

# Spectral Analysis – Fourier Decomposition

Adding together different sine waves

PHY103

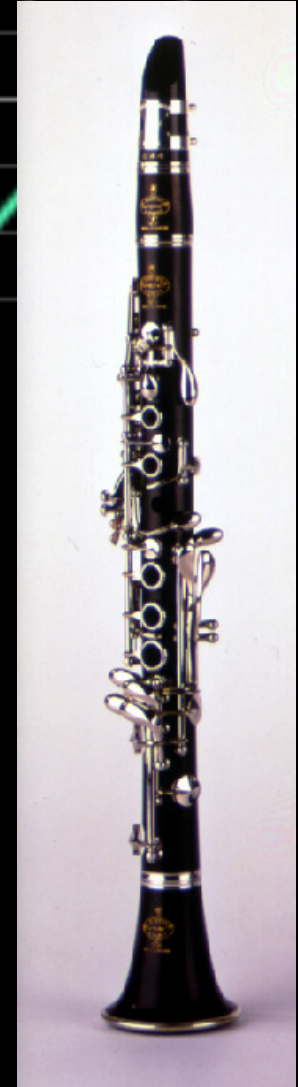


image from [http://hem.passagen.se/eriahl/e\\_flat.htm](http://hem.passagen.se/eriahl/e_flat.htm)

# Spectral decomposition

## Fourier decomposition

- Previous lectures we focused on a single **sine** wave.
- With an *amplitude* and a *frequency*
- Basic spectral unit ----

How do we take a complex signal and describe its frequency mix?

We can take any function of time and describe it as a sum of sine waves each with different amplitudes and frequencies

Sine waves – one amplitude/ one frequency

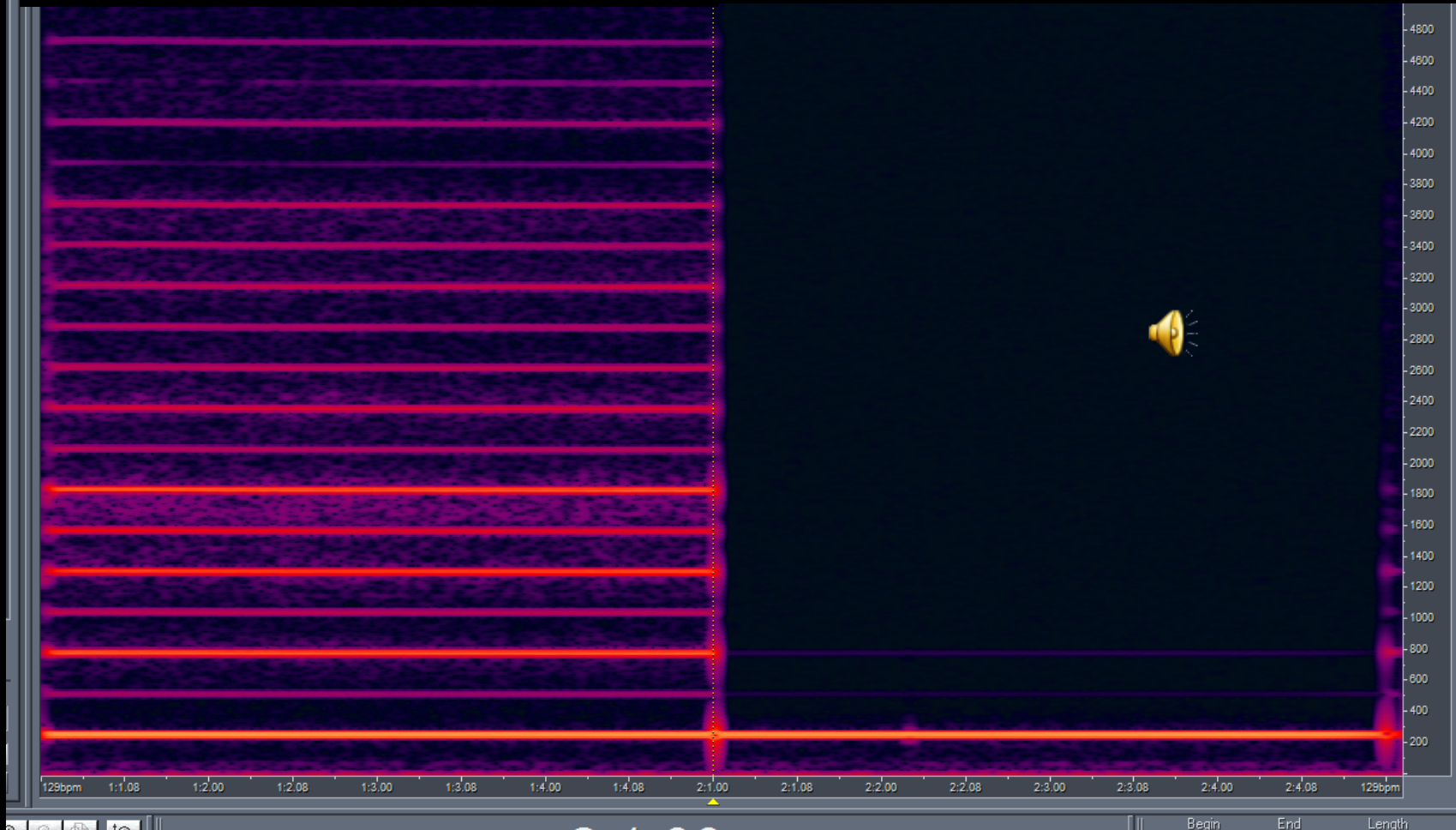
Sounds as a series of pressure or motion variations in air.

Sounds as a sum of different amplitude signals each with a different frequency.

Waveform vs Spectral view in Audition

Spectral view  
Clarinet spectrum

Clarinet spectrum with only  
the lowest harmonic remaining



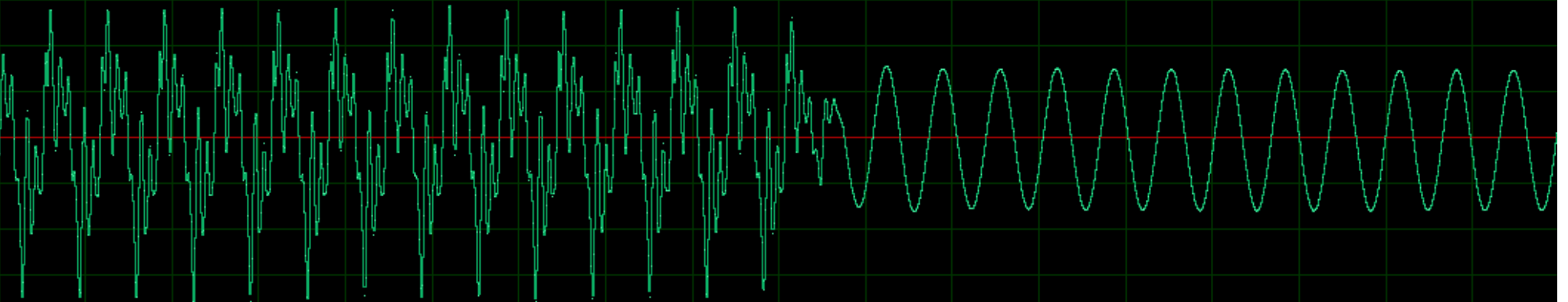
Frequency →

Time →

# Waveform view

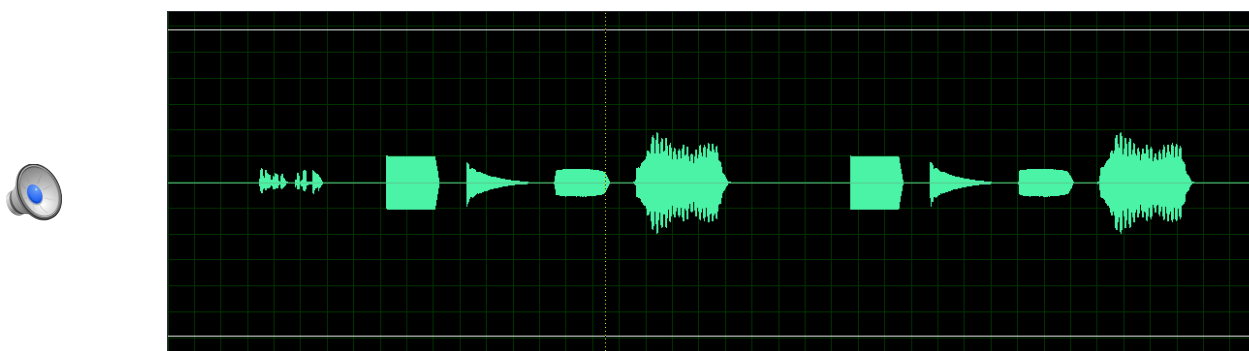
Full sound

Only lowest harmonic

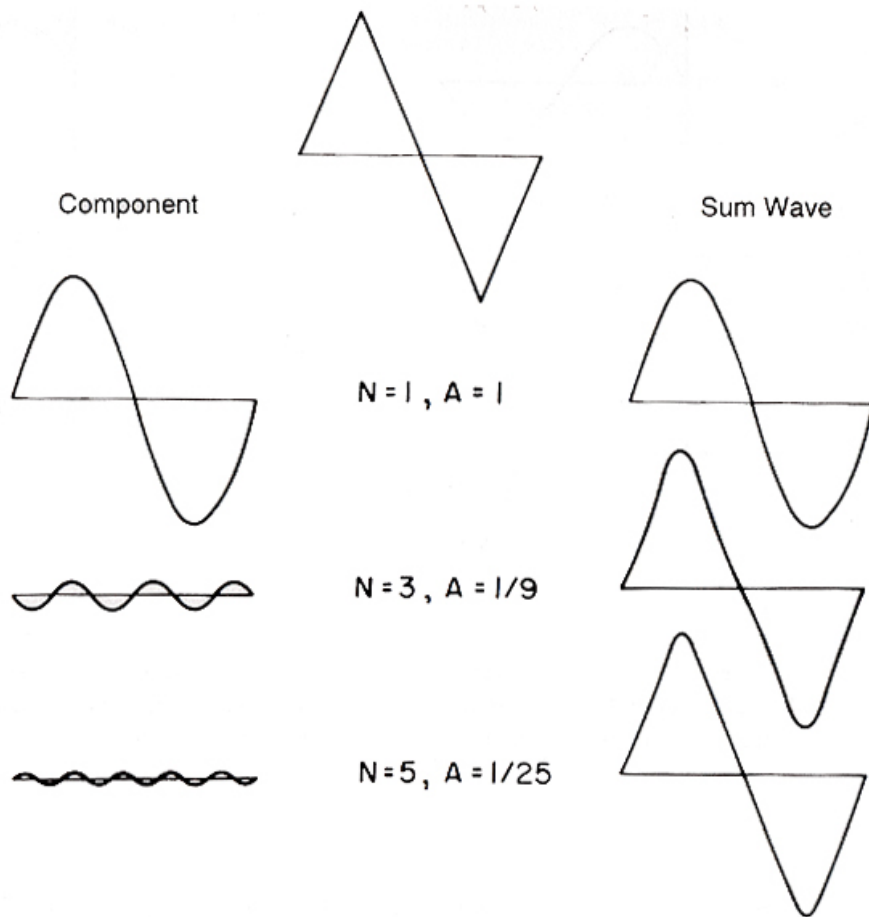


# Four complex tones in which all partials have been removed by filtering (Butler Example 2.5)

One is a French horn, one is a violin, one is a pure sine, one is a piano (but out of order)



It's hard to identify the instruments. However clues remain (attack, vibrato, decay)



**Figure 4-4** Fourier synthesis of a triangular wave. At the left are the successive harmonics; at the right are the sum waves including each successive harmonic. The graph at the top is the wave being synthesized.

