



Short Time Fourier Transform (STFT)

Phan Duy Hùng



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)

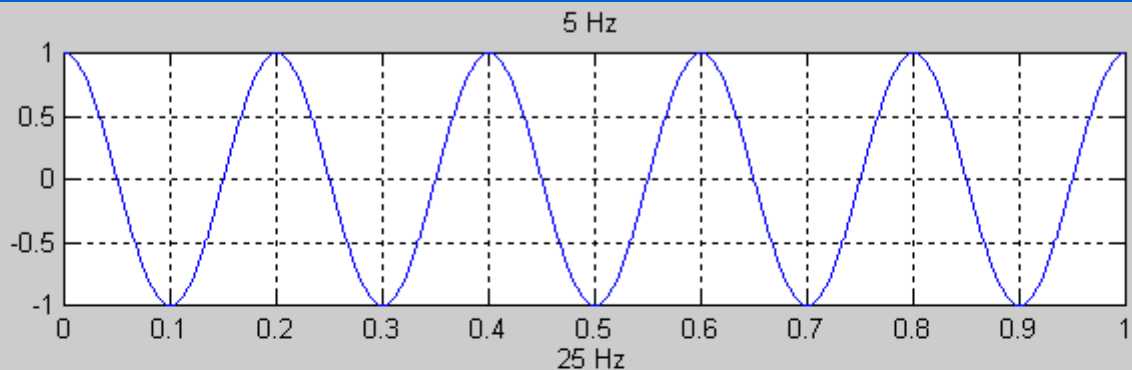
where:

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

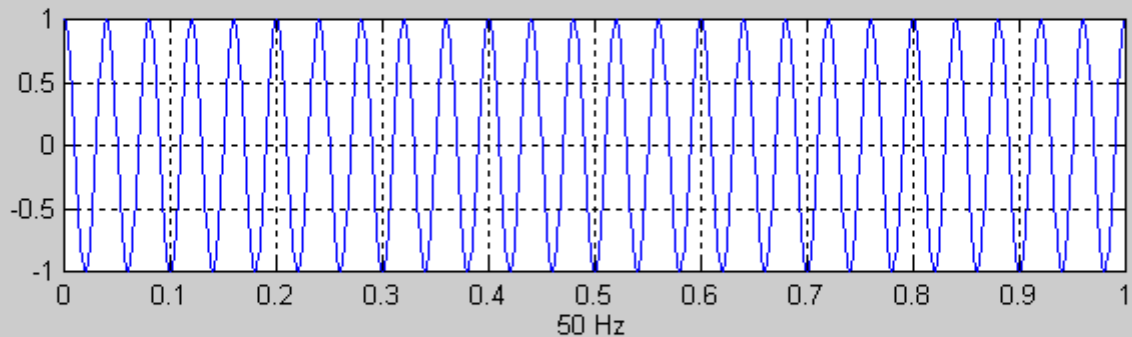
(forward DFT)

Examples

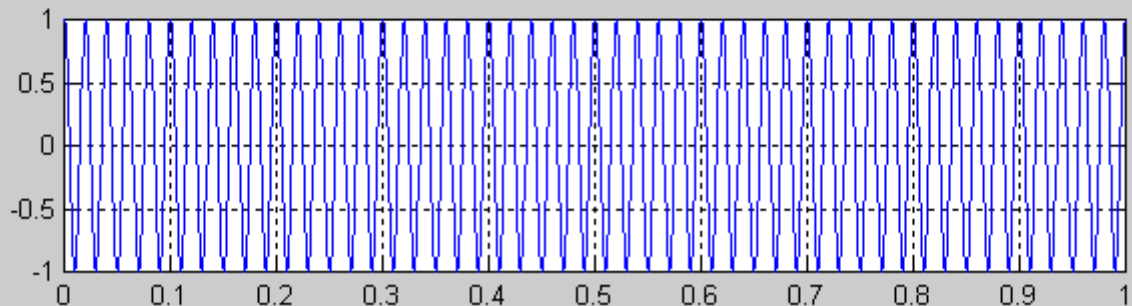
$$f_1(t) = \cos(2\pi \cdot 5 \cdot t)$$



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

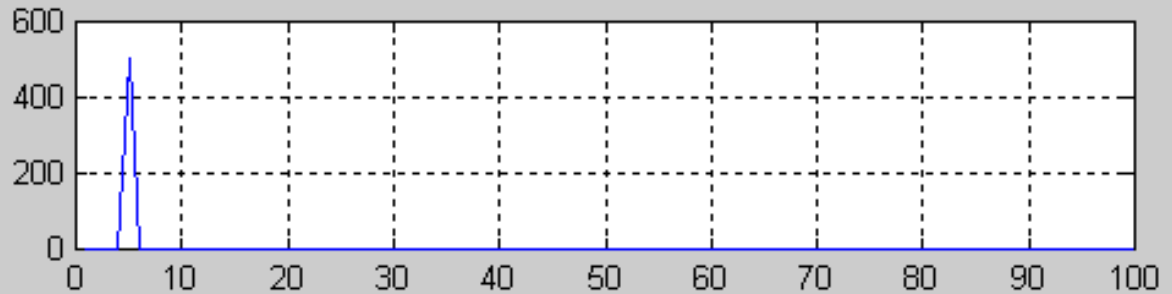


$$f_3(t) = \cos(2\pi \cdot 50 \cdot t)$$

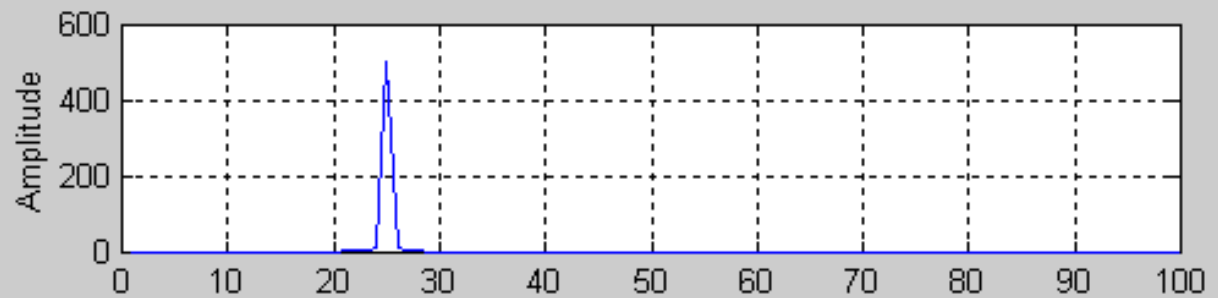


Examples (cont'd)

