Applied Signal Processing Lecture 6

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Agenda

- Frequency Analysis DTFT of a window of data
- Window functions
- FIR filters
 - Specifications
 - Linear phase
 - The Window design method
 - FIR-LS design method
 - FIR-PM design method
- Design demo

DTFT results

Multiplication of two time domain signals

$$y(n) = x(n)w(n)$$

$$\Leftrightarrow$$

$$Y(\omega) = \frac{1}{\omega_s} \int_0^{\omega_s} X(\lambda)W(\omega - \lambda) d\lambda$$

Frequency domain convolution of the DTFTs

Frequency analysis

We observe a signal x(n) in an observation window of length N.

We have the sequence $x(0), x(1), \dots, x(N-1)$ available for analysis

$$\hat{x}(n) \triangleq r_N(n)x(n), \quad \forall n$$

where

$$r_N(n) = \begin{cases} 1 & n = 0, \dots, N-1 \\ 0 & \text{otherwise} \end{cases}$$

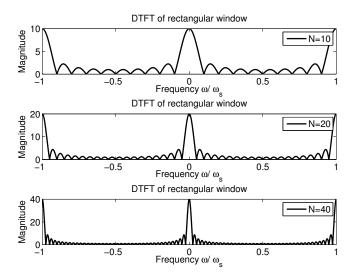
$$\hat{x}(n) = x(n)r_N(n) \Leftrightarrow \hat{X}(\omega) = \frac{1}{\omega_s} \int_0^{\omega_s} R_N(\lambda)X(\omega - \lambda) d\lambda$$

where

$$R_{N}(\omega) = \sum_{n=-\infty}^{\infty} r_{N}(n)e^{-j\omega\Delta tn} = \sum_{n=0}^{N-1} e^{-j\omega\Delta tn} =$$

$$= \frac{1 - e^{-j\omega\Delta tN}}{1 - e^{-j\omega\Delta t}} = e^{-j\frac{N-1}{2}\omega\Delta t} \frac{\sin(\frac{N\omega\Delta t}{2})}{\sin(\frac{\omega\Delta t}{2})}$$

The rectangular window $R_N(\omega)$



Observations

- $R_N(\omega) = 0$ for $\omega = \pm \omega_s/N, \pm 2\omega_s/N$
- Main lobe:
 - High and narrow is good
 - Width $\sim 1/N$
- Side lobes: Low levels are good
- As $N o \infty$ then $R_N(\omega) pprox \omega_s ilde{\delta}(\omega)$

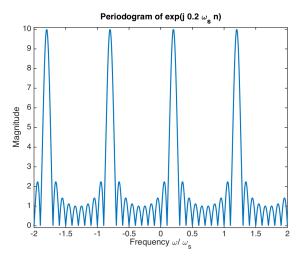
Example

Consider a discrete time signal $x(n) \triangleq e^{j\omega_0 n}$ with $\Delta t = 1$

Then
$$X(\omega) = 2\pi \tilde{\delta}(\omega - \omega_0)$$

$$\hat{X}(\omega) = rac{1}{2\pi} \int_0^{2\pi} X(\omega - \lambda) R_N(\lambda) d\lambda =$$

$$= rac{1}{2\pi} \int_0^{2\pi} 2\pi \tilde{\delta}(\omega - \omega_0 - \lambda) R_N(\lambda) d\lambda = R_N(\omega - \omega_0)$$



Periodogram based on N=10 samples of the signal $x(n)=e^{j0.2\omega_s\Delta tn}$

Window functions

We are not restricted to the rectangular window

$$\hat{x}(n) = w(n)x(n)$$

Window	-3dB bw [Hz]	PSLL [dB]	SLRO [dB/oct.]
Rectangular	<u>0.89</u> N∆t	-13	-6
Hanning	<u>N∆t</u> 1.4 <u>N</u> ∆t	-32	-18
Hamming	1.5	-43	-6
Dolph-Chebyshev	$\frac{\overline{N\Delta t}}{1.44}$ $\overline{N\Delta t}$	-60	0

BW - Band width

PSLL - Peak side-lobe level

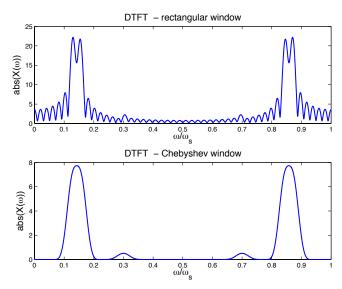
SLRO - Side lobe roll off

Example

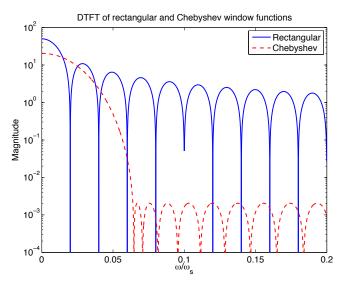
$$x(n) = \sin(2\pi f_0 n) + \sin(2\pi f_1 n) + 0.05\sin(2\pi f_2 n)$$

where $f_0=0.135,\ f_1=0.150$ and $f_2=0.3$. The number of samples are N=50

- Rectangular window
- Chebyshev window



Frequency analysis using DTFT.