Making a triangle wave with a sum of harmonics.

Adding in higher frequencies makes the triangle tips sharper and sharper.

# Sum of waves

- Complex wave forms can be reproduced with a sum of different amplitude sine waves
  - Any waveform can be turned into a sum of different amplitude sine waves
- “Fourier decomposition - Fourier series”



# What does a triangle wave sound like compared to the square wave and pure sine wave?

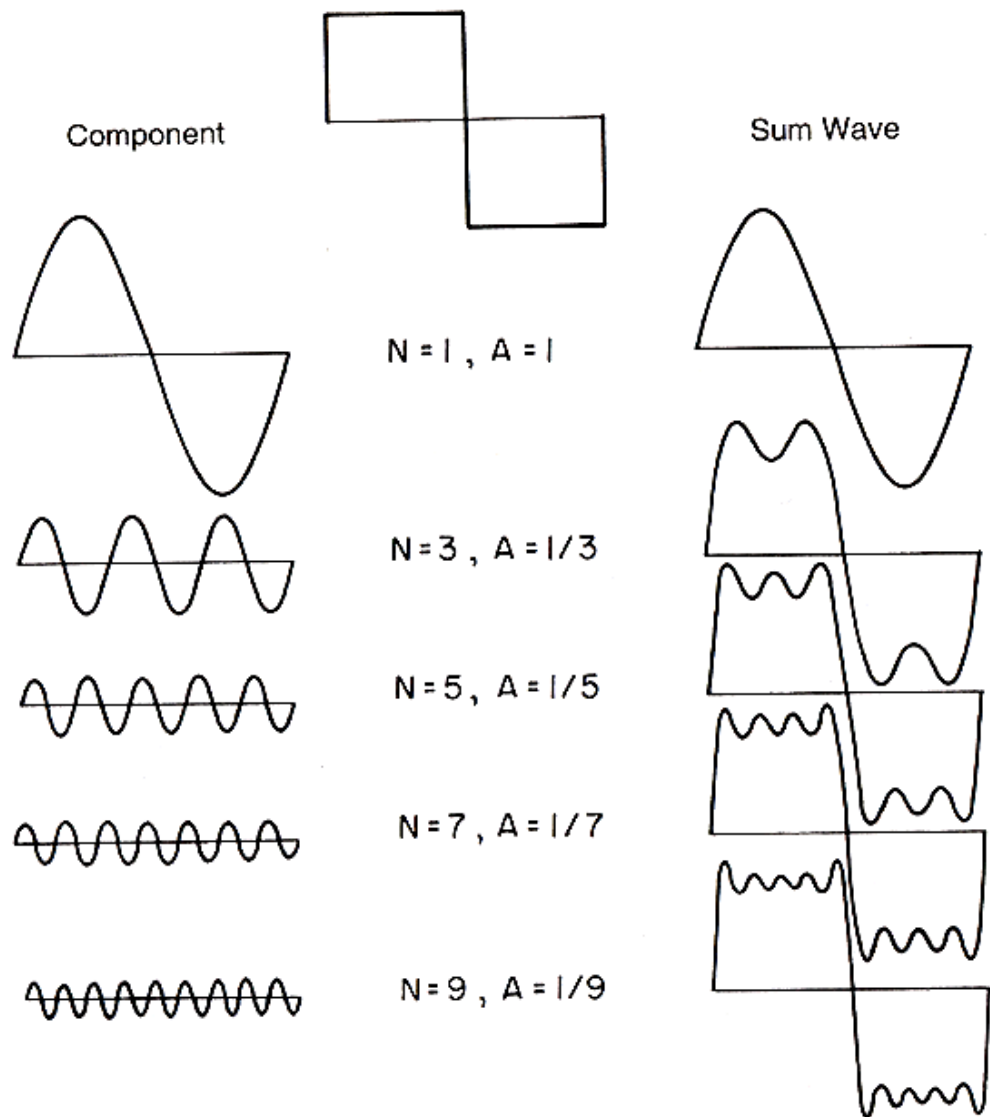
- (Done in lab and previously in class)
- Function generators often carry sine, triangle and square waves (and often sawtooths too)

If we keep the frequency the same the pitch of these three sounds is the same.

However they sound different.

**Timbre** --- that character of the note that enables us to identify different instruments from their sound.

Timbre is related to the frequency spectrum.



**Figure 4-5** Fourier synthesis of a square wave. At the left are the successive harmonics; at the right are the sum waves including each successive harmonic. The graph at the top is the wave being synthesized.

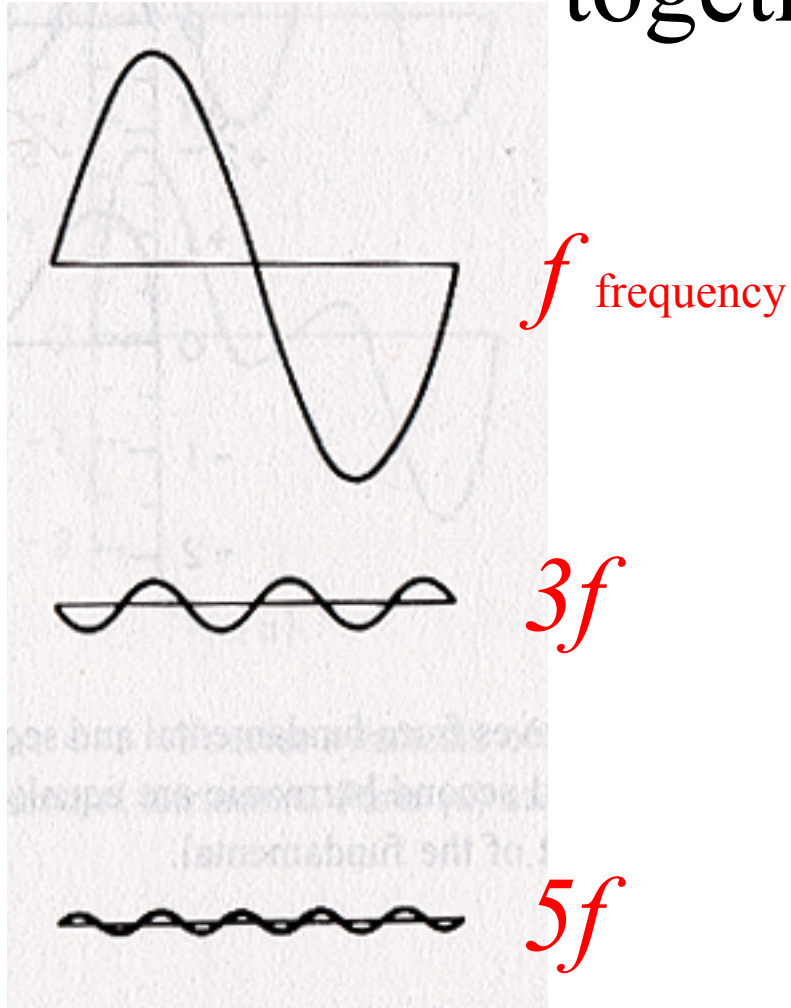
From Berg and Stork

# Square wave

Same harmonics however the higher order harmonics are stronger.

Square wave sounds shriller than the triangle which sounds shriller than the sine wave

# Which frequencies are added together?

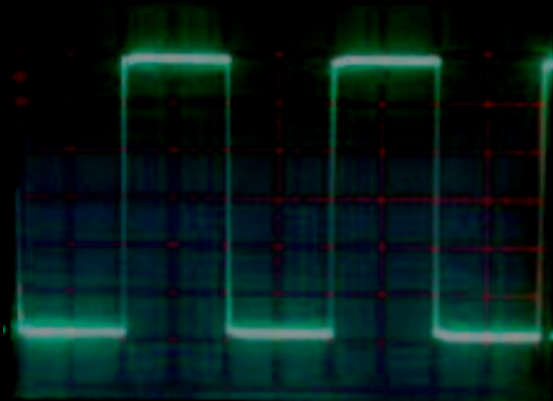
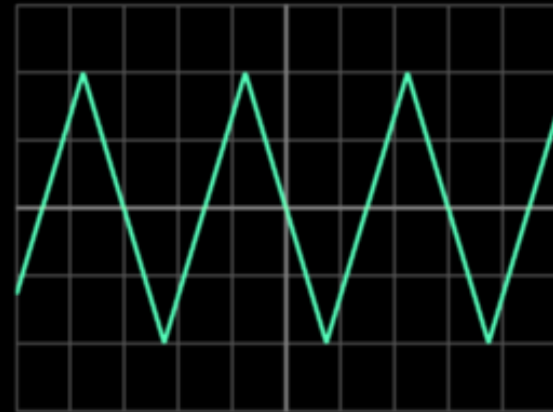


To get a triangle or square wave we only add sine waves that fit exactly in one period. They cross zero at the beginning and end of the interval.

These are harmonics.

# Periodic Waves

- Both the triangle and square wave cross zero at the beginning and end of the interval.
  - We can repeat the signal
- Is “Periodic”
- Periodic waves can be decomposed into a sum of harmonics or sine waves with **frequencies** that are **multiples** of the biggest one that fits in the interval.



