

# **The Short-Time Fourier Transform**

## **Part 1**

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# The Short-Time Fourier Transform

- The short-time Fourier transform (or **windowed Fourier transform**) is the widely used method for studying *transient signals*
- The concept behind it is simple and powerful
- Suppose we listen to a piece of music that lasts **an hour** where in the beginning there are violins and at the end drums
- If we Fourier analyze the whole hour, the Fourier analysis will show peaks at the frequencies corresponding to the violins and drums.
- That will tell us that there were violins and drums, but will not give us any indication of *when* the violins and drums were played

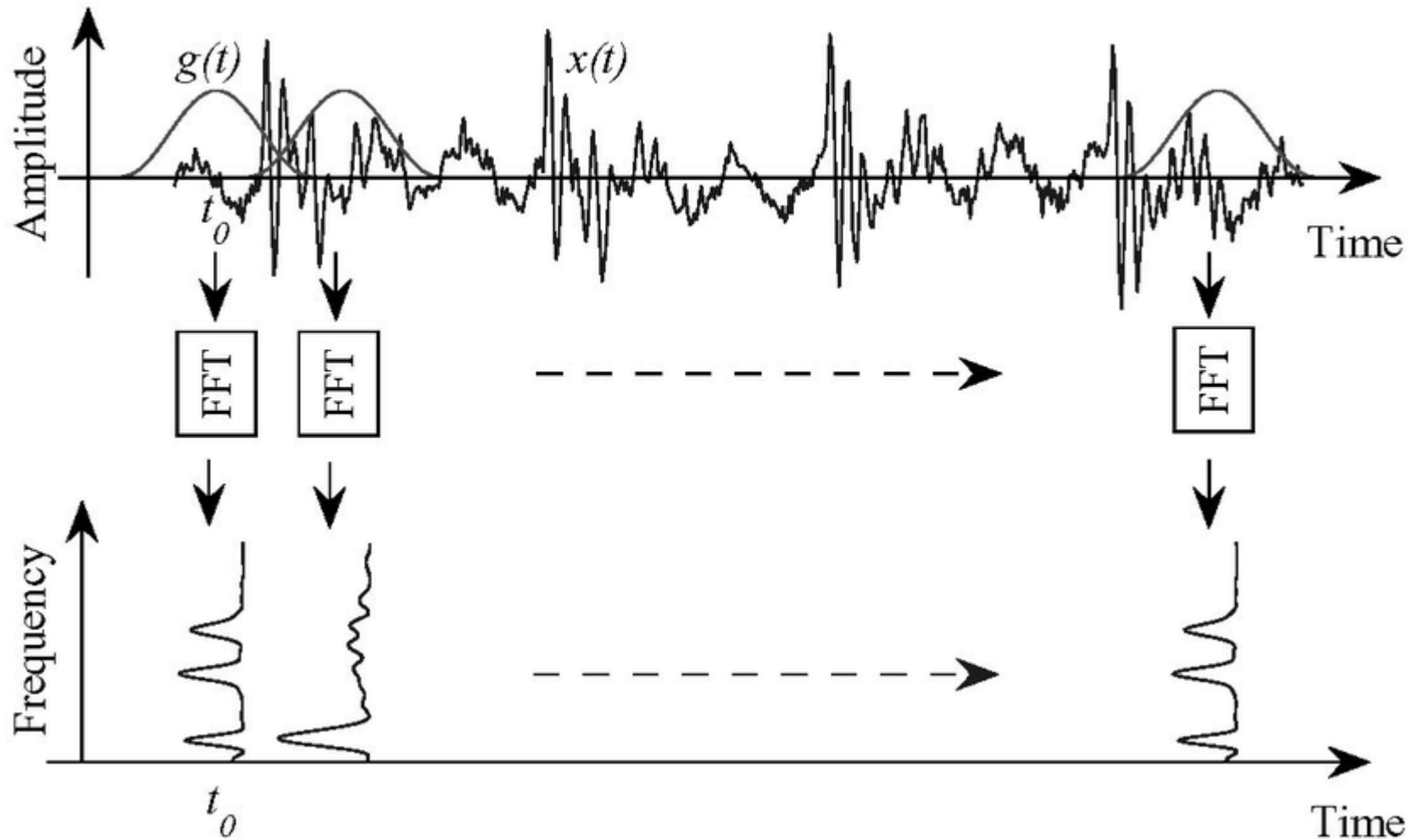
# The Short-Time Fourier Transform

- The most straightforward thing to do is to break up the hour into segments (for example, the duration of each segment is **five minutes**) and Fourier analyze each segment
- Upon examining the Fourier transform of each segment we will see in which five minutes intervals the violins and drums occurred
- If we want to localize even better, we break up the hour into **one minute segments** or even smaller time intervals and Fourier analyze each segment