DTFT of the rectangular and Chebyshev window of length 50

Zero padding

To calculate the DTFT for arbitrary frequencies ω is straight-forward. It is just to implement the definition

$$\hat{X}(\omega) = \sum_{n=0}^{N-1} x(n)e^{-j\omega\Delta tn}$$

However, using an equidistantly spaced frequency grid the calculations can be made more efficiently by employing the FFT algorithm.

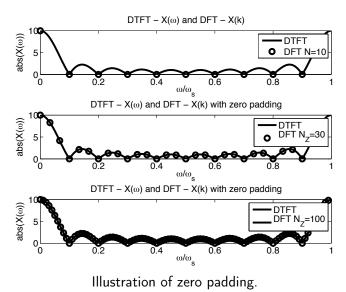
Assume the signal is of length N

We want to calculate the DTFT at frequencies $\omega_k = \omega_s k/N_Z$, $k = 0, ..., N_Z - 1$ where $N_Z > N$.

We get the desired result if we calculate the FFT (i.e. DFT) for the extended signal

$$\hat{x}_{z}(n) = \begin{cases} \hat{x}(n), & n = 0, ..., N-1 \\ 0 & n = N, ..., N_{Z}-1 \end{cases}$$

Zero padding example



FIR Filter Design

- FIR Finite impulse response filters h(n)
- Easy to design and analyze
- Fast filtering algorithms with FFT

Convolution

$$y(n) = \sum_{k=0}^{M-1} h(k)x(n-k)$$

Frequency function

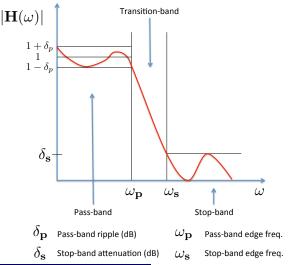
$$H(\omega) = \sum_{n=0}^{M-1} h(n) e^{j\omega \Delta t n}$$

The filter *shape* the DTFT of the output:

$$Y(\omega) = H(\omega)X(\omega)$$

Frequency selective filters

Objective of the filtering are to *enhance* or *suppress* certain bands of the frequency content of the signal.



Typical filter types

- Low-pass filter (LP)
- High-pass filter (HP)
- Band-bass filter (BP)
- Band-stop filter (BS)

Ideal LP filter

Frequency function

$$H_{LP}(\omega) = egin{cases} 1 & |\omega| < \omega_c \ 0 & \omega_c < |\omega| < \omega_s/2. \end{cases}$$

Impulse response

$$h_{LP}(n) = \begin{cases} \frac{2\omega_c}{\omega_s} & n = 0\\ \frac{2\omega_c}{\omega_s} \frac{\sin \omega_c \Delta tn}{\omega_c \Delta tn} & n \neq 0 \end{cases}$$

- Infinte long. ⇒ Truncate the length (windowing)
 - $h_T(n) = w(n)h_{LP}(n), \quad n = 0, \pm 1, \dots, \pm (M-1)/2$
- Non-causal. ⇒ Shift it to make it causal (introduces a delay)
 - $h(n) = h_T(n (M-1)/2), \quad n = 0, ..., M-1)$

Linear phase FIR filters

Non-causal truncated filters

- Odd filters $h_T(-n) = -h_T(n)$
- Even filters $h_T(-n) = h_T(n)$

Have a zero (or constant π) phase function

Shifted filters
$$h(n) = h_T(n - (M-1)/2), \quad n = 0, \dots, M-1$$

have linear phase $-\omega \Delta t(M-1)/2$.

Linear phase filters act as a delay in the passband and preserves pulse shapes.

Analysis of the filter

Truncation yieleds

$$H_T(\omega) = \frac{1}{\omega_s} \int_{-\omega_s/2}^{\omega_s/2} H_{LP}(\lambda) W(\omega - \lambda) d\lambda$$

The delay introduces a phase shift and the final result is

$$H(\omega) = e^{-j\omega\Delta t \frac{M-1}{2}} H_T(\omega) = e^{-j\omega\Delta t \frac{M-1}{2}} \frac{1}{\omega_s} \int_{-\omega_s/2}^{\omega_s/2} H_{LP}(\lambda) W(\omega - \lambda) d\lambda$$

The choice of window function will influence the stop-band attenuation and the width of the transition band.

Window function properties

Window's name	Trans band	Peak 20 $\log_{10} \delta_s$
Rectangular	1/M	-21 dB
Hamming	3.3/M	-53 dB
Blackman	5.5/M	-74 dB

Optimal FIR design methods

- FIR-I S:
 - Minimizes the sum of the squared frequency function deviation from the spec.
- FIR-PM·
 - Minimizes the maximum frequency deviation from the spec.

FIR-LS and FIR-PM

FIR-LS

$$\min_{\mathbf{h}} \sum_{k=1}^{N_{\text{spec}}} W_K |H_D(\omega_k) - H(\omega_k, h)|^2$$

FIR-PM

$$\min_{\mathbf{h}} \max_{\omega} |H_D(\omega) - H(\omega, h)|^2.$$

Efficent algoritrhms exist to solve both problems.