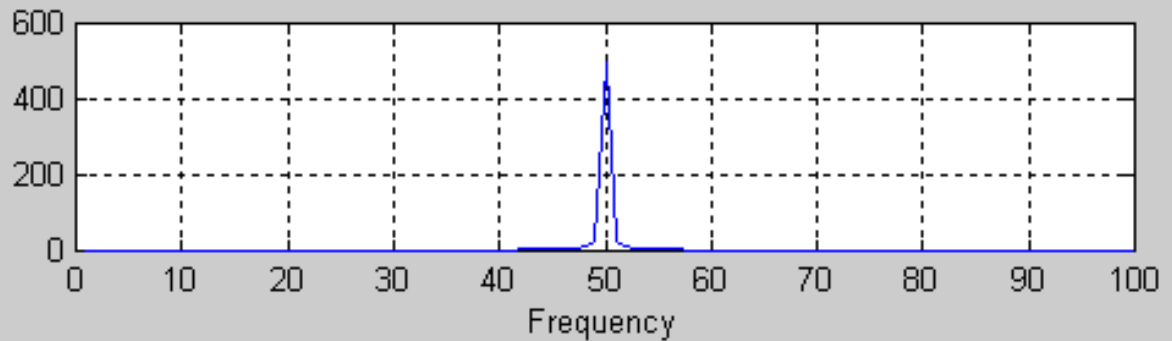
$F_1(u)$



$F_2(u)$



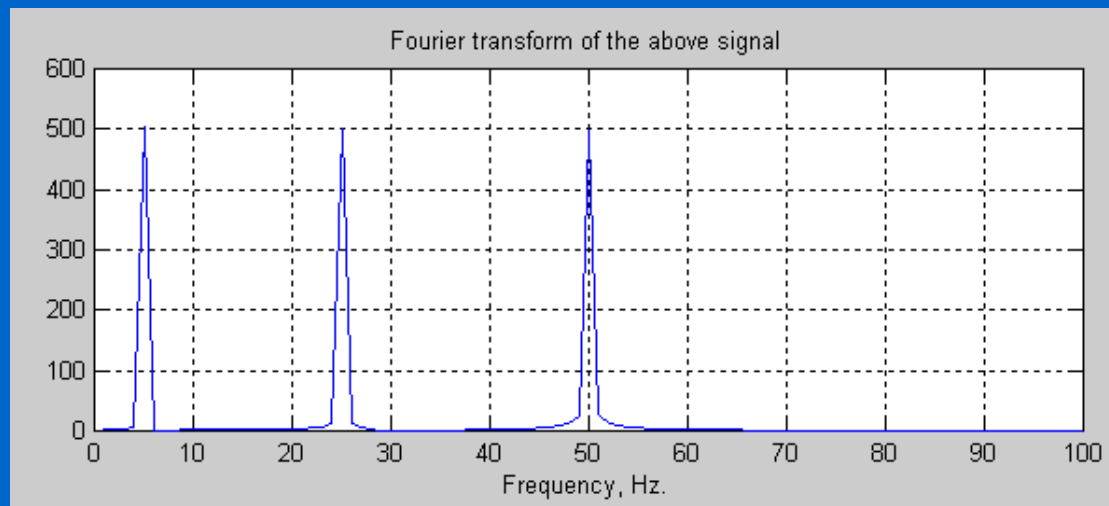
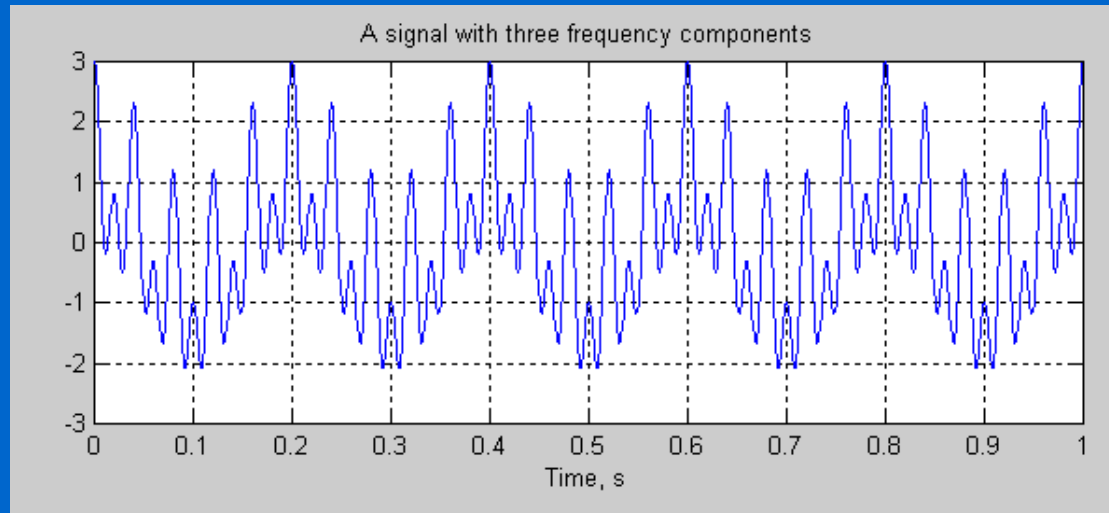
$F_3(u)$



Fourier Analysis – Examples (cont'd)

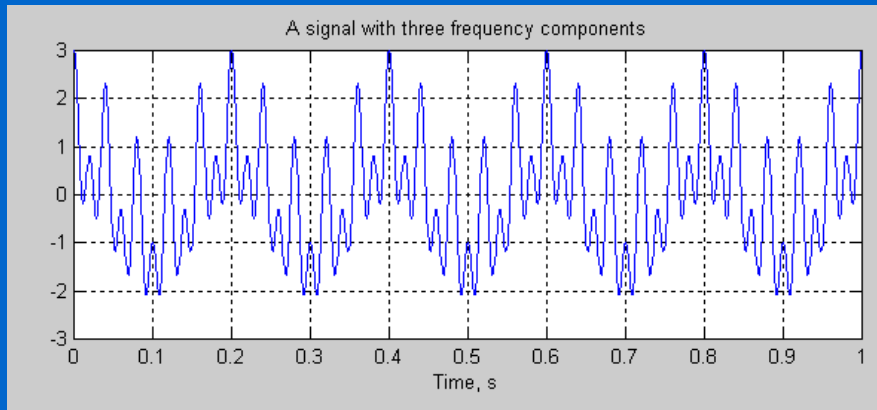
$$f_4(t) = \cos(2\pi \cdot 5 \cdot t) \\ + \cos(2\pi \cdot 25 \cdot t) \\ + \cos(2\pi \cdot 50 \cdot t)$$

$F_4(u)$?

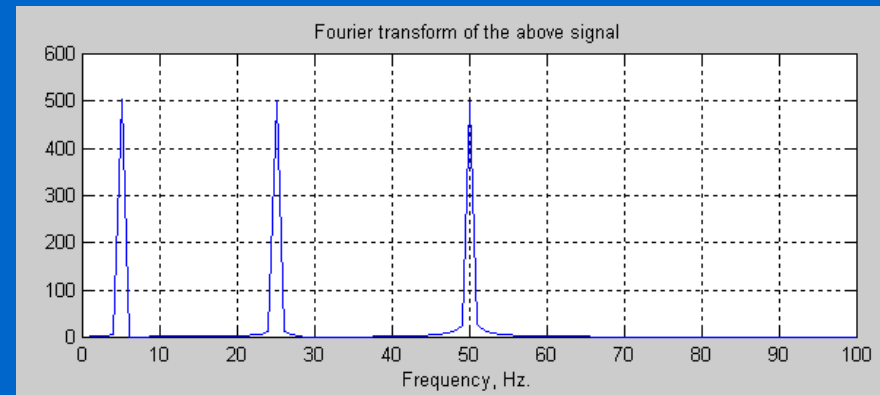


Limitations of Fourier Transform

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



Poor localization in freq domain!



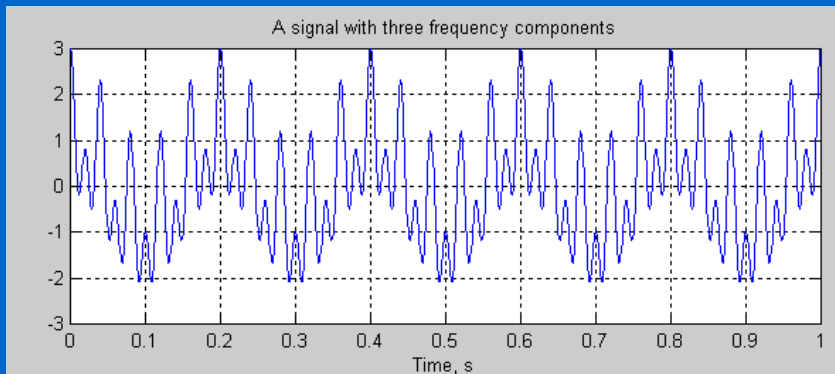
Great localization in freq domain!

Limitations of Fourier Transform (cont'd)

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

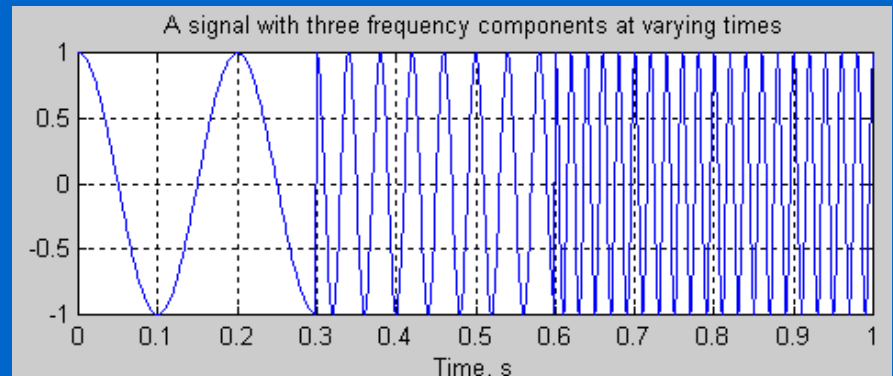
$f_4(t)$

Stationary signal
(non-varying frequency)