# Sum of harmonics

- Also known as the **Fourier series**
- Is a sum of sine and cosine waves which have frequencies  $f, 2f, 3f, 4f, 5f, \dots$
- Any **periodic** wave can be decomposed in a Fourier series

# Building a sawtooth by waves

- Cookdemo7
  - a. top down
  - b. bottom up



# Light spectrum

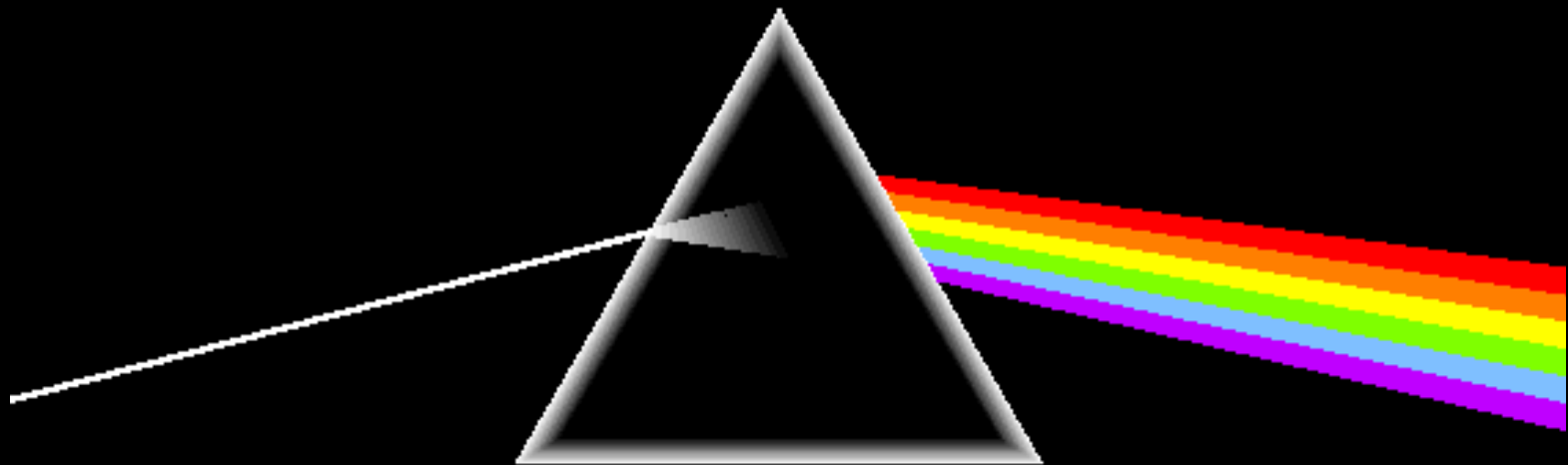
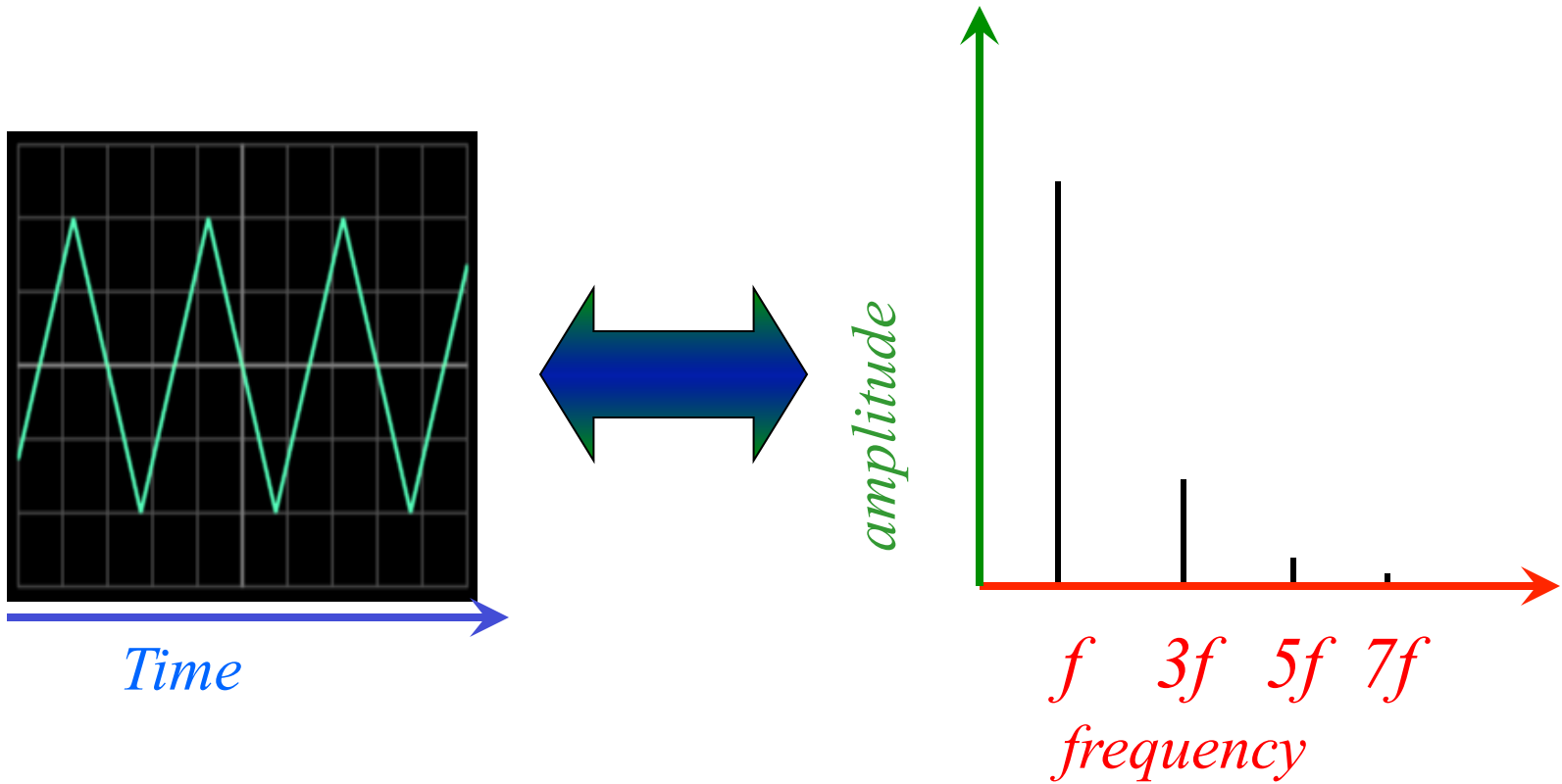
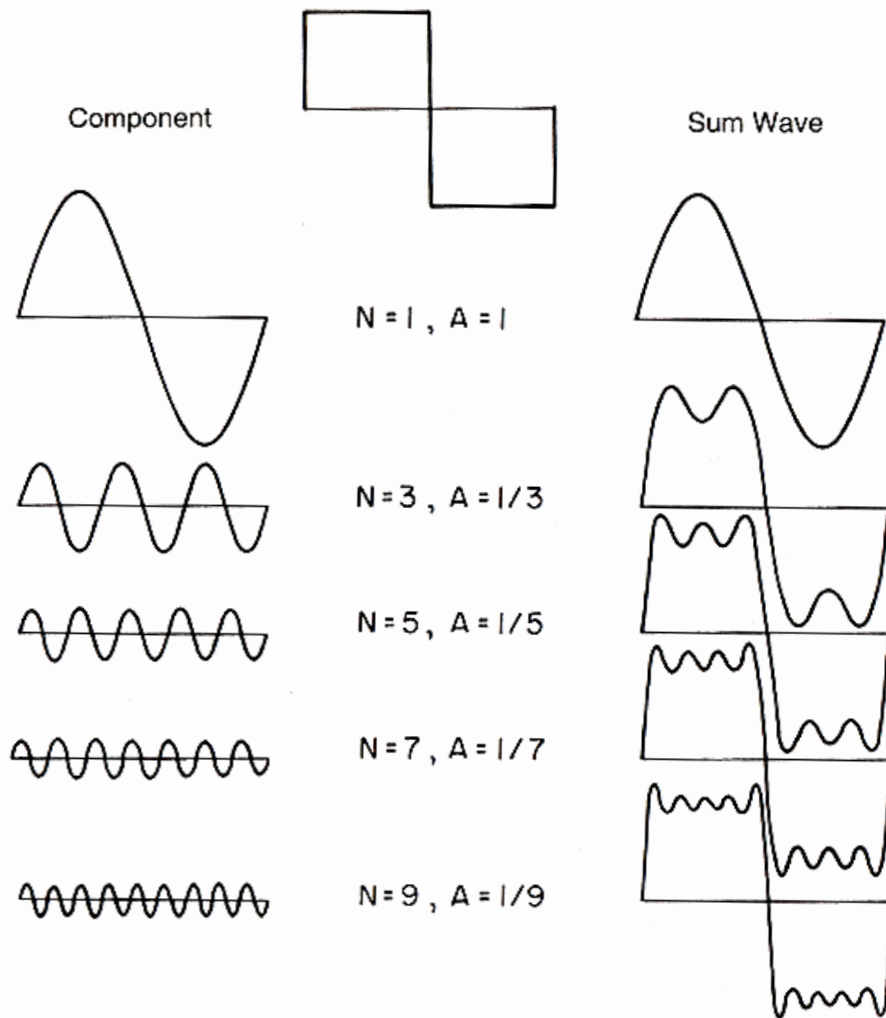


Image from <http://scv.bu.edu/~aarondf/avgal.html>

# Sound spectrum





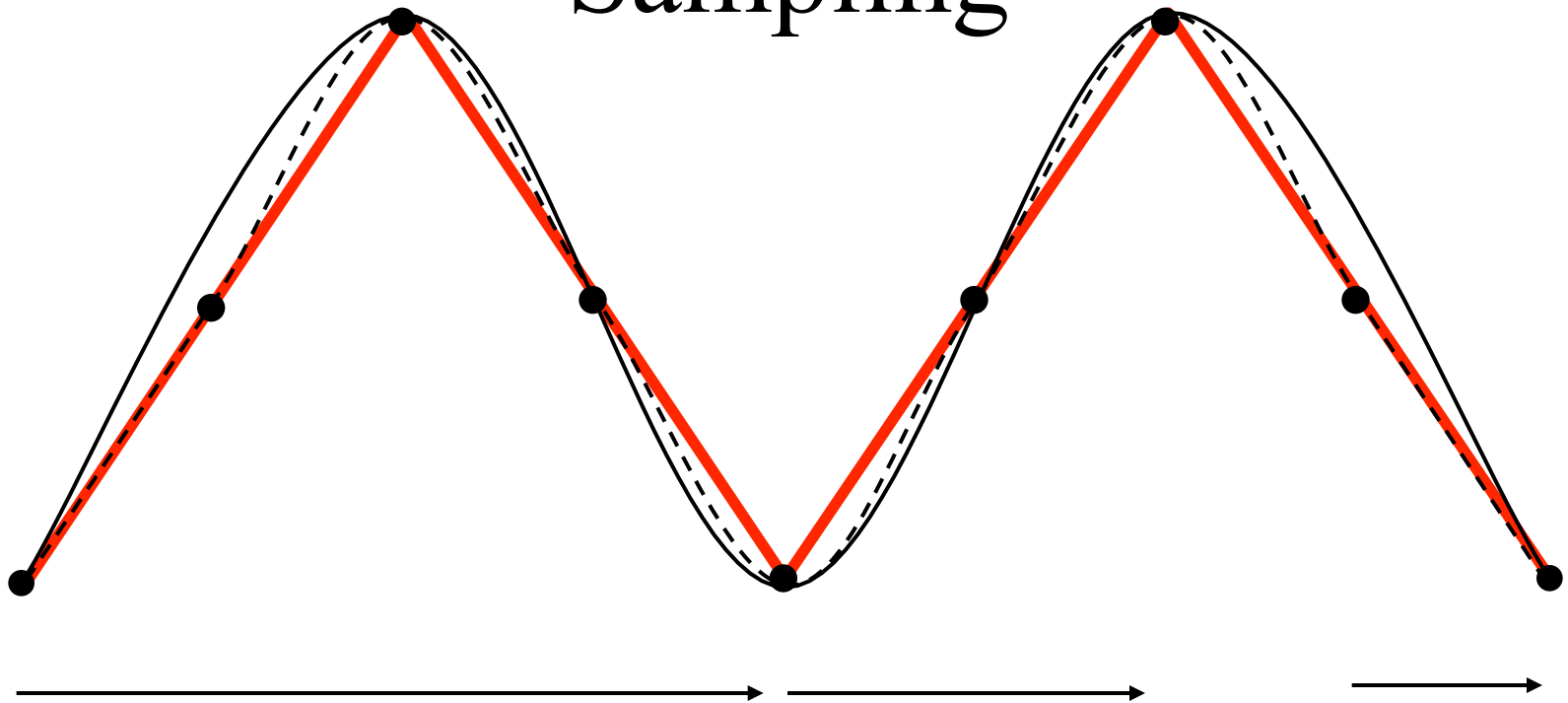


**Figure 4-5** Fourier synthesis of a square wave. At the left are the successive harmonics; at the right are the sum waves including each successive harmonic. The graph at the top is the wave being synthesized.