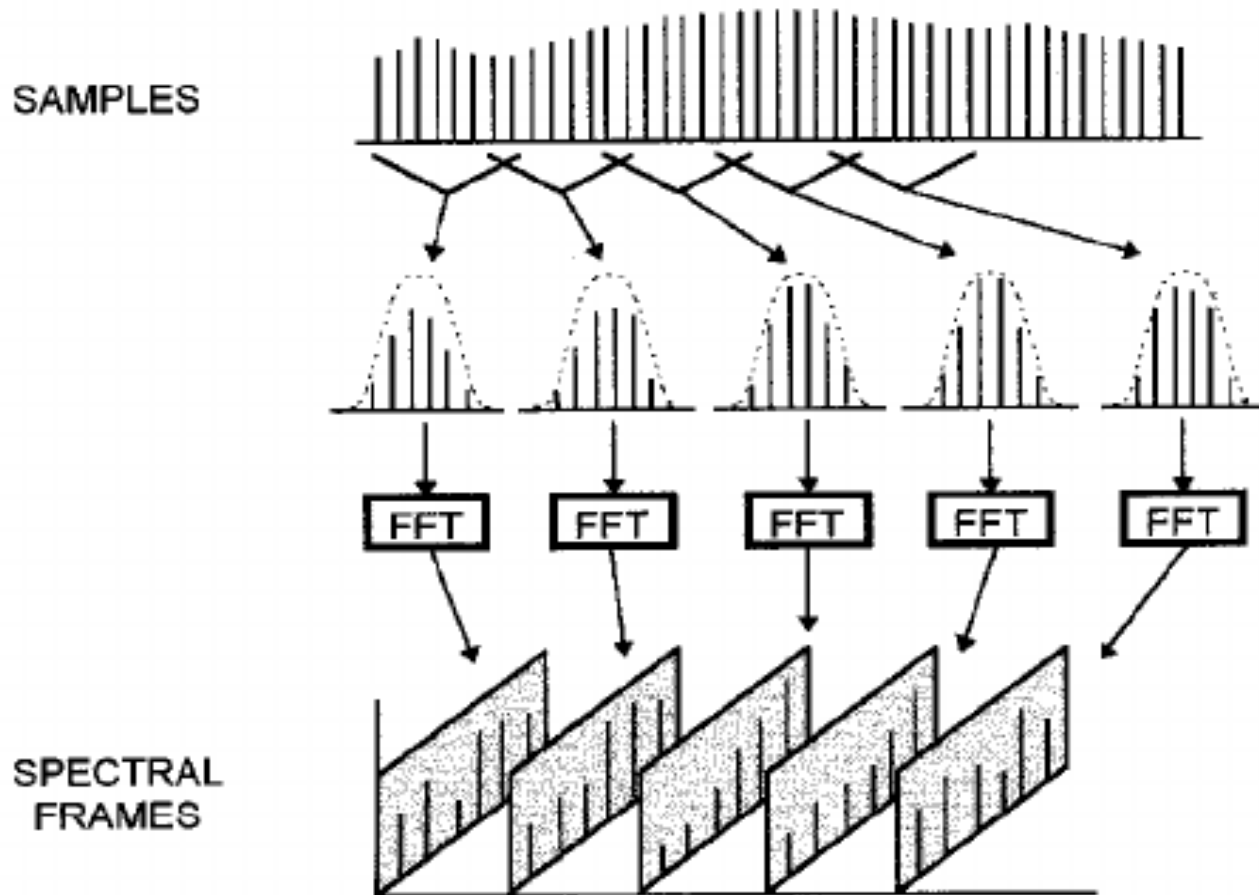
# The Short-Time Fourier Transform

- That is the basic idea of the short-time Fourier transform: break up the signal into small time segments by using a window and Fourier analyze each time segment to ascertain the signals that existed in that segment
- The *totality of transforms* indicates how the Fourier transform is varying in time

