

## ESE 531: Digital Signal Processing

Lec 25: April 23, 2019  
Wavelet Transform



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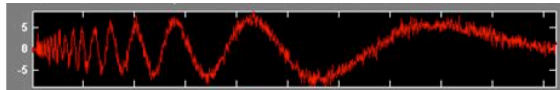
## Wavelet Transform



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### Motivation

- Some signals obviously have spectral characteristics that vary with time



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### Criticism of Fourier Spectrum

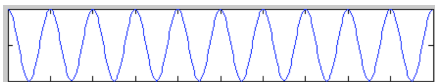
- It's giving you the spectrum of the 'whole time-series'
- Which is OK if the time-series is stationary. But what if it's not?
- We need a technique that can "march along" a time series and that is capable of:
  - Analyzing spectral content in different places
  - Detecting sharp changes in spectral character

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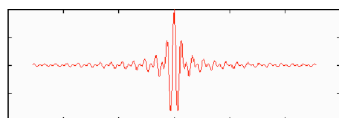
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### Fourier vs. Wavelet

- Fourier Analysis is based on an indefinitely long cosine wave of a specific frequency



- Wavelet Analysis is based on an short duration wavelet of a specific center frequency



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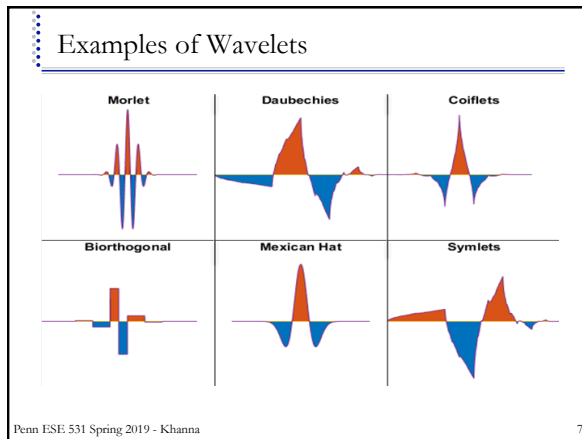
### Wavelet Transform

- All wavelet derived from *mother* wavelet

$$\psi_{s,\tau}(t) = \frac{1}{\sqrt{s}} \psi\left(\frac{t-\tau}{s}\right)$$

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### Wavelet – Scaled and Shifted

$$\Psi_{s,\tau}(t) = \frac{1}{\sqrt{s}} \Psi\left(\frac{t-\tau}{s}\right)$$

Annotations for the equation:

- $\frac{1}{\sqrt{s}}$ : normalization
- $t - \tau$ : shift in time
- $s$ : change in scale: big s means long wavelength
- $\Psi$ : Mother wavelet
- $\Psi_{s,\tau}(t)$ : wavelet with scale, s and time,  $\tau$

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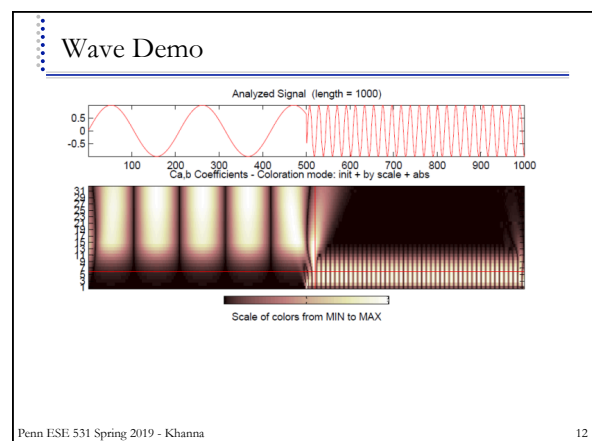
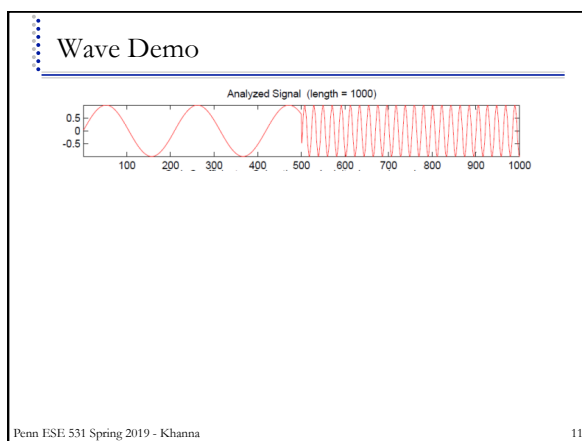
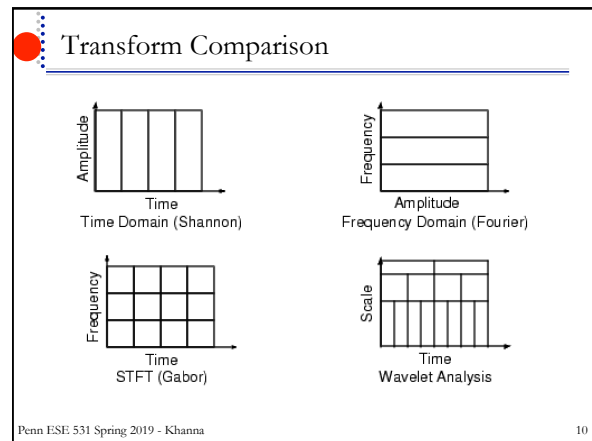
### Continuous Wavelet Transform