Sharp bends  
imply high  
frequencies

Leaving out the high  
frequency  
components  
smoothes the curves  
**Low pass filter**  
removes high  
frequencies –  
Makes the sound less  
shrill or bright

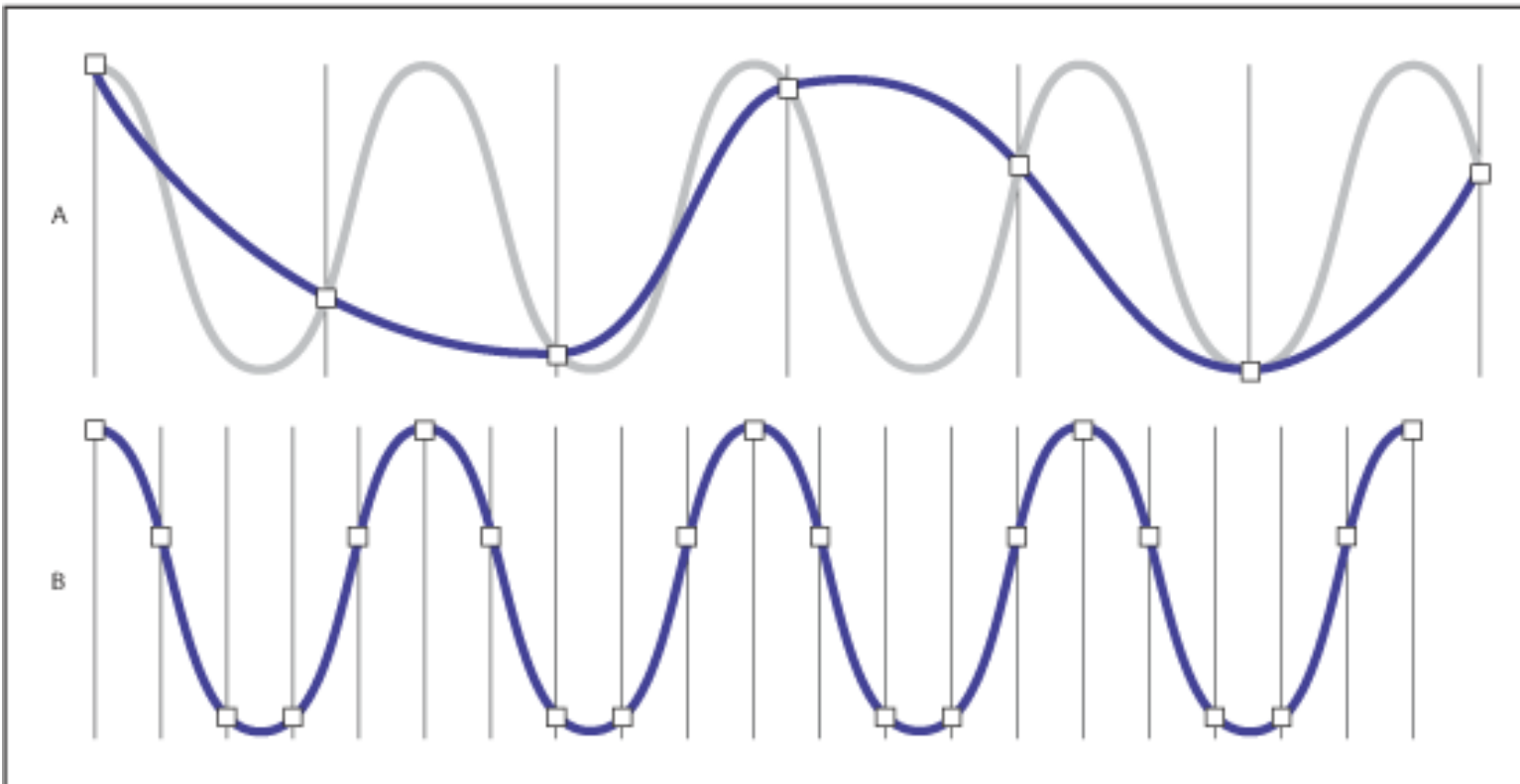
# Sampling



If sampled every  
period then the  
entire wave is lost

The shorter the sampling  
spacing, the better the wave  
is measured --- more high  
frequency information

# More on sampling



Two sample rates A. Low sample rate that distorts the original sound wave. B. High sample rate that perfectly reproduces the original sound wave. Image from Adobe Audition Help.

# Guideline for sampling rate

- Turning a sound wave into digital data: you must measure the voltage (pressure) as a function of time. But at what times?
- Sampling rate (in seconds) should be a few times faster than the period (in seconds) of the fastest frequency you would like to be able to measure
- To capture the sharp bends in the signal you need short sampling spacing
- What is the relation between frequency and period?

# Guideline for choosing a digital sampling rate

*Period is 1/frequency*

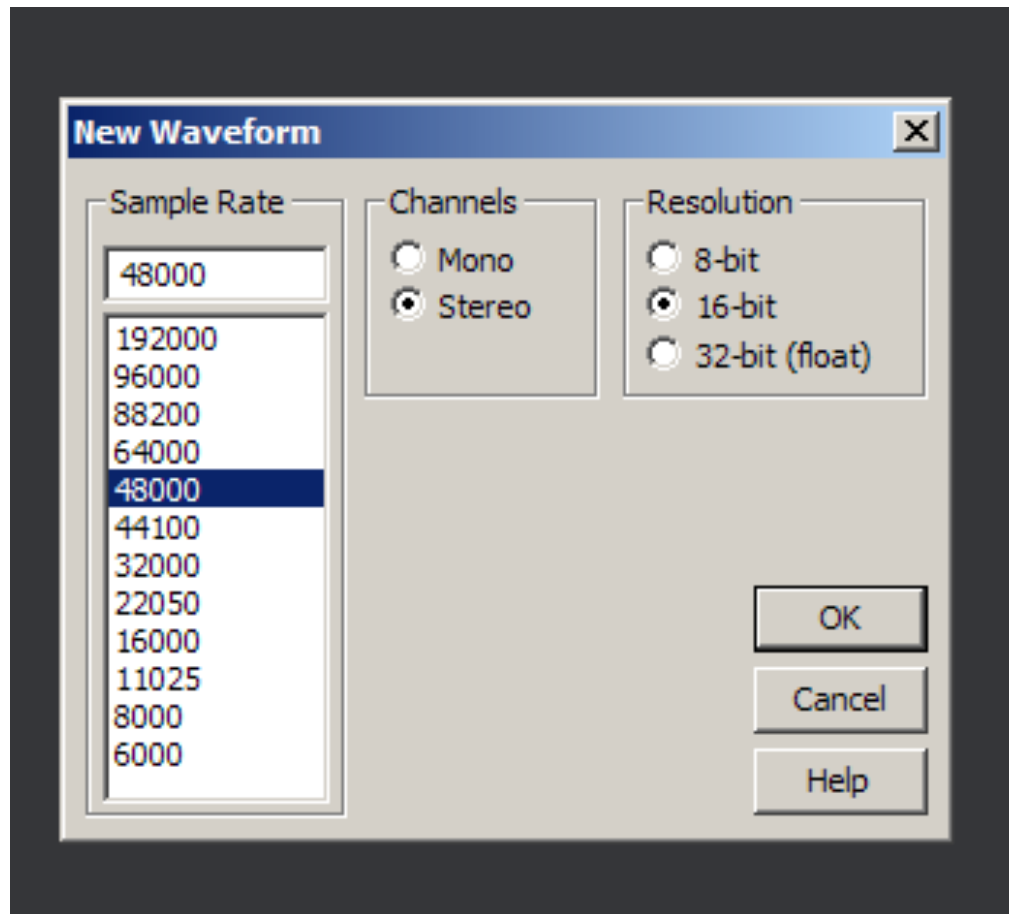
$$P = 1 / f$$

**Sampling rate should be a few times shorter than 1/(maximum frequency) you would like to measure**

For example. If you want to measure up to 10k Hz. The period of this is  $1/10^4$  seconds or 0.1ms.

You would want to sample at a rate a few times less than this or at  $\sim 0.02$ ms.




# Recording in Audition



The most common sample rates for digital audio editing are as follows:

- 11,025 Hz Poor AM Radio Quality/Speech (low-end multimedia)
- 22,050 Hz Near FM Radio Quality (high-end multimedia)
- 32,000 Hz Better than FM Radio Quality (standard broadcast rate)
- 44,100 Hz CD Quality
- 48,000 Hz DAT Quality
- 96,000 Hz DVD Quality

# Demo –degrading sampling and resolution

- Clip of song by Lynda Williams sampling is 48kHz resolution 16 bit 
- 48kHz sampling , 8 bit 
- 11kHz sampling, 16bit 

