



# Short Time Fourier Transform (STFT)



# Fourier Transform

- Fourier Transform reveals **which** frequency components are present in a given function.

$$f(x) = \sum_{u=0}^{N-1} F(u) e^{\frac{j2\pi ux}{N}}, x = 0, 1, \dots, N-1$$

(inverse DFT)

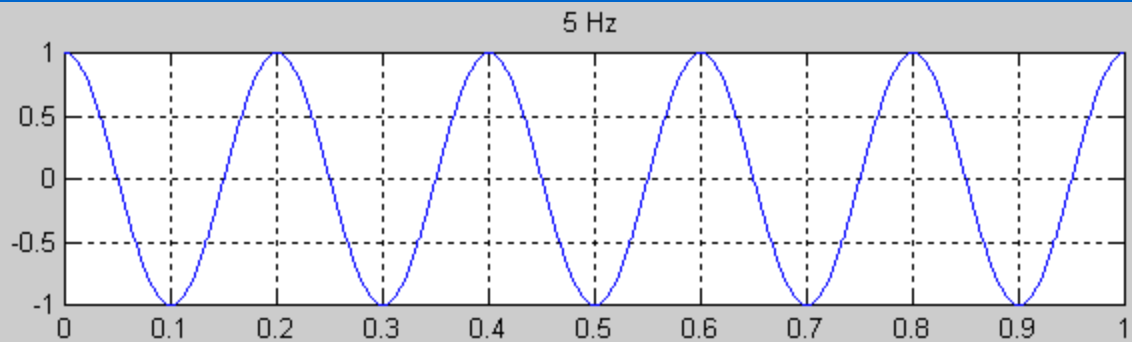
where:

$$F(u) = \frac{1}{N} \sum_{x=0}^{N-1} f(x) e^{\frac{-j2\pi ux}{N}}, u = 0, 1, \dots, N-1$$

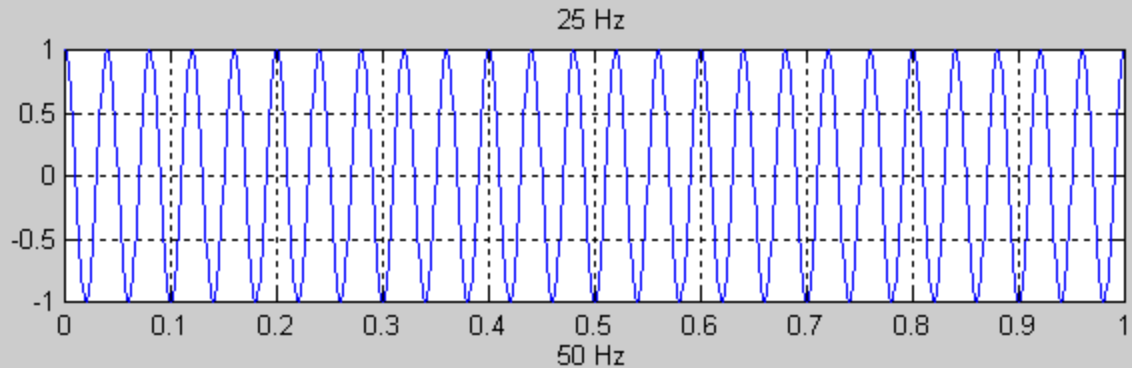
(forward DFT)