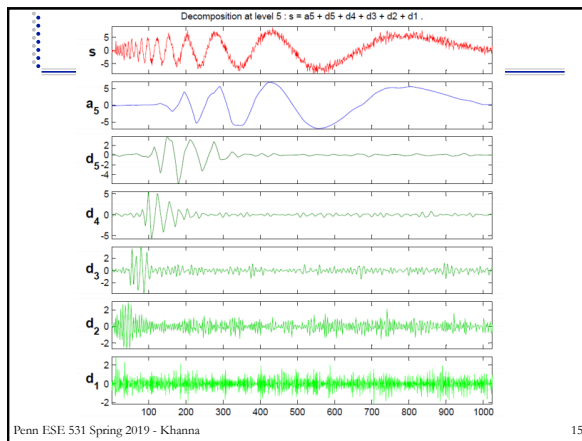
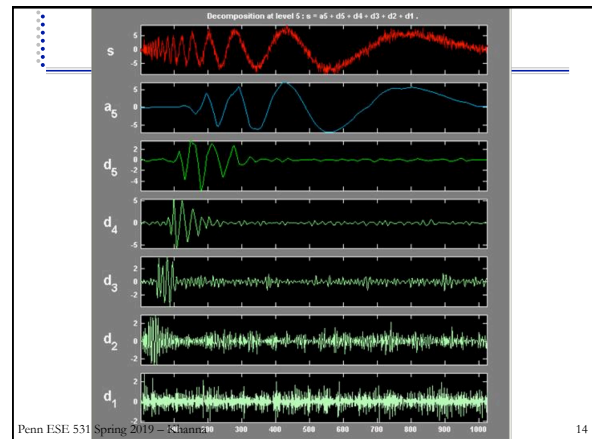
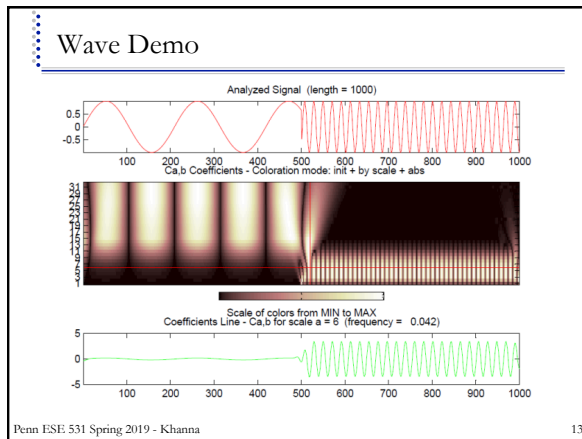
$$\gamma(s, \tau) = \int f(t) \Psi_{s,\tau}(t) dt$$

Annotations for the equation:

- $f(t)$ : time-series
- $\gamma(s, \tau)$ : coefficient of wavelet with scale, s and time,  $\tau$
- $\Psi_{s,\tau}(t)$ : wavelet with scale, s, and shift,  $\tau$

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### Inverse Wavelet Transform

- Build up a time-series as sum of wavelets of different scales,  $s$ , and positions,  $t$

$$f(t) = \int \int \gamma(s, \tau) \psi_{s, \tau}(t) d\tau ds$$

time-series

coefficients of wavelets

wavelet with scale,  $s$  and time,  $\tau$

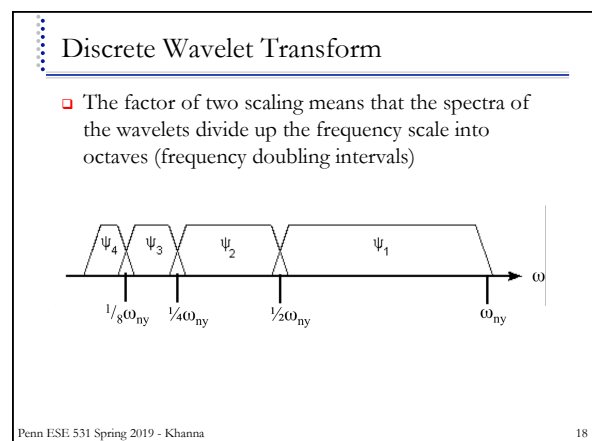
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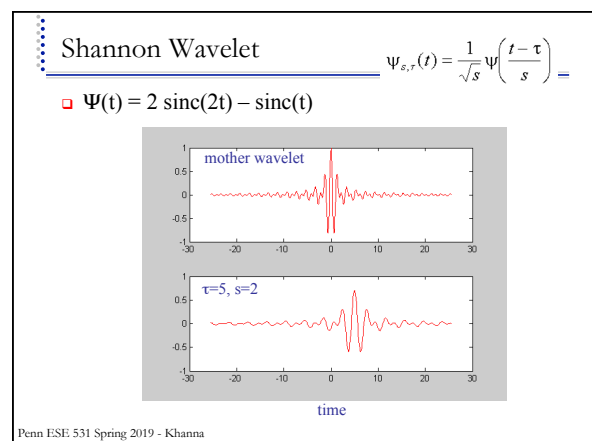
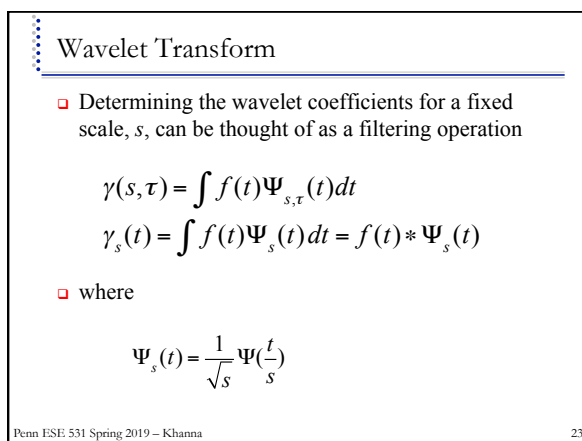
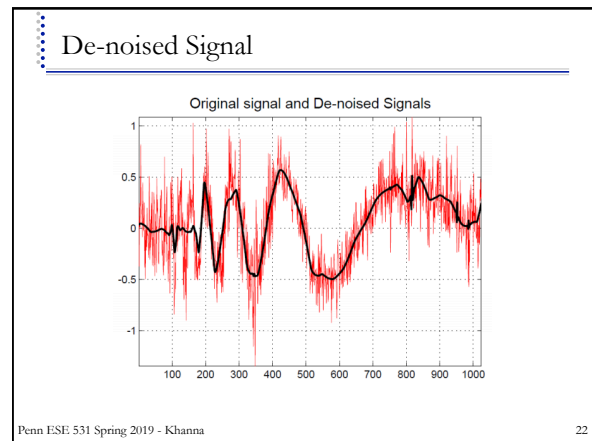
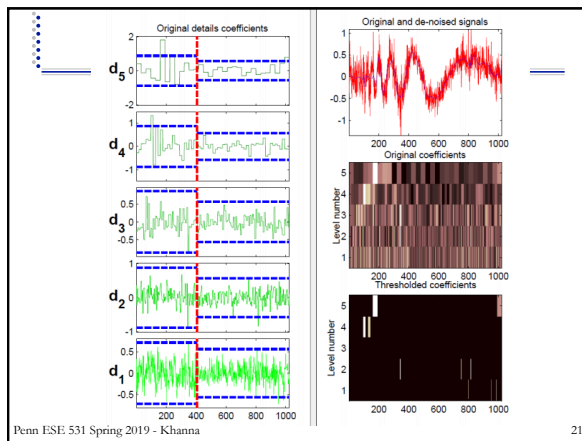
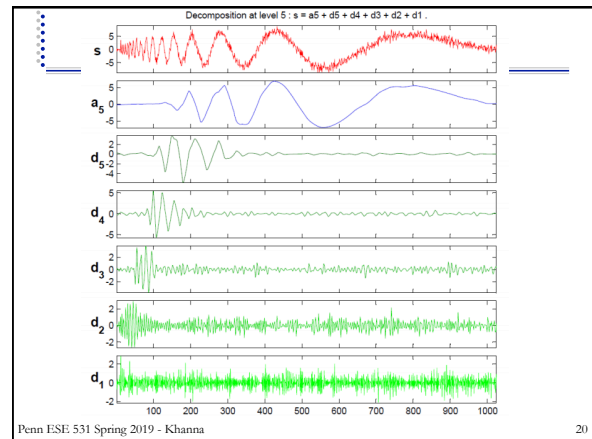
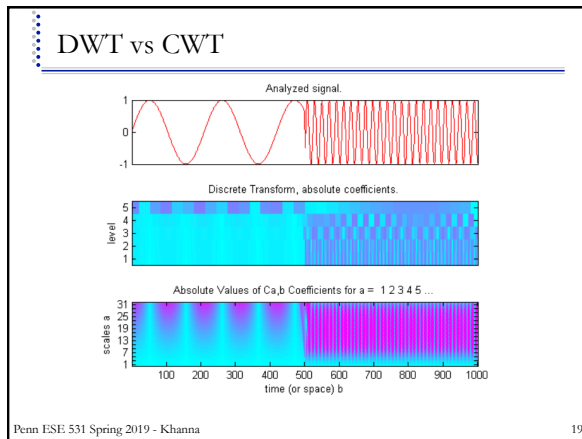
### Discrete wavelets:

- Scale wavelets only by powers of 2
  - $s_j = 2^j$
- And shifting by multiples of  $s_j$  for each successive scale
  - $\tau_{j,k} = 2^j k$
- Then  $\gamma(s_j, \tau_{j,k}) = \gamma_{j,k}$ 
  - where  $j = 1, 2, \dots, \infty, k = -\infty, \dots, -2, -1, 0, 1, 2, \dots, \infty$

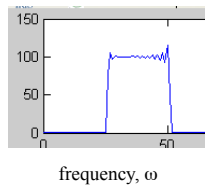
$$\gamma_{j,k} = \frac{1}{\sqrt{2^j}} \int f(t) \psi\left(\frac{t - k2^j}{2^j}\right) dt$$

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## Fourier spectrum of Shannon Wavelet



- Wavelet coefficients are a result of bandpass filtering

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## Discrete Wavelet Transform

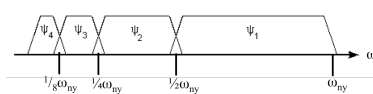
- The coefficients of  $\Psi$  is just the band-pass filtered time-series, where  $\Psi$  is the wavelet, now viewed as the impulse response of a bandpass filter.

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## Discrete Wavelet Transform

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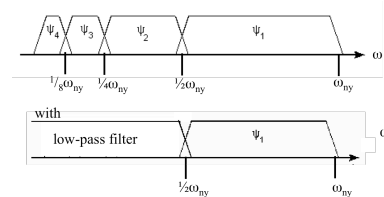


