



# 신호처리 (FFT and STFT)

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# Signal Representation by Harmonic Sinusoids



- Question
  - How to decompose it to harmonic sinusoids
  - How to find the coefficients of harmonic sinusoids
  - How to find the frequency components

→ Fourier Transform by FFT

# Fast Fourier Transform (FFT)

# FFT with Sampling Frequency

- You can think this signal as a vibration signal from a rotating machinery with 60 Hz

$$x[n] = 2 \cos(2\pi 60t) \implies f = 60$$

```
import numpy as np
import matplotlib.pyplot as plt
%matplotlib inline

from scipy.fftpack import fft, fftshift
from scipy.signal import spectrogram
```

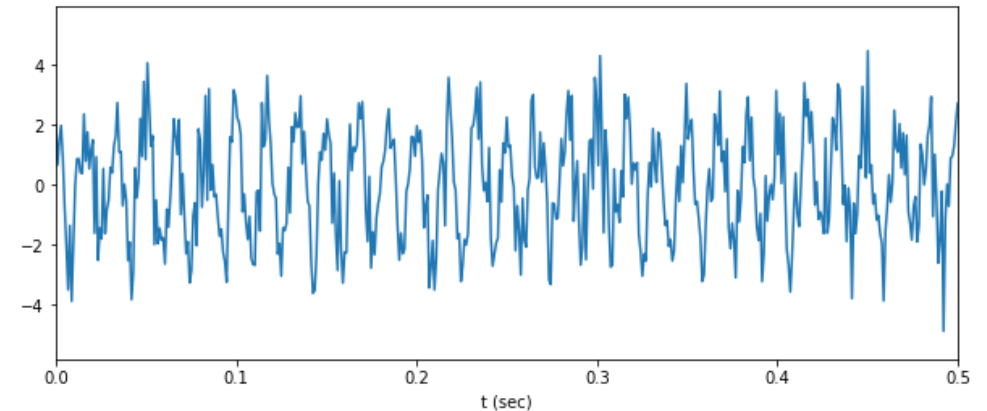
```
Fs = 2**10           # Sampling frequency
T = 1/Fs             # Sampling period (or sampling interval)

N = 5000            # Total data points (signal length)

t = np.arange(0, N)*T # Time vector (time range)

k = np.arange(0, N)  # vector from 0 to N-1
f = (Fs/N)*k         # frequency range
```

```
x = 2*np.cos(2*np.pi*60*t) + np.random.randn(N)
```

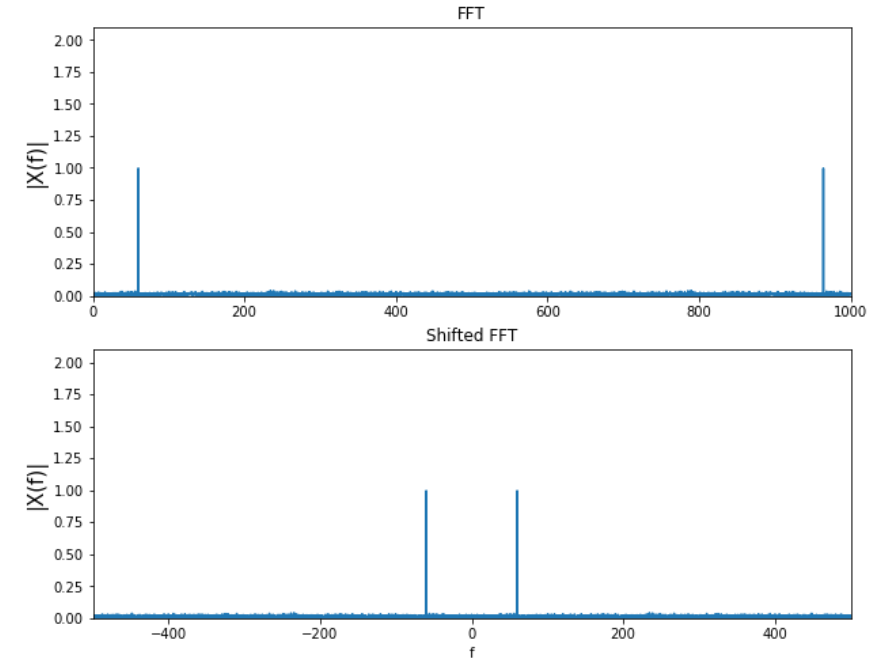
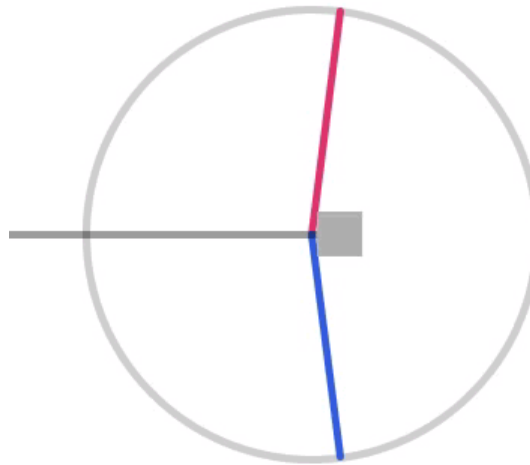


