blackman-Harris

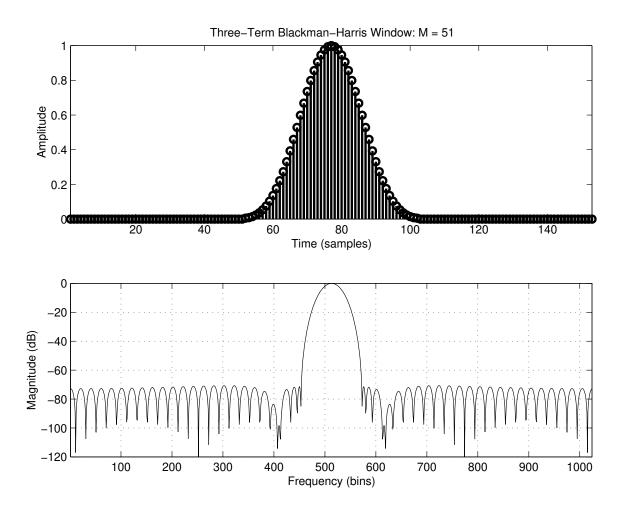
### **Properties:**

- $\alpha_0 = 0.4243801$   $\alpha_1 = 0.4973406$ , and  $\alpha_2 = 0.0782793$ .
- Side-lobe level -71.5 dB.
- Side lobes roll off  $\approx 6$  dB per octave in the absence of aliasing (like rectangular and Hamming).
- All degrees of freedom (scaling aside) are used to minimize side lobes (like Chebyshev-Hamming  $\approx$  Hamming).

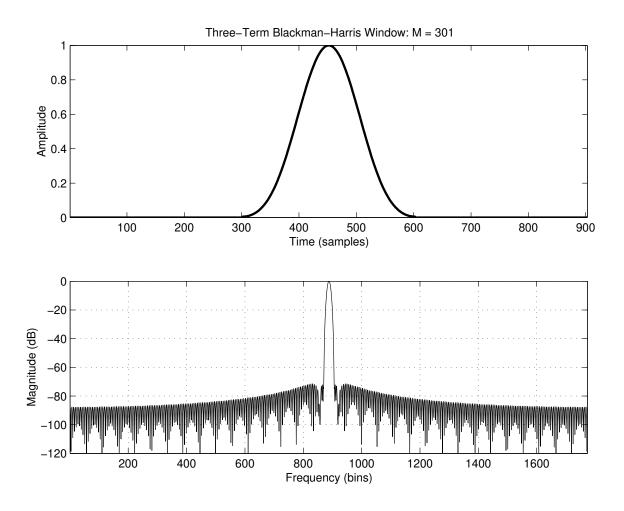
#### Matlab:

```
N = 101; L = 3; No2 = (N-1)/2; n=-No2:No2;
ws = zeros(L,3*N); z = zeros(1,N);
for l=0:L-1
   ws(l+1,:) = [z,cos(l*2*pi*n/N),z];
end
% 3-term Blackman-Harris(-Nuttall):
alpha = [0.4243801, 0.4973406, 0.0782793];
w = alpha * ws;
```

# Three-Term Blackman-Harris Window and Transform



# Longer Three-Term Blackman-Harris Window and Transform



#### Power-of-Cosine

$$w(n) = w_R(n) \cos^P\left(\frac{\pi n}{M}\right), \quad n \in \left[-\frac{M-1}{2}, \frac{M-1}{2}\right]$$

- $P = 0, 1, 2, \dots$
- first P terms of its Taylor expansion, evaluated at the endpoints (1/2 sample beyond last sample) are 0
- $\bullet$  roll-off rate  $\approx 6(P+1) \ \mathrm{dB/octave}$
- $P = 0 \Rightarrow$  Rectangular window
- $P = 1 \Rightarrow \mathsf{MLT}$  sine window (shifted to zero-phase)
- $P = 2 \Rightarrow \text{Hann window ("raised cosine"} = \text{"}\cos^2\text{"})$
- ullet  $P=4\Rightarrow$  Alternate Blackman (max roll-off rate in Blackman family)

• • • •

Thus,  $\cos^P$  windows parametrize Lth-order Blackman-Harris windows configured to use all degrees of freedom to maximize roll-off rate (L=P/2+1)

# **Spline Windows**

A spline window of order N is a repeated convolution of rectangular windows:

$$w_{\mathsf{Spline}(N)}(n) = (\underbrace{w_R * w_R * \cdots * w_R}_{N+1})(n)$$
 $\leftrightarrow W_{\mathsf{Spline}(N)}(\omega) = \operatorname{asinc}^{N+1}$ 

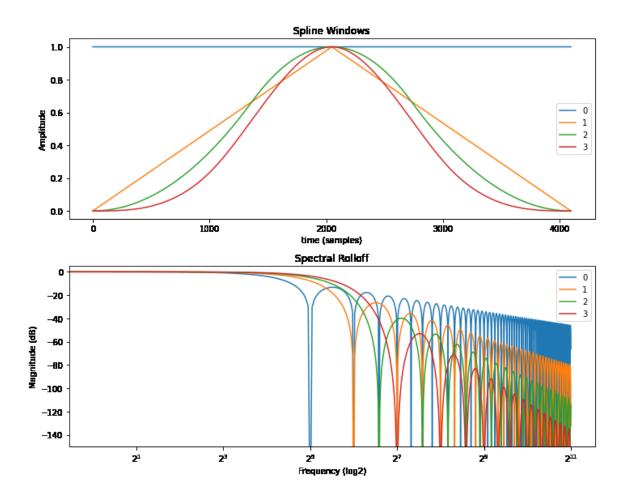
## **Special Cases:**

- $N = 0 \Rightarrow \text{Rectangular (constant)}$
- $N = 1 \Rightarrow$  Triangular (linear)
- $N = 2 \Rightarrow \mathsf{Quadratic}$
- $N=3 \Rightarrow \mathsf{Cubic}$

#### **Roll-Off Rate:**

As N increases, the window becomes smoother.  $w_{\mathsf{Spline}(N)}$  is (N-1)-times continuously differentiable, and has roll-off rate 6(N+1) dB per octave.

