# STANFORD UNIVERSITY EE 102B Spring-2013

Lecture 19
Block Processing and Spectrum Analysis
May 15, 2013

#### **ASSIGNMENTS**

- Reading for this Lecture:
  - SPF: Section 12-3, Chapter 66-6 thru 66-9
  - S&S:
  - HW#06 is due by 5pm today, May 15, in Packard 263. No late penalty if handed in by 5pm Friday, May 17.
- Lab #05 is due by 5pm, Friday, May 17, in Packard 263. Disregard the "verfied/signed" line on the report page. New Lab 05 is posted.

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## Office Hours for Course Staff – Come see us.

- Ron Schafer: Mon. and Weds. 2:00-4:00 pm, Packard 211.
- Dookun Park: Tues. 1:30 ~ 3:30 pm in Packard 107
- Ruo Yu Gu (Roy): Mon. 4:00 ~ 6:00 pm in Packard 106

### **Lecture Objective**

- Review sampling theorem for the DTFT
- Review discrete-time convolution and the DFT
- Block processing with the DFT
  - Segmenting a signal
  - Convolution with a segmented signal
- Spectrum analysis

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### A SAMPLING THEOREM FOR THE DTFT

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### USING THE DTFT SAMPLING THEOREM FOR TIME-DOMAIN CONVOLUTION

# **Summary of DTFT Sampling Theorem**

Sample DTFT to get DFT

$$X[k] = \sum_{n=-\infty}^{\infty} x[n]e^{-j(2\pi/N)kn} = X(e^{j\omega_k}) \quad k = 0,1,...,N-1$$

$$x[k] = \sum_{n=-\infty}^{\infty} x[n]e^{-j(2\pi/N)kn} = X(e^{j\omega_k}) \quad k = 0,1,...,N-1$$

Reconstruction by IDFT

$$\tilde{x}[n] = \frac{1}{N} \sum_{k=0}^{N-1} X[k] e^{j(2\pi/N)kn} = \sum_{r=-\infty}^{\infty} x[n-rN]$$

Exact reconstruction if no overlap; i.e.,

$$\tilde{x}[n] = x[n]$$
  $0 \le n \le N-1$ , if  $x[n] \ne 0$  only for  $0 \le n \le N-1$ 

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### Convolution of Finite-Length Sequences

Let h[n] be of length P and x[n] of length L

$$y[n] = x[n] * h[n] = \sum_{m=0}^{L-1} x[m]h[n-m]$$

y[n] also has finite length, L+P-1 samples

$$y[n] = 0$$
 for  $n < 0$  and for  $n > L + P - 2$ 

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## Sampling the DTFT of a Convolution

$$y[n] = h[n] * x[n] \Leftrightarrow Y(e^{j\hat{\omega}}) = H(e^{j\hat{\omega}})X(e^{j\hat{\omega}})$$
$$Y(e^{j(2\pi/N)k}) = H(e^{j(2\pi/N)k})X(e^{j(2\pi/N)k})$$

Compute the sampled DTFT by computing the DFTs

$$X[k] = \sum_{n=0}^{L-1} x[n]e^{-j(2\pi/N)kn} \qquad H[k] = \sum_{n=0}^{P-1} h[n]e^{-j(2\pi/N)kn}$$

$$\tilde{y}[n] = \frac{1}{N} \sum_{k=0}^{N-1} \underbrace{H[k]X[k]}_{Y[k]} e^{j(2\pi/N)kn} = \sum_{r=-\infty}^{\infty} y[n-rN]$$

 $\tilde{y}[n] = y[n]$   $0 \le n \le N-1$  if  $N \ge L+P-1$ 

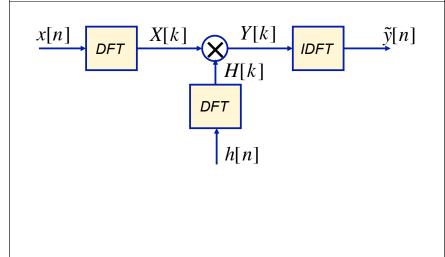
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### Convolution Using the DFT



### BLOCK PROCESSING OF LONG SIGNALS

### **Long (Continuing) Signals**

- Sometimes we have signals that go on for a long time; e.g., symphony performance, news broadcast, lecture in EE102B, seismograph signal, ...
- Usually it is not the entire signal but rather its local variations that interest us.
- This leads to trying to track variations in a signal; e.g., speech encodes information in a changing acoustic pattern.