# Implementing FFT Routine

$$x[n] = 2 \cos(2\pi 60t) \implies f = 60$$

```
# original fft  
  
xt = fft(x)/N  
xtshift = fftshift(xt)  
  
kr = np.hstack([np.arange(0, N/2), np.arange(-N/2, 0)])  
fr = (Fs/N)*kr  
fs = fftshift(fr)
```

$$\cos \omega t = \frac{e^{i\omega t} + e^{-i\omega t}}{2}$$



# Single-sided FFT (or Positive FFT)

$$x[n] = 2 \cos(2\pi 60t) \implies f = 60$$

```
# original fft

xt = fft(x)/N
xtshift = fftshift(xt)

kr = np.hstack([np.arange(0, N/2), np.arange(-N/2, 0)])
fr = (Fs/N)*kr
fs = fftshift(fr)
```

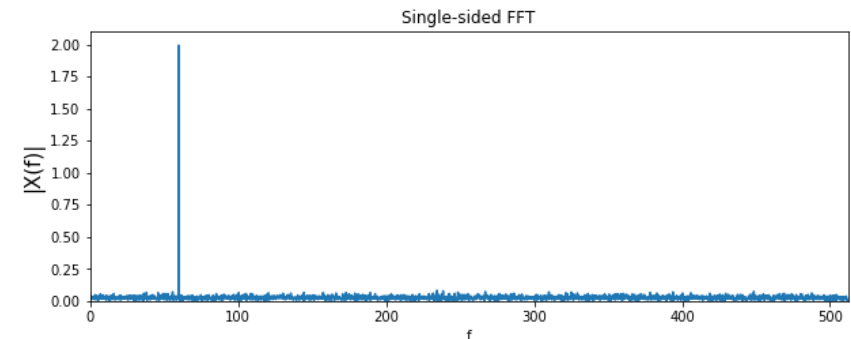
- Only want the second half of the FFT, since the last is redundant
- 2xamplitude except the DC component

```
# single-sides fft

xt = fft(x)/N
xtss = xt[0:int(N/2)+1]      # 0:N/2
xtss[1:-1] = 2*xtss[1:-1]

fss = f[0:int(N/2)+1]
```

- Only want the second half of the FFT, since the last is redundant
- 2xamplitude except the DC component