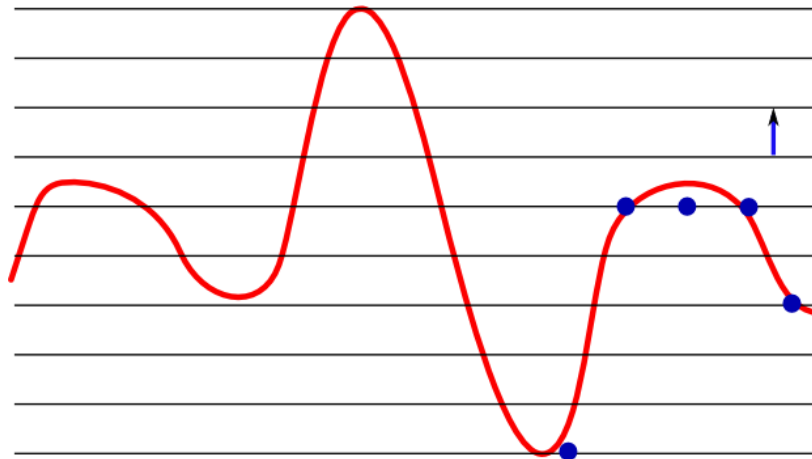
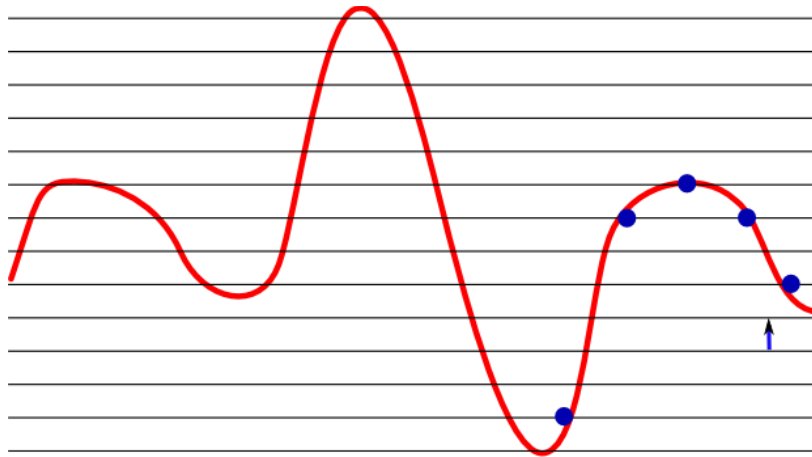
# Bits of measurement

8 bit binary number	00000000b = 0d
	00000001b = 1d
	00000010b = 2d
	00000011b = 3d
	00000100b = 4d
	...
	11111111b = 511d

can describe  $2^8 = 512$  different levels



sampling  
↔



# Bit of precision

Error in amplitude of signal  
loudness error  
error in recording the strength of signal

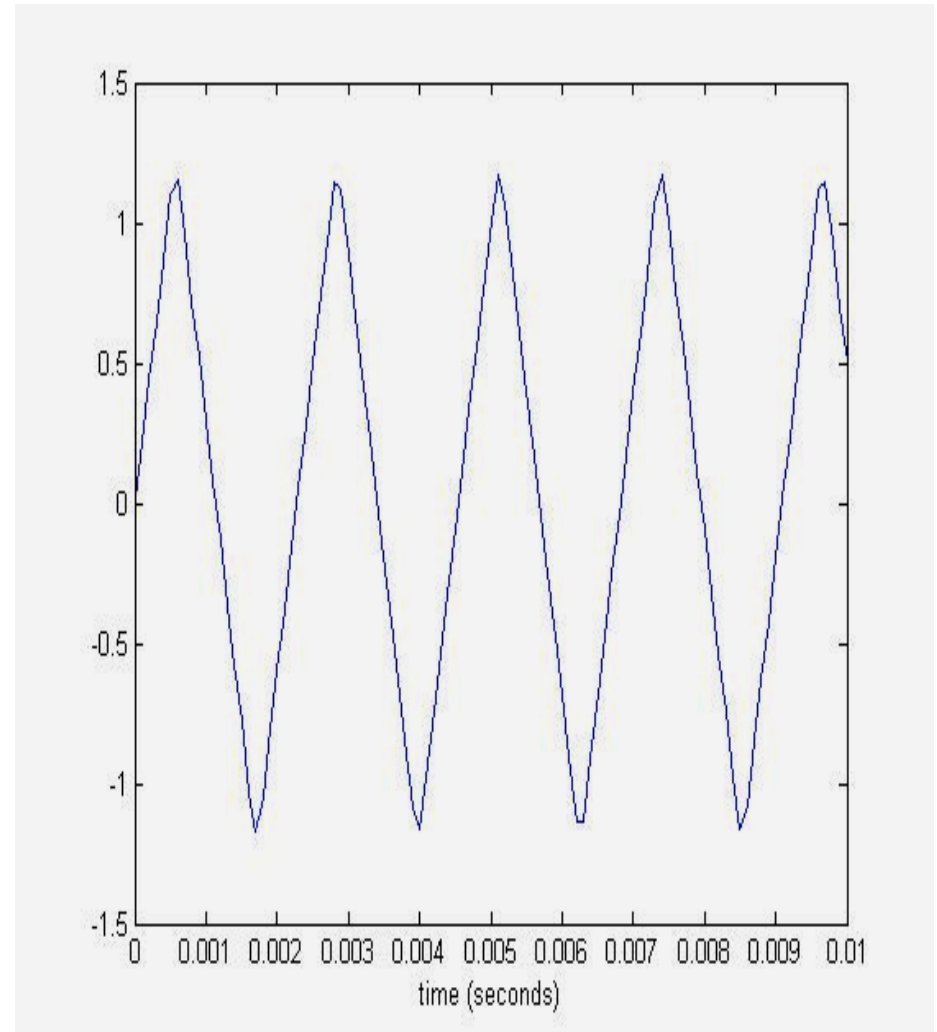
# Bits of measurement

A signal that goes between 0Volt and 1Volt

- 8 bits of information
- You can measure  $1V/512 = 0.002V = 2mV$   
accuracy
- 16bits of information  $2^{16} = 65536$
- $1V/65536 = 0.000015V = 0.015mV =$   
15micro Volt accuracy

# Creating a triangle wave with Matlab using a Fourier series

```
dt = 0.0001; % sampling
time = 0:dt:0.01; % from 0 to 0.01 seconds total
with sampling interval dt
% Here my sample interval is 0.0001sec or a
frequency of 10^4Hz
frequency1 = 440.0; % This should be the note A
% harmonics of this odd ones only
frequency2 = frequency1*3.0;
frequency3 = frequency1*5.0;
frequency4 = frequency1*7.0;
% here are some amplitudes
a1 = 1.0;
a2 = 1.0/9.0;
a3 = 1.0/25.0;
a4 = 1.0/49.0;
% here are some sine waves
y1 = sin(2.0*pi*frequency1*time);
y2 = sin(2.0*pi*frequency2*time);
y3 = sin(2.0*pi*frequency3*time);
y4 = sin(2.0*pi*frequency4*time);
% now let's add some together
y = a1*y1 - a2*y2 + a3*y3 - a4*y4;
plot(time, y); % plot it out
```



# Playing the sound

```
%Modify the file so the second line has  
time = 0:dt:2; %(2 seconds)
```

```
%Last line: play it:
```

```
sound(y, 1/dt)
```

Save it as a .wav file for later

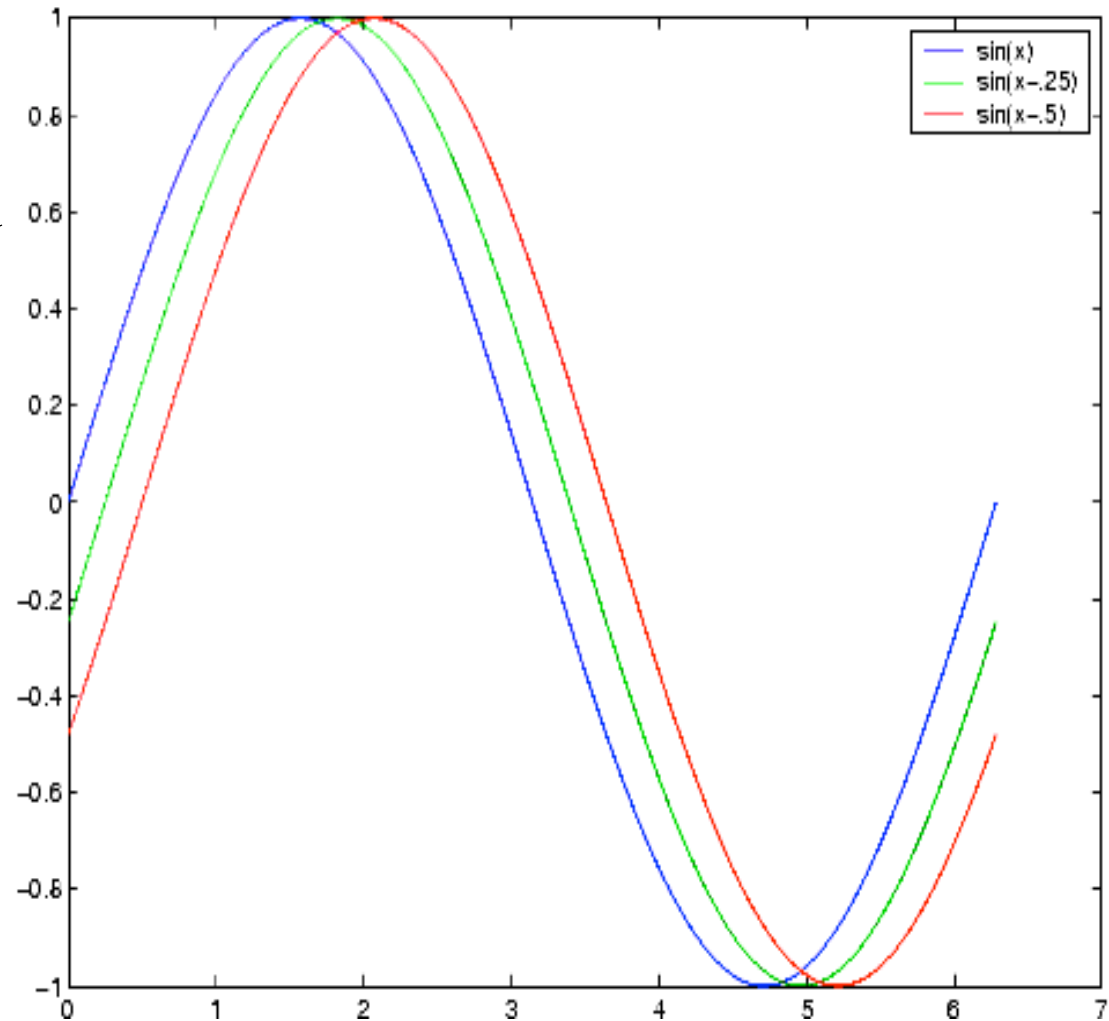
```
wavwrite(0.8*y,1/dt,'triangle.wav')
```