# Examples

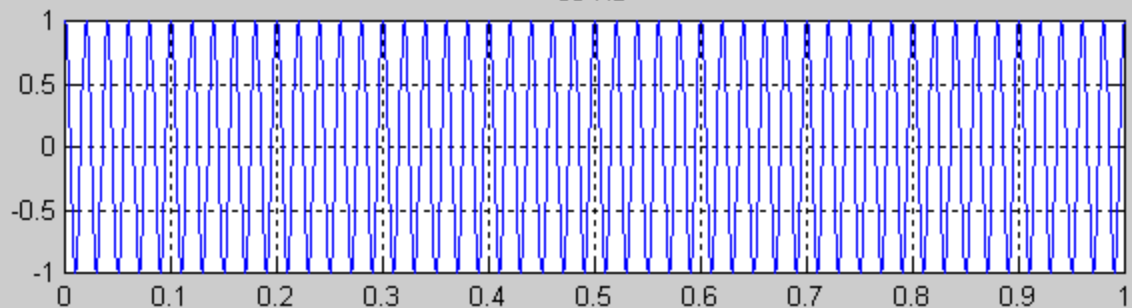
$$f_1(t) = \cos(2\pi \cdot 5 \cdot t)$$



$$f_2(t) = \cos(2\pi \cdot 25 \cdot t)$$

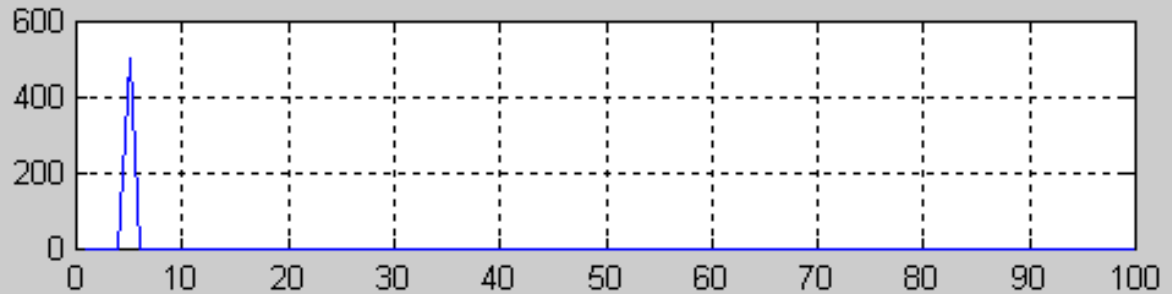


$$f_3(t) = \cos(2\pi \cdot 50 \cdot t)$$

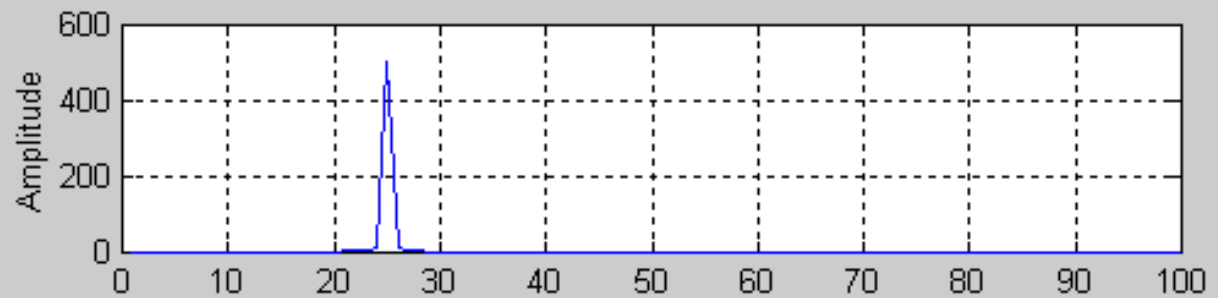


## Examples (cont'd)

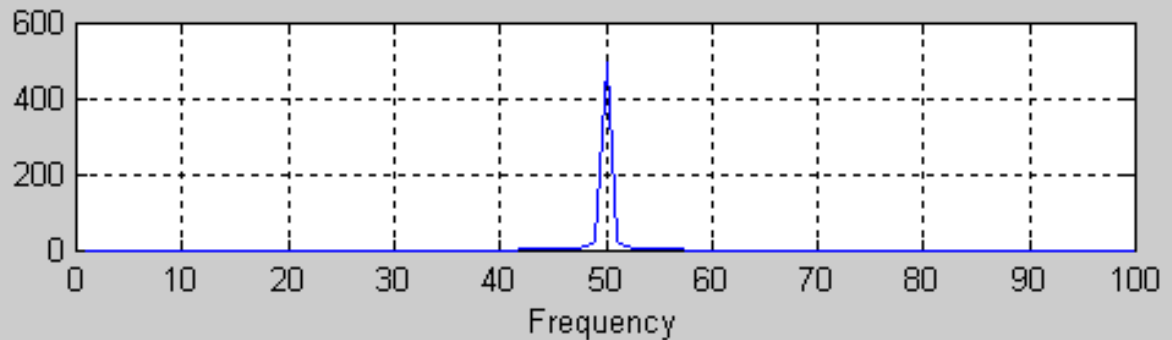
$F_1(u)$



$F_2(u)$



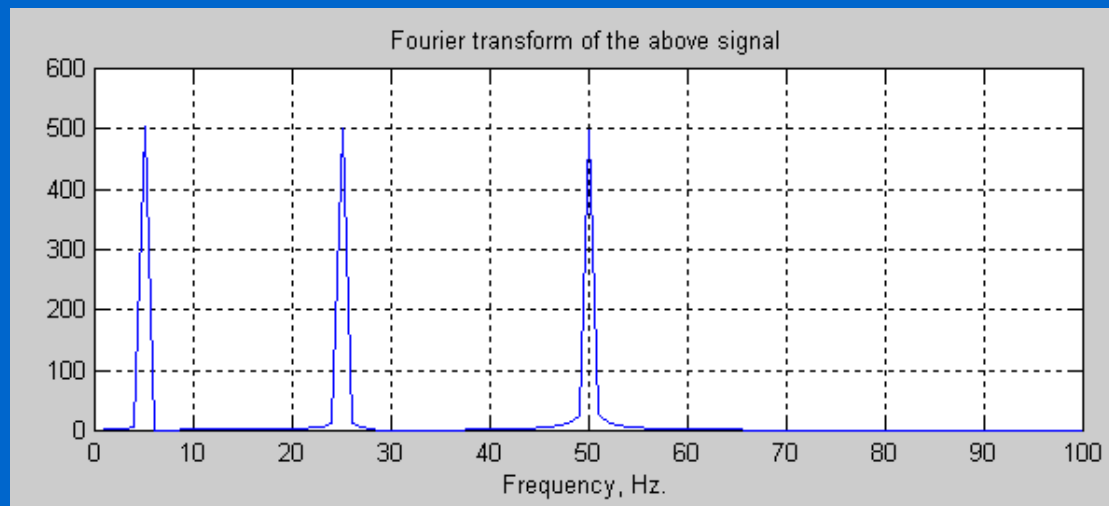
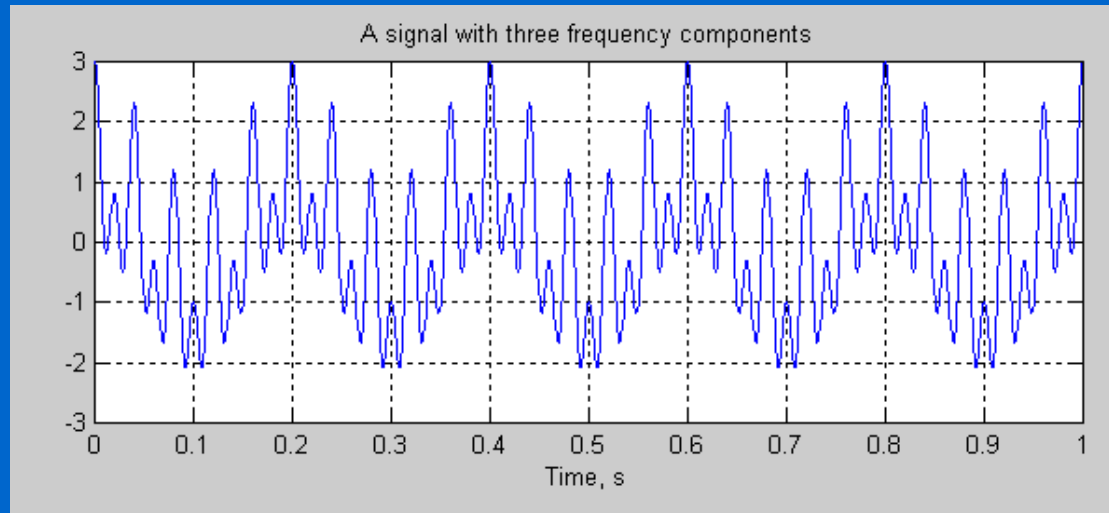
$F_3(u)$



# Fourier Analysis – Examples (cont'd)

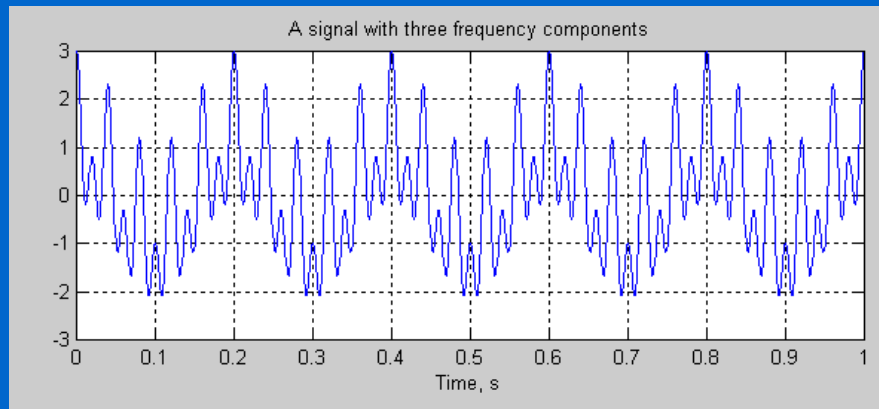
$$\begin{aligned} f_4(t) = & \cos(2\pi \cdot 5 \cdot t) \\ & + \cos(2\pi \cdot 25 \cdot t) \\ & + \cos(2\pi \cdot 50 \cdot t) \end{aligned}$$

$F_4(u)$  ?

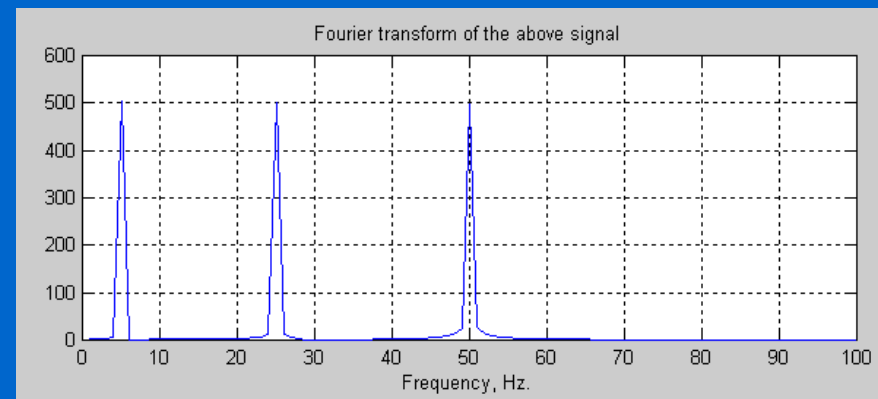


# Limitations of Fourier Transform

1. Cannot not provide **simultaneous** time and frequency localization.



Poor localization in freq domain!



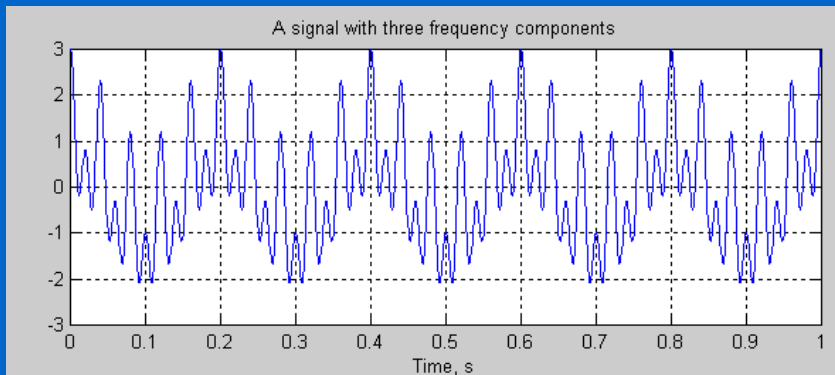
Great localization in freq domain!