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# Time-Dependent (Short Time) DTFT and DFT

Definition: short-time DTFT

$$X(e^{j\hat{\omega}}, n] = \sum_{m=0}^{L-1} w[m]x[m+n]e^{-j\hat{\omega}n} \quad 0 \le \hat{\omega} < 2\pi$$

Definition: short-time DFT

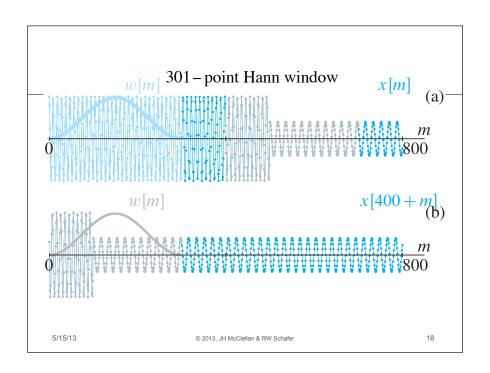
$$X[k,n] = \sum_{m=0}^{L-1} w[m]x[m+n]e^{-j(2\pi/N)kn} \quad k = 0,1,...,N-1$$

w[m]x[m+n] focuses attention on the signal around time n.

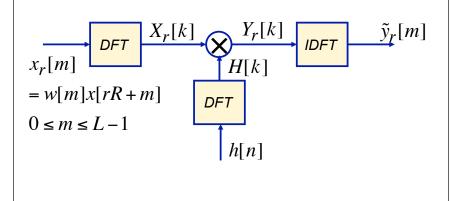
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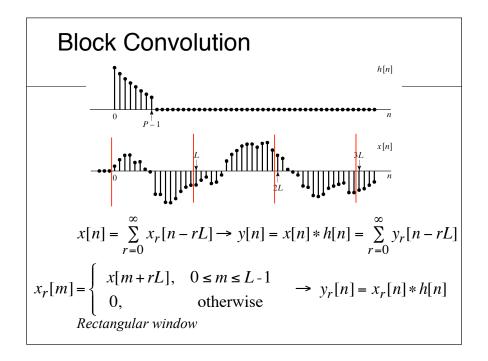
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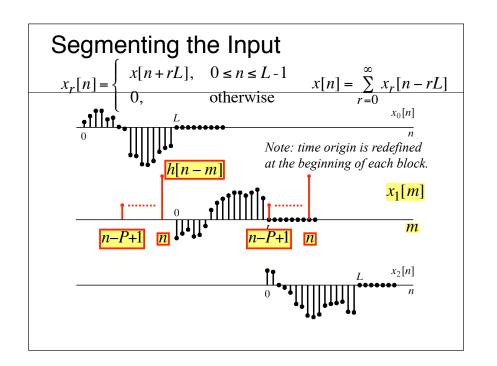
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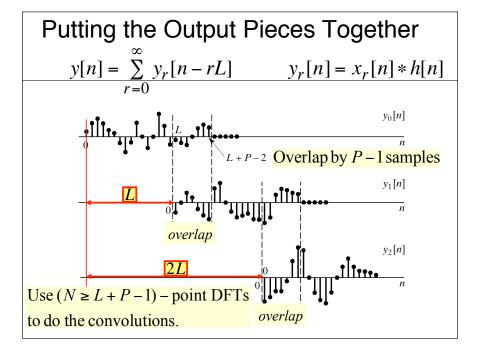


### Convolution Using the DFT









#### **WINDOWING**

### Multiplication in the Time-Domain

Multiplication in the time domain

$$y[n] = w[n]x[n]$$

Convolution in the frequency domain

$$Y(e^{j\hat{\omega}}) = \frac{1}{2\pi} \int_{-\pi}^{\pi} X(e^{j\theta}) W(e^{j(\hat{\omega} - \theta)}) d\theta$$