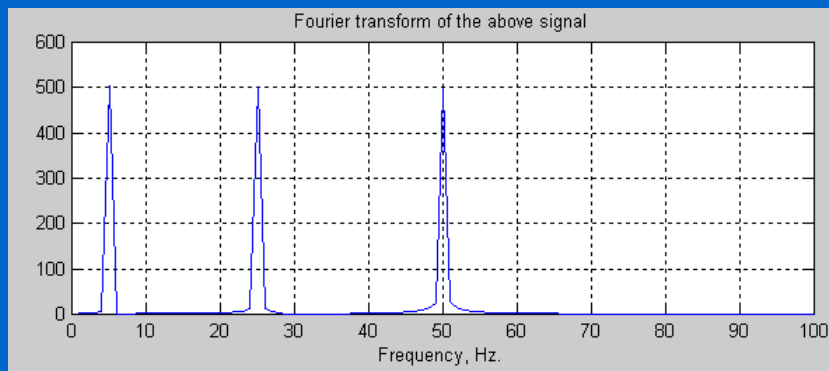
$f_5(t)$

Non-stationary signal
(varying frequency)

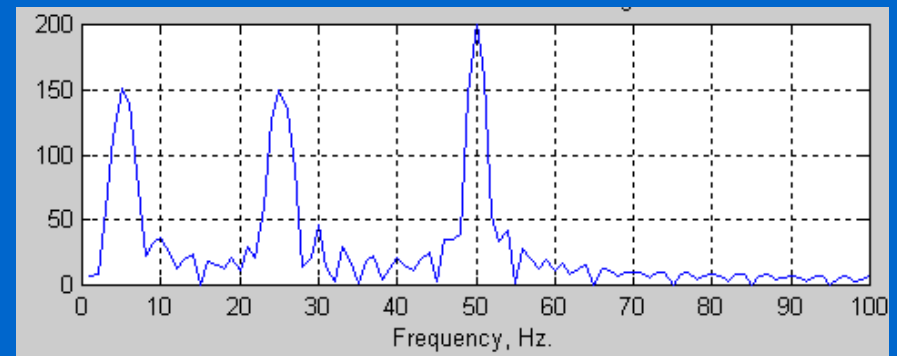


Limitations of Fourier Transform (cont'd)

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



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

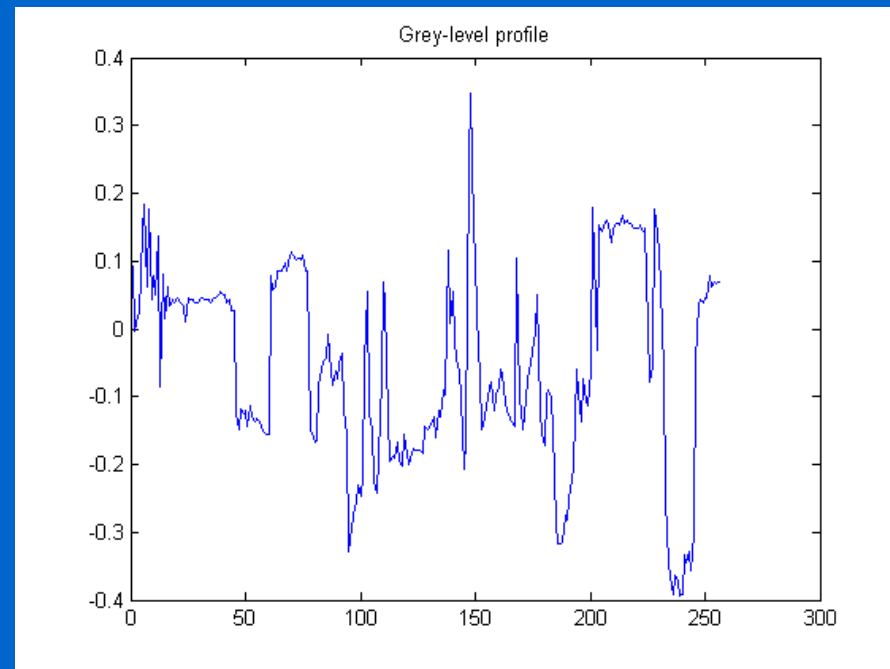
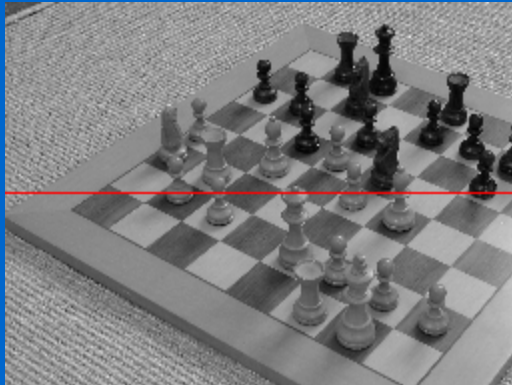


Perfect knowledge of **what** frequencies exist, but **no** information about **where** these frequencies are located in **time!**

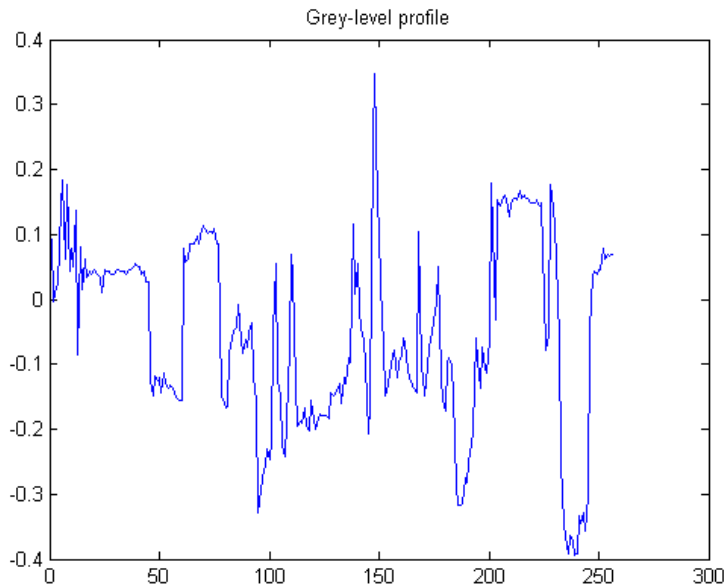
-
-
-

Limitations of Fourier Transform (cont'd)

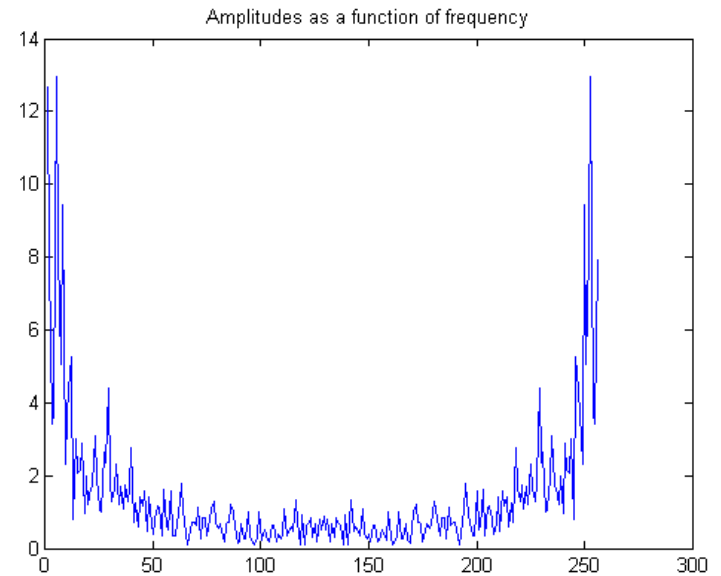
3. **Not efficient** for representing non-smooth functions.



Representing discontinuities or sharp corners (cont'd)



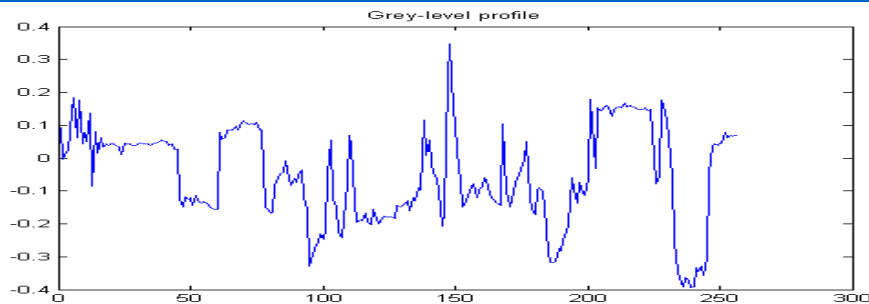
FT
→



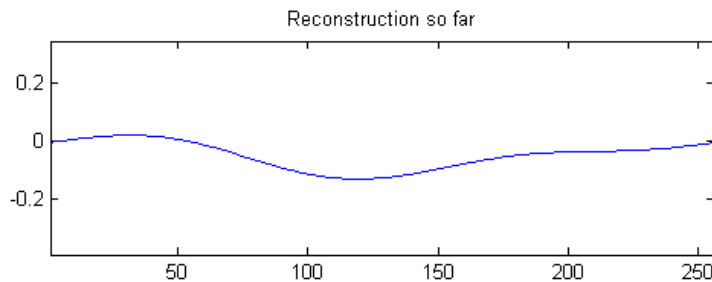
$$F(u) = \frac{1}{N} \sum_{x=0}^{N-1} f(x) e^{\frac{-j2\pi ux}{N}}, u = 0, 1, \dots, N-1$$