# Limitations of Fourier Transform (cont'd)

2. Not very useful for analyzing **time-variant, non-stationary** signals.

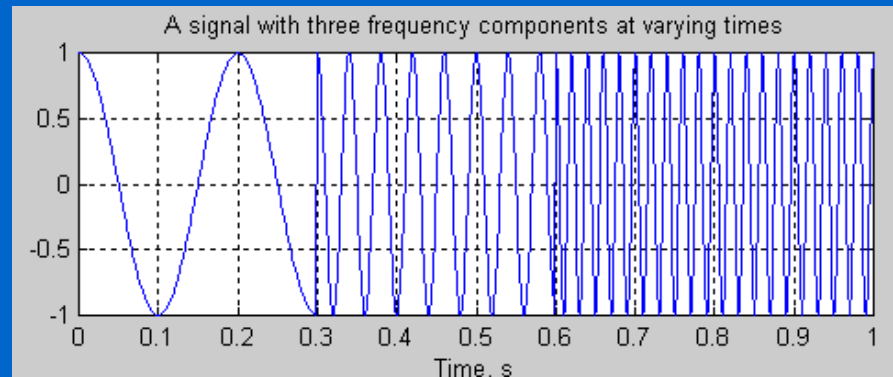
$f_4(t)$

**Stationary signal**  
(non-varying frequency)



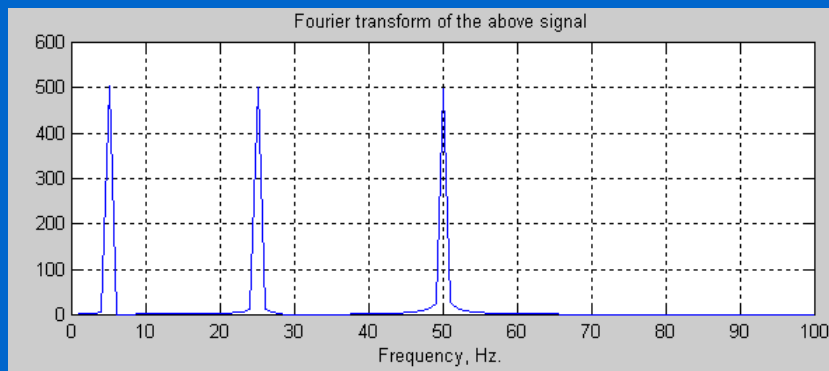
$f_5(t)$

**Non-stationary signal**  
(varying frequency)

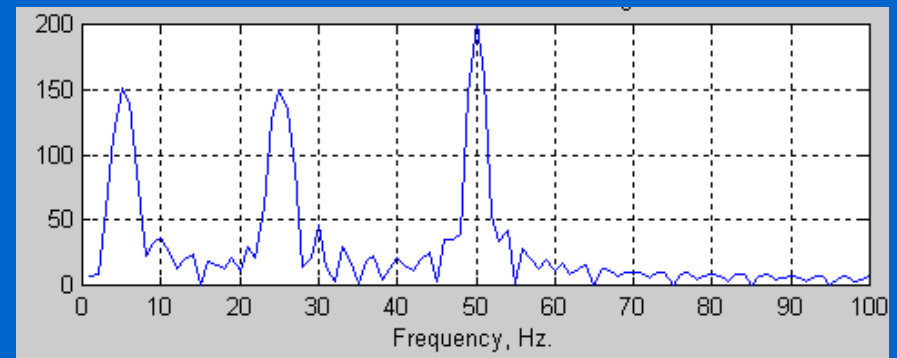


# Limitations of Fourier Transform (cont'd)

$F_4(u)$  Three frequency components, present at **all times!**



$F_5(u)$  Three frequency components, **NOT** present at all times!



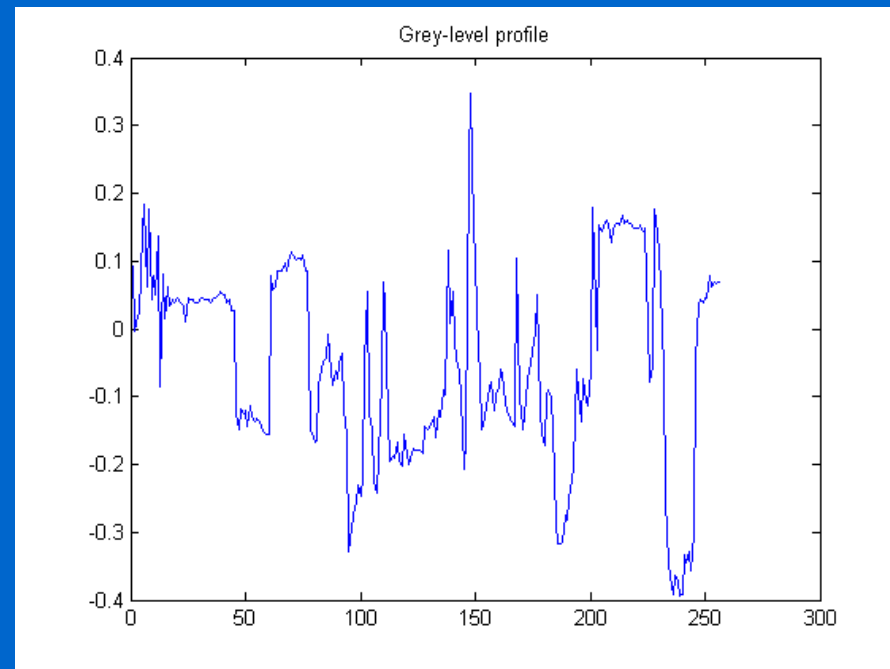
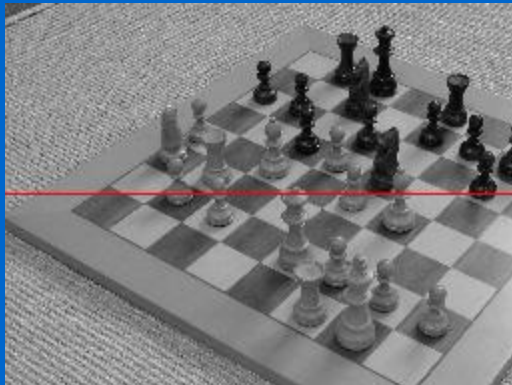
Perfect knowledge of **what** frequencies exist, but **no** information about **where** these frequencies are located in **time!**



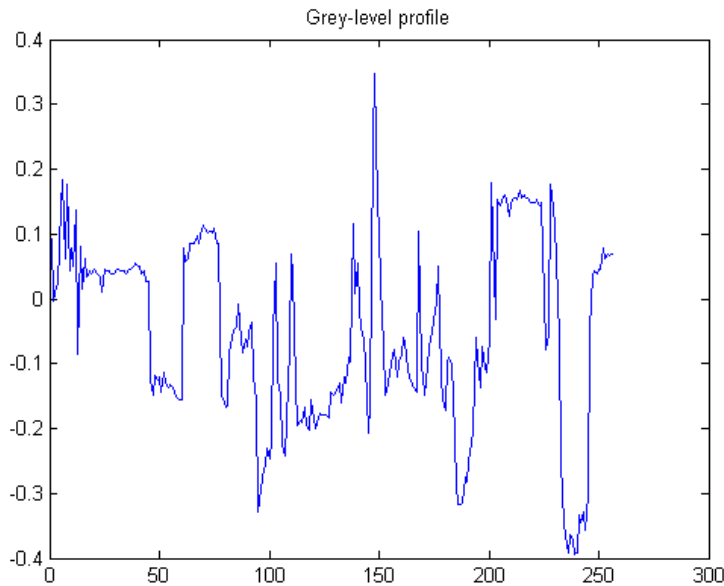
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# Limitations of Fourier Transform (cont'd)

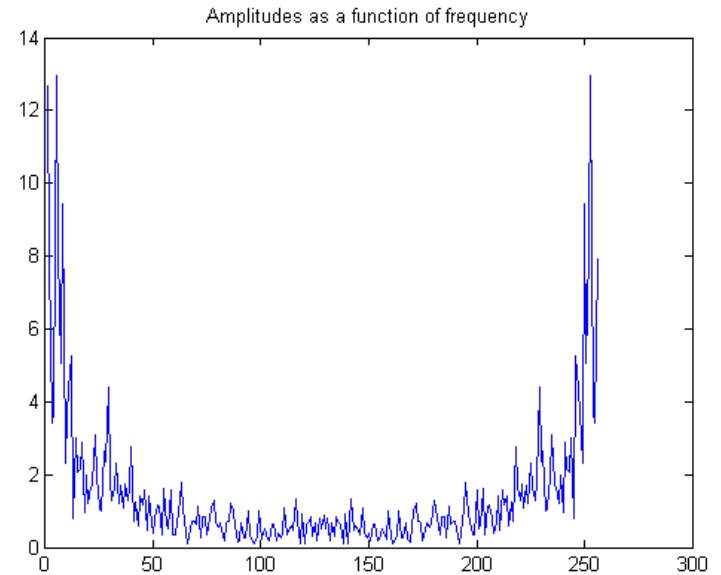
3. **Not efficient** for representing non-smooth functions.



# Representing discontinuities or sharp corners (cont'd)



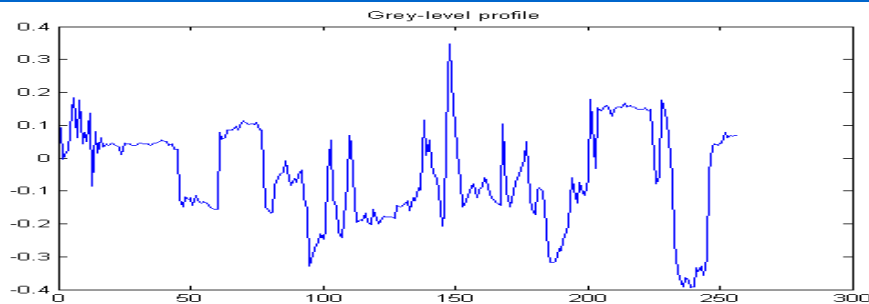
FT  
→



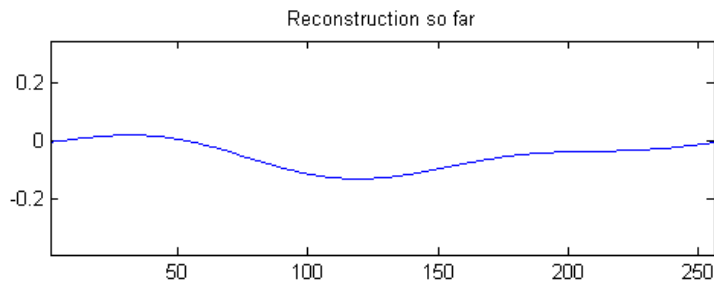
$$F(u) = \frac{1}{N} \sum_{x=0}^{N-1} f(x) e^{\frac{-j2\pi ux}{N}}, u = 0, 1, \dots, N-1$$