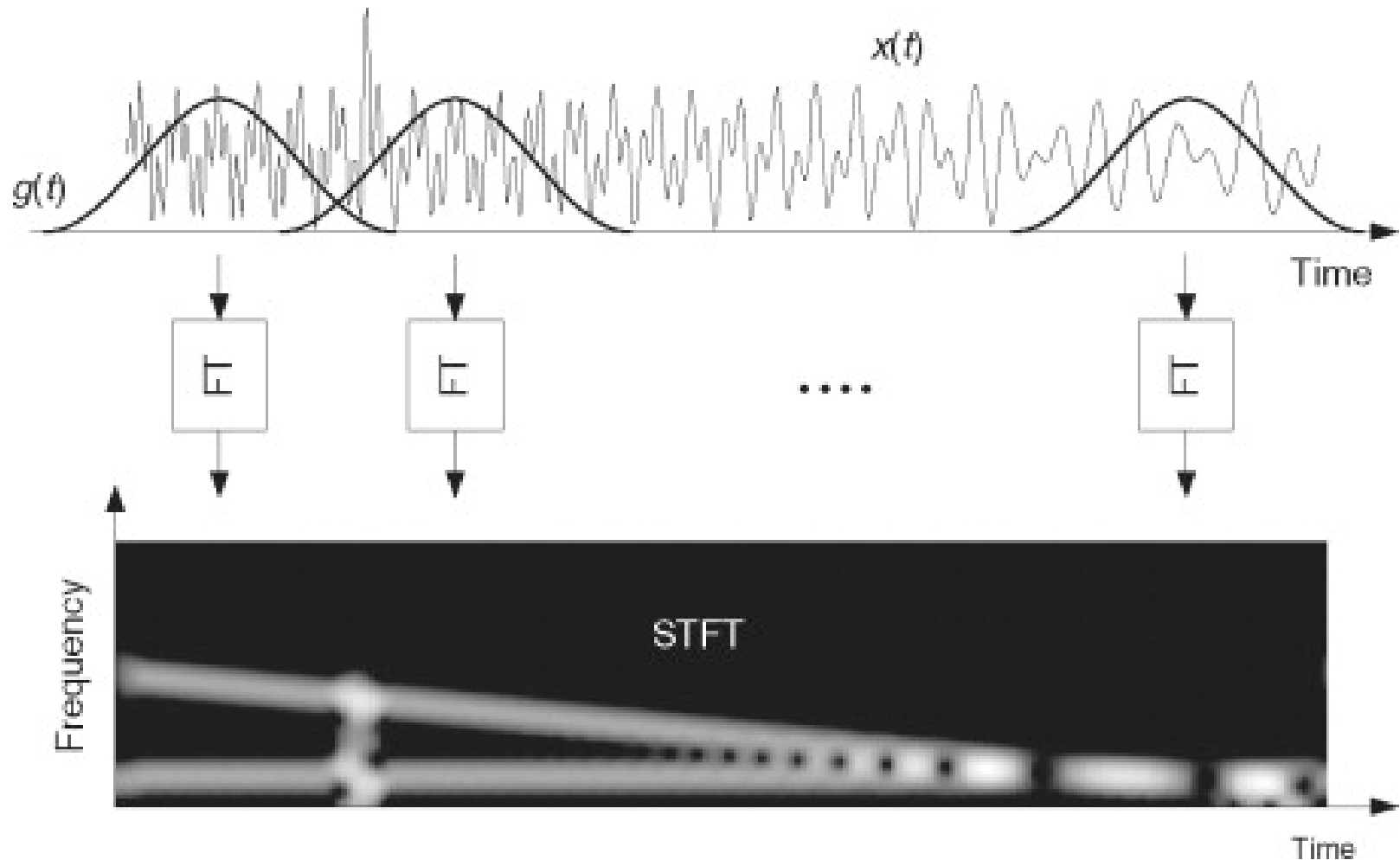
# The Short-Time Fourier Transform



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# The Short-Time Fourier Transform



# The Short-Time Fourier Transform

- Can this process be continued to achieve finer and finer time localization? Can we make the intervals as short as we want?
- The answer is “no”, because after a certain narrowing the answers we get for the Fourier transform **become meaningless** and show no relation to the Fourier transform of the signal



# The Short-Time Fourier Transform

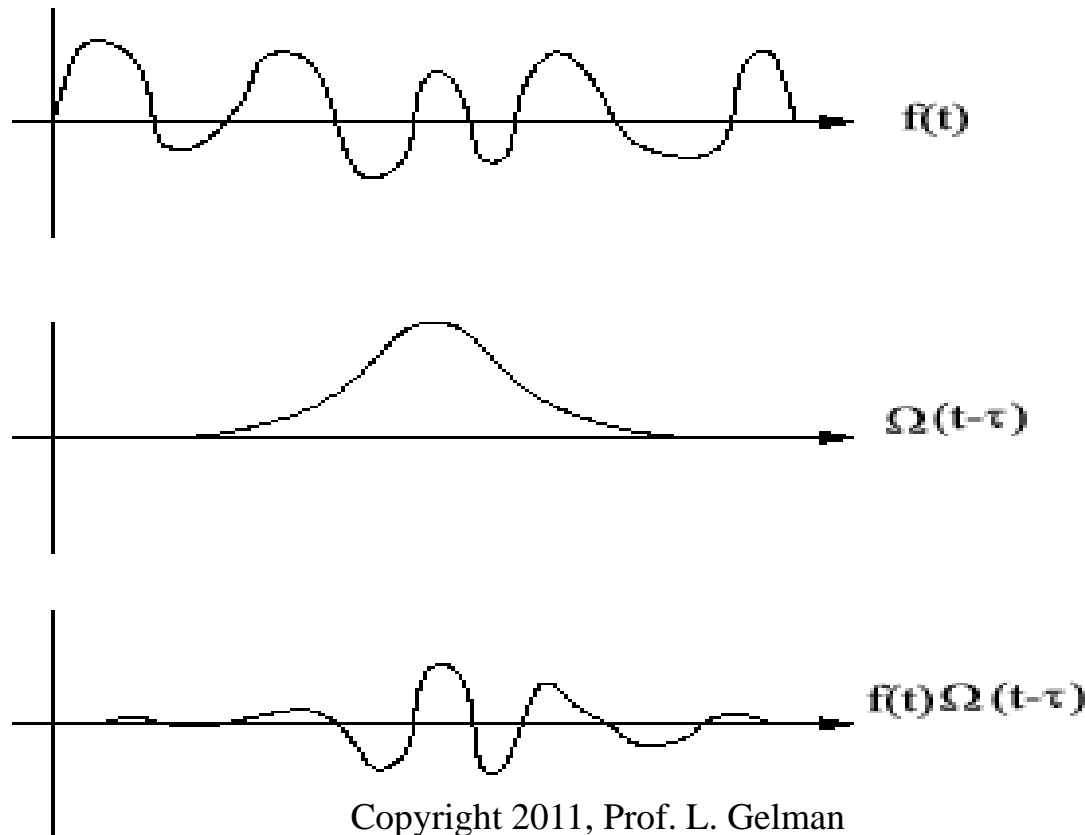
- The reason is that short duration signals have inherently **large bandwidths**, and the Fourier transform of such short duration signals **have very little to do with the properties of the original signal**
- It is the well-known **uncertainty principle** as applied to the small time intervals that we have created for the purpose of analysis
- We should always keep in mind that in the short-time Fourier transform the properties of the signal **are scrambled with the properties of the window function**; unscrambling is required for proper interpretation and estimation of the original signal

# The Short-Time Fourier Transform

- The above difficulty notwithstanding; the short-time Fourier transform is very good in many respects
- It is **well defined**, based on reasonable physical principles, and for many signals it gives an excellent time-frequency structure consistent with our intuition
- However, for certain situations it may not be the best method in the sense that it does not always give us the clearest possible picture of frequency content