Representing discontinuities or sharp corners (cont'd)

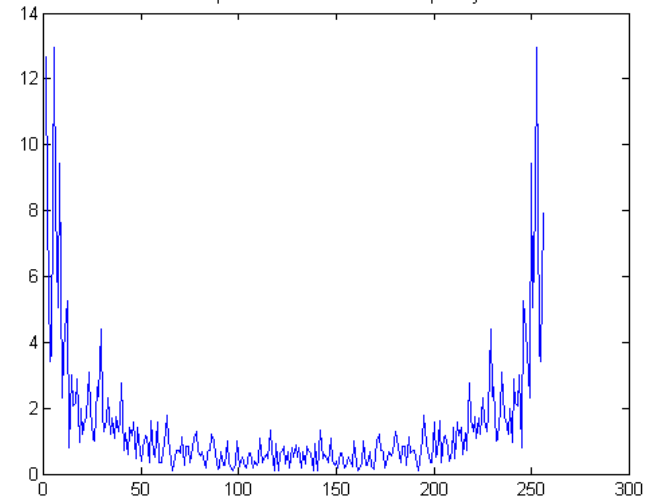
Original



Reconstructed



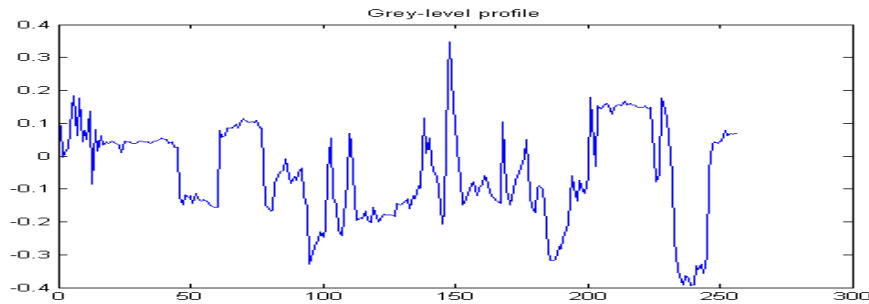
Amplitudes as a function of frequency



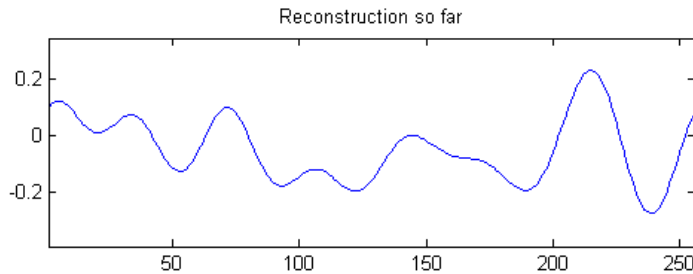
$$f(x) = \sum_{u=0}^1 F(u) e^{\frac{j2\pi ux}{N}}, x = 0, 1, \dots, N-1$$

Representing discontinuities or sharp corners (cont'd)

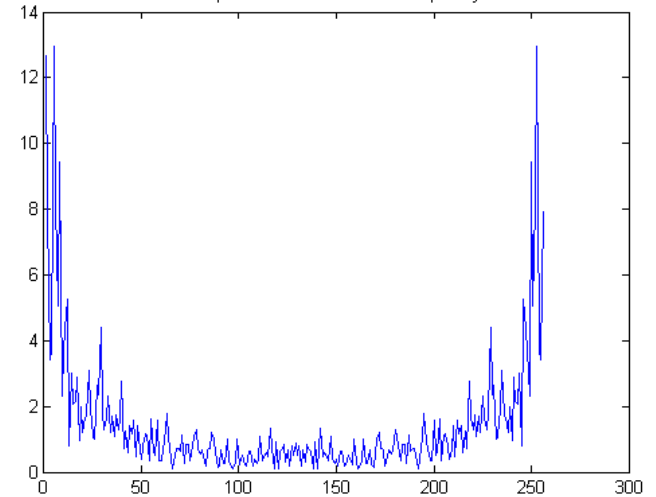
Original



Reconstructed



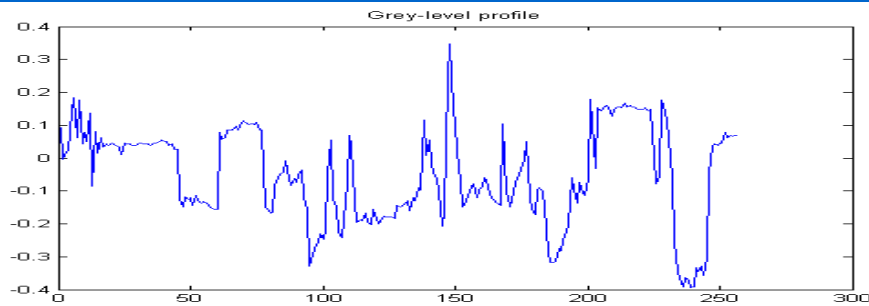
Amplitudes as a function of frequency



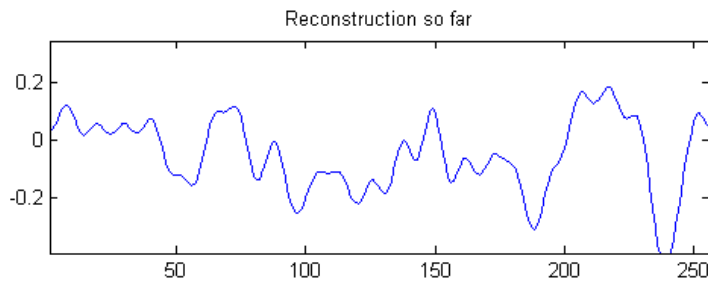
$$f(x) = \sum_{u=0}^7 F(u) e^{\frac{j2\pi ux}{N}}, x = 0, 1, \dots, N-1$$

Representing discontinuities or sharp corners (cont'd)

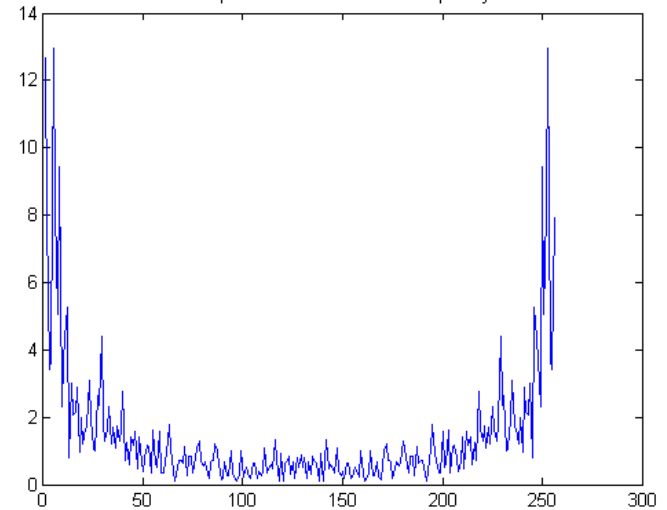
Original



Reconstructed



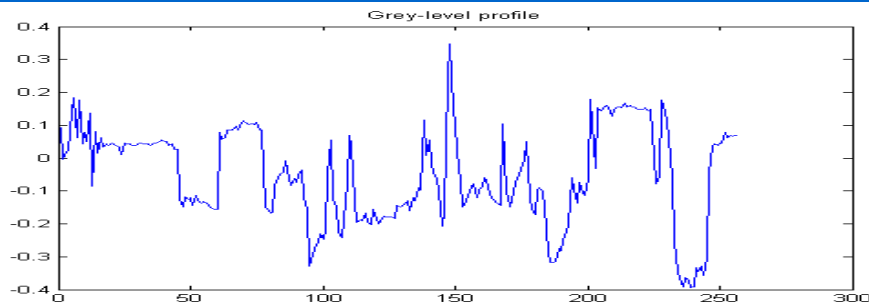
Amplitudes as a function of frequency



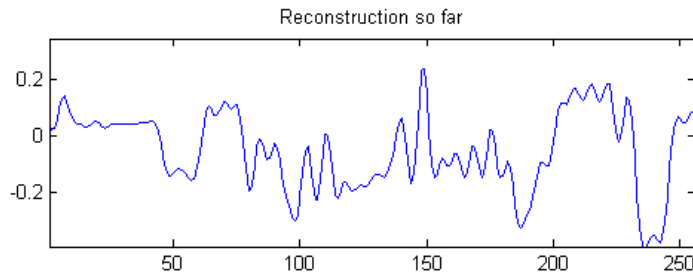
$$f(x) = \sum_{u=0}^{23} F(u) e^{\frac{j2\pi ux}{N}}, x = 0, 1, \dots, N-1$$

