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Short Time Fourier Transform (STFT)



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Fourier Transform

- Fourier Transform reveals which frequency components are present in a function:

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

(inverse DFT)

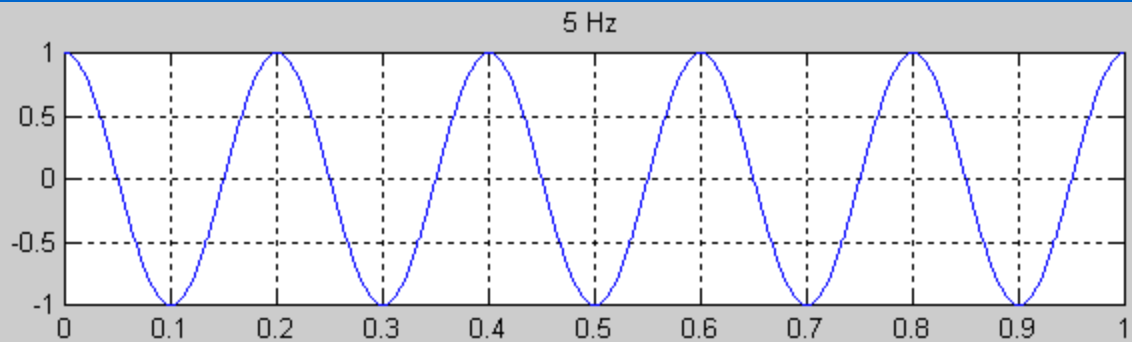
where:

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

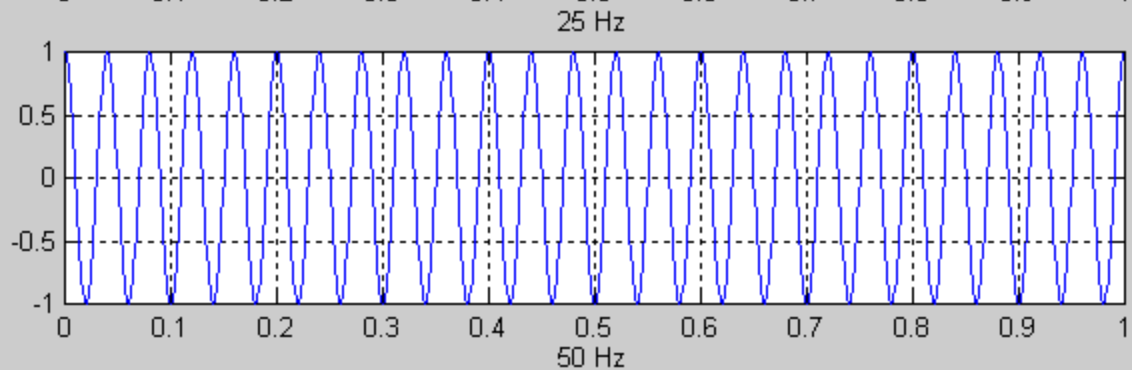
(forward DFT)