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## Discrete Wavelet Transform

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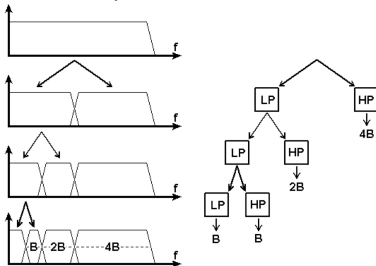


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## Digital Wavelet as Multirate Filter Bank

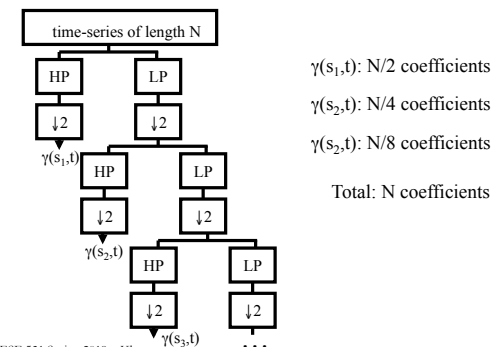
- Repeat recursively!



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## Digital Wavelet as Multirate Filter Bank

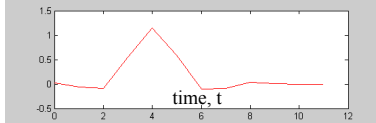


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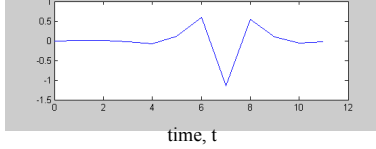
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## Impulse Responses

Coiflet low pass filter



Coiflet high-pass filter

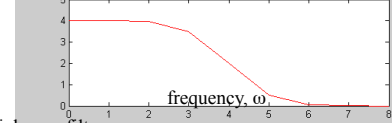


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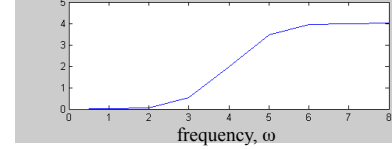
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## Filter Responses

Spectrum of low pass filter

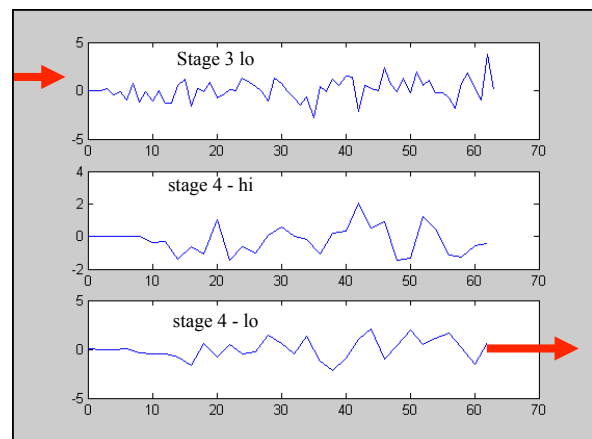
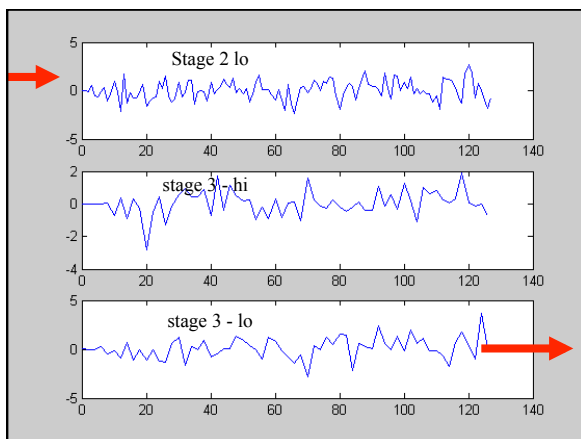
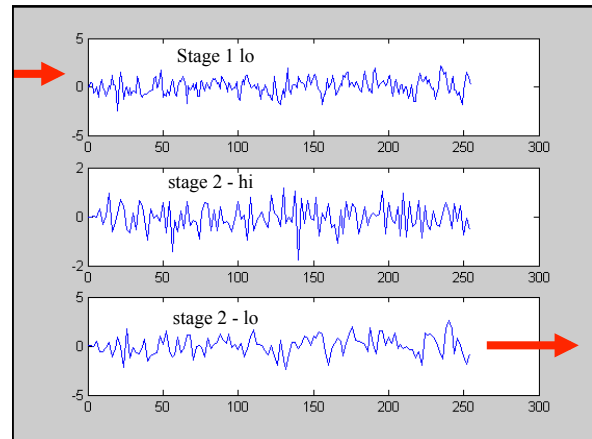
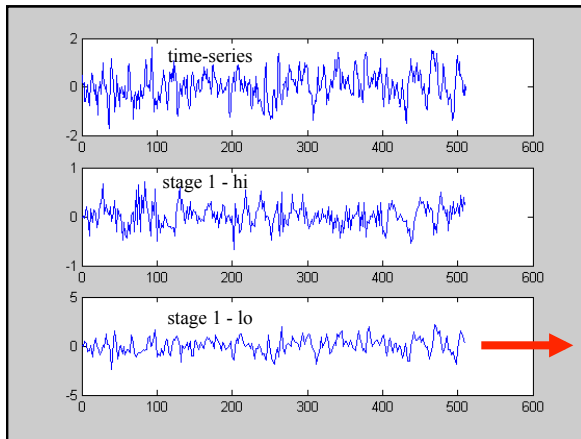