# **Spline Window Examples**



### Miscellaneous Windows

# Bartlett ("Triangular")

$$w(n) = w_R(n) \left[ 1 - \frac{|n|}{(M-1)/2} \right]$$

- Convolution of two half-length rectangular windows
- Window transform is  $sinc^2 \implies$ 
  - First sidelobe twice as far down as rect (-26 dB)
  - Main lobe twice as wide as that of a rectangular window having the same length (same as that of a half-length rect used to make it)
- Often applied to sample correlations of finite data
- Also called the "tent function"
- ullet M-1 often replaced by M or M+1 to avoid including endpoint zeros

# **Using Any Window as a Tapering Function**

Sometimes we need a wide rectangular window with tapered edges:

- 1. Split any window into halves, inserting the rectangle between
- 2. *Convolve* the rectangular window with any desired window

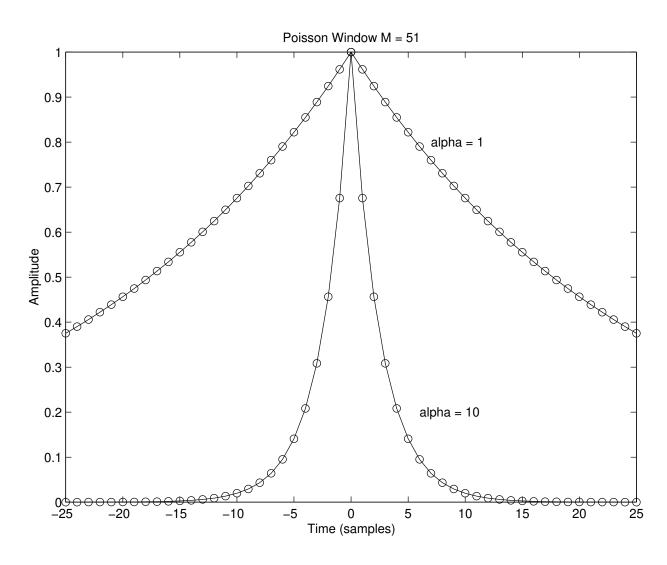
Method 2 preserves *smoothness* of the tapering window and hence its *roll-off rate*.

# Poisson ("Exponential")

$$w_P(n) = w_R(n)e^{-\alpha \frac{|n|}{(M-1)/2}}$$

where  $\alpha$  determines the time constant  $\tau$ :

$$\frac{\tau}{T} = \frac{M-1}{2\alpha} \quad \text{samples}$$



## **Poisson Window in System Identification**

In the z-plane, the Poisson window has the effect of contracting the spectrum toward zero inside unit circle. Consider an *infinitely long* Poisson window (no truncation by a rectangular window  $w_R$ ) applied to a causal signal h(n) having z transform H(z):

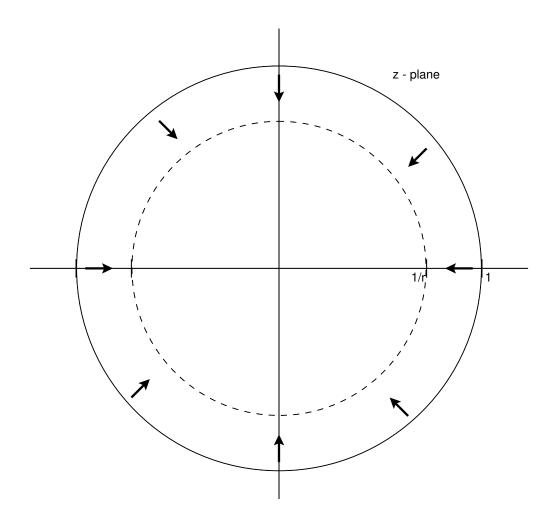
$$H_{P}(z) = \sum_{n=0}^{\infty} [w(n)h(n)]z^{-n}$$

$$= \sum_{n=0}^{\infty} \left[h(n)e^{-\frac{\alpha n}{M/2}}\right]z^{-n} \quad \text{(let } r \stackrel{\Delta}{=} e^{\frac{\alpha}{M/2}}\text{)}$$

$$= \sum_{n=0}^{\infty} h(n)z^{-n}r^{-n} = \sum_{n=0}^{\infty} h(n)(zr)^{-n}$$

$$= H(zr)$$

- $\bullet$  Unit-circle response moved to |z|=1/r<1
- Marginally stable poles now decay as  $r^{-n} = e^{-\alpha n/(M/2)}$



The Poisson window can be useful for impulse-response modeling by poles and/or zeros ("system identification"). In such applications, the window length is best chosen to include substantially all of the impulse-response data.