# Representing discontinuities or sharp corners (cont'd)

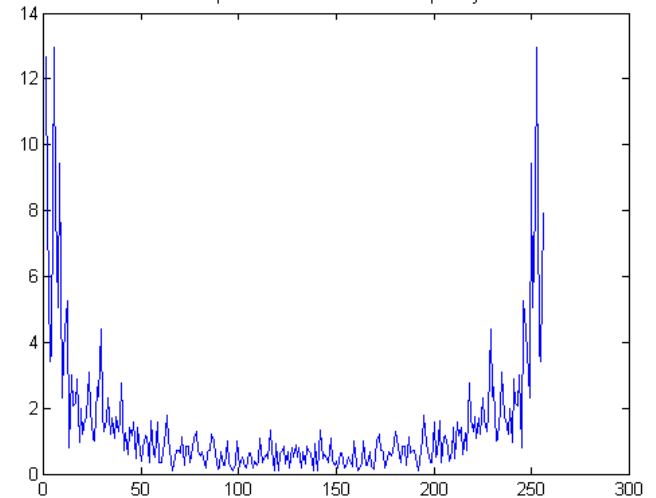
Original



Reconstructed



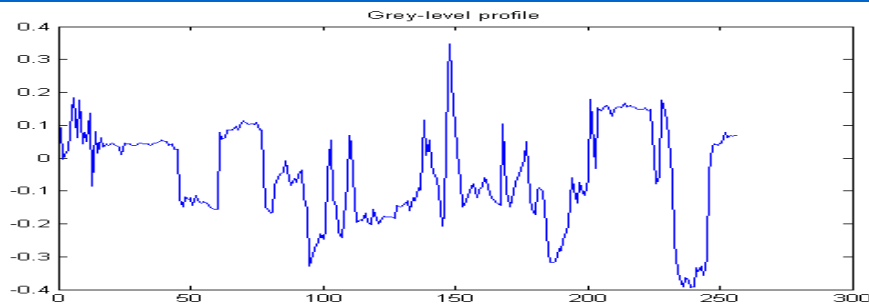
Amplitudes as a function of frequency



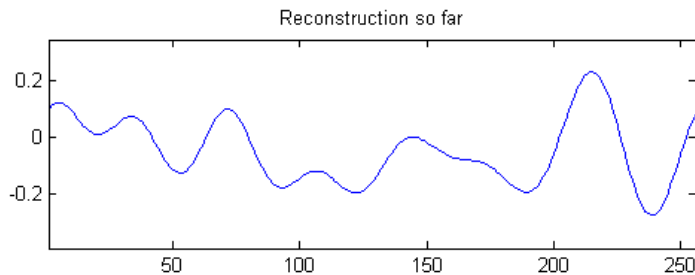
$$f(x) = \sum_{u=0}^1 F(u) e^{\frac{j2\pi ux}{N}}, x = 0, 1, \dots, N-1$$

# Representing discontinuities or sharp corners (cont'd)

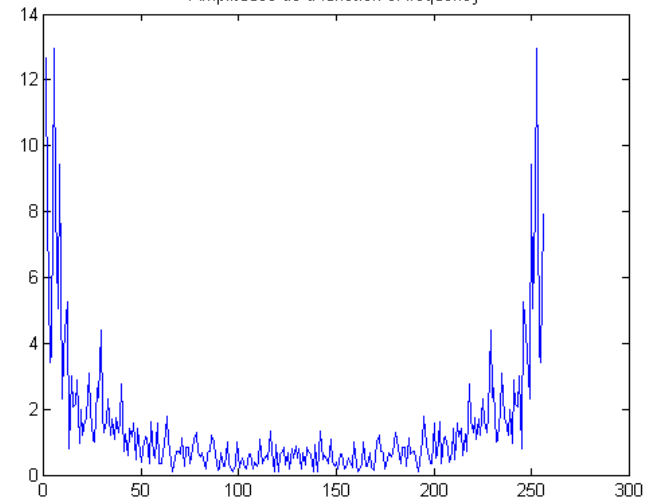
Original



Reconstructed



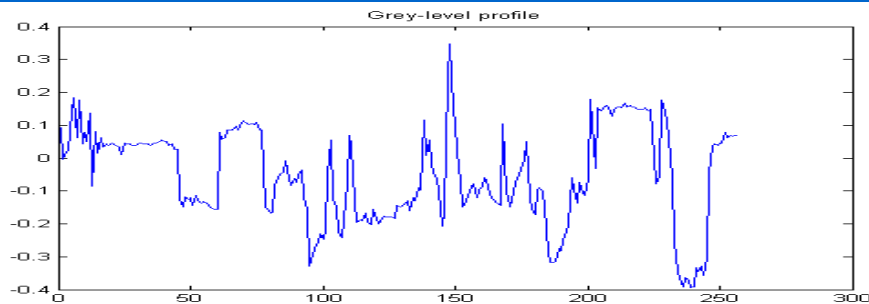
Amplitudes as a function of frequency



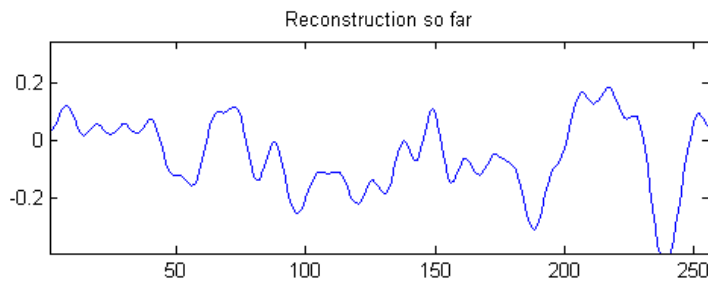
$$f(x) = \sum_{u=0}^7 F(u) e^{\frac{j2\pi ux}{N}}, x = 0, 1, \dots, N-1$$

# Representing discontinuities or sharp corners (cont'd)

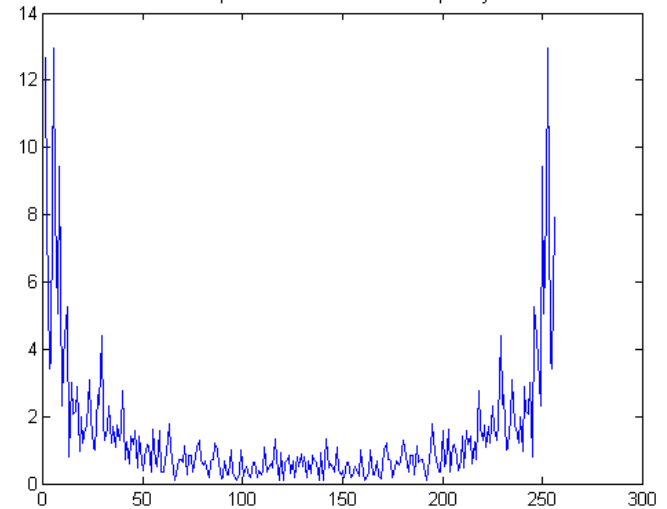
Original



Reconstructed



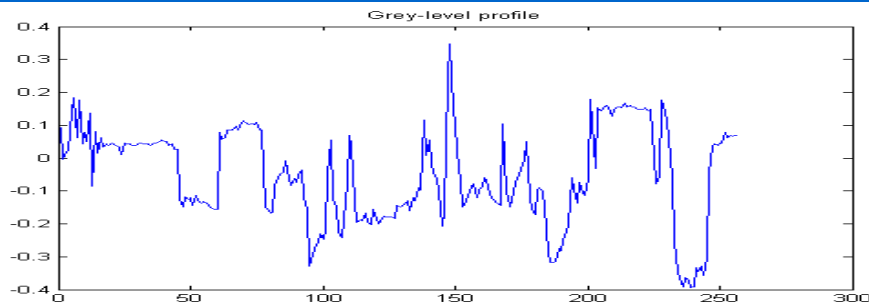
Amplitudes as a function of frequency



$$f(x) = \sum_{u=0}^{23} F(u) e^{\frac{j2\pi ux}{N}}, \quad x = 0, 1, \dots, N-1$$