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### **Rectangular Window**

Time window

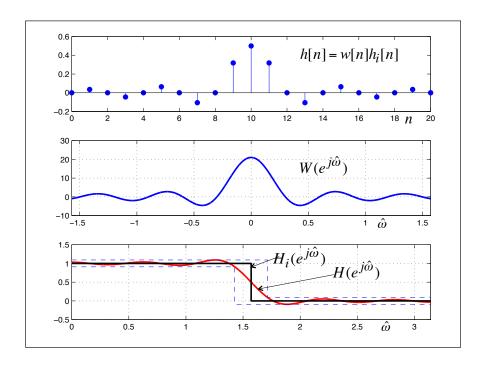
$$w[n] = \begin{cases} 1 & 0 \le n \le L - 1 \\ 0 & \text{otherwise} \end{cases}$$

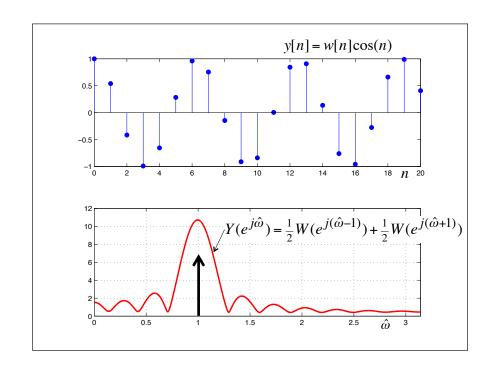
DTFT of Window

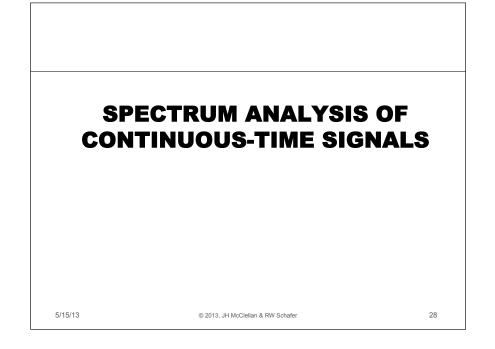
$$W(e^{j\hat{\omega}}) = \frac{\sin(\hat{\omega}L/2)}{\sin(\hat{\omega}/2)} e^{-j\hat{\omega}(L-1)/2}$$

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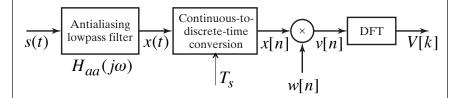
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# **Spectrum Analysis of Continuous-Time Signals**



Work out the details of the relationship between V[k] and  $S(j\omega)$  to verify your understanding of all parts of this important DSP system as well as how the parts fit together to give an estimate of the spectrum.

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Definition: short-time DFT

$$X[k,n] = \sum_{m=0}^{L-1} w[m]x[m+n]e^{-j(2\pi/N)kn} \quad k = 0,1,...,N-1$$

 w[m]x[m+n] focuses attention on the signal around time n.

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## Examples to try with spectrum\_analysis\_demo

- W=[2\*pi/16], A=[1], TH=[0], B=0, N=32, L=32, swin=r
- W=[2\*pi/16], A=[1], TH=[0], B=0, N=32, L=32, swin=h
- W=[2\*pi/16], A=[1], TH=[0], B=0, N=32, L=16, swin=r
- W=[2\*pi/14, 4\*pi/15], A=[1,.75], TH=[0,0], B=0, N=64, L=64, swin=r
- W=[2\*pi/16, 4\*pi/16], A=[1,.75], TH=[0,0], B=0, N=64, L=64, swin=r
- W=[2\*pi/14, 4\*pi/15], A=[1,.75], TH=[0,0], B=0, N=64, L=32, swin=k
- W=[2\*pi/14, 4\*pi/15], A=[1,.75], TH=[0,0], B=0, N=64, L=64, swin=k

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