# Hann-Poisson ("No Sidelobes")

$$w(n) = \frac{1}{2} \left[ 1 + \cos \left( \pi \frac{n}{(M-1)/2} \right) \right] e^{-\alpha \frac{|n|}{(M-1)/2}}$$

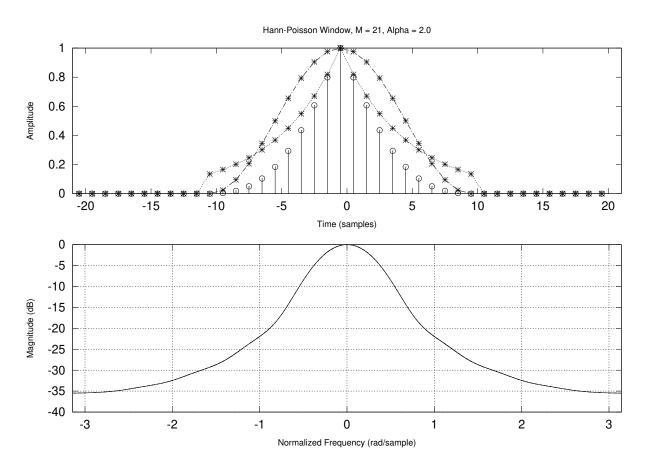
- Poisson window times Hann window (exponential times raised cosine)
- "No sidelobes" for  $\alpha \geq 2$
- Valuable for "hill climbing" optimization methods (gradient-based)

#### Matlab:

```
function [w,h,p] = hannpoisson(M,alpha)
%HANNPOISSON - Length M Hann-Poisson window
```

```
Mo2 = (M-1)/2; n=(-Mo2:Mo2)';
scl = alpha / Mo2;
p = exp(-scl*abs(n));
scl2 = pi / Mo2;
h = 0.5*(1+cos(scl2*n));
w = p.*h;
```

#### Hann-Poisson Window and Transform



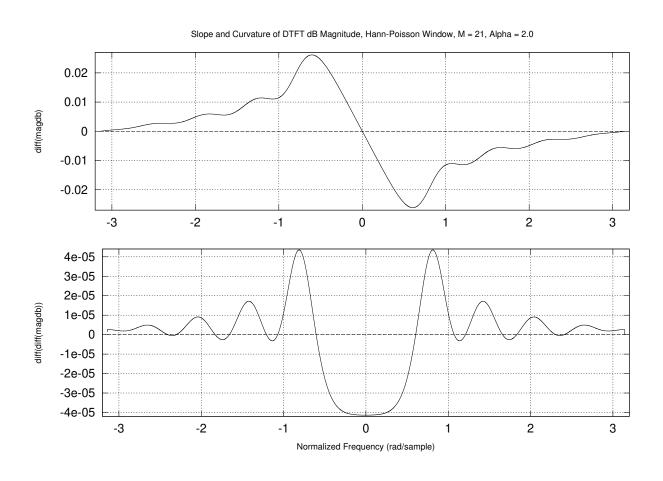
https://ccrma.stanford.edu/~jos/Windows/Hann\_Poisson\_Window\_Transform.html

## Question:

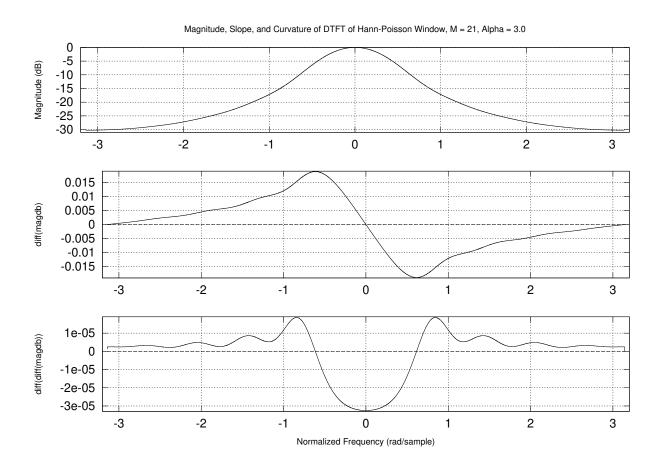
How can a truncated *anything* avoid having ripples in its spectrum? ("Gibbs ripples")

Let's look at the *derivatives* of the window:

# Hann-Poisson Slope and Curvature



# Slope and Curvature for Larger Alpha



### Maximum Main-Lobe Energy Window: DPSS

Question: How do we use all M degrees of freedom (sample values) in an M-point window w(n) to obtain  $W(\omega) \approx \delta(\omega)$  in some optimal sense?

That is, we wish to perform the following optimization:

$$\max_{w} \left[ \frac{\text{main lobe energy}}{\text{total energy}} \right]$$

In the continuous-time case  $[\omega \in (-\infty, \infty)]$ , this problem is solved by a *prolate spheroidal wave function*, an eigenfunction of the integral equation

$$\int_{-\omega_c}^{\omega_c} W(\nu) \frac{\sin[\pi D \cdot (\omega - \nu)]}{\pi(\omega - \nu)} d\omega = \lambda W(\omega), \ |\omega| \le \omega_c$$

where D is the nonzero time-duration of w(t) in seconds.

#### Interpretation:

$$\begin{split} & \left[ \mathrm{Chop}_{2\omega_c}(W) \right] * \left[ D \operatorname{sinc}(D\omega) \right] \\ &= \mathrm{FT}(\mathrm{Chop}_D(\mathrm{IFT}(\mathrm{Chop}_{2\omega_c}(W)))) \; = \; \lambda W \end{split}$$

where  $Chop_D(w)$  is a rectangular windowing operation which zeros w outside the interval  $t \in [-D/2, D/2]$ .