Examples

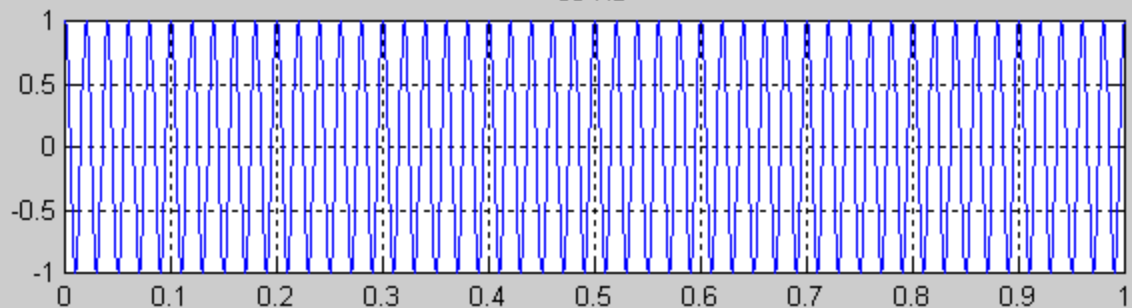
$$f_1(t) = \cos(2\pi \cdot 5 \cdot t)$$



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

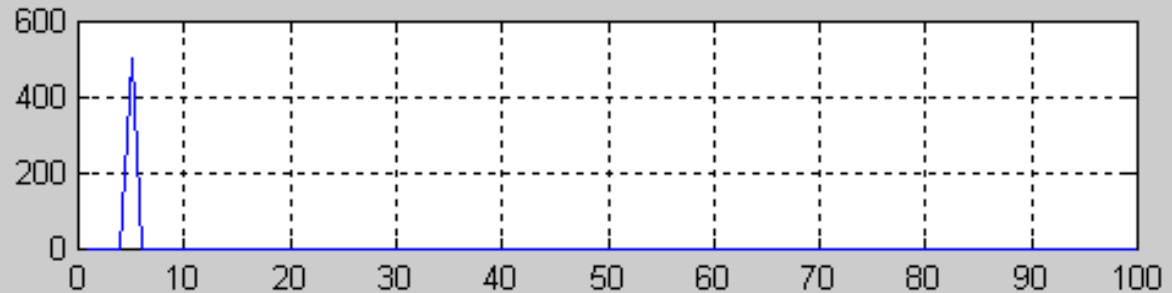


$$f_3(t) = \cos(2\pi \cdot 50 \cdot t)$$

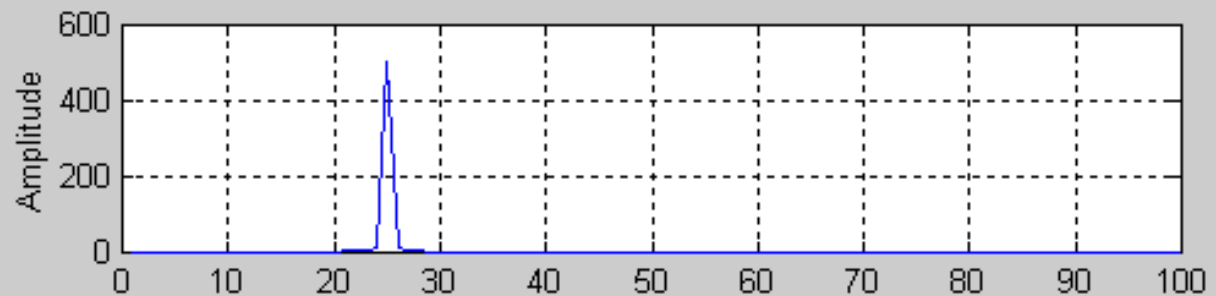


The plots below indicate that the continuous signals have been sampled at $F_s = \dots\dots$ Hz (or possibly larger)