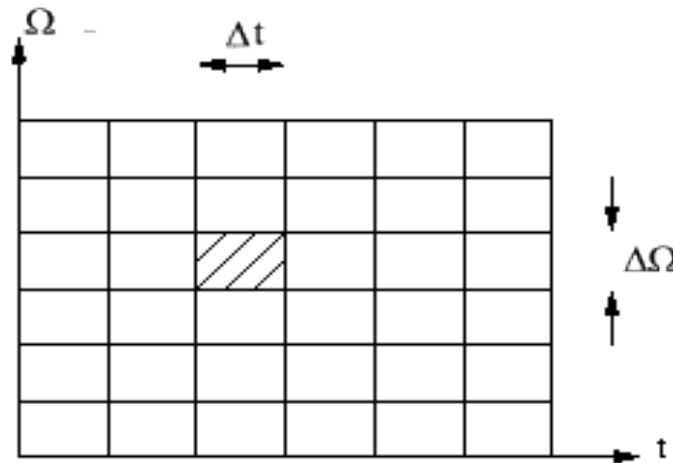
# The Short-Time Fourier Transform

The short-time Fourier transform can be interpreted as a **"sliding window Fourier transform"**: to slide the time center of window in time, window the input signal, and compute the Fourier transform of the result



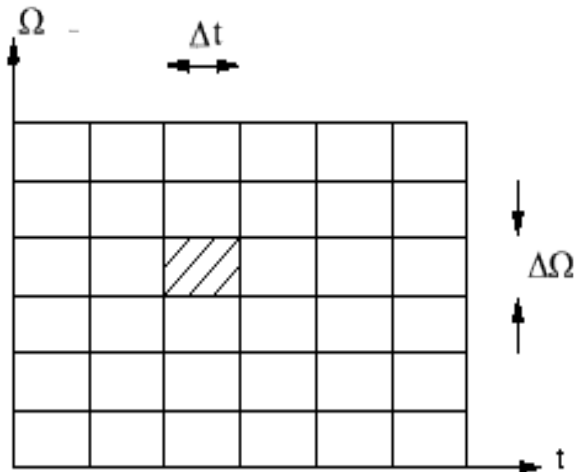
# The Short-Time Fourier Transform

- The idea is to isolate the signal in the **vicinity of time**, then perform the Fourier transform in order to estimate the "local" frequency content at time.
- This can be understood as time and frequency shifts of the window function. The short-time Fourier transform basis is often illustrated by a **tiling of the time-frequency plane**, where each tile represents a particular basis element



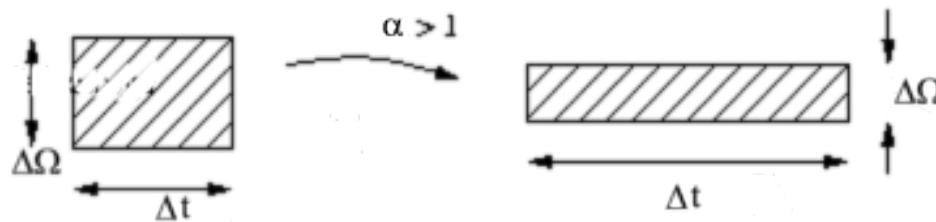
# The Short-Time Fourier Transform

- The height and width of a tile represent the **frequency and temporal widths** of the basis element, respectively, and the position of a tile represents the spectral and temporal centers of the basis element
- Note that, while the tiling diagram suggests that the short-time Fourier transform uses a discrete set of time/frequency shifts, the short-time Fourier transform basis is really constructed from a **continuum** of time/frequency shifts



# Uncertainty Principle

- Note that we can decrease spectral width at the cost of increased temporal width by stretching basis waveforms in time, **although the time-bandwidth product (i.e., the area of each tile) will remain constant**



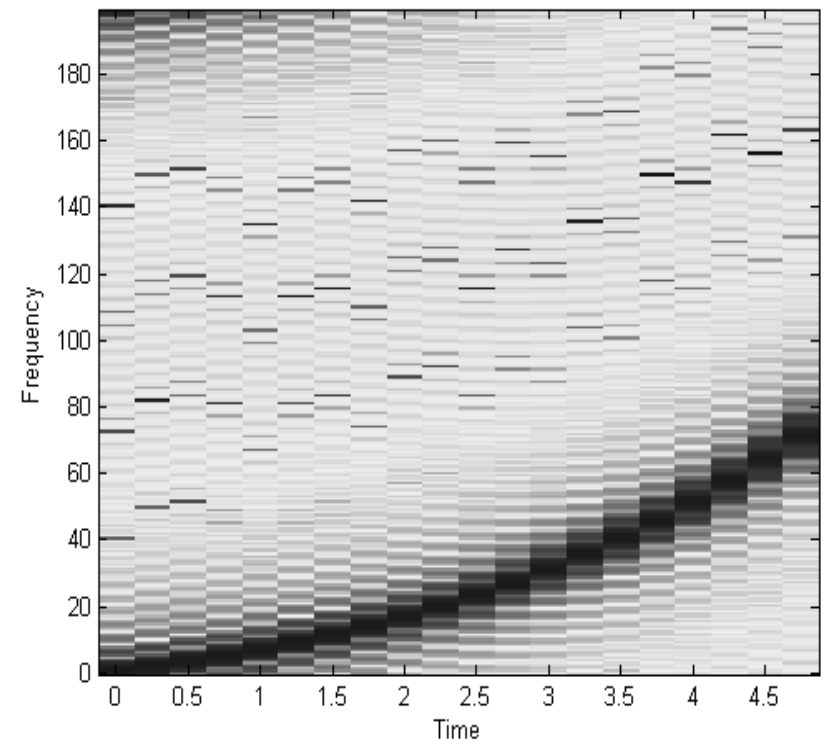
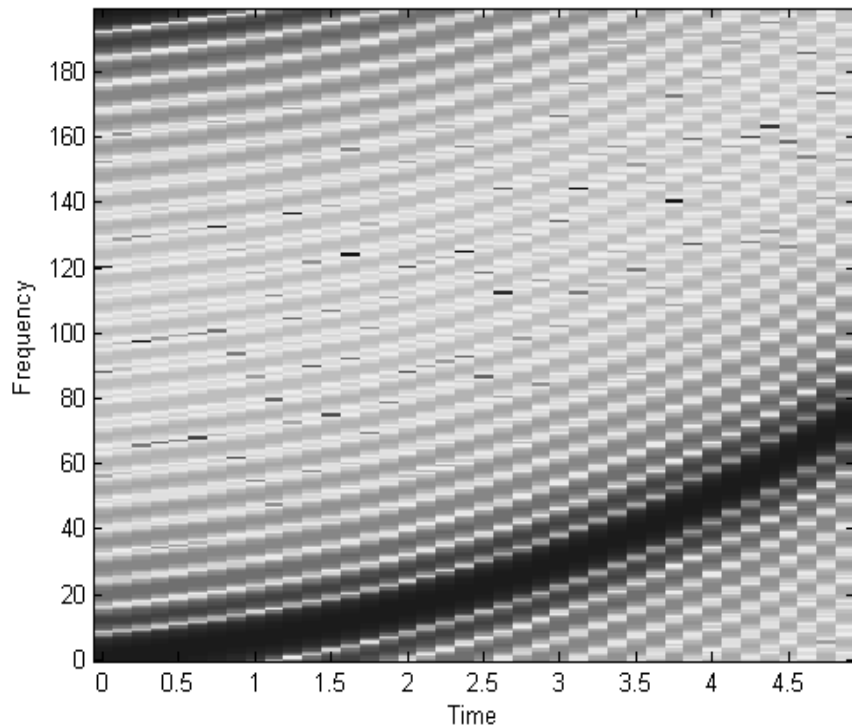
# Uncertainty Principle