W is thus the bandlimited extrapolation of its main lobe  $(\omega \in [-\omega_c, \omega_c])$ 

The optimal window transform W is an eigenfunction of this operation sequence corresponding to the *largest* eigenvalue.

The resulting optimal window w has maximum main-lobe energy as a fraction of total energy.

It may be called the *Slepian window*, or *prolate spheroidal* window in the continuous-time case.

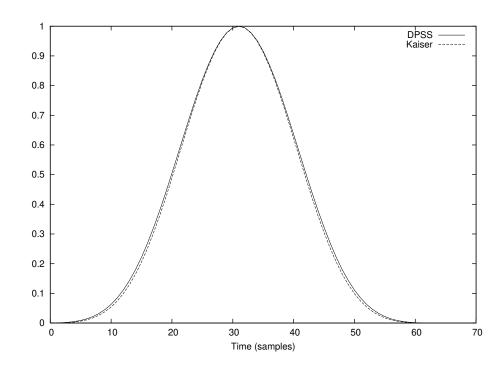
In discrete time, we need *Discrete Prolate Spheroidal Sequences (DPSS)*, eigenvectors of the following symmetric Toeplitz matrix constructed from a sampled sinc function:

$$S[k, l] = \frac{\sin[\omega_c T(k-l)]}{k-l}, \quad k, l = 0, 1, 2, \dots, M-1$$

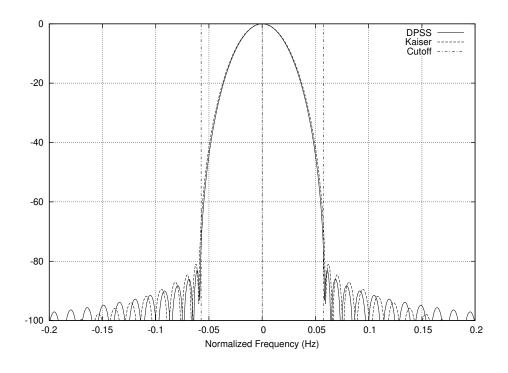
- ullet M= window length in samples
- $\omega_c = \text{main-lobe cut-off frequency (rad/sec)}$
- $\bullet$  T= sampling period in seconds.

The DPSS window (digital Slepian window) is then given by the eigenvector corresponding to the largest eigenvalue.

#### Matlab for the DPSS Window



## Kaiser and DPSS Window Transforms Compared



- Kaiser  $\approx$  DPSS
- DPSS window has a slightly narrower main lobe
- DPSS window has lower overall side-lobe levels
- Kaiser window side lobes roll off faster
- Otherwise they are very similar

## Kaiser (Kaiser-Bessel)

Kaiser discovered a very good approximation to prolate spheroidal wave functions using Bessel functions:

$$w_K(n) \stackrel{\Delta}{=} \left\{ \begin{array}{l} \frac{I_0\left(\beta\sqrt{1-\left(\frac{n}{M/2}\right)^2}\right)}{I_0(\beta)}, \ -\frac{M-1}{2} \leq n \leq \frac{M-1}{2} \\ 0, \ \qquad \text{elsewhere} \end{array} \right.$$

$$w_K(n) \stackrel{\Delta}{=} w_R(n) \frac{I_0 \left(\beta \sqrt{1 - \left(\frac{n}{M/2}\right)^2}\right)}{I_0(\beta)}$$

This is called the Kaiser (or Kaiser-Bessel) window.

The Fourier transform of the Kaiser window  $w_K(t)$  (where t is treated as continuous) is given by

$$W(\omega) = \frac{M}{I_0(\beta)} \frac{\sinh\left(\sqrt{\beta^2 - \left(\frac{M\omega}{2}\right)^2}\right)}{\sqrt{\beta^2 - \left(\frac{M\omega}{2}\right)^2}}$$
$$= \frac{M}{I_0(\beta)} \frac{\sin\left(\sqrt{\left(\frac{M\omega}{2}\right)^2 - \beta^2}\right)}{\sqrt{\left(\frac{M\omega}{2}\right)^2 - \beta^2}}$$

where  $I_0$  is the zero-order modified Bessel function of the first kind.

#### Modified Bessel Function of the 1st Kind

A series expansion for the order zero, modified Bessel function of the first kind is given by

$$I_0(x) = \sum_{k=0}^{\infty} \left[ \frac{(x/2)^k}{k!} \right]^2$$

Compare this with

$$e^{x/2} = \sum_{k=0}^{\infty} \frac{(x/2)^k}{k!}.$$

#### Kaiser-Bessel Window Notes

$$w_K(n) \stackrel{\Delta}{=} w_R(n) \frac{I_0 \left(\beta \sqrt{1 - \left(\frac{n}{M/2}\right)^2}\right)}{I_0(\beta)}$$

- "Closed form" (given  $I_0$  series or table)
- $\bullet$  Reduces to rectangular window for  $\beta=0$
- Asymptotic roll-off is 6 dB/octave
- For  $\beta\gg 0$ , first null in window transform is at  $\omega_0\approx 2\beta/M$   $\Rightarrow \beta=M\omega_0/2$
- ullet Sometimes the Kaiser window is parameterized by lpha:

$$\beta \stackrel{\Delta}{=} \pi \alpha$$

#### Kaiser Window Time-Bandwidth Product

• Define the main-lobe "cutoff frequency" as half-way to the first zero in  $W(\omega)$ :