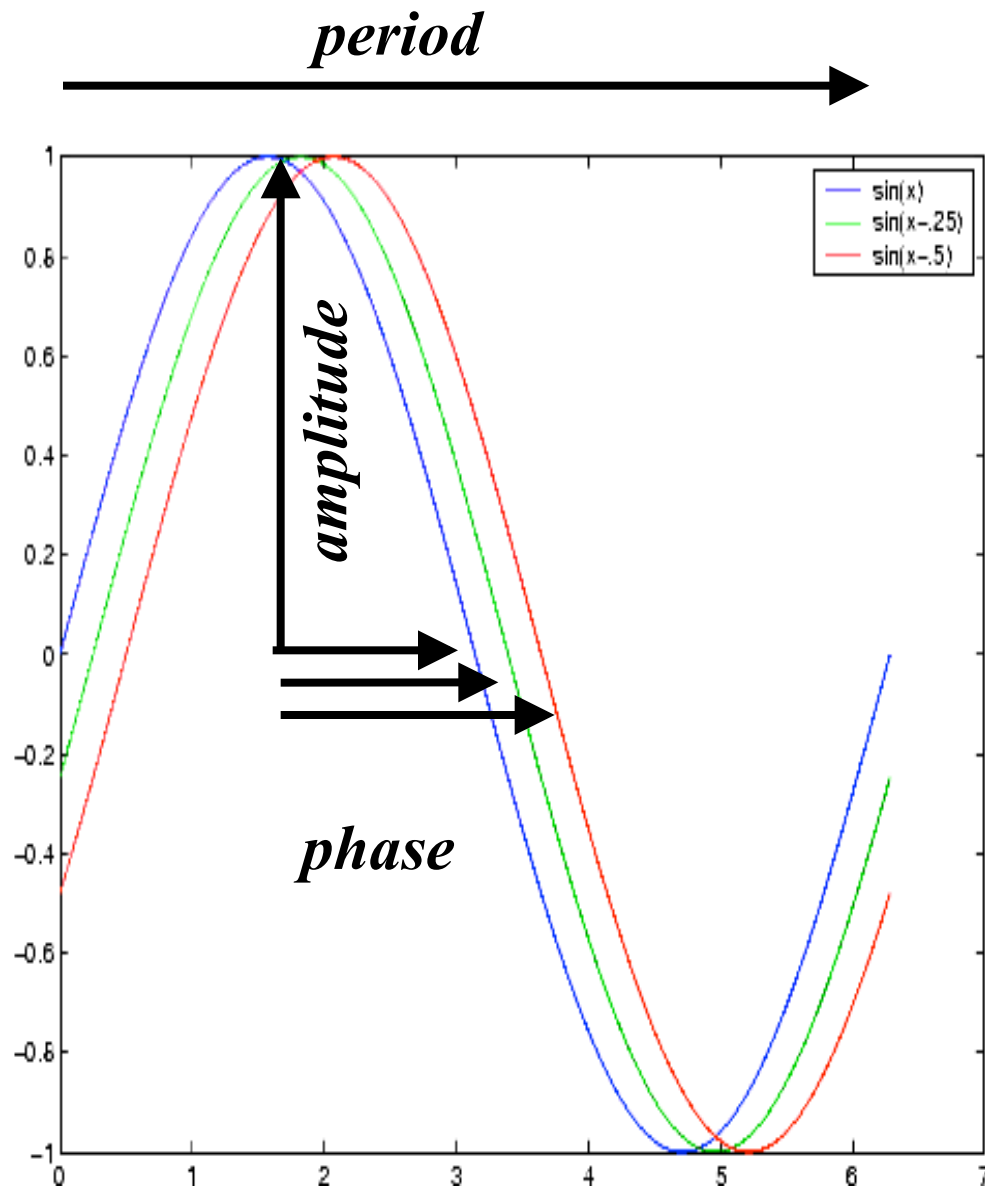


# Phase

Up to this point we have only discussed amplitude and frequency

```
x = 0:pi/100:2*pi;  
y = sin(x);  
y2 = sin(x-.25);  
y3 = sin(x-.5);  
plot(x,y,x,y2,x,y3)
```





## Sine wave

$$A \sin(2\pi f t + \phi)$$

$A$  is amplitude

$f$  is frequency,

$$f = 1/P$$

$\phi$  the phase

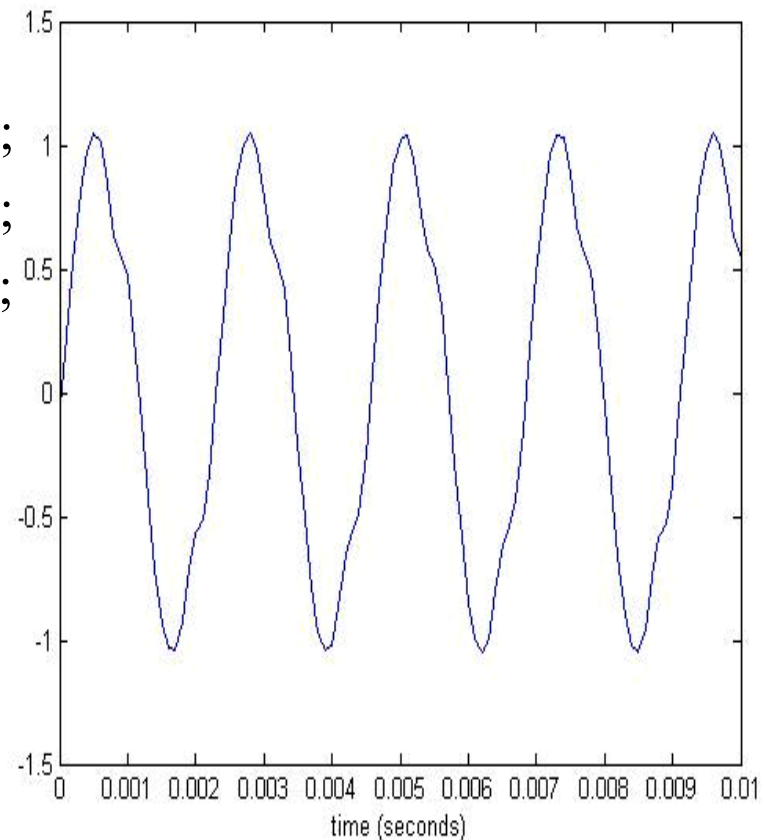
Units:  $f$  Hz

Units:  $\phi$  degrees  
or radians

# What happens if we vary the phase of the components we used to make the triangle wave?

```
y1 = sin(2.0*pi*frequency1*time);  
y2 = sin(2.0*pi*frequency2*time - 1.6);  
y3 = sin(2.0*pi*frequency3*time - 0.1);  
y4 = sin(2.0*pi*frequency4*time + 1.3);  
y = a1*y1 + a2*y2 + a3*y3 + a4*y4;
```

Shape of wave is changed even though frequency spectrum is the same



# Is there a difference in the sound?



These two are sums with the same amplitude sine waves components, however the **phases** of the sine waves differ.



# Another example

This sound file has varying phases of its frequencies.

Do we hear any difference in time?



Sound file from

<http://webphysics.davidson.edu/faculty/dmb/py115/MusTechS05.htm>

ohmslaw1.wav - Adobe Audition

Edit View Effects Generate Analyze Favorites Options Window Help

# Spectrum of this sound

Files Effects Favorites

ohmslaw1.wav  
Untitled \*

Auto [Volume Icon] Preview Volume

Show File Types: Sort By: Recent Accents Show Cues Full Paths

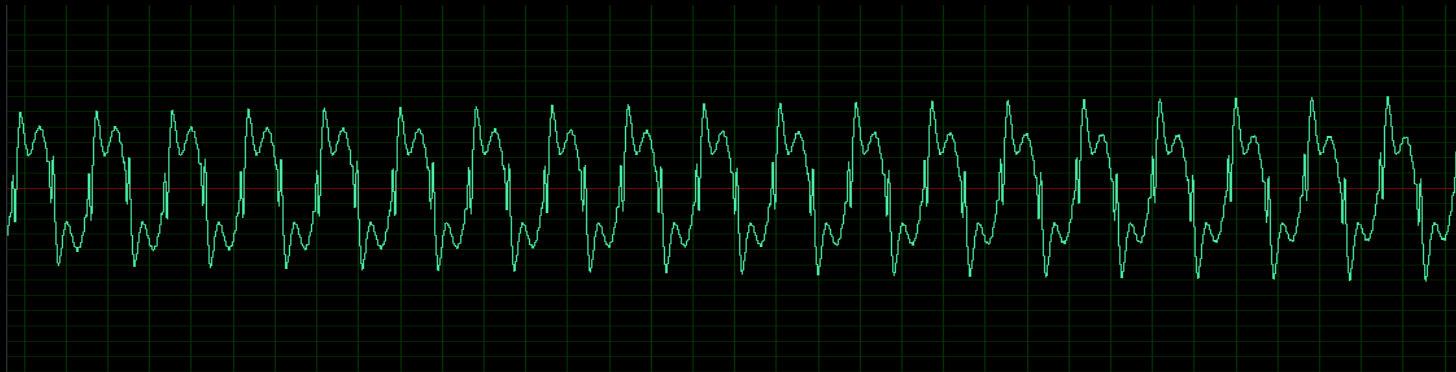
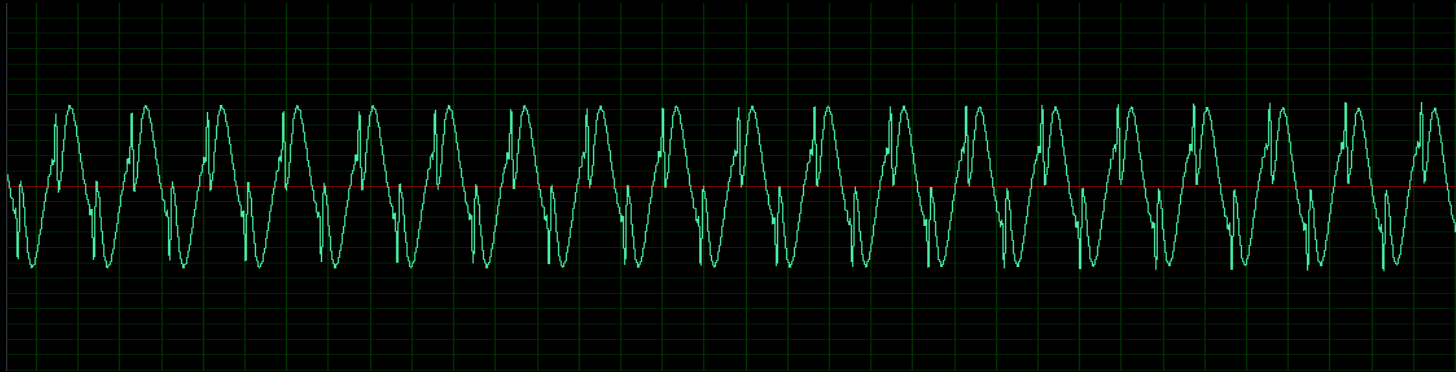
0:02.500

	Begin	End	Length
Sel	0:02.500		0:00.000
View	0:00.938	0:05.679	0:04.740

821.4Hz @ 0:02.676 44100 • 16-bit • Mono 688 K 17.83 GB free

Start [Taskbar Icons] Lectures Overlap Proces... Command Prompt ohmslaw1.wa... Cygwin Microsoft Powe... 1:51 PM

# Waveform views at different times



# Do we hear phase?

Helmholtz and Ohm argued that our ear and brain are only sensitive to the frequencies of sounds.

Timbre is a result of frequency mix.