# Representing discontinuities or sharp corners (cont'd)

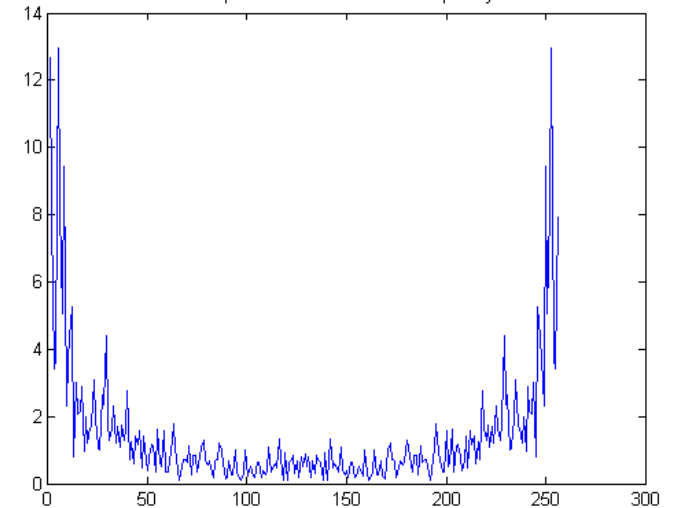
Original



Reconstructed



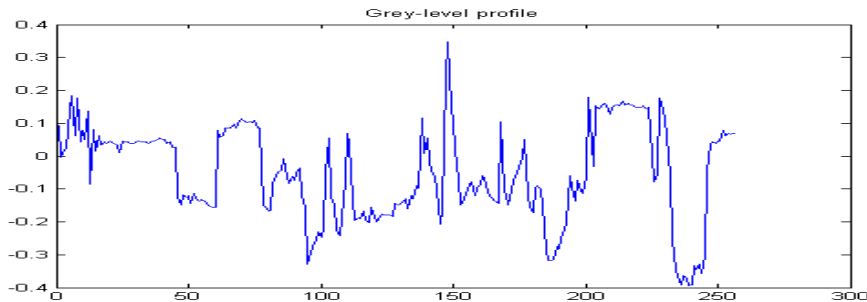
Amplitudes as a function of frequency



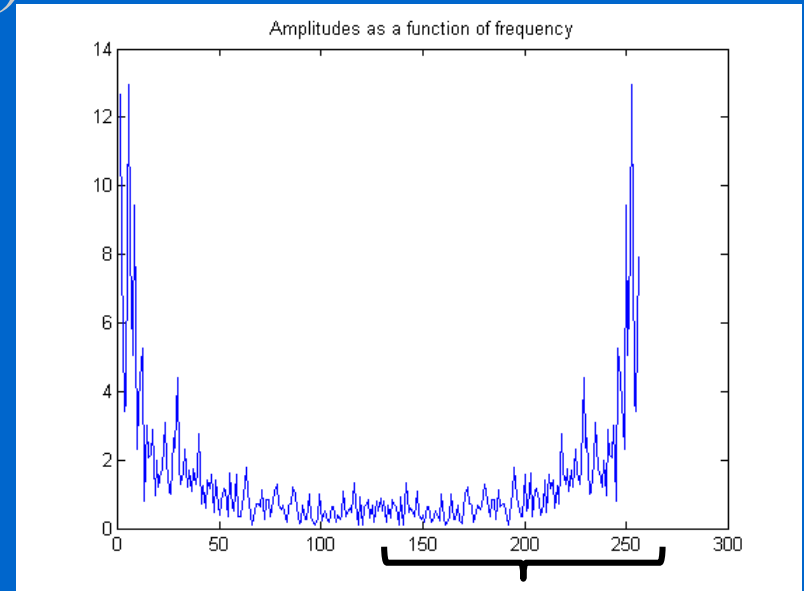
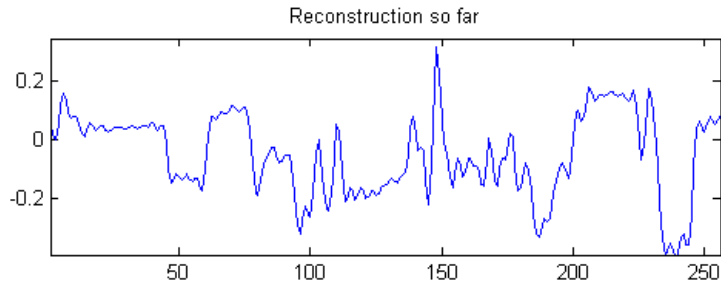
$$f(x) = \sum_{u=0}^{39} F(u) e^{\frac{j2\pi ux}{N}}, \quad x = 0, 1, \dots, N-1$$

# Representing discontinuities or sharp corners (cont'd)

Original



Reconstructed

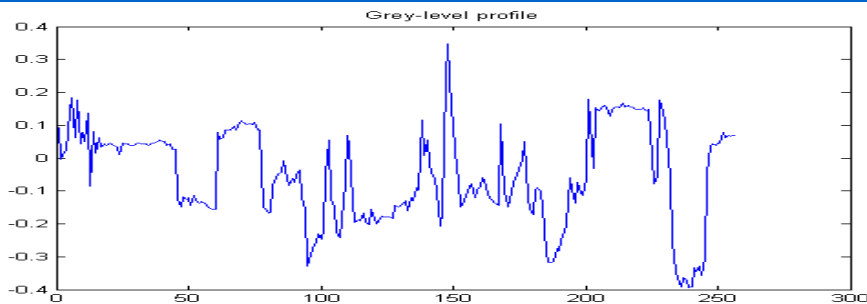


128 coefficients

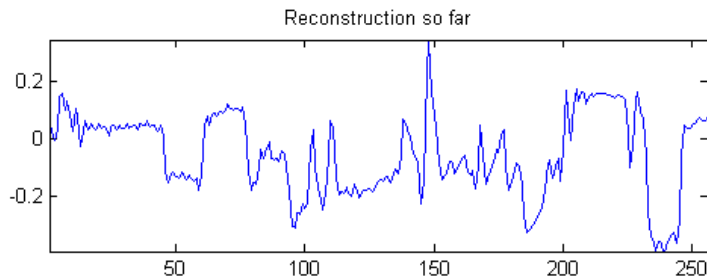
$$f(x) = \sum_{u=0}^{63} F(u) e^{\frac{j2\pi ux}{N}}, x = 0, 1, \dots, N-1$$

# Representing discontinuities or sharp corners (cont'd)

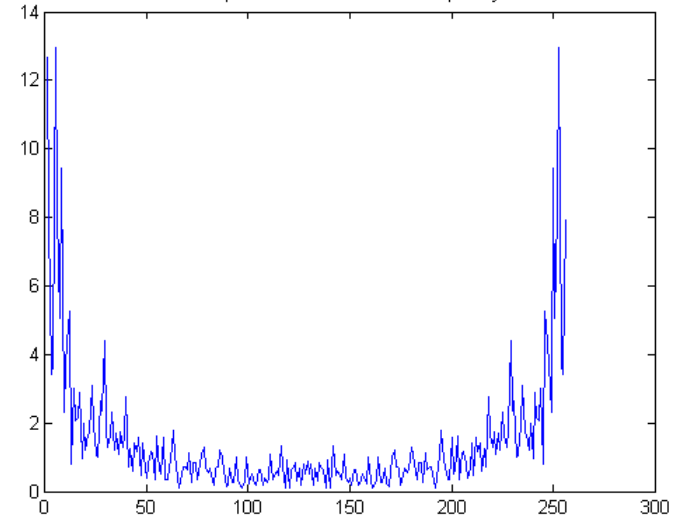
Original



Reconstructed



Amplitudes as a function of frequency

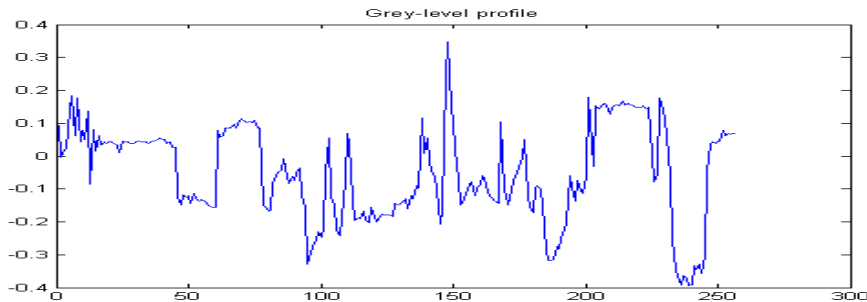


$$f(x) = \sum_{u=0}^{95} F(u) e^{\frac{j2\pi ux}{N}}, \quad x = 0, 1, \dots, N-1$$