$$\omega_c = \frac{\omega_0}{2} = \frac{\beta}{M} = \frac{\pi \alpha}{M}$$

Then

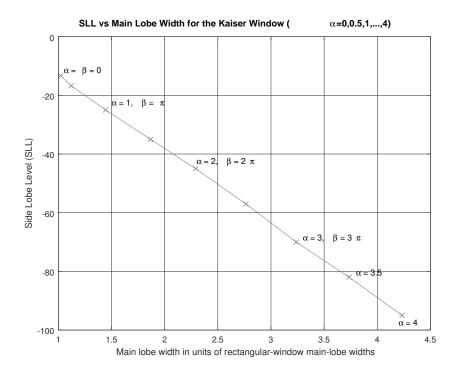
$$eta = M\omega_c = rac{1}{2}M\cdot(2\omega_c)$$

$$= rac{1}{2} \, ext{duration (samples)} \, imes \, ext{bandwidth (rad/sample)}$$
 $lpha = rac{eta}{\pi} = rac{2eta}{2\pi}$ 

$$= ext{duration (samples)} \, imes \, ext{bandwidth (cycles/sample)}$$

- $\beta=M\omega_c T$  is equal to 1/2 'time-bandwidth product'  $\beta=\frac{1}{2}\Delta t\cdot\Delta\omega\Rightarrow \quad \alpha=\Delta t\cdot\Delta f$
- In this definition of time-bandwidth product, the "cut-off frequency  $\omega_c$  of the Kaiser-window transform is defined as *half* of the first null frequency, *i.e.*,  $\omega_c = \omega_0/2$ .
- $\beta$  trades off side lobe level for main lobe width larger  $\beta \Rightarrow$  lower S.L.L., wider mainlobe