Representing discontinuities or sharp corners (cont'd)

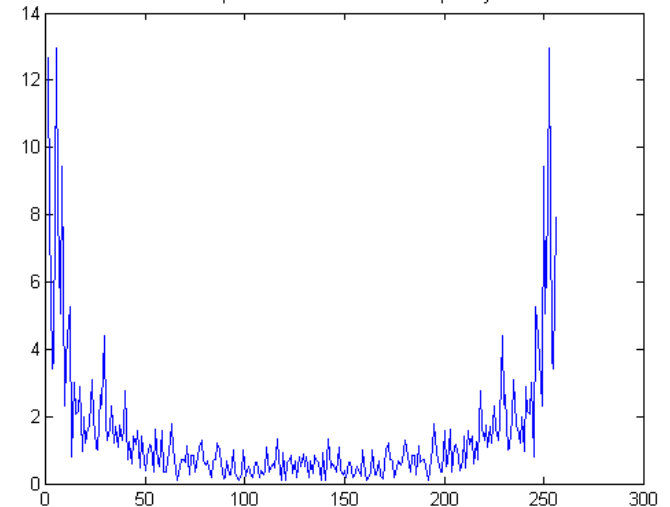
Original



Reconstructed



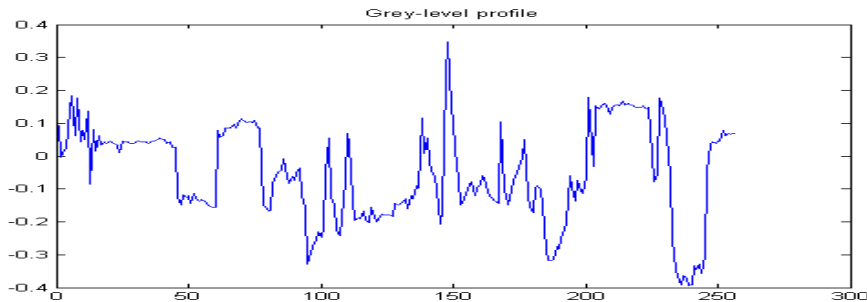
Amplitudes as a function of frequency



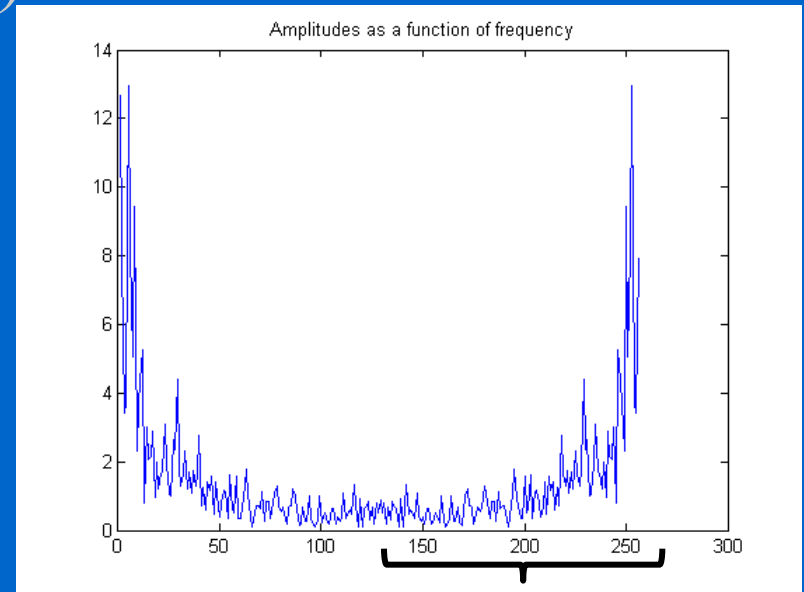
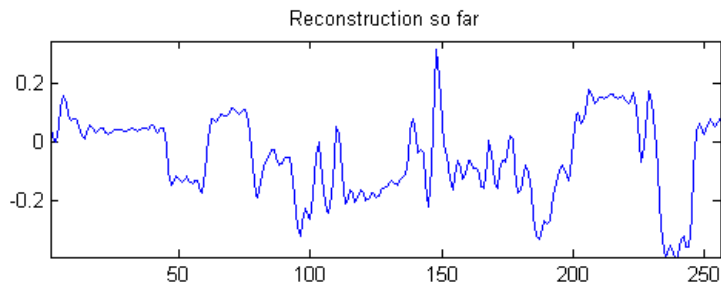
$$f(x) = \sum_{u=0}^{39} F(u) e^{\frac{j2\pi ux}{N}}, \quad x = 0, 1, \dots, N-1$$

Representing discontinuities or sharp corners (cont'd)

Original



Reconstructed

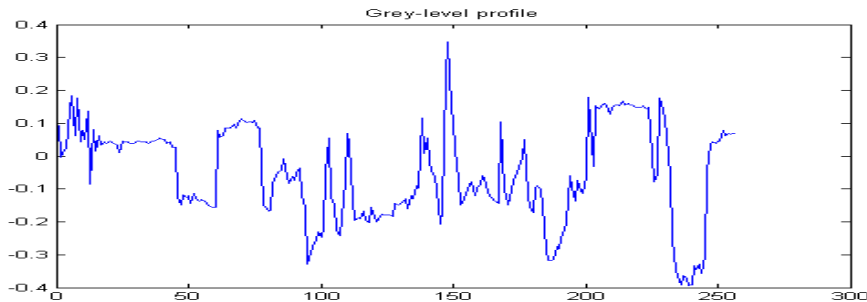


128 coefficients

$$f(x) = \sum_{u=0}^{63} F(u) e^{j \frac{2\pi ux}{N}}, \quad x = 0, 1, \dots, N-1$$

Representing discontinuities or sharp corners (cont'd)

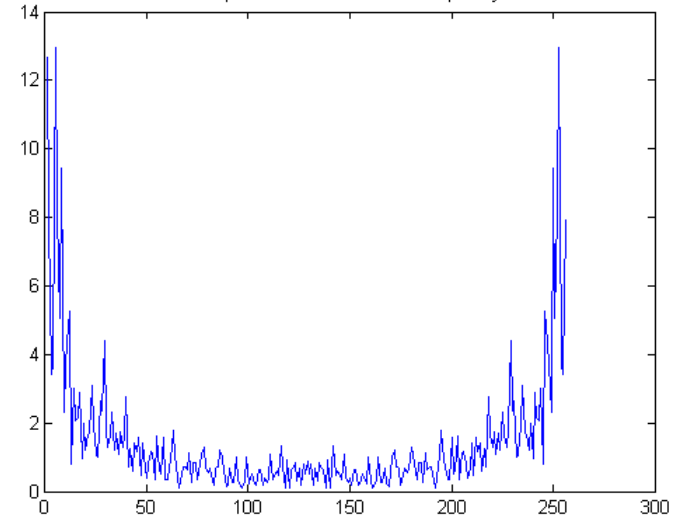
Original



Reconstructed



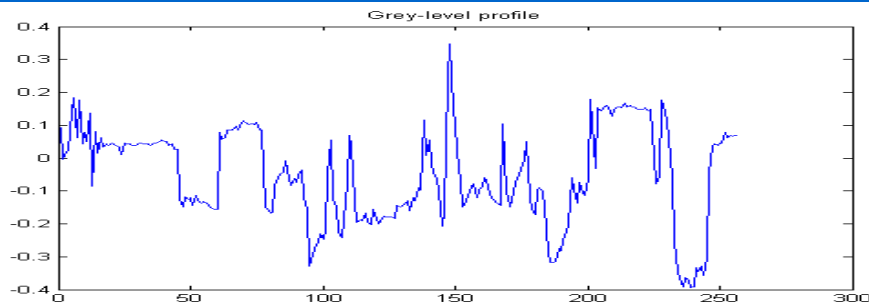
Amplitudes as a function of frequency



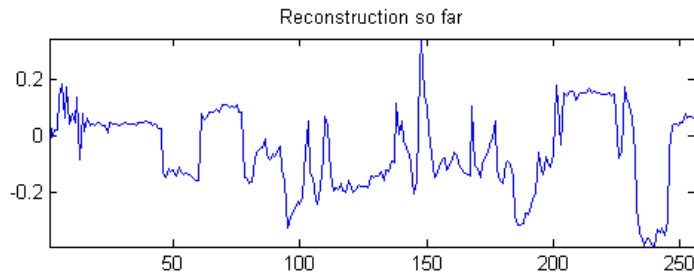
$$f(x) = \sum_{u=0}^{95} F(u) e^{\frac{j2\pi ux}{N}}, \quad x = 0, 1, \dots, N-1$$