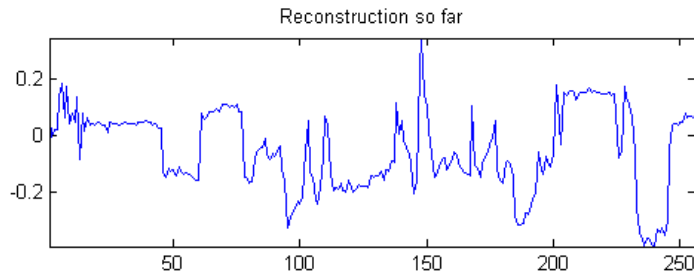


# Representing discontinuities or sharp corners (cont'd)

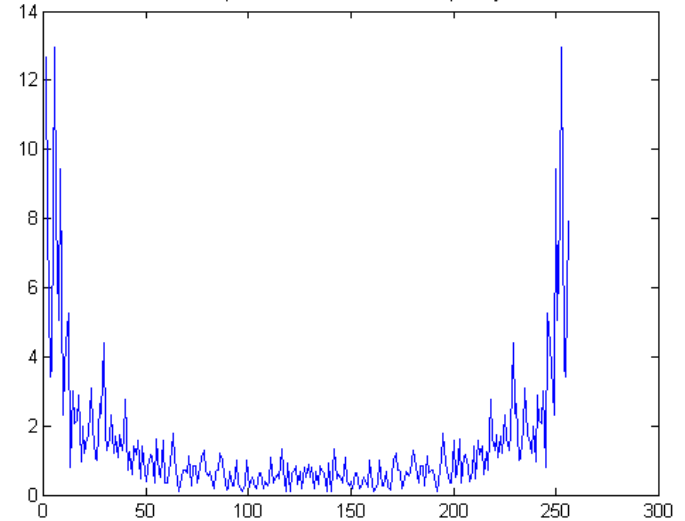
Original



Reconstructed



Amplitudes as a function of frequency

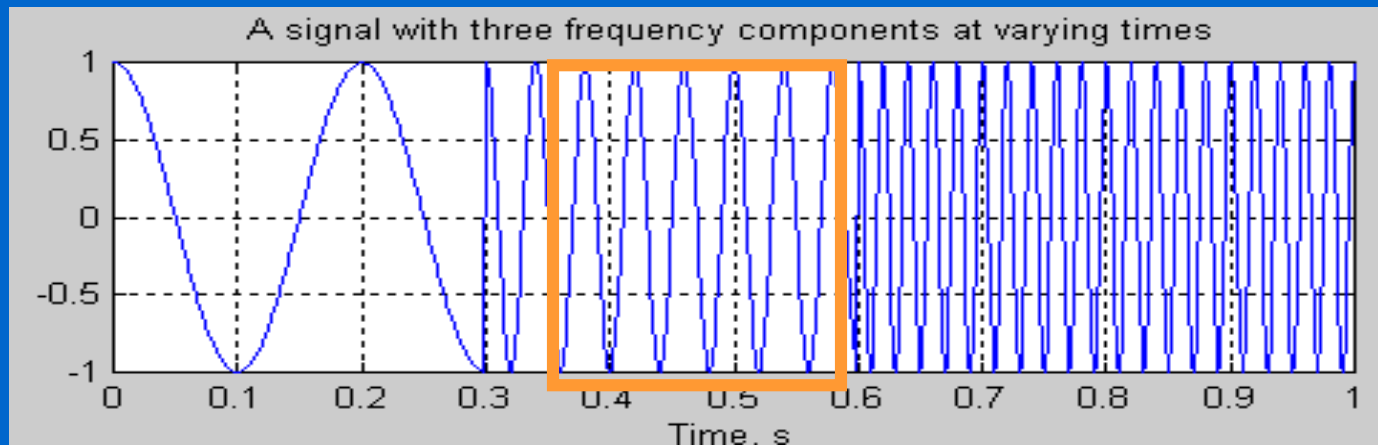


$$f(x) = \sum_{u=0}^{127} F(u) e^{\frac{j2\pi ux}{N}}, \quad x = 0, 1, \dots, N-1$$

A **large number** of Fourier components  
is needed to represent discontinuities.

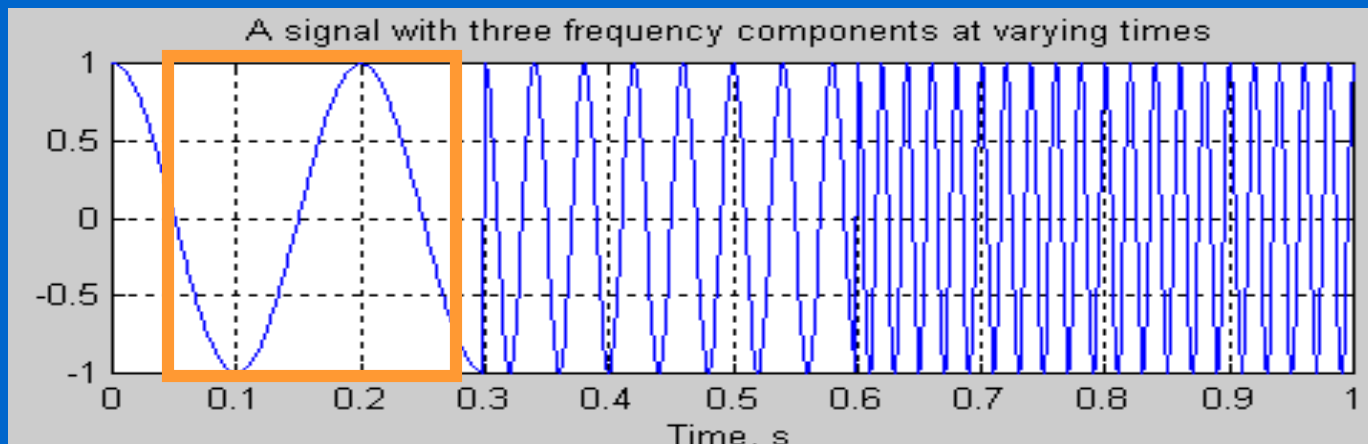
# Short Time Fourier Transform (STFT)

- Segment signal into narrow time intervals (i.e., narrow enough to be considered stationary) and take the FT of each segment.
- Each FT provides the spectral information of a separate time-slice of the signal, providing simultaneous time and frequency information.



# STFT - Steps

- (1) Choose a window of finite length
- (2) Place the window on top of the signal at  $t=0$
- (3) Truncate the signal using this window
- (4) Compute the FT of the truncated signal, save results.
- (5) Incrementally slide the window to the right
- (6) Go to step 3, until window reaches the end of the signal



# STFT - Definition

time parameter      frequency parameter

$$STFT_f^u(t', u) = \int_t [f(t) \cdot W(t - t')] \cdot e^{-j2\pi ut} dt$$

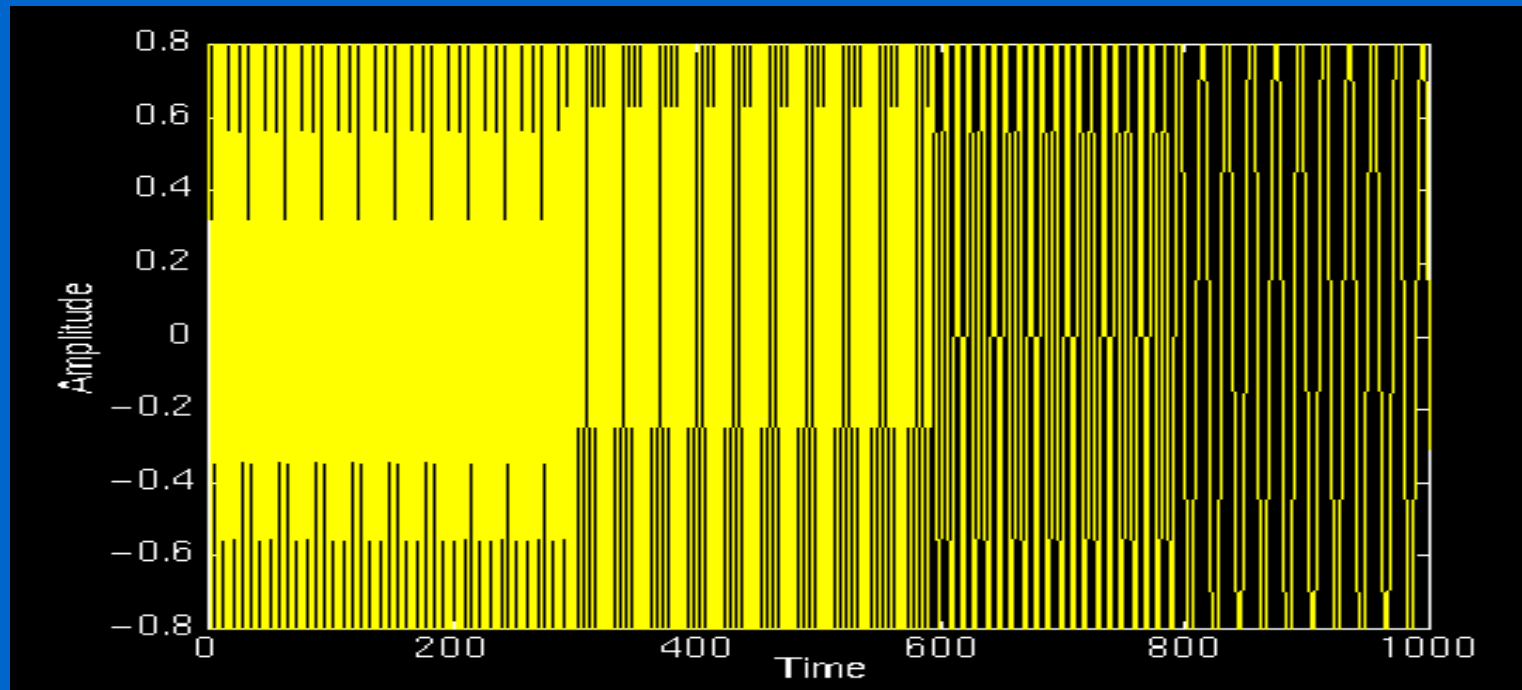
2D function

windowing function      centered at  $t=t'$

The diagram illustrates the Short-Time Fourier Transform (STFT) definition. The equation is presented on a light blue background. Annotations include: 'time parameter' pointing to  $t'$ , 'frequency parameter' pointing to  $u$ , '2D function' pointing to the entire left-hand side  $STFT_f^u(t', u)$ , 'windowing function' pointing to  $W(t - t')$ , and 'centered at  $t=t'$ ' pointing to the argument  $t - t'$  of the windowing function. The integration variable  $t$  is indicated by a subscript below the integral sign.