- Unfortunately both spectral and time widths cannot be made arbitrary narrow; hence there is an inherent trade-off between time and frequency resolutions for a particular window (uncertainty principle), that

$$\Delta_t \Delta_f \geq a$$

- where  $\Delta_t$  is duration of time window, is  $\Delta_f$  bandwidth of frequency window
- The Gaussian window function  $h(\tau) = e^{-a\tau^2/2}$  is *optimal* in terms of the uncertainty principle; i. e.  $a = 0.5$
- The short-time Fourier transform using the Gaussian window is known as the Gabor transform

# The Short-Time Fourier Transform





# The Short-Time Fourier Transform: Definition

- In the conventional Fourier transform, the signal is compared with complex sinusoidal functions
- Because sinusoidal functions are spread over the entire time domain and are not concentrated in time, the Fourier transform does not explicitly indicate how a signal's frequency **contents evolve in time**
- To study the properties of the signal at time , one emphasizes the signal *at that time* and suppresses the signal at other times

# The Short-Time Fourier Transform

- This is achieved by multiplying the signal by a window function, centered at time  $t$  :

$$x_t(\tau) = x(\tau)h(\tau - t)$$

- The modified signal is a function of **two times**, the fixed time we are interested in,  $t$  , and the running time,  $\tau$
- The window function is chosen to leave the signal more or less unaltered around the time of interest  $t$  but to suppress the signal for times distant from the time of interest

# The Short-Time Fourier Transform

- Since the modified signal emphasizes the signal around the time  $t$ , the short-time Fourier transform will reflect the distribution of frequency around that time:

$$X(t, f) = \int x(\tau)h(\tau - t)e^{-i2\pi f\tau} d\tau$$

- The power spectral density at that time is therefore:

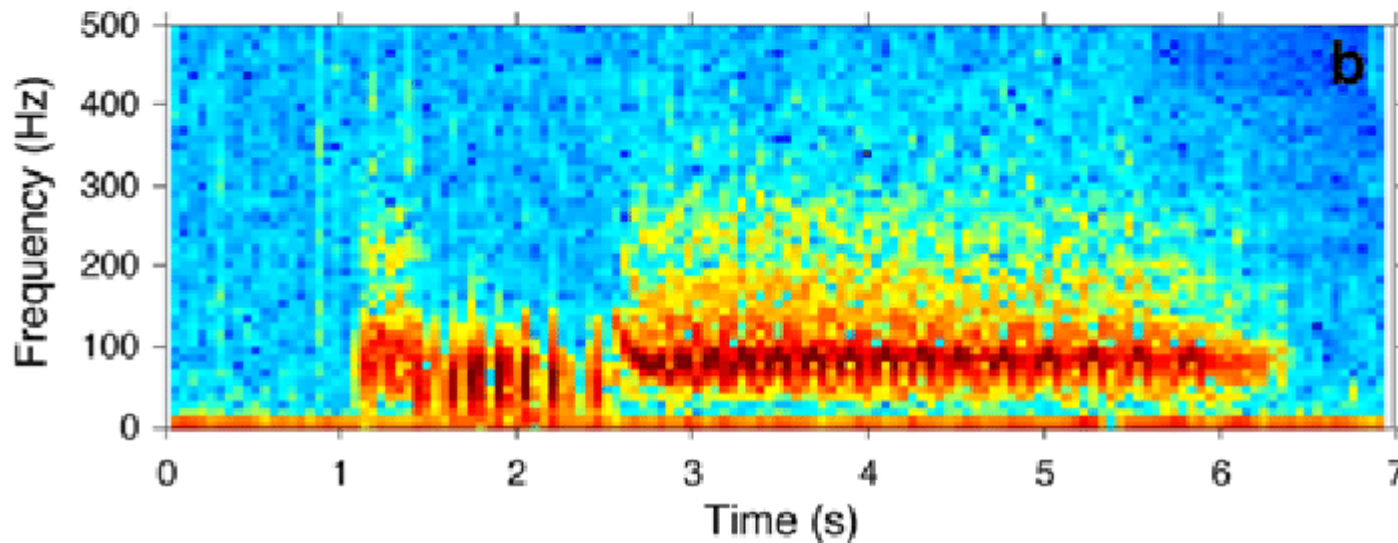
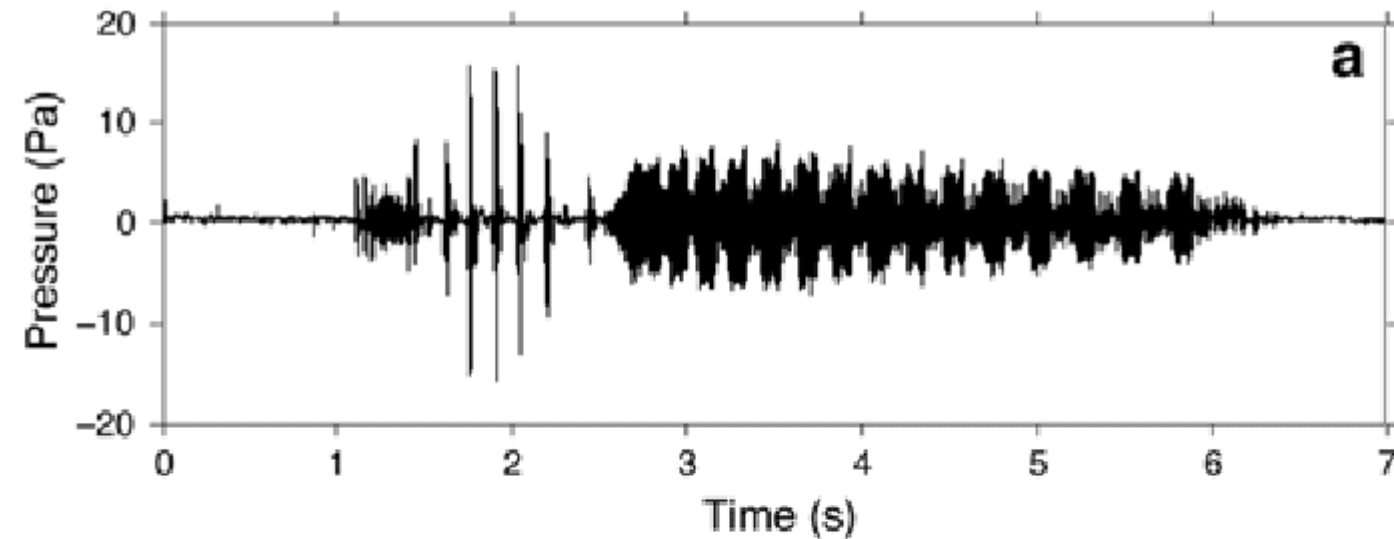
$$S_{xx}(t, f) = |X(t, f)|^2$$

- The magnitude of the short-time Fourier transform is called the *spectrogram*
- For each different time we get a different spectrogram and the *totality* of these spectrograms is the *time-frequency distribution*

# The Short-Time Fourier Transform

- These spectrograms can be plotted one behind the other in the **“waterfall” diagram**
- Since we are interested in analyzing the signal around the time  $t$ , we presumably have chosen a narrow window that is peaked around that time