There are exceptions to this (e.g., low frequencies)

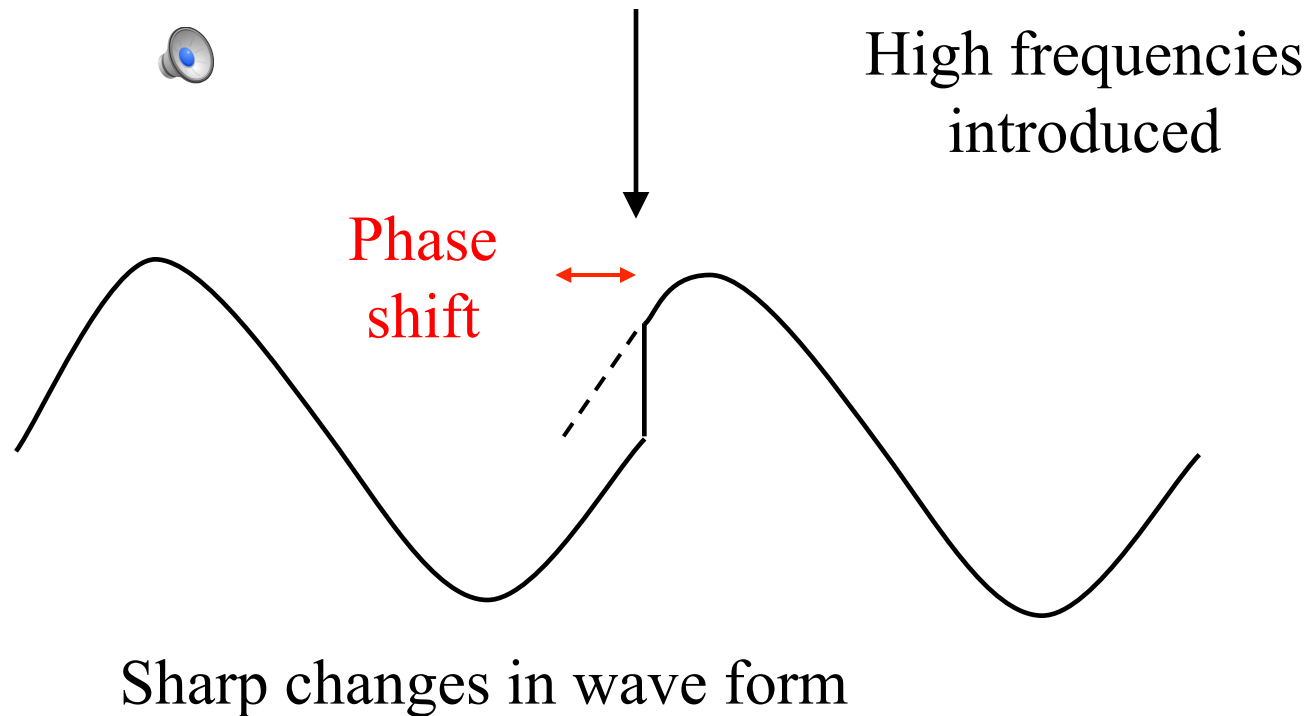
Two major psycho-acoustic models

1) Place theory – each spot in basal membrane is sensitive to a different frequency

2) Timing – rate of firing of neurons is important and gives us phase information

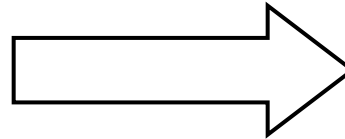
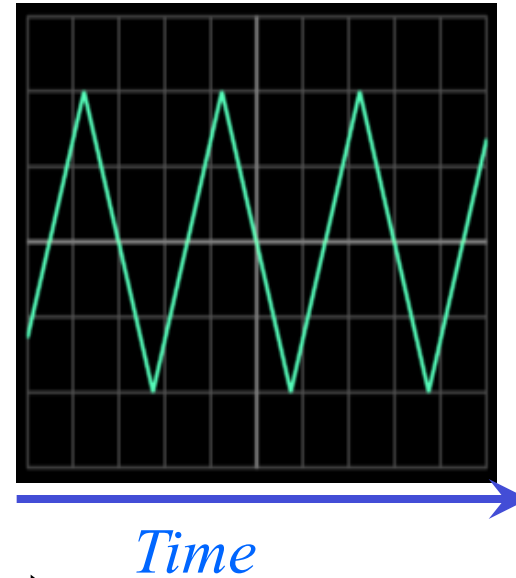
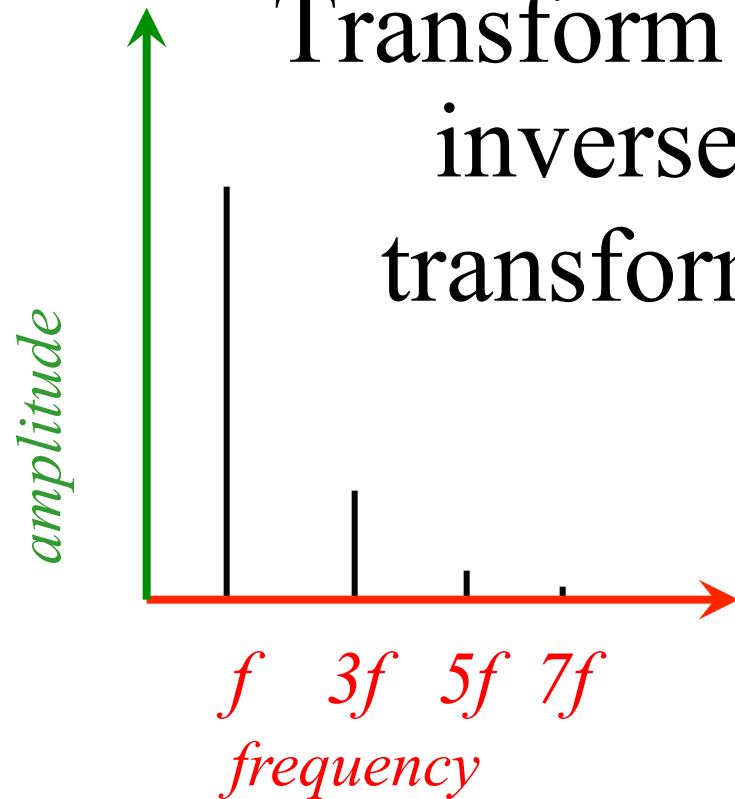
What is the role of each in how our ear and brains process information? Open questions remain on this.

# Cutting and pasting audio



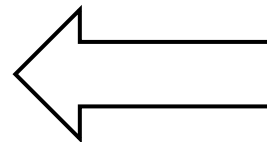
Demo with a cut and paste in Audition/Audacity of a generated sine. Note: the effect in spectral view depends on the length of the FFT used, also you need to be fairly zoomed out horizontal to see the noise.

# Transform and inverse transform

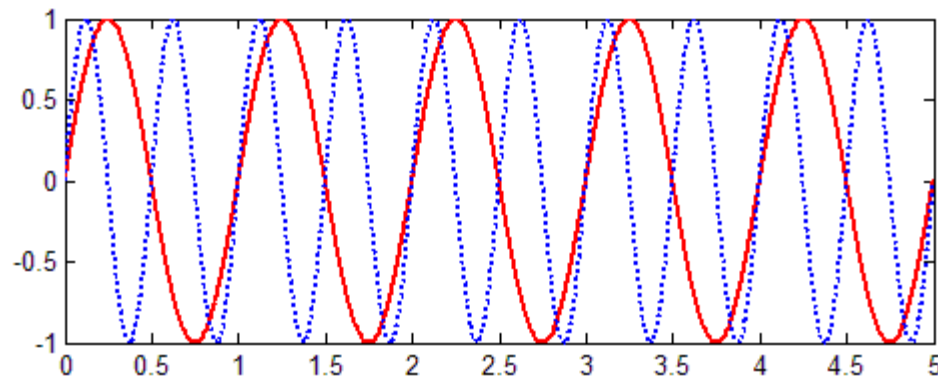


I have shown how to go this way

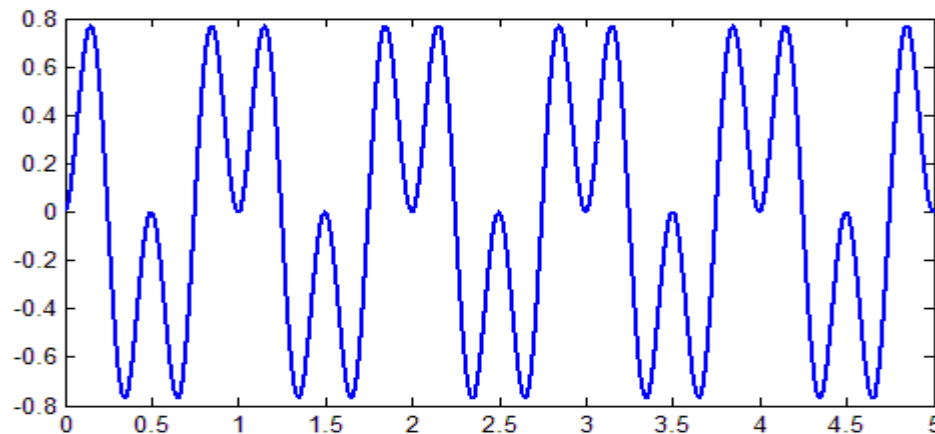
How we will talk about how to take a  
signal and estimate the strength of its  
frequency components