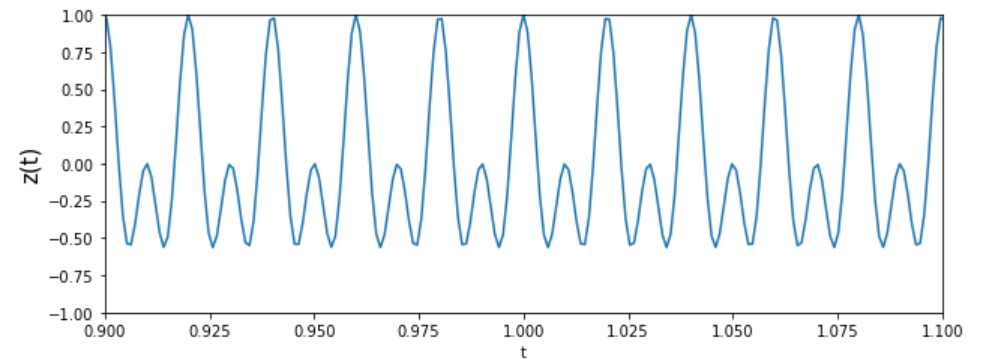
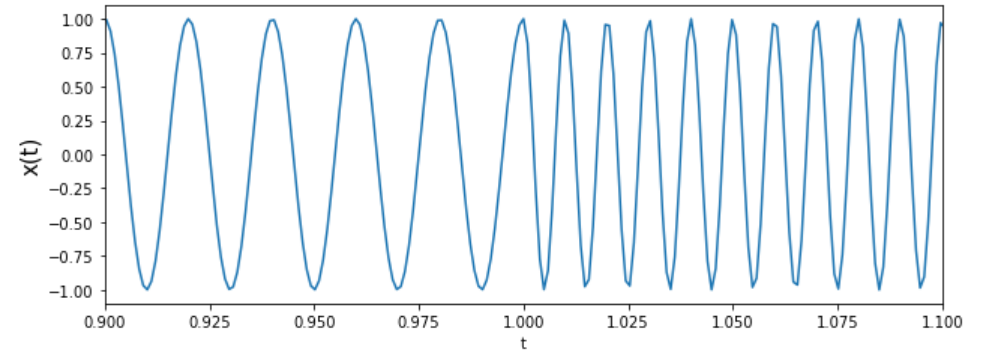
# STFT (Short-Time Fourier Transformation)

# Issue on FFT

```
x1 = np.cos(2*np.pi*50*t)
x2 = np.cos(2*np.pi*100*t)

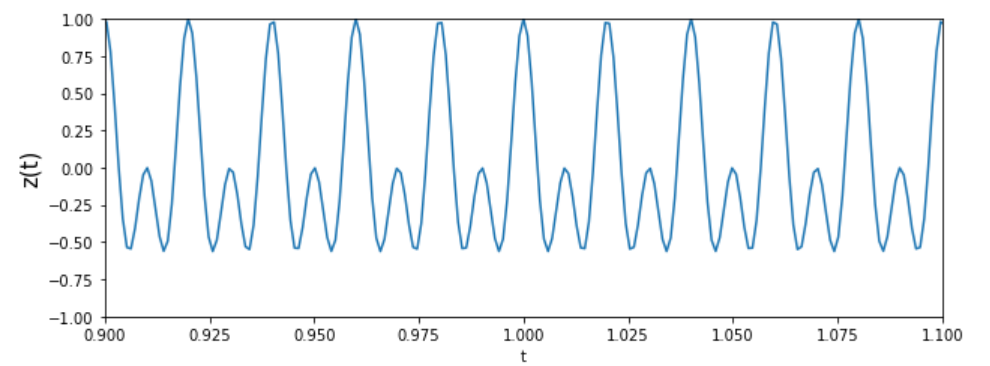