#### Kaiser Side-Lobe Level vs. Main-Lobe Width

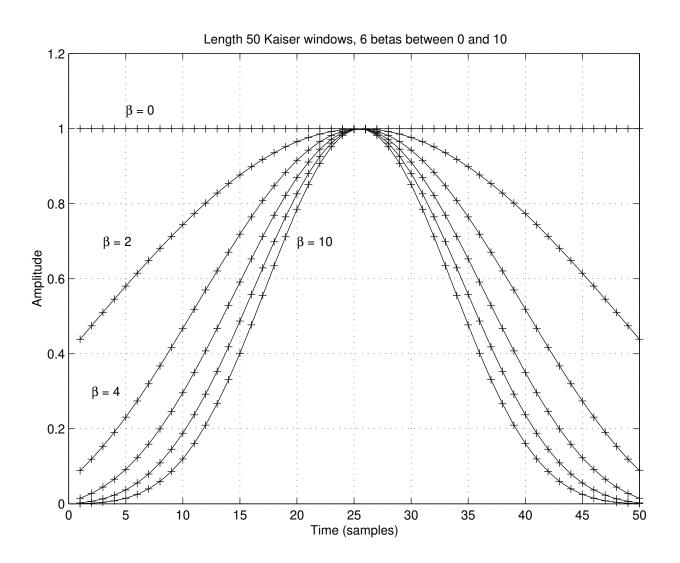


#### **Mathematica Demonstration:**

http://demonstrations.wolfram.com/KaiserWindowTransform/

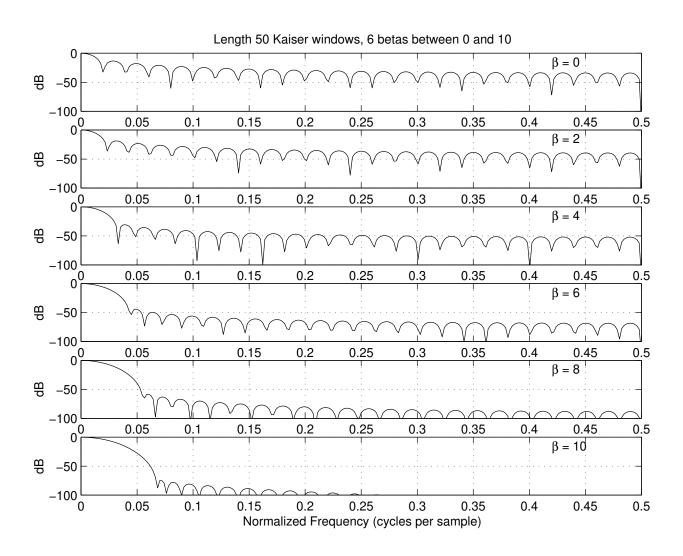
## **Kaiser Window Examples**

$$\beta = [0, 2, 4, 6, 8, 10]$$



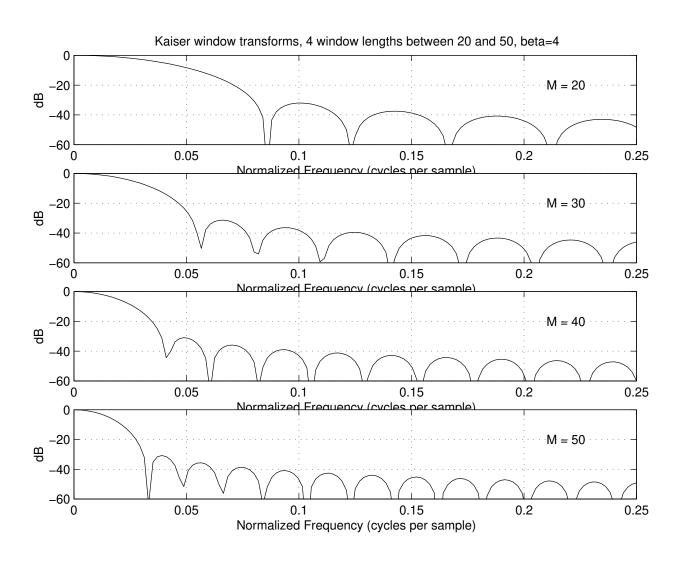
## **Kaiser Window Transform Examples**

$$\beta = [0, 2, 4, 6, 8, 10]$$



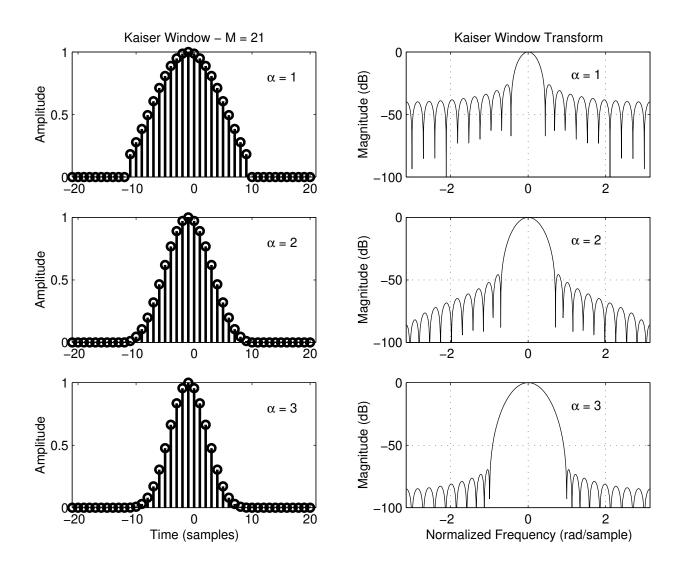
# Kaiser Window: Different Lengths M

$$M = [20, 30, 40, 50]$$

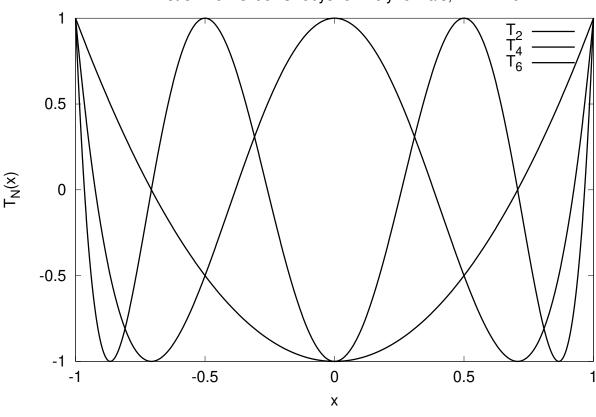


## More Kaiser-Window Examples

$$\alpha = [1, 2, 3] \quad (\beta = [\pi, 2\pi, 3\pi])$$



#### **Chebyshev Polynomials**



First 3 Even-Order Chebyshev Polynomials, N = 2:2:6

The nth Chebyshev polynomial may be defined by

$$T_n(x) = \begin{cases} \cos[n\cos^{-1}(x)], & |x| \le 1\\ \cosh[n\cosh^{-1}(x)], & |x| > 1 \end{cases}.$$

Clearly,  $T_0(x) = 1$  and  $T_1(x) = x$ .

Using the double-angle trig formula

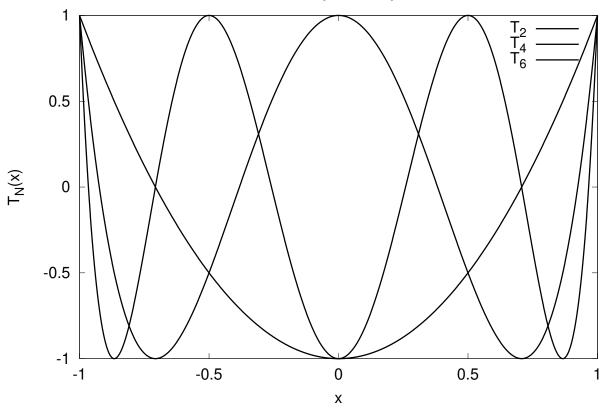
 $cos(2\theta) = 2cos^2(\theta) - 1$ , it can be verified that

$$T_n(x) = 2xT_{n-1}(x) - T_{n-2}(x)$$
  $(n \ge 2)$ 

### **Chebyshev Polynomial Properties**

- ullet  $T_n(x)$  is an nth-order polynomial in x
- ullet  $T_n(x)$  is an even function when n is an even integer, and odd when n is odd
- $T_n(x)$  has n zeros in the open interval (-1,1), and n+1 extrema in the closed interval [-1,1]
- $T_n(x) > 1$  for x > 1

First 3 Even-Order Chebyshev Polynomials, N = 2:2:6



$$T_n(x) = \cos[n\cos^{-1}(x)], \quad |x| \le 1$$

### **Dolph-Chebyshev Window**

Minimize the Chebyshev norm of the side lobes, e.g.,

$$\begin{aligned} & \min_{w,\sum w=1} \| \operatorname{sidelobes}(W) \|_{\infty} \\ & \equiv & \min_{w,\sum w=1} \left\{ \max_{\omega > \omega_c} |W(\omega)| \right\} \end{aligned}$$

Alternatively, *minimize main lobe width* subject to a sidelobe spec:

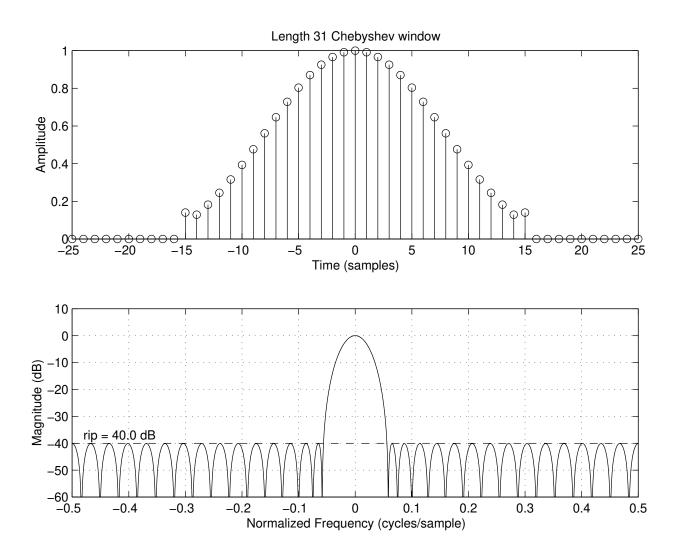
$$\min_{w,W(0)=1} (\omega_c) \bigg|_{|W(\omega)| \le c_{\alpha}, \forall |\omega| \ge \omega_c}$$

Closed-Form Window Transform (Dolph):

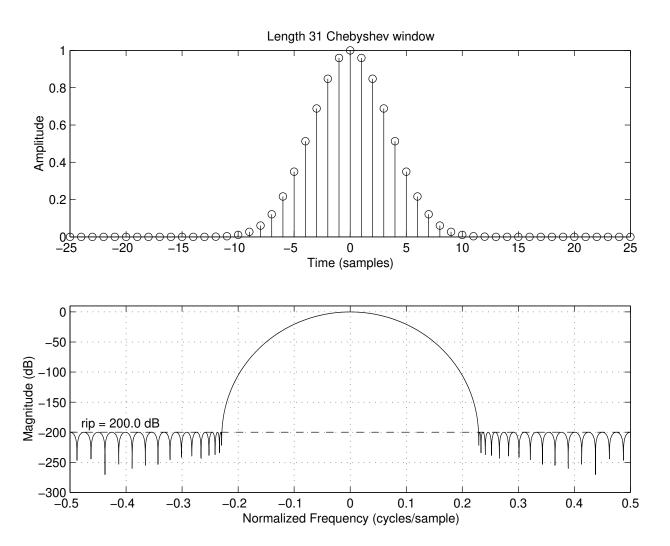
$$W(\omega_k) = \frac{\cos\left\{M\cos^{-1}\left[\Gamma\cos\left(\frac{\pi k}{M}\right)\right]\right\}}{\cosh\left[M\cosh^{-1}(\Gamma)\right]}, \quad (|k| \le M - 1)$$
  
$$\Gamma = \cosh\left[\frac{1}{M}\cosh^{-1}(10^{\alpha})\right] \ge 1, \quad (\alpha \approx 2, 3, 4)$$

- Window  $w = \mathrm{IDFT}(W)$  [zero-centered case] or IDFT of  $(-1)^k W(\omega_k)$  for causal case
- $\alpha$  controls sidelobe level ("stopband ripple"): Side-Lobe Level in dB =  $-20\alpha$ .
- smaller ripple  $\Rightarrow$  larger  $\omega_c$
- see matlab function "chebwin(M,ripple)"

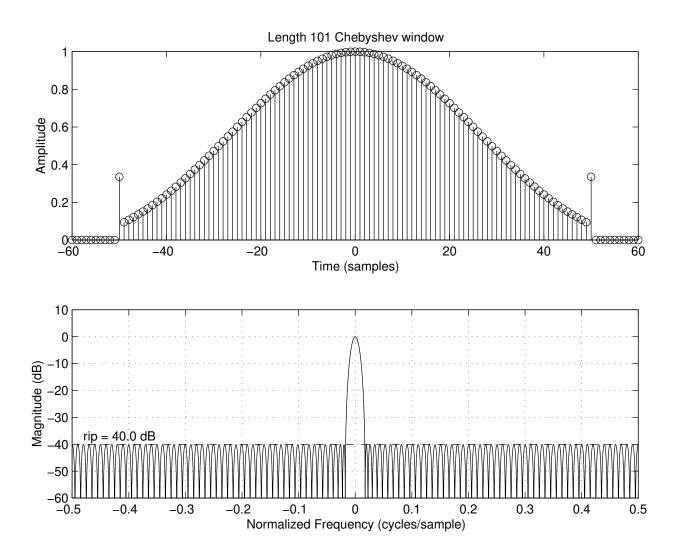
# Dolph-Chebyshev Window, Length 31, Ripple -40 dB



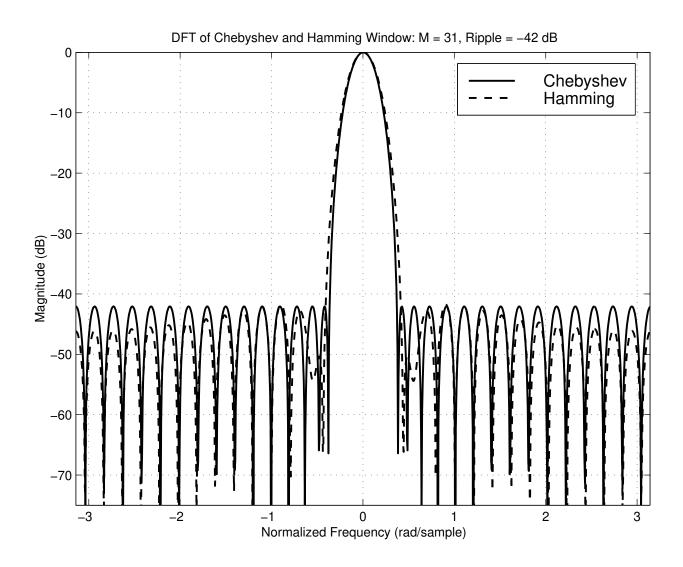
# Dolph-Chebyshev Window, Length 31, Ripple -200 dB