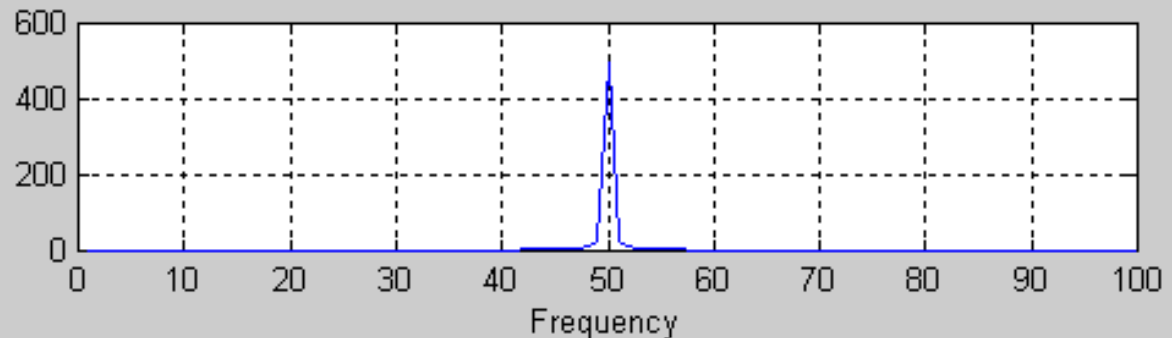
$F_1(u)$



$F_2(u)$



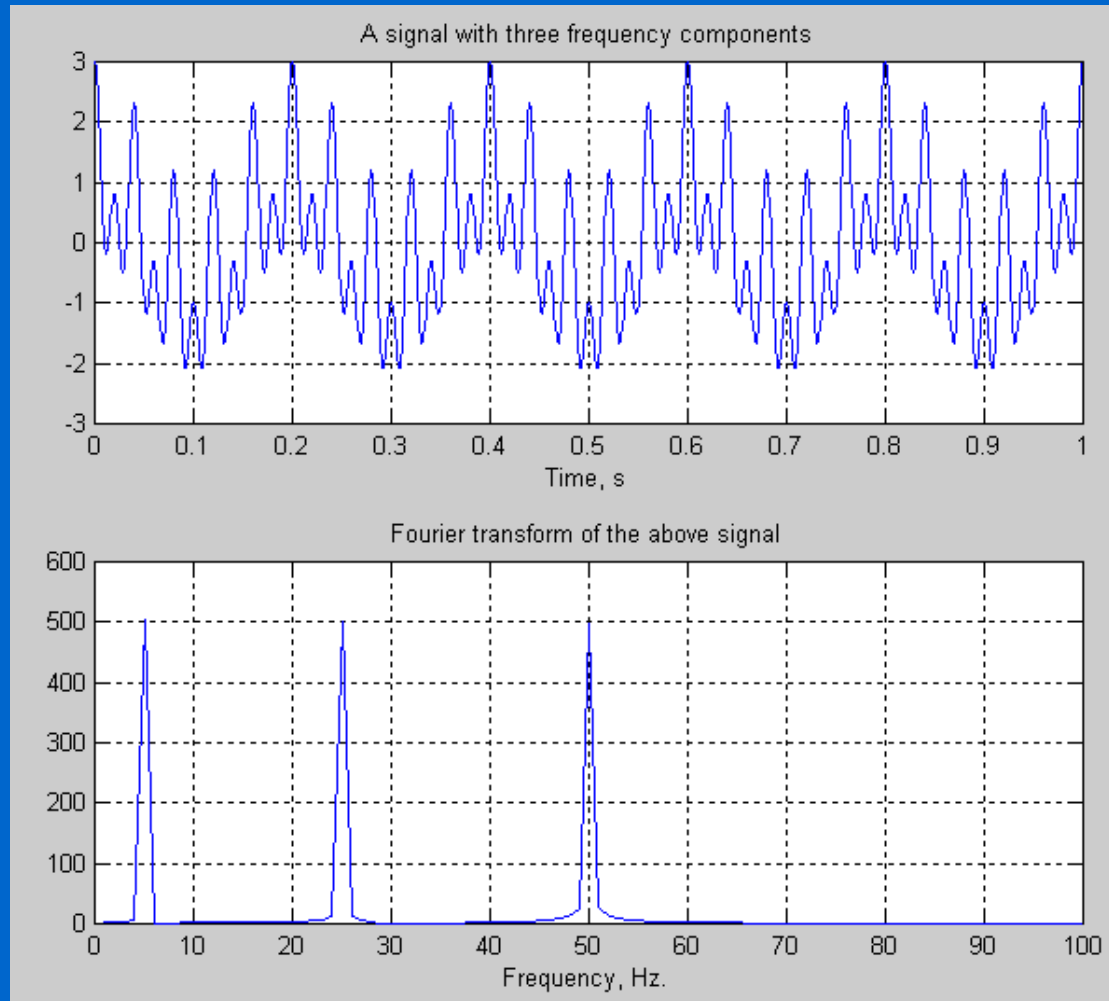
$F_3(u)$



Fourier Analysis – Examples (cont'd)

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

$F_4(u)$



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Limitations of Fourier Transform

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

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Limitations of Fourier Transform (cont'd)

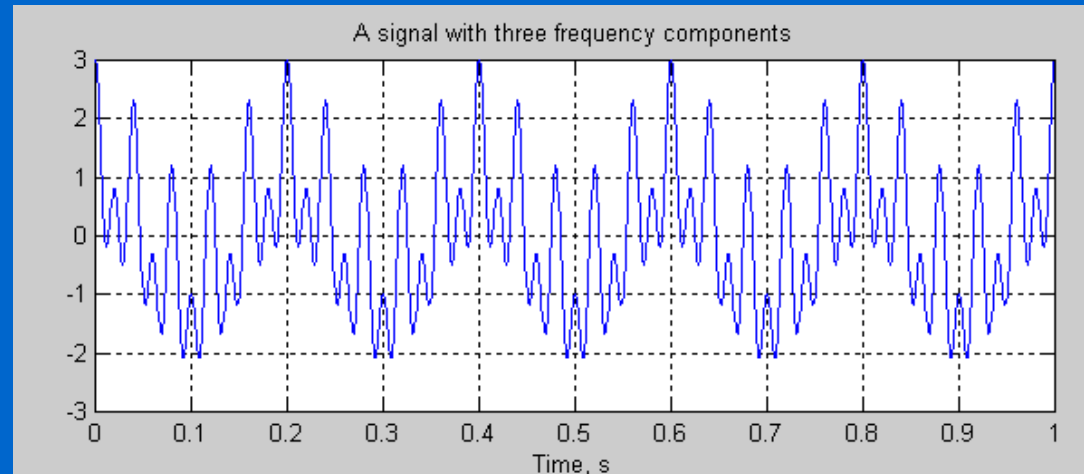
1. Cannot provide **simultaneous** time and frequency localization.