Representing discontinuities or sharp corners (cont'd)

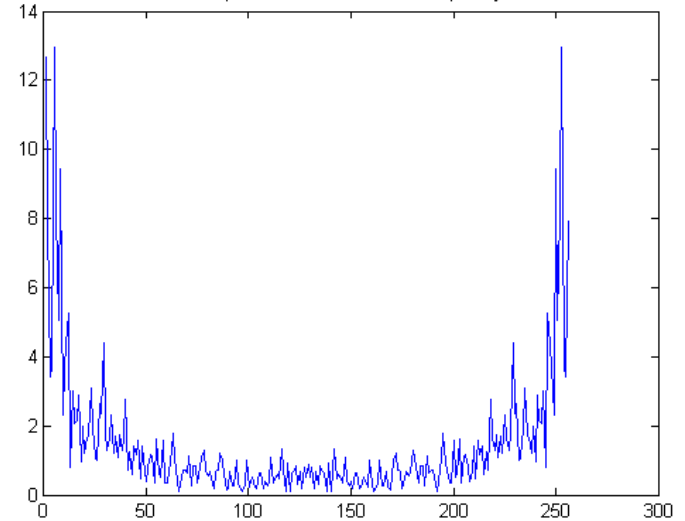
Original



Reconstructed



Amplitudes as a function of frequency

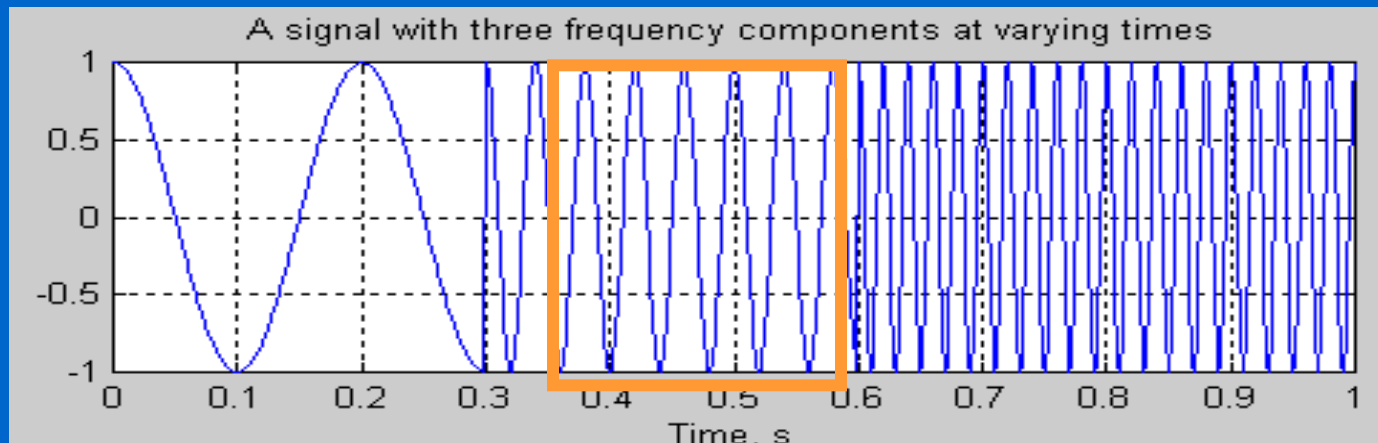


$$f(x) = \sum_{u=0}^{127} F(u) e^{\frac{j2\pi ux}{N}}, \quad x = 0, 1, \dots, N-1$$

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

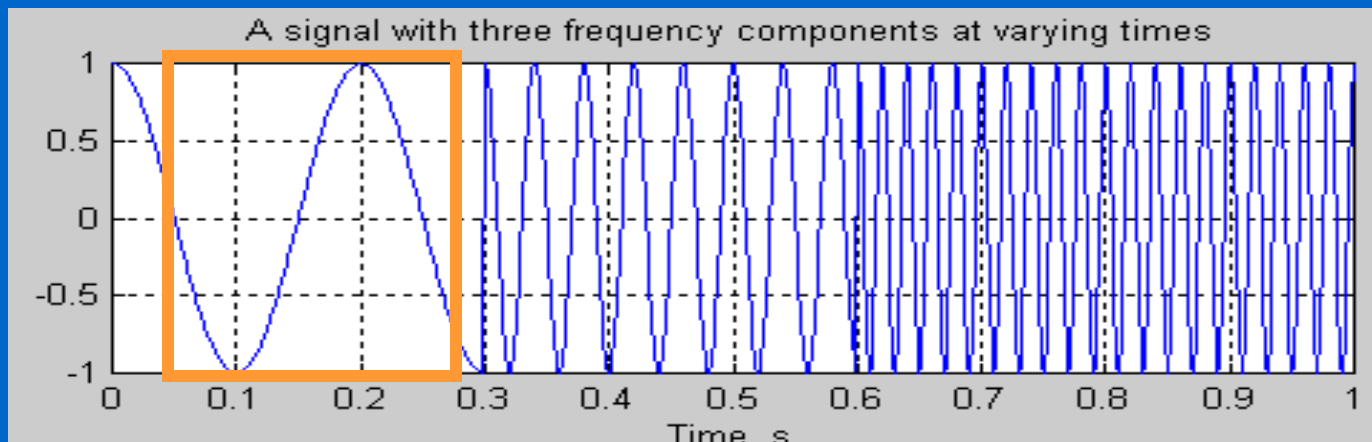
Short Time Fourier Transform (STFT)

- 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

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

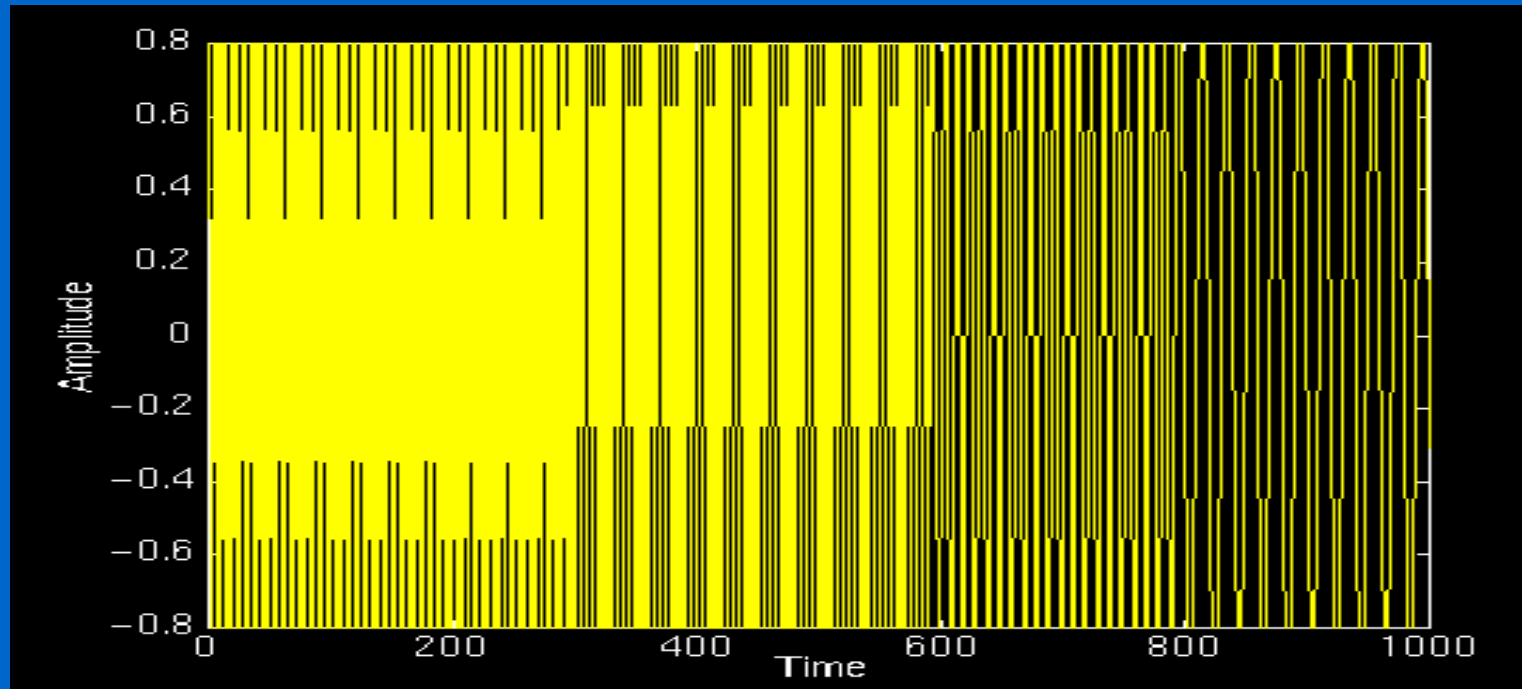
2D function

windowing function centered at $t=t'$

The diagram illustrates the Short-Time Fourier Transform (STFT) definition. The equation is $STFT_f^u(t', u) = \int_t [f(t) \cdot W(t - t')] \cdot e^{-j2\pi ut} dt$. Annotations include: 'time parameter' pointing to t' , 'frequency parameter' pointing to u , '2D function' pointing to the entire left-hand side, '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 shown as a subscript of the integral sign.

Example

$f(t)$



[0 – 300] ms \rightarrow 75 Hz sinusoid

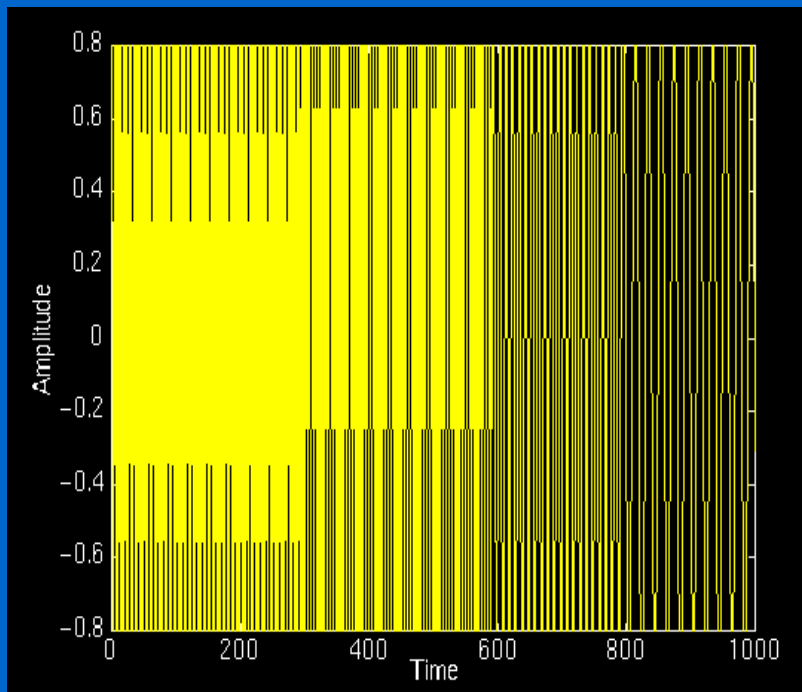
[300 – 600] ms \rightarrow 50 Hz sinusoid

[600 – 800] ms \rightarrow 25 Hz sinusoid

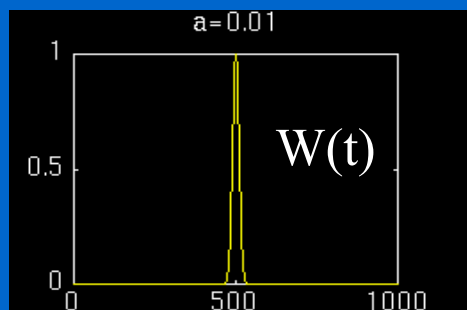
[800 – 1000] ms \rightarrow 10 Hz sinusoid

Example

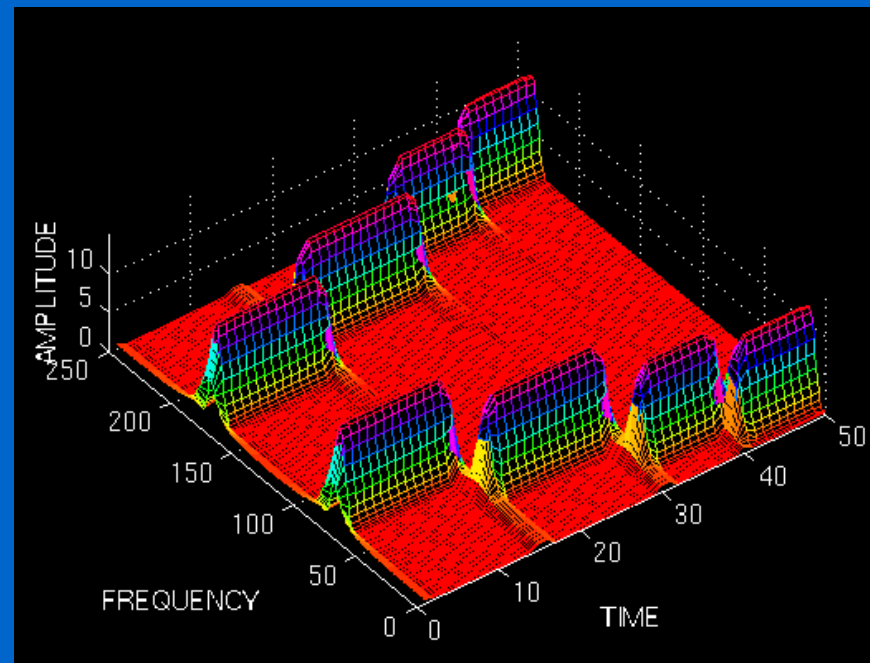
$f(t)$



$[0 - 300] \text{ ms} \rightarrow 75 \text{ Hz}$
 $[300 - 600] \text{ ms} \rightarrow 50 \text{ Hz}$
 $[600 - 800] \text{ ms} \rightarrow 25 \text{ Hz}$
 $[800 - 1000] \text{ ms} \rightarrow 10 \text{ Hz}$



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



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.

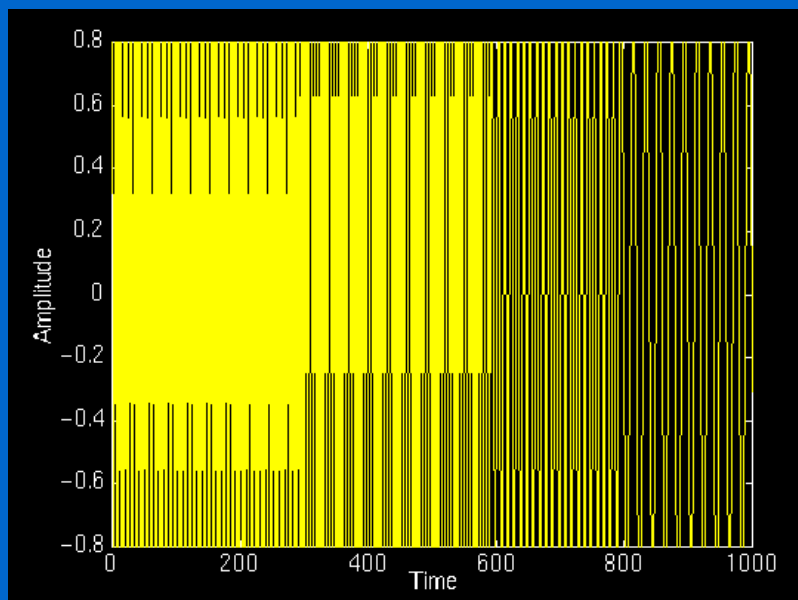
$$STFT_f^u(t', u) = \int_t [f(t) \cdot \delta(t - t')] \cdot e^{-j2\pi ut} dt = f(t') \cdot e^{-jut'}$$

STFT Window Size (cont'd)