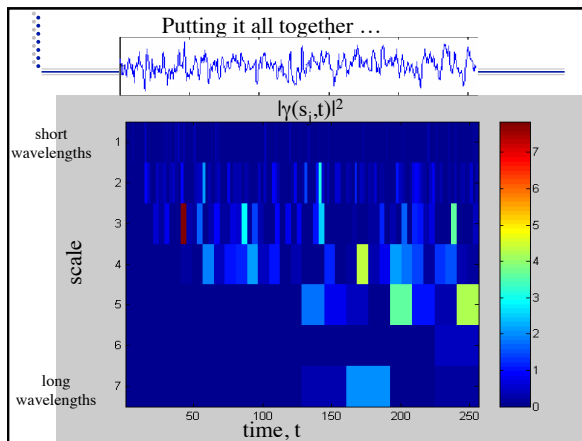
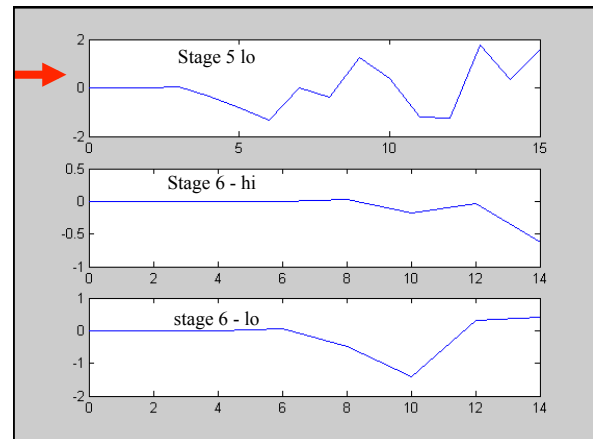
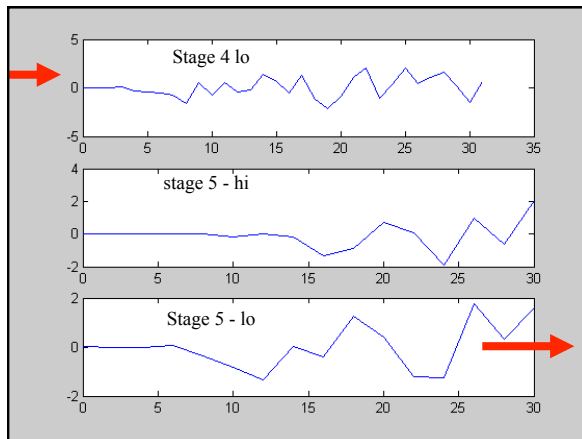
Spectrum of high pass filter



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### Big Ideas

- Wavelet transform
  - Capture temporal data with fewer coefficients than STFT
  - Use scaling and translation to get different resolution at different levels

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### Admin

- Project
  - Due 4/30

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