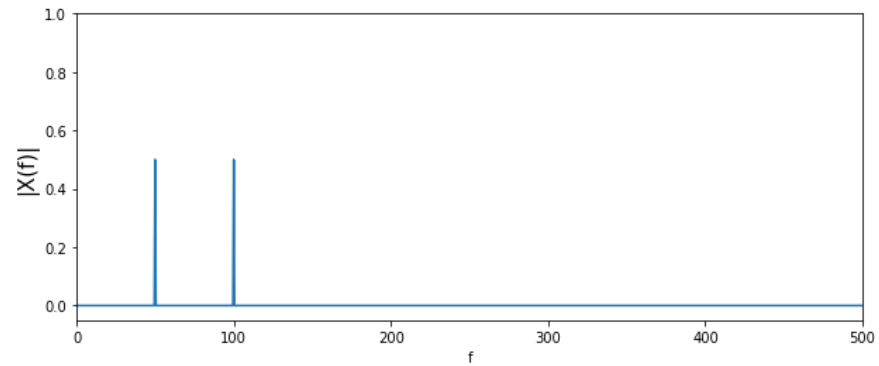
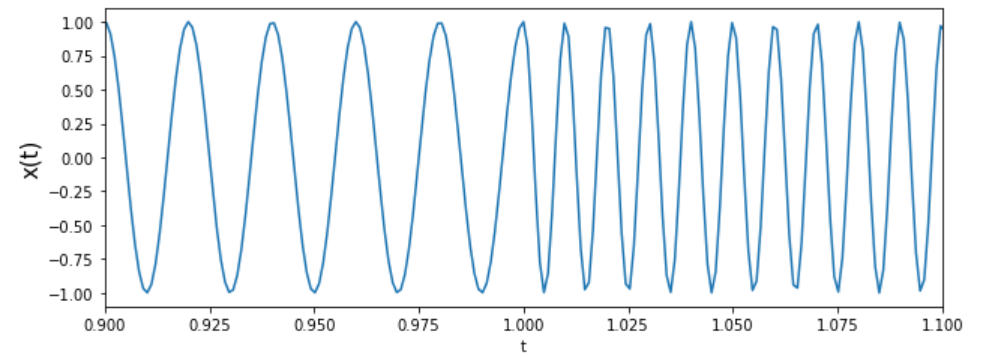
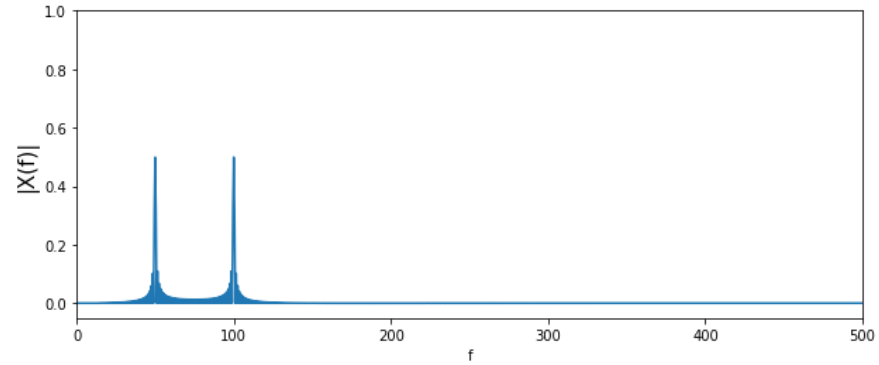
x = np.zeros(t.shape)
x[0:int(N/2)] = x1[0:int(N/2)]
x[int(N/2):-1] = x2[int(N/2):-1]
```

```
z = 1/2*(x1 + x2)
```