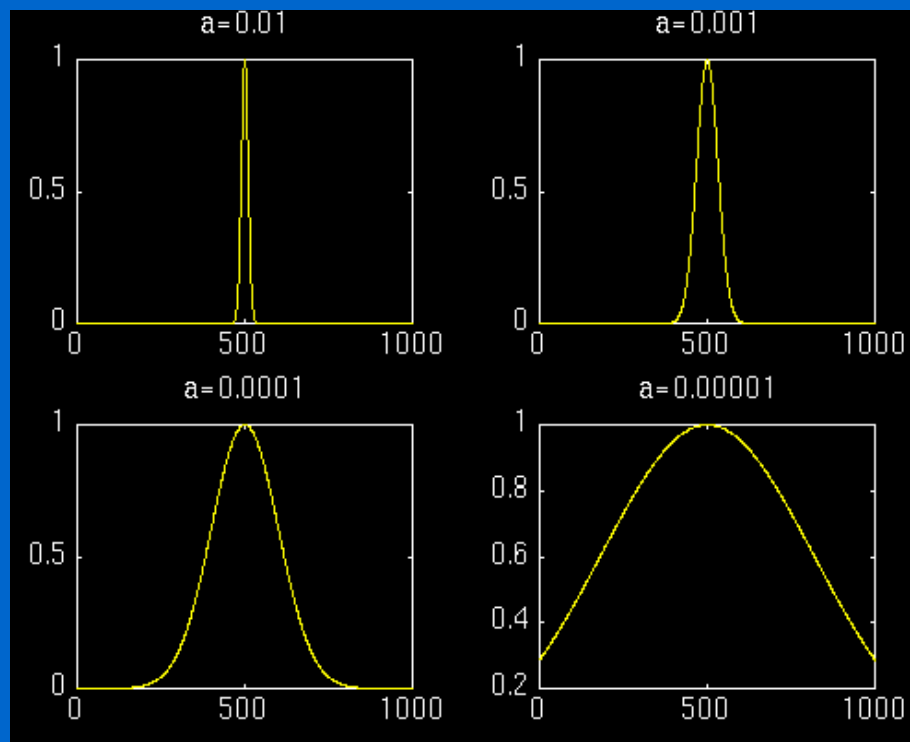
- **Wide window** → good frequency resolution, poor time resolution.
- **Narrow window** → good time resolution, poor frequency resolution.

Example

different size windows

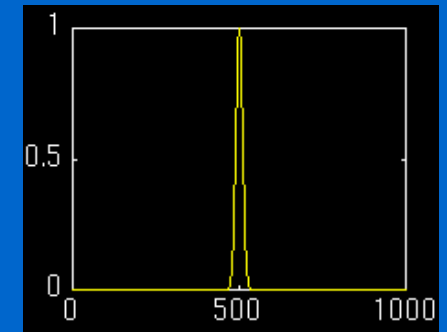
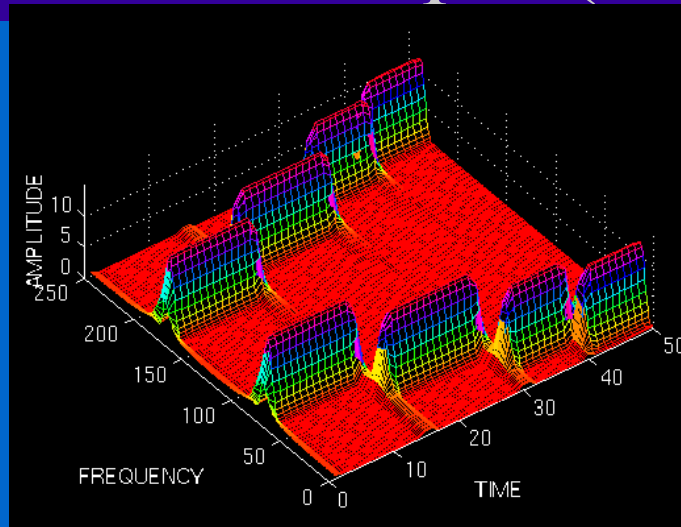


[0 – 300] ms → 75 Hz
[300 – 600] ms → 50 Hz
[600 – 800] ms → 25 Hz
[800 – 1000] ms → 10 Hz

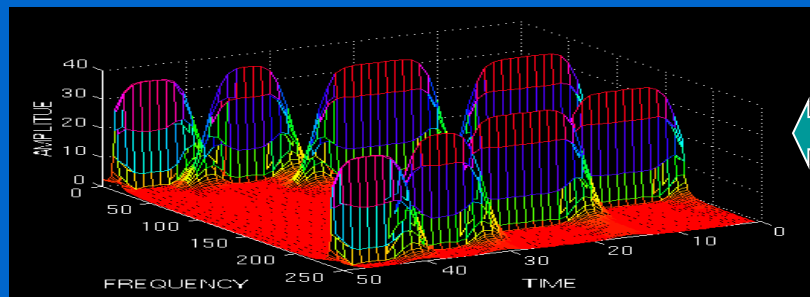


Example (cont'd)

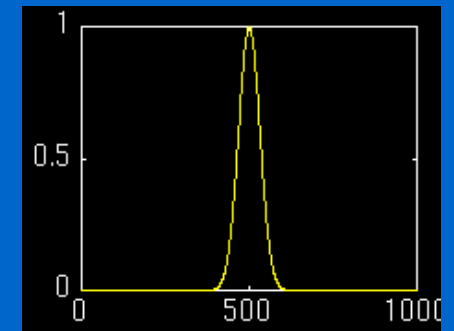
$$STFT_f^u(t', u)$$



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

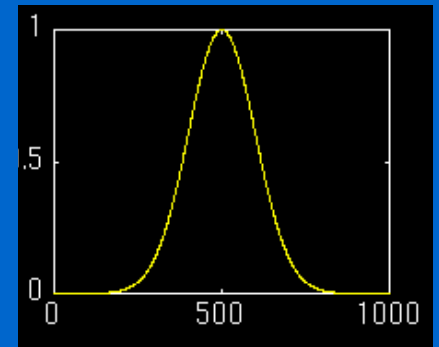
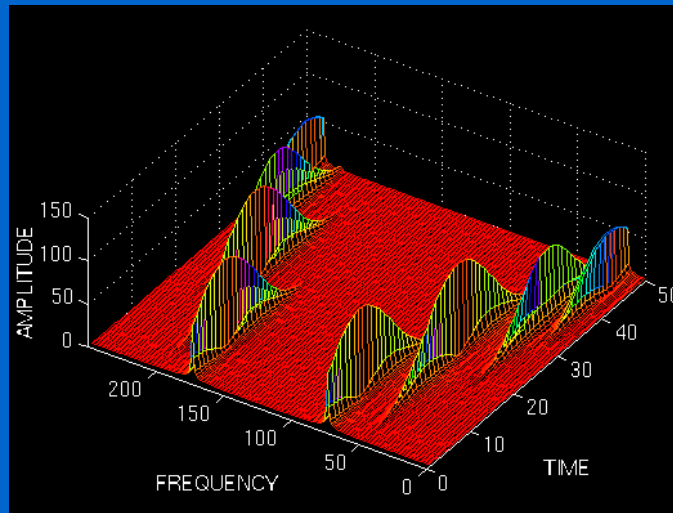


scaled: $t/20$

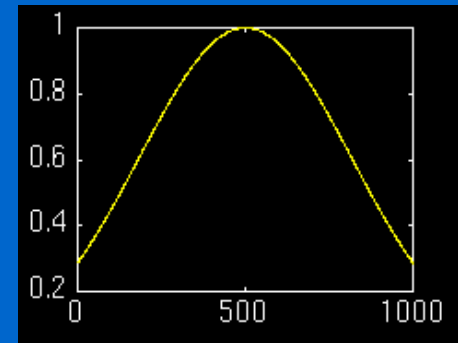
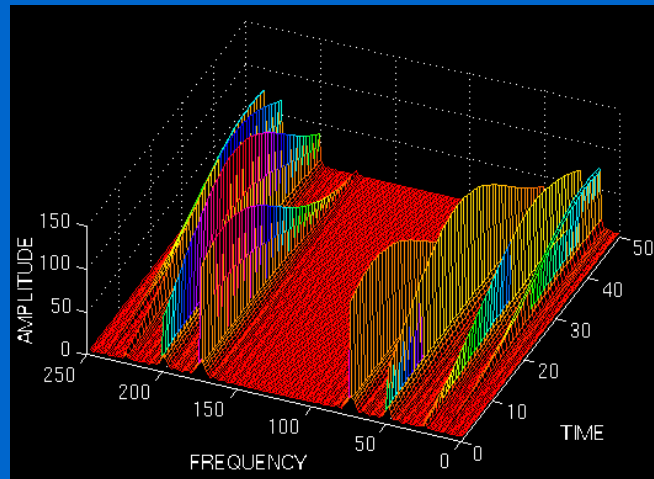


Example (cont'd)

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




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



scaled: $t/20$

Heisenberg (or Uncertainty) Principle

$$\Delta t \cdot \Delta f \geq \frac{1}{4\pi}$$
Two white arrows originate from the text below. One arrow points from 'Time resolution' to the Δt term in the equation. The other arrow points from 'Frequency resolution' to the Δf term in the equation.

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

Δt and Δf 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*.