# Example

$f(t)$



[0 – 300] ms  $\rightarrow$  75 Hz sinusoid

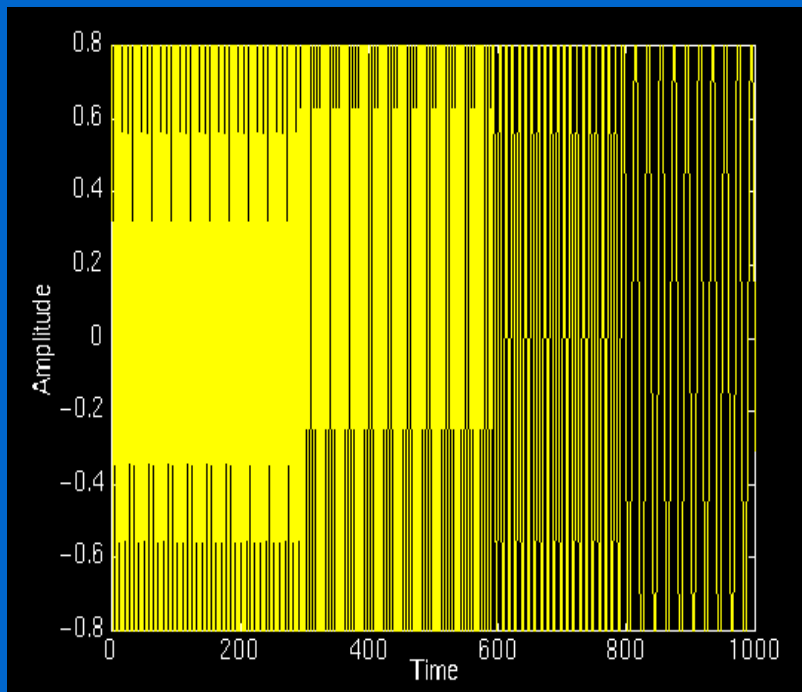
[300 – 600] ms  $\rightarrow$  50 Hz sinusoid

[600 – 800] ms  $\rightarrow$  25 Hz sinusoid

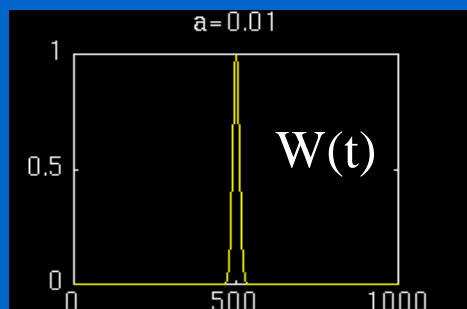
[800 – 1000] ms  $\rightarrow$  10 Hz sinusoid

# Example

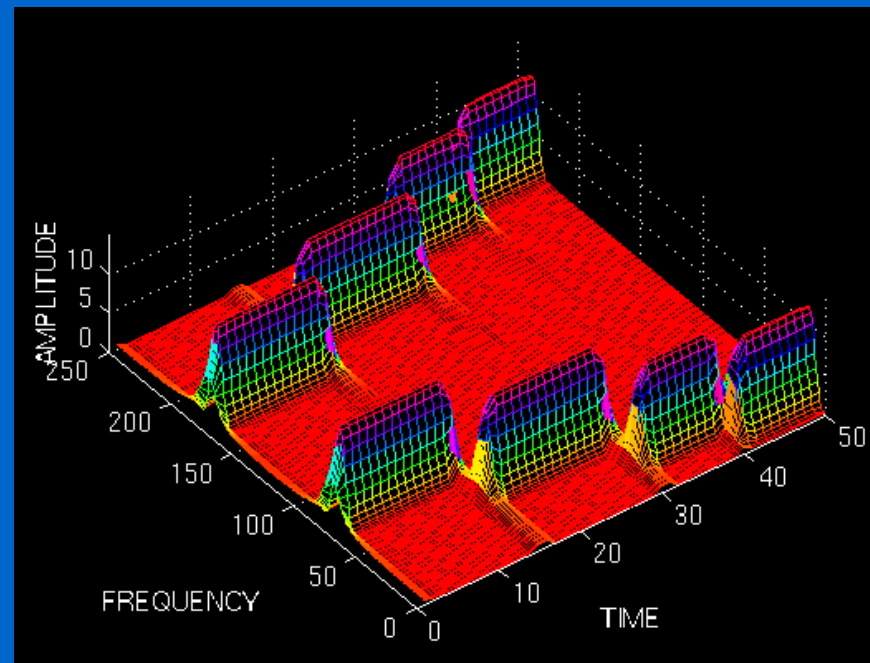
$f(t)$



[0 – 300] ms → 75 Hz  
 [300 – 600] ms → 50 Hz  
 [600 – 800] ms → 25 Hz  
 [800 – 1000] ms → 10 Hz



$$STFT_f^u(t', u)$$



scaled:  $t/20$

## Choosing Window $W(t)$

- What shape should it have?
  - Rectangular, Gaussian, Elliptic ...
- How wide should it be?
  - Should be **narrow** enough to ensure that the portion of the signal falling within the window is stationary.
  - Very narrow windows, however, do not offer good **localization** in the frequency domain.

# STFT Window Size

$$STFT_f^u(t', u) = \int_t [f(t) \cdot W(t - t')] \cdot e^{-j2\pi ut} dt$$

**$W(t)$  infinitely long:**  $W(t) = u(t)$   $\rightarrow$  STFT turns into FT, providing excellent frequency localization, but no time localization.

**$W(t)$  infinitely short:**  $W(t) = \delta(t)$   $\rightarrow$  results in the time signal (with a phase factor), providing excellent time localization but no frequency localization.

$$STFT_f^u(t', u) = \int_t [f(t) \cdot \delta(t - t')] \cdot e^{-j2\pi ut} dt = f(t') \cdot e^{-jut'}$$