# Dolph-Chebyshev Window, Length 101, SLL -40 dB



# Dolph-Chebyshev and Hamming Windows Compared



For the comparison, we set the ripple parameter for chebwin to  $42~\mathrm{dB}$ :

window = [ chebwin(31,42), zeros(1,1024-31) ];

#### Gaussian

The Gaussian "bell curve" is the only smooth function that transforms to itself:

$$\frac{1}{\sigma\sqrt{2\pi}}e^{-t^2/2\sigma^2} \leftrightarrow e^{-\omega^2/2(1/\sigma)^2}$$

It also achieves the minimum time-bandwidth product

$$\sigma_t \sigma_\omega = \sigma \times (1/\sigma) = 1$$

when "width" of a function is defined as the square root of its second central moment. For even functions w(t),

$$\sigma_t \stackrel{\Delta}{=} \sqrt{\int_{-\infty}^{\infty} t^2 w(t) dt}.$$

- Since the true Gaussian function has infinite duration, in practice we must window it with some finite window.
- ullet Philippe Depalle suggests using a triangular window raised to some power lpha for this purpose.
  - This choice preserves the absence of sidelobes for sufficiently large  $\alpha$ .
  - It also preserves non-negativity of the transform

### The Gaussian Window in Spectral Modeling

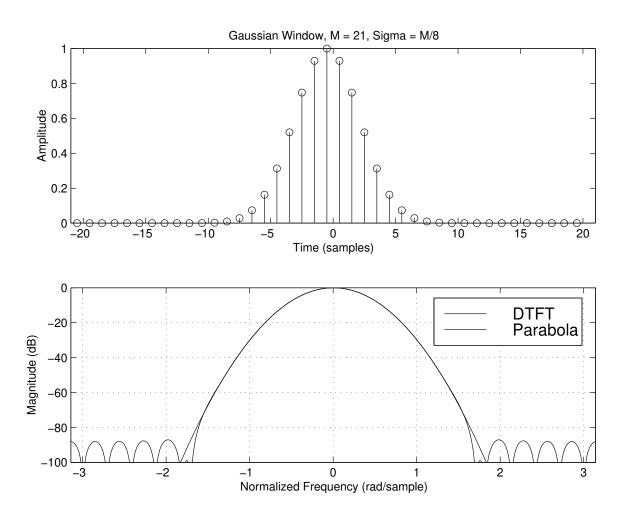
**Special Property:** On a dB scale, the Gaussian is quadratic  $\Rightarrow$  parabolic interpolation of a sampled Gaussian transform is exact.

Conjecture: Quadratic interpolation of spectral peaks is generally more accurate on a *log-magnitude scale* (e.g., dB) than on a linear magnitude scale. This has been verified in a number of cases, and no counter-examples are yet known. Exercise: Prove this is true for the rectangular window.

#### Matlab for the Gaussian Window

```
function [w] = gausswin(M,sigma)
n=(-(M-1)/2:(M-1)/2);
w = exp(-n.*n./(2*sigma.*sigma));
```

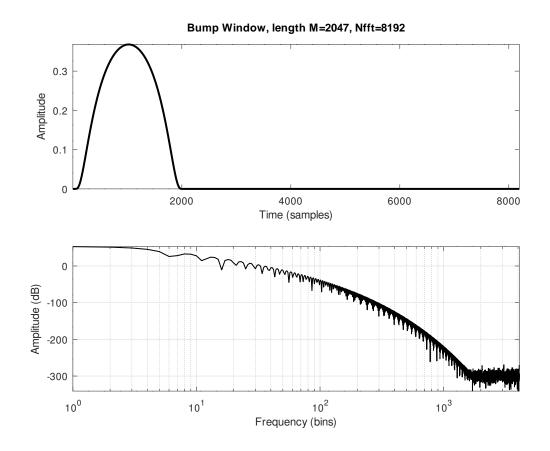
## **Gaussian Window and Transform**



## **Bump Window**

$$w(t) = \begin{cases} e^{-\frac{1}{1-t^2}} & |t| < 1, \\ 0 & \text{otherwise} \end{cases}$$

- Infinitely differentiable everywhere (then sample)
- Roll-off rate unbounded (faster than any polynomial)
- Aliasing progressively slows the decay



#### **More Windows**

There is a nice collection of window definitions and citations on Wikipedia: https://en.wikipedia.org/wiki/Window\_function

# **Optimal Windows**

### Generally we desire

$$W(\omega) \approx \delta(\omega)$$

- Best results are obtained by formulating this as an *FIR filter design problem*.
- In general, both time-domain and frequency-domain specifications are needed.
- Equivalently, both *magnitude* and *phase* specifications are necessary in the frequency domain.

## **Optimal Windows for Audio Coding**

Recently, numerically optimized windows have been developed by Dolby which achieve the following objectives:

- Narrow the window in time
- Smooth the onset and decay in time
- Reduce sidelobes below the worst-case masking threshold

#### **Conclusion**

- There is rarely a closed form expression for an optimal window in practice.
- The hardest task is formulating the *ideal error* criterion.
- ullet Given an error criterion, it is usually straightforward to minimize it numerically with respect to the window samples w.