# Multiplying two cosines with different frequencies

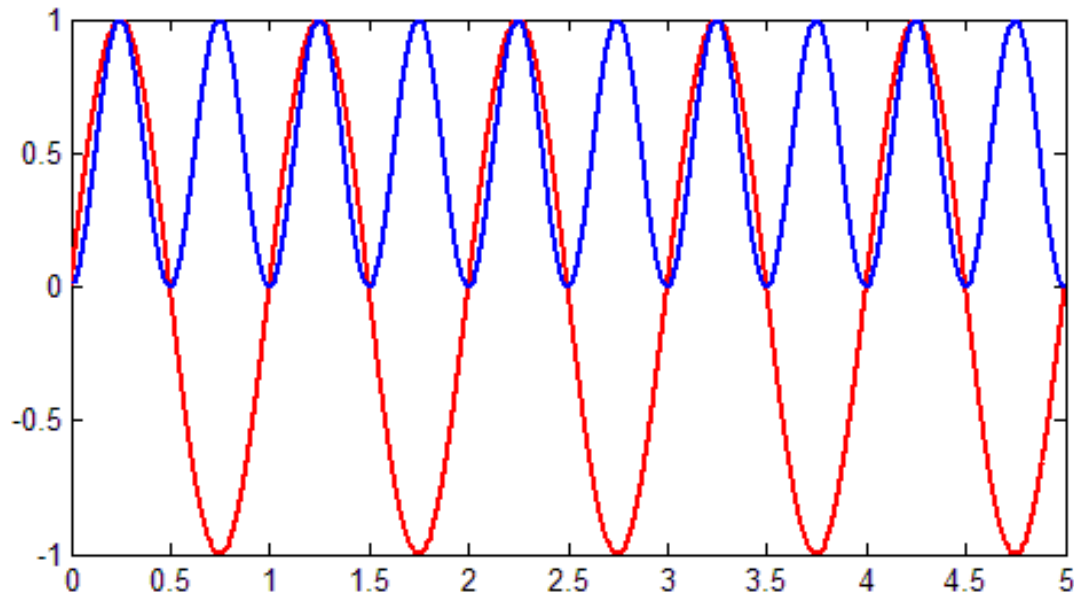


$\cos(2\pi t)$  and  
 $\cos(4\pi t)$



Multiplied together  
Average to zero

# Multiplying two cosines with the same frequency

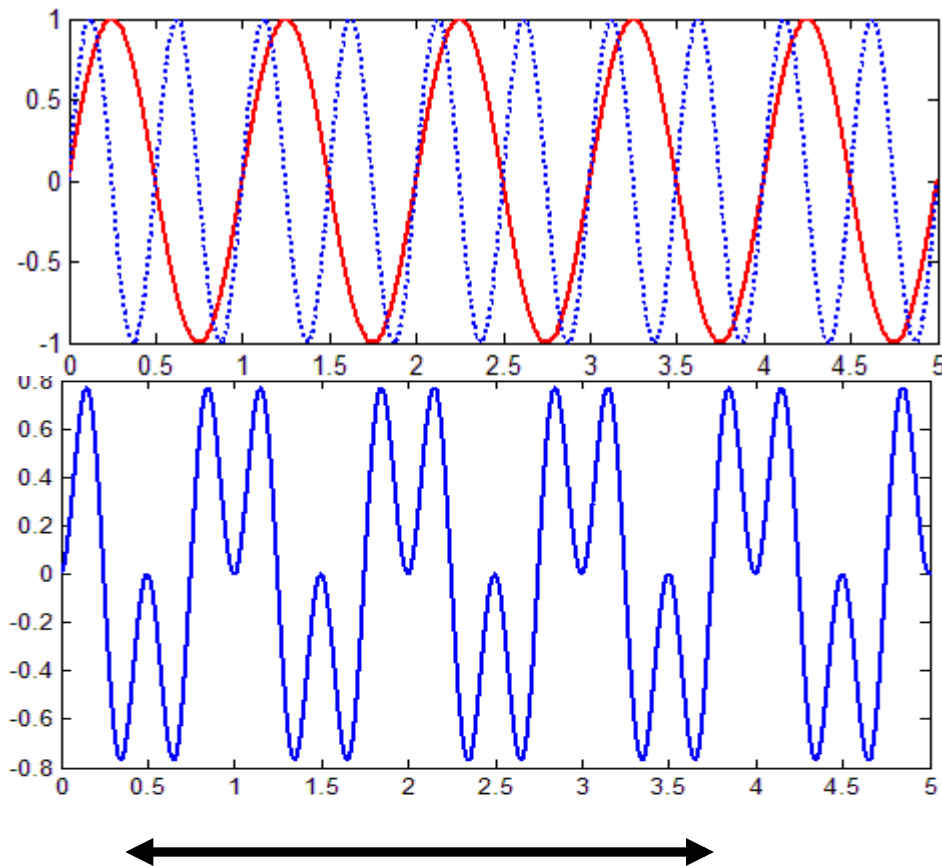


The  
average is  
not zero.

The  
average is  
 $1/2$



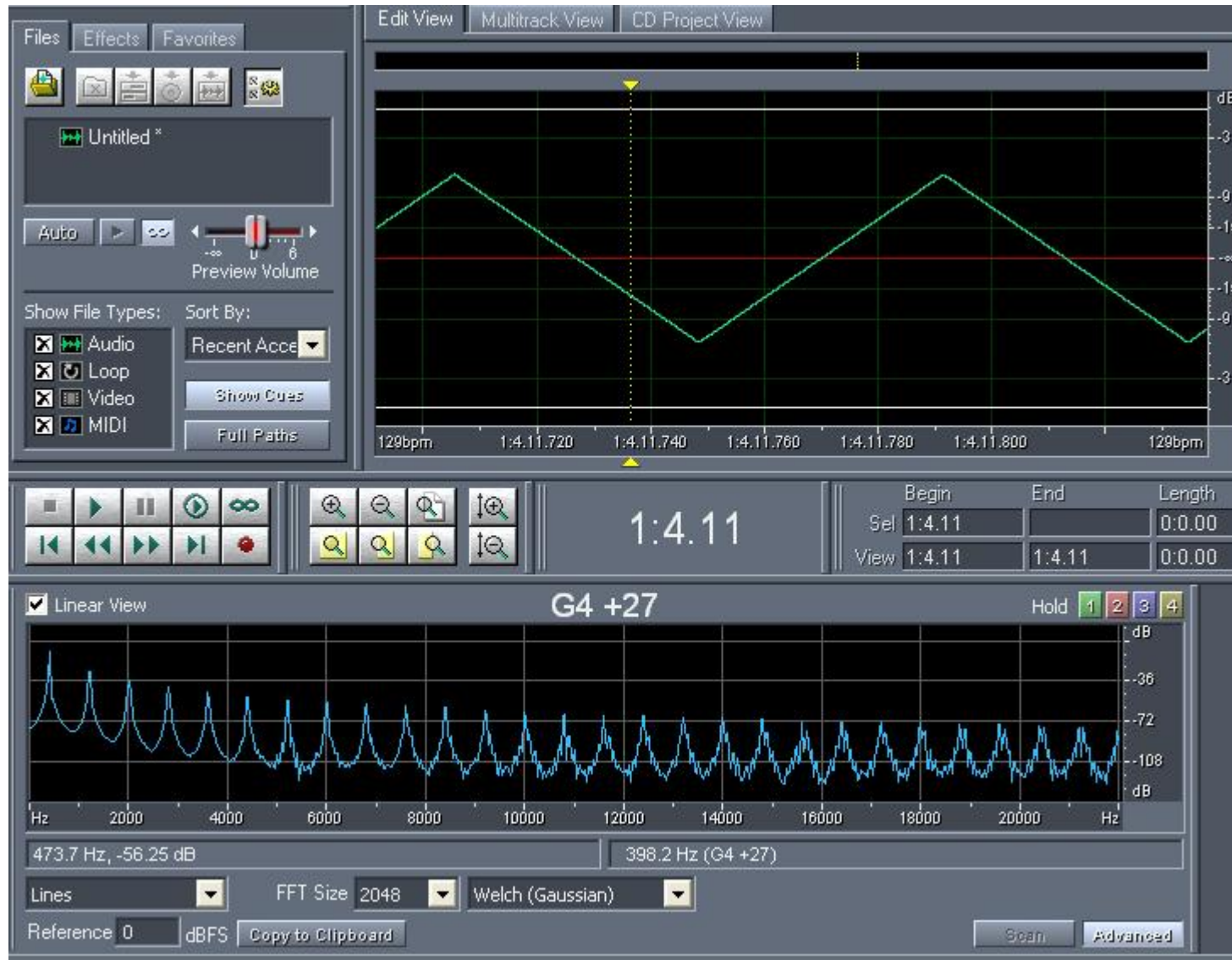
# Multiplying two cosines with different frequencies



$\cos(2\pi t)$  and  
 $\cos(4\pi t)$

What if your  
window fits  
here?

# Windowing and errors



# Calculating the amplitude of each Fourier component

What is the average of

$$\cos(2\pi ft) \cos(2\pi gt)?$$

Over a long interval this averages to zero  
unless

$$f=g$$

Sine/Cosine functions are “orthogonal”

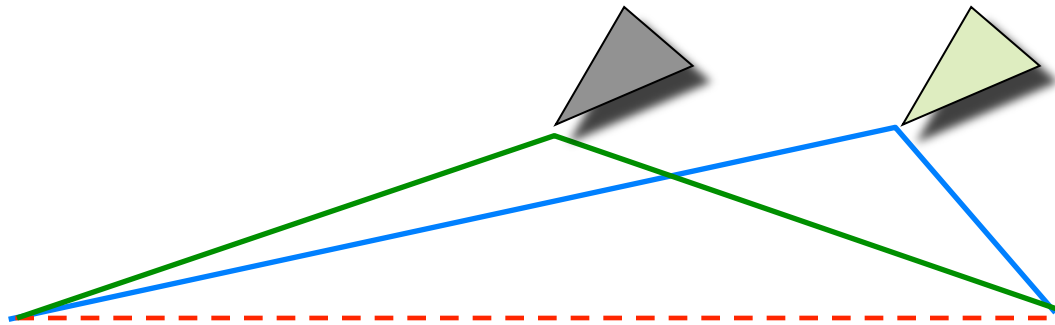
# Calculating the amplitude of each Fourier component

$$f(t) = \sum_m [A_m \sin(2\pi m f t) + B_m \cos(2\pi m f t)]$$

- Procedure: multiply the waveform  $f(t)$  by a cosine or sine and take the average.
- Multiply by 2. This gives you the coefficient  $A_m$  or  $B_m$ .

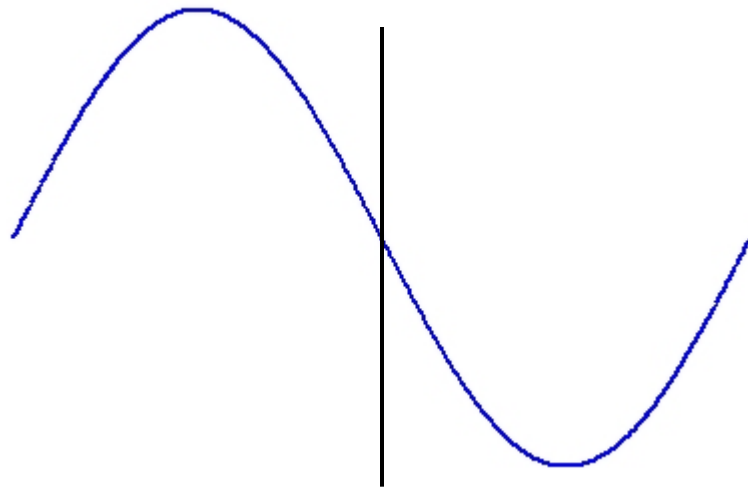
# Predicting the spectrum of a plucked string

- Can one predict the amplitude of each mode (overtone/harmonic?) following plucking?



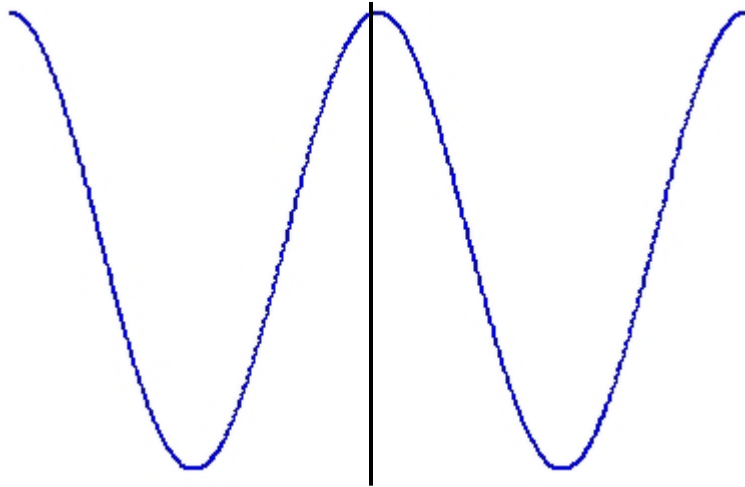