# Issue on FFT

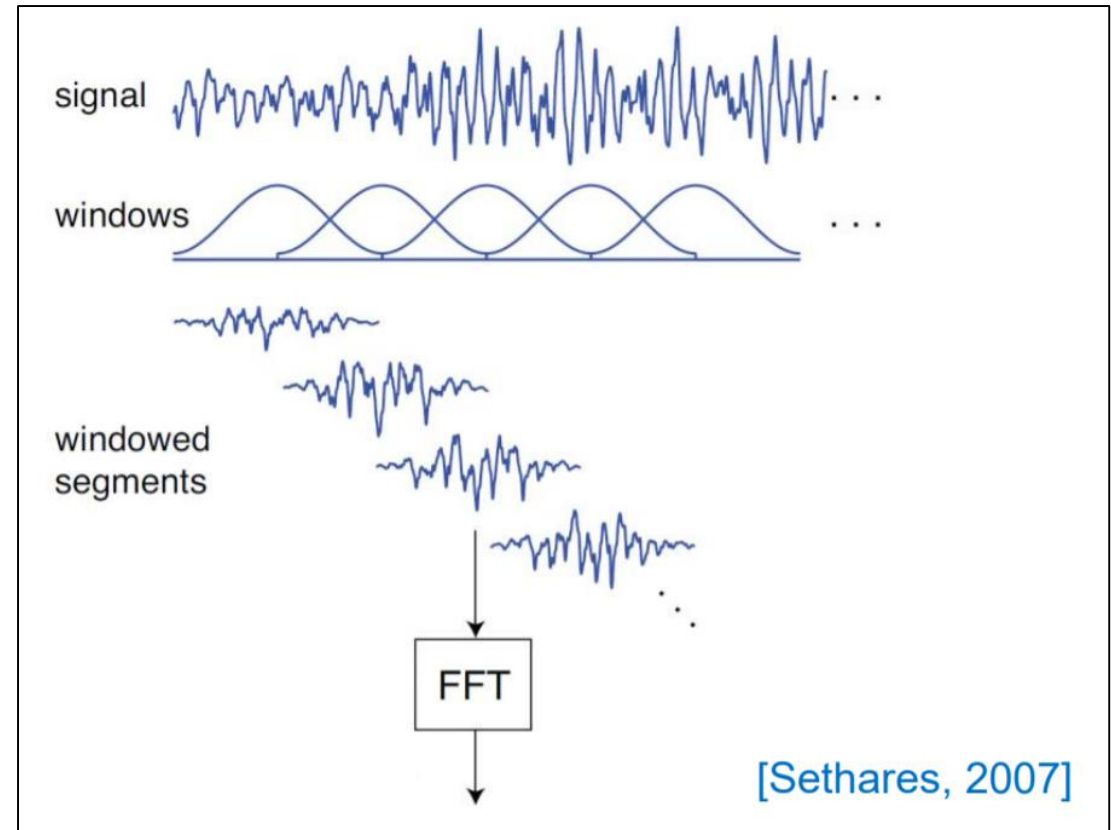


# Short-Time Fourier Transformation

- The spectral content of speech changes over time (non-stationary)
  - As an example, formants change as a function of the spoken phonemes
  - Applying the DFT over a long window does not reveal transitions in spectral content
- To avoid this issue, we apply the DFT over short periods of time
  - For short enough windows, speech can be considered to be stationary
  - Remember, though, that there is a time-frequency tradeoff here

# Short-Time Fourier Transformation

- Define analysis window
- Define the amount of overlap between windows
  - e.g., 30%
- Define a windowing function
  - e.g., Hann, Gaussian
- Generate windowed segments
  - Multiply signal by windowing function
- Apply the FFT to each windowed segment