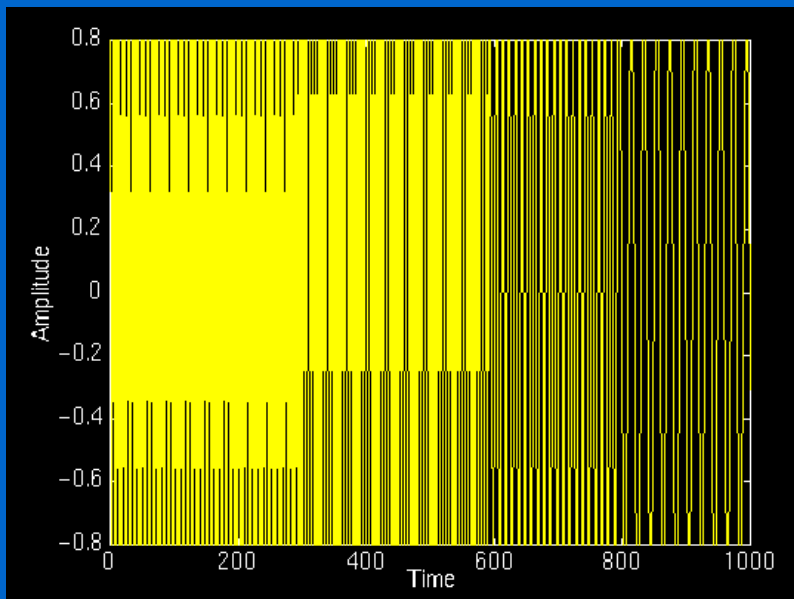


## STFT Window Size (cont'd)

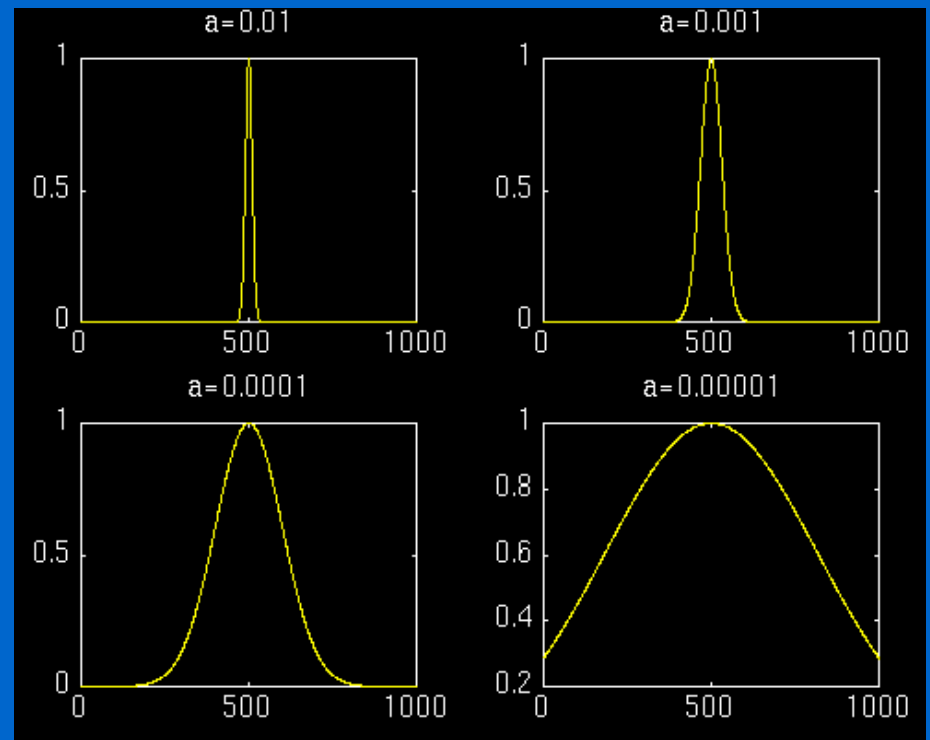
- **Wide window** → good frequency resolution, poor time resolution.
- **Narrow window** → good time resolution, poor frequency resolution.

# Example

different size windows

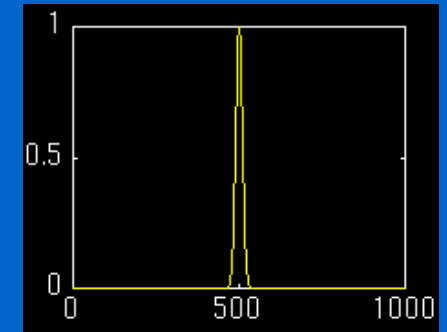
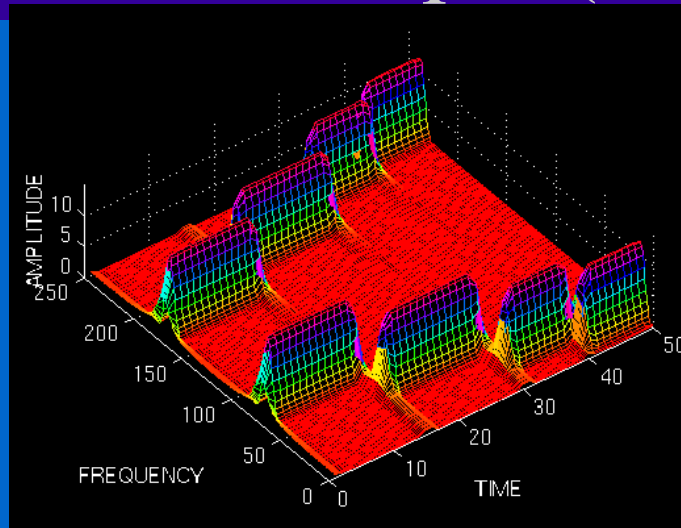


[0 – 300] ms → 75 Hz  
[300 – 600] ms → 50 Hz  
[600 – 800] ms → 25 Hz  
[800 – 1000] ms → 10 Hz

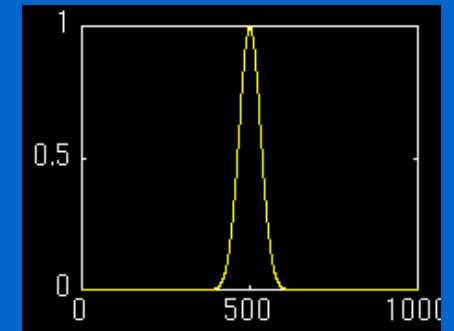
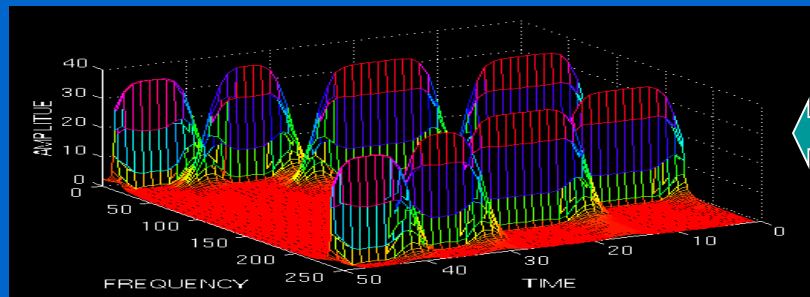


# Example (cont'd)

$$STFT_f^u(t', u)$$



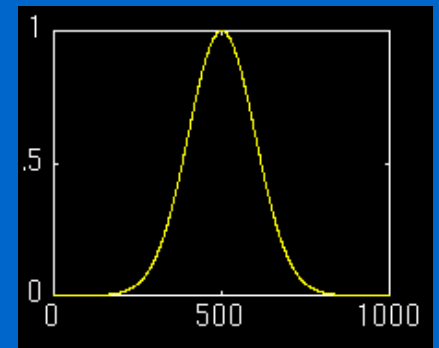
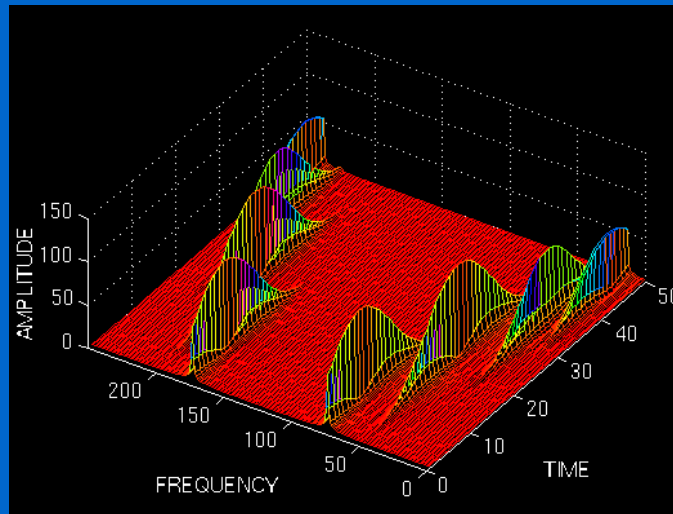
$$STFT_f^u(t', u)$$



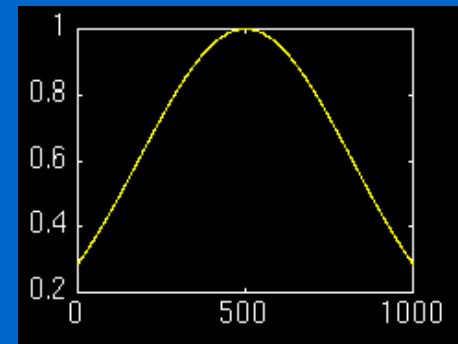
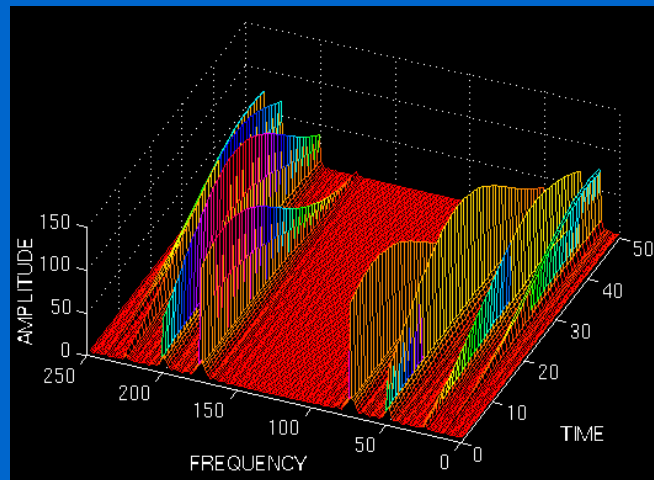
scaled:  $t/20$

## Example (cont'd)

$$STFT_f^u(t', u)$$




$$STFT_f^u(t', u)$$



scaled:  $t/20$

# Heisenberg (or Uncertainty) Principle

$$\Delta t \cdot \Delta f \geq \frac{1}{4\pi}$$


**Time resolution:** How well two spikes in time can be separated from each other in the frequency domain.

**Frequency resolution:** How well two spectral components can be separated from each other in the time domain

*$\Delta t$  and  $\Delta f$  cannot be made arbitrarily small !*

•  
•  
•

# Heisenberg (or Uncertainty) Principle

- We cannot know the **exact** time-frequency representation of a signal.
- We can only know what *interval of frequencies* are present in which *time intervals*.