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

Stationary vs non-stationary signals

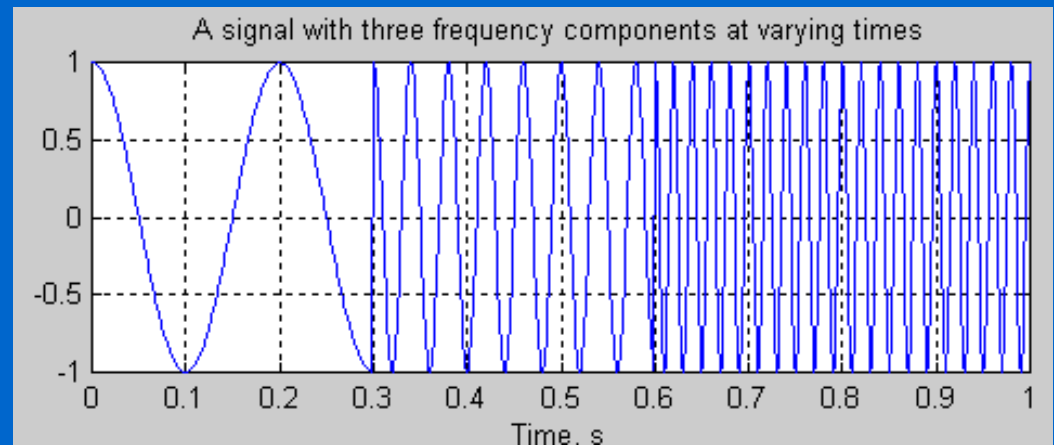
- Stationary signals:
time-invariant spectra

$$f_4(t)$$



- Non-stationary signals: time-varying spectra

$$f_5(t)$$



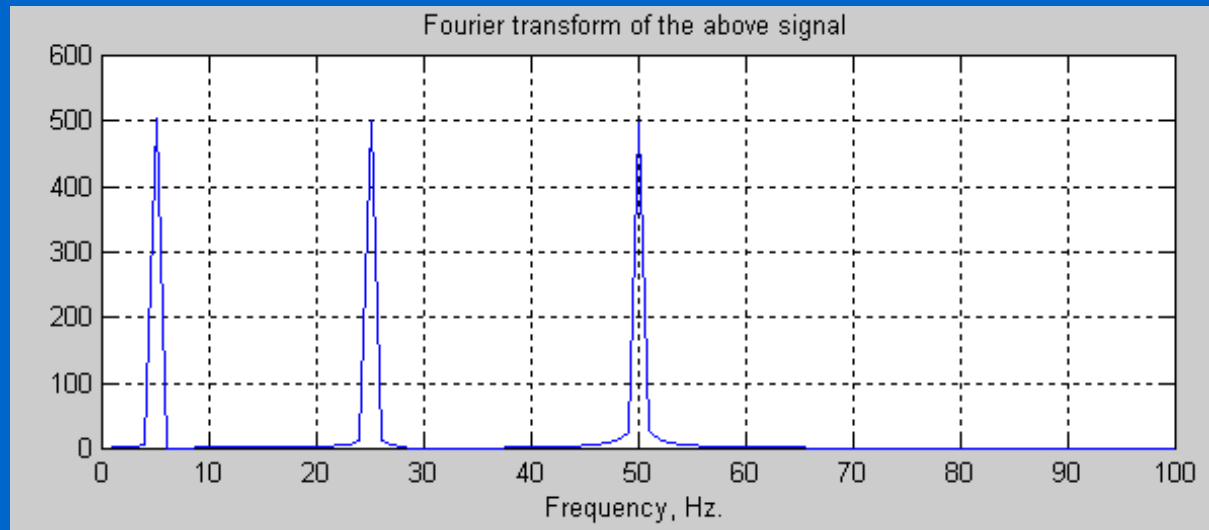
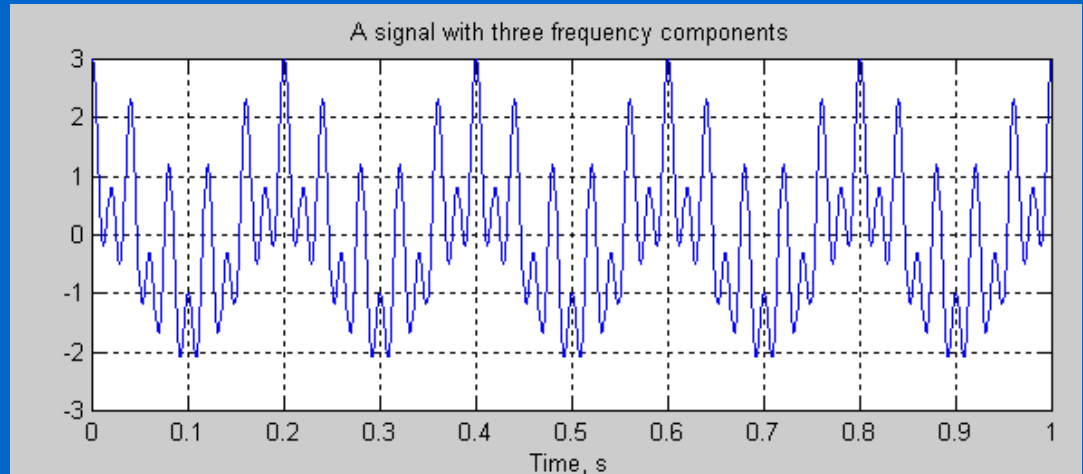
Stationary vs non-stationary signals (cont'd)

Stationary signal:

Three frequency components, present at all times!

$$f_4(t)$$

$$F_4(u)$$

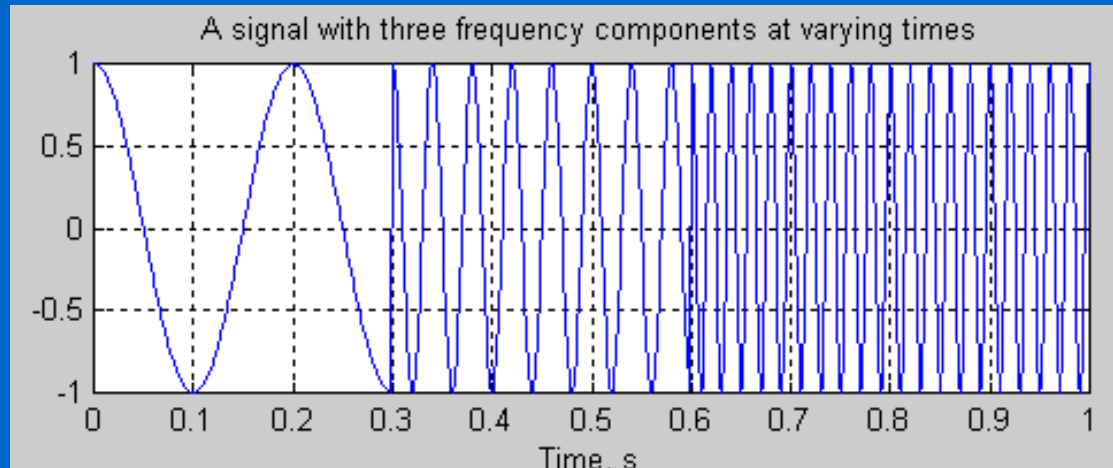


Stationary vs non-stationary signals (cont'd)

Non-stationary signal:

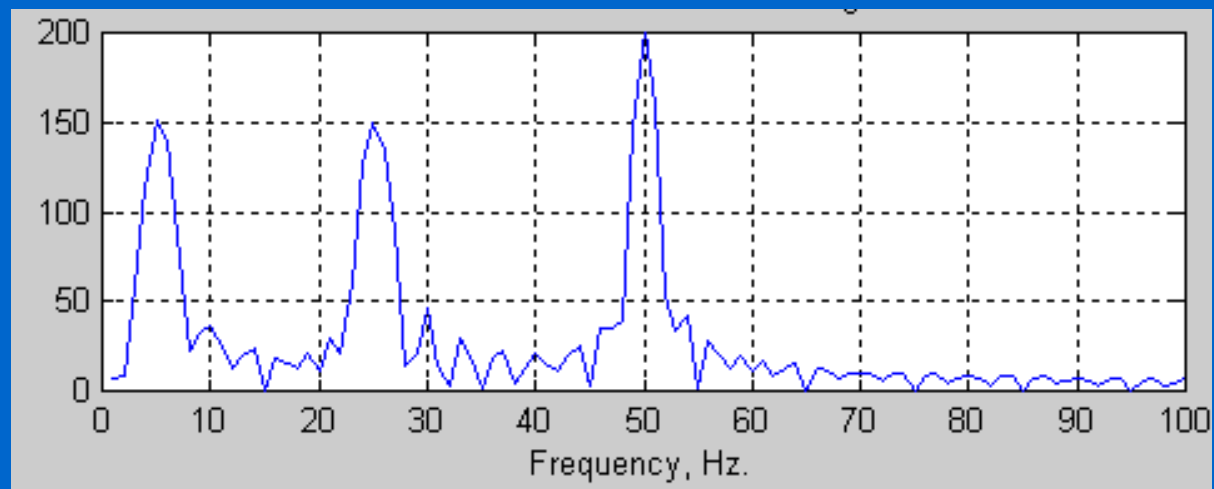
Three frequency components, NOT present at all times!

$f_5(t)$



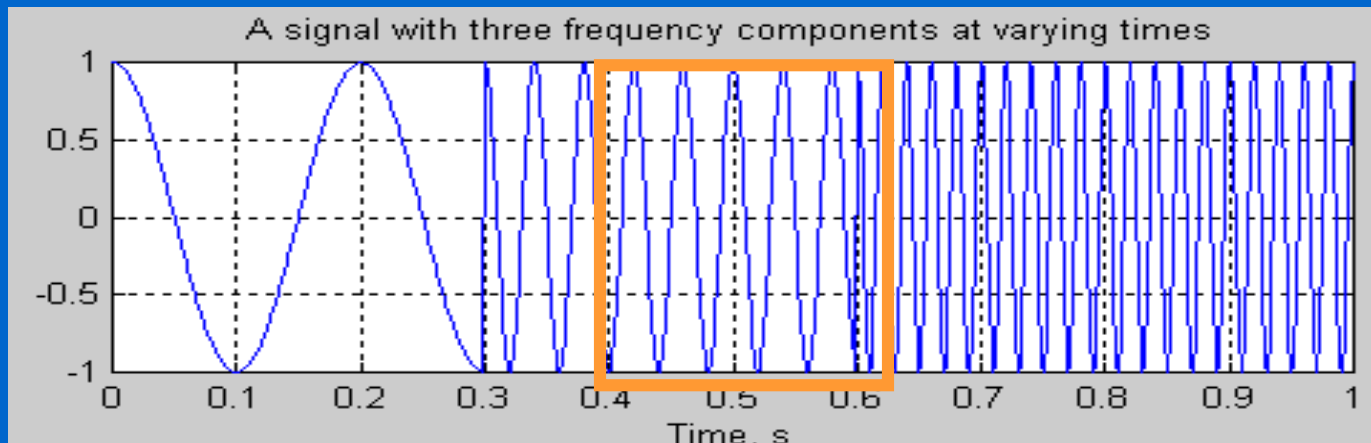
(note: *more freq. components are present due to signal transitions*)

$F_5(u)$



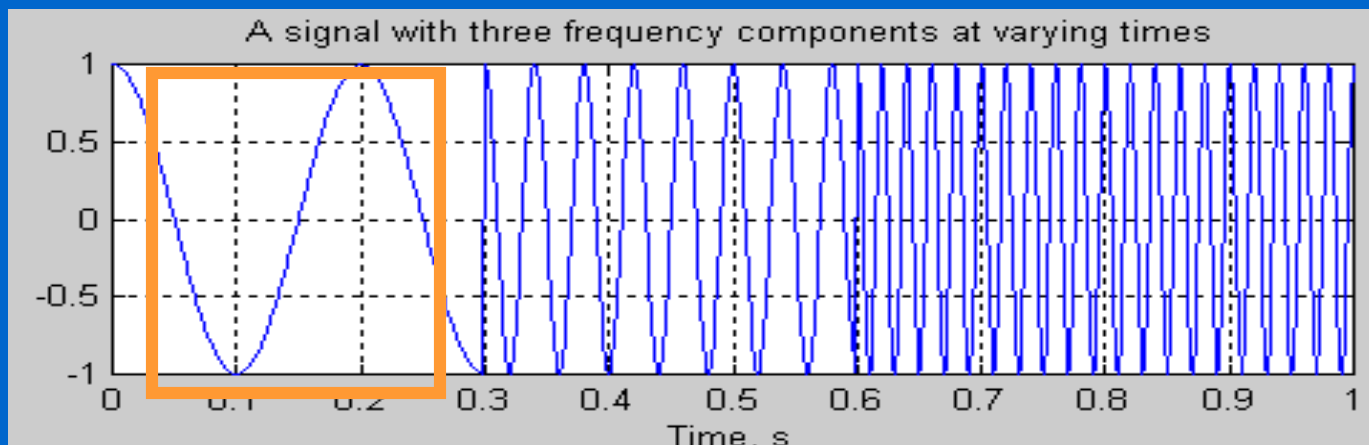
Short Time Fourier Transform (STFT)