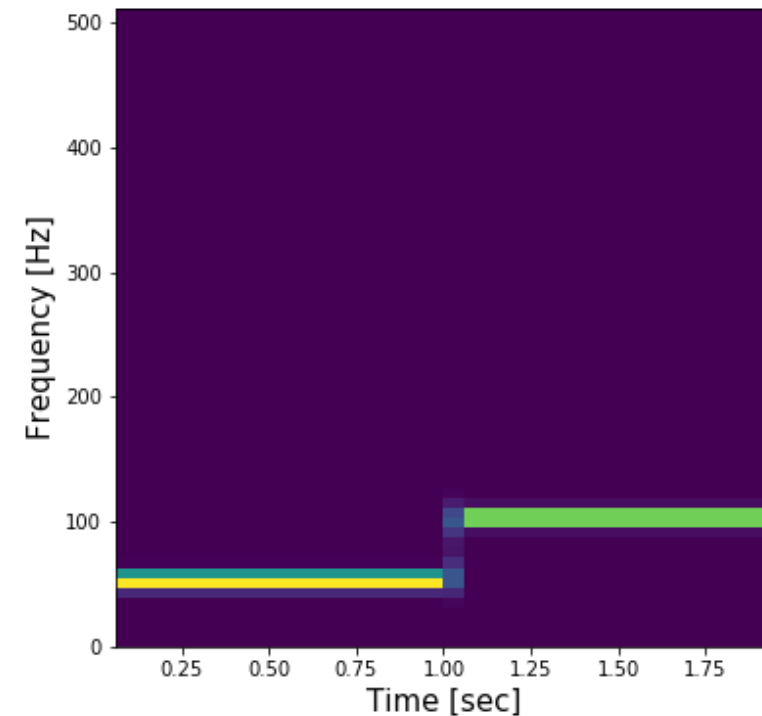


# STFT in Python

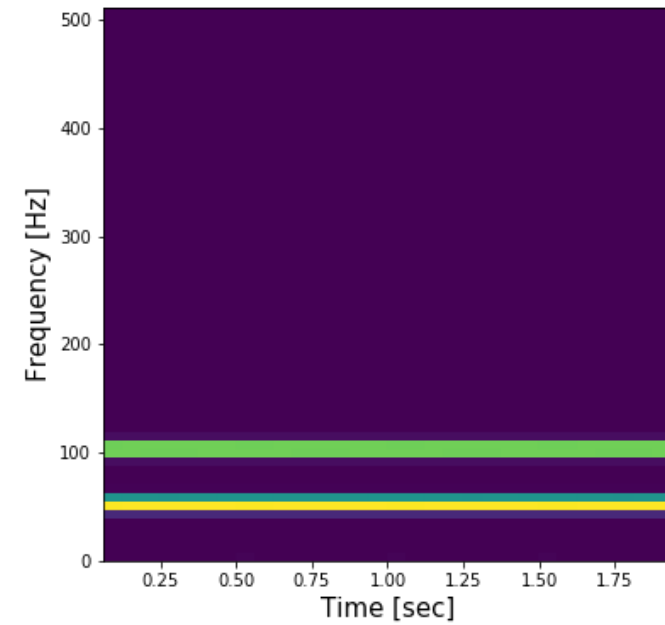
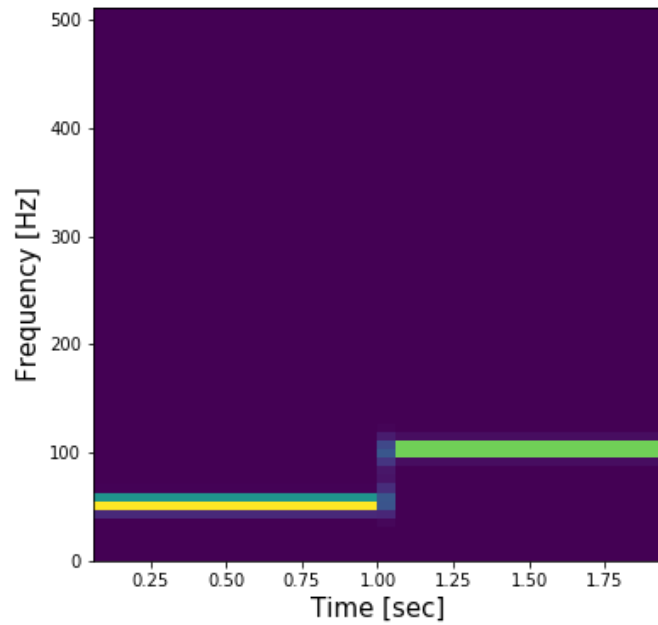
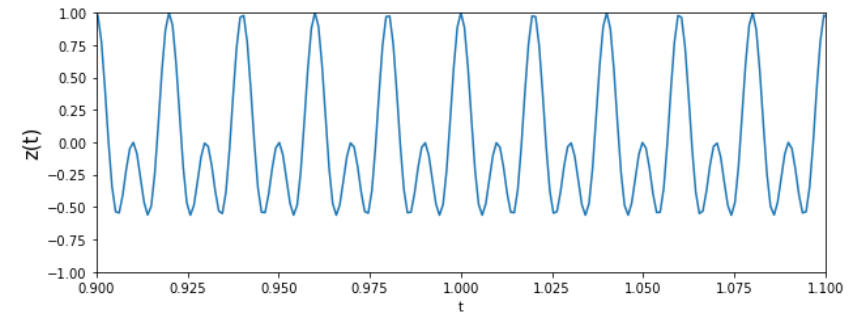
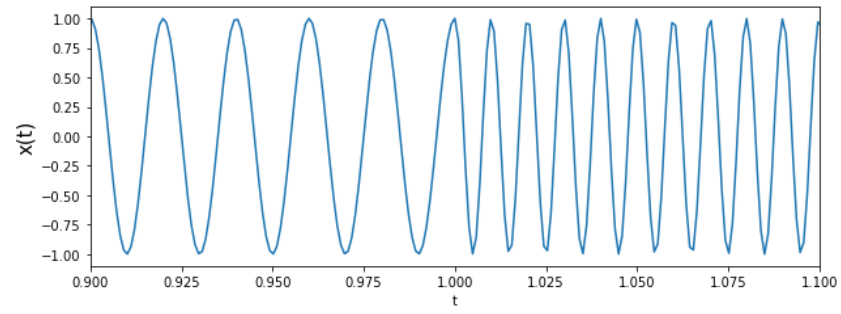
- Python function 'spectrogram' is useful for easily computing STFTs

```
windowsize = 2**7  
window = np.hanning(windowsize)  
nfft = windowsize  
noverlap = windowsize/2
```

```
f, t, Sxx = spectrogram(x,  
                        Fs,  
                        window = window,  
                        noverlap = noverlap,  
                        nfft = nfft,  
                        scaling = 'spectrum',  
                        mode = 'psd')
```



# STFT: Time and Frequency



# STFT Resolution Trade-off