# The Short-Time Fourier Transform



# The Short-Time Fourier Transform: Waterfall Plots

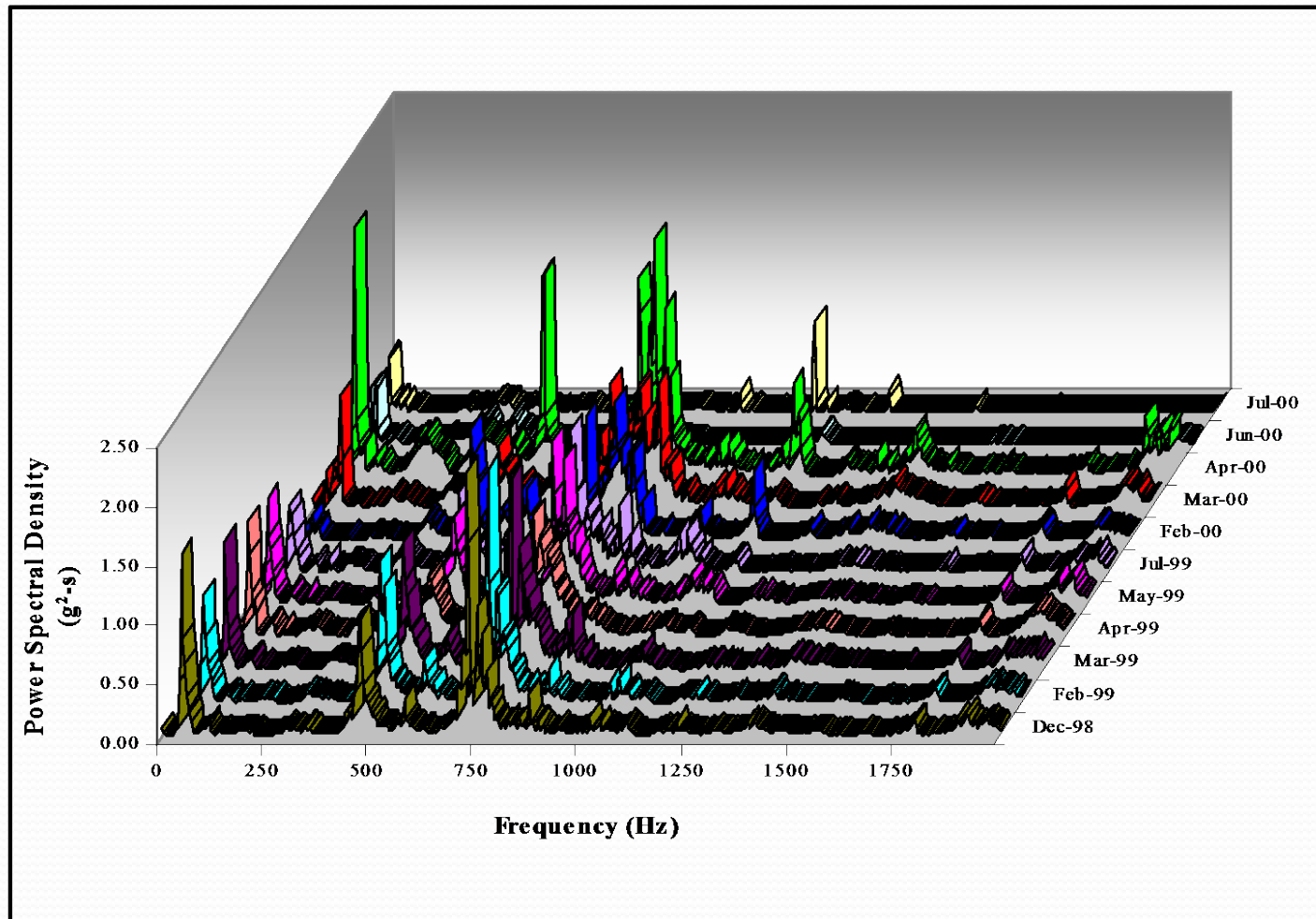
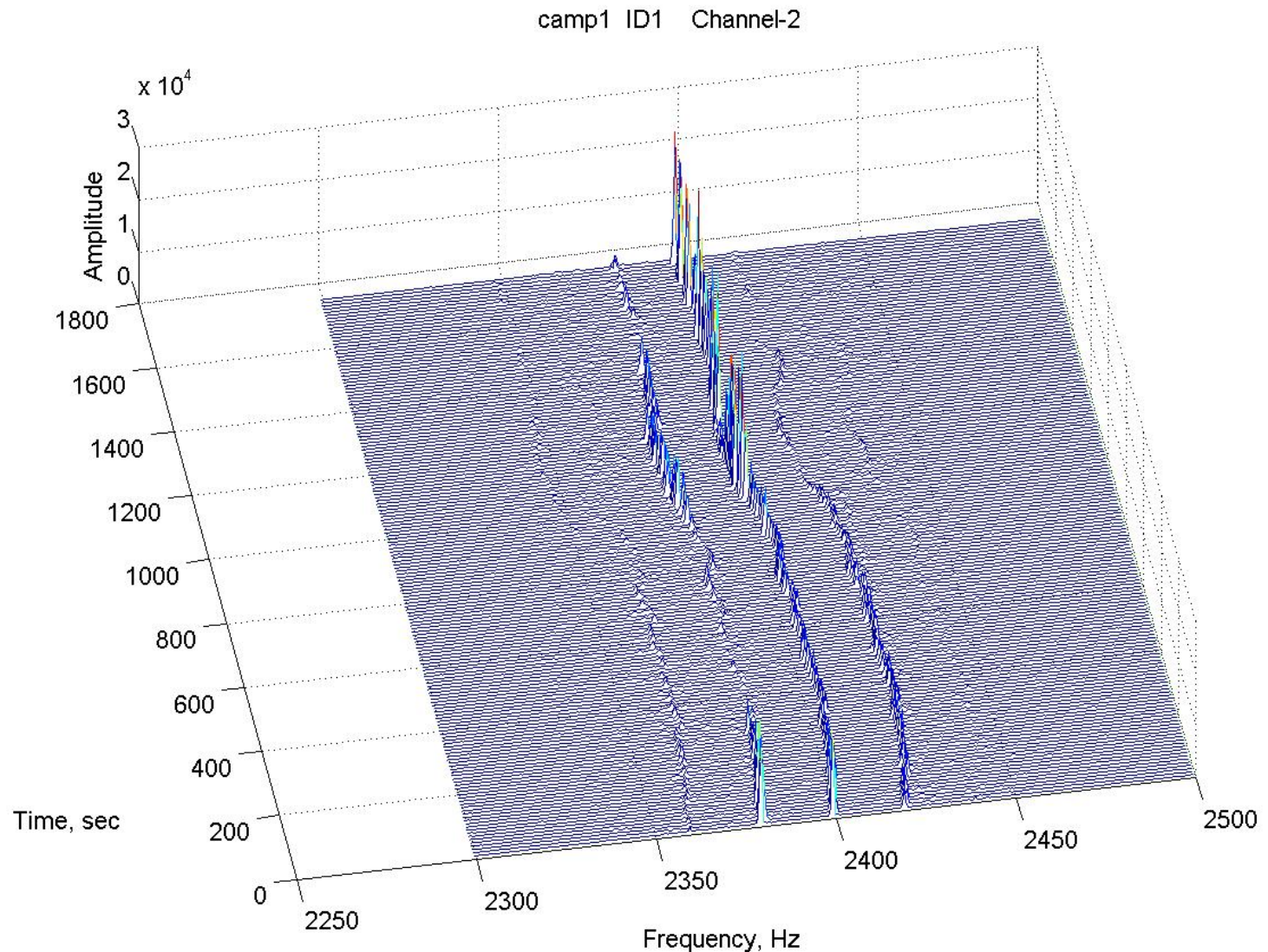