- Segment the 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 function 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 in the continuous case

2D function:

Time
parameter

Frequency
parameter

Signal to
be analyzed

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

STFT of $f(t)$:
computed for each
window centered at $t=t'$

Windowing
function

Centered at $t=t'$

Choosing Window $W(t)$

- What shape should it have?
 - Rectangular, Gaussian, ...
- How wide should it be?
 - Window should be **narrow** enough to ensure that the portion of the signal falling within the window is stationary.
 - But ... very narrow windows 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: $W(t) = 1 \rightarrow$ 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.
- **Wavelets** (next year): use multiple window sizes.

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(1927: Position vs. velocity of an object)

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

Δt and Δf cannot be made arbitrarily small!

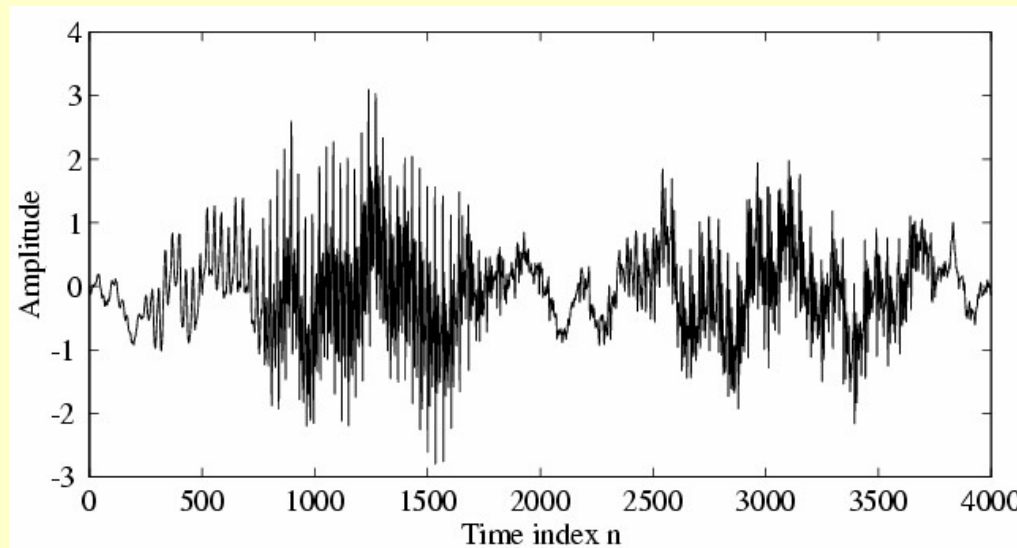
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Heisenberg (or Uncertainty) Principle

- 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*.

STFT Computation Using MATLAB

- The M-file `specgram` can be used to compute the STFT of a signal
- The application of `specgram` is illustrated next

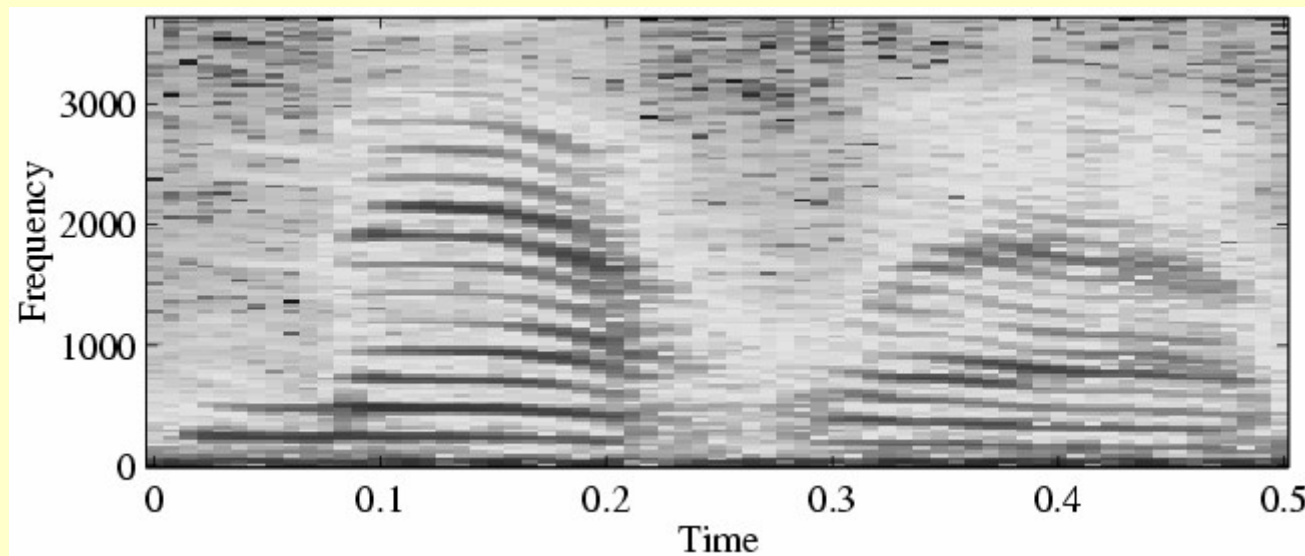


A speech signal of duration 4001 samples

STFT Computation Using MATLAB

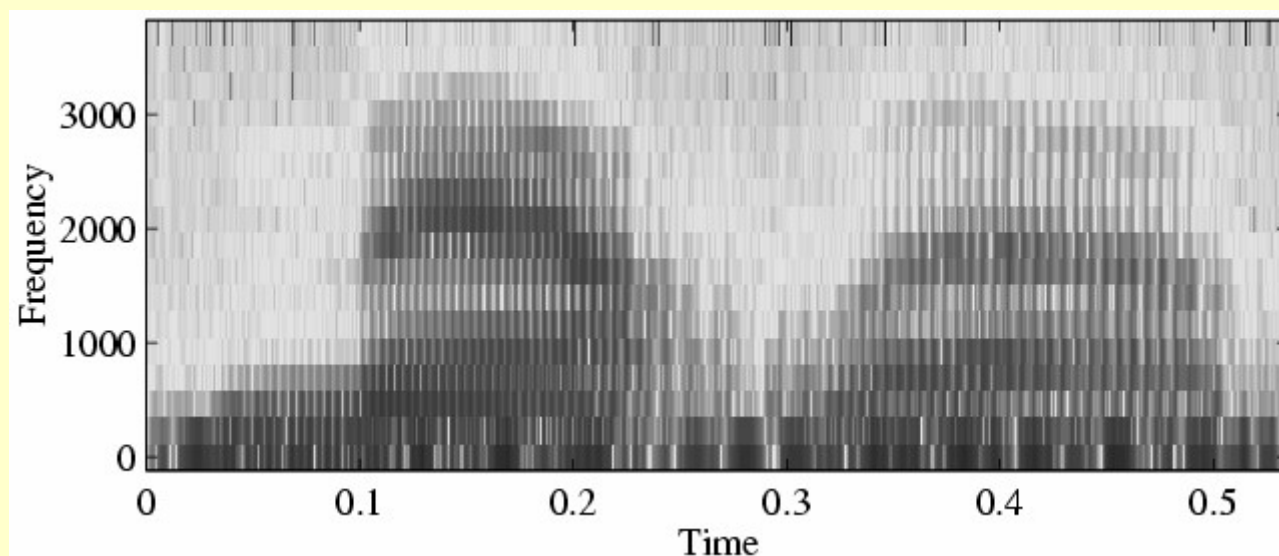
sampling
freq. ?

- Using Program 11_4 we compute the narrowband spectrogram of this speech signal



STFT Computation Using MATLAB

- The wideband spectrogram of the speech signal is shown below



- The frequency and time resolution tradeoff between the two spectrograms can be seen