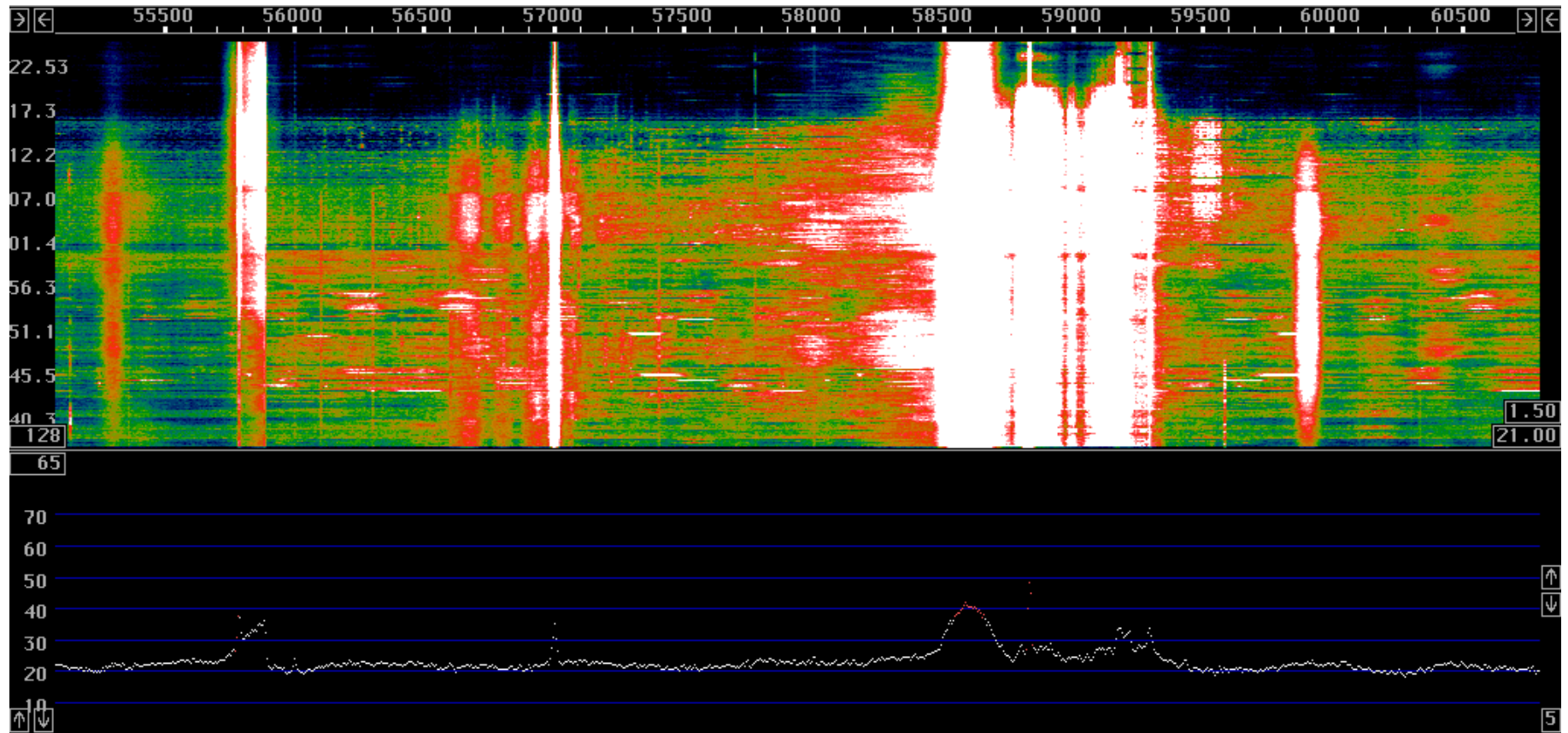
Figure 12.15 Waterfall plot of the signals in the radial direction for the bottom bearing on the pump

# The Short-Time Fourier Transform





# The Short-Time Fourier Transform





# Properties

- The spectrogram is a **real-valued function**, even if the signal is complex
- A time shift in the signal has the same time shift in the short-time Fourier transform

# Case Study 1

- **Impulse and application of the Gaussian window for STFT**

$$x(t) = \sqrt{2\pi} \delta(t - t_0) \quad h(t) = \left(\frac{a}{\pi}\right)^{0.25} e^{-at^2/2}$$

- **The short-time Fourier transform is:**

$$X(t, f) = \left(\frac{a}{\pi}\right)^{0.25} e^{-i2\pi t_0 f} \exp\left[-\frac{a(t - t_0)^2}{2}\right]$$

**which yields the following spectrogram (modulus):**

$$S_{xx}(t, f) = \left(\frac{a}{\pi}\right)^{0.5} \exp\left[-a(t - t_0)^2\right]$$

# Case Study 2

- The sum of a sinusoid and impulse; application of the Gaussian window for STFT
- Using previous results we obtain the short-time Fourier transform (spectrogram):

$$S_{xx}(t, f) = \frac{1}{(a\pi)^{0.5}} \exp\left[-\frac{4\pi^2(f-f_0)^2}{a}\right] + \left(\frac{a}{\pi}\right)^{0.5} \exp[-a(t-t_0)^2] \\ + \frac{2}{\sqrt{\pi}} e^{-\frac{4\pi^2(f-f_0)^2}{a-a(t-t_0)^2}} \cos 2\pi(f(t-t_0) - f_0 t)$$

- This example illustrates one of the fundamental difficulties with the spectrograms, i. e. for one window we cannot have high resolution in time and frequency

# The Short-Frequency Time Transform

- We may wish to study time properties at a *particular frequency*
- We then window the Fourier transform with a **frequency window**,  $H(f)$ , and take the **time transform**, which, of course, is the inverse Fourier transform:

$$x(t, f) = \int X(f') H(f - f') e^{-i2\pi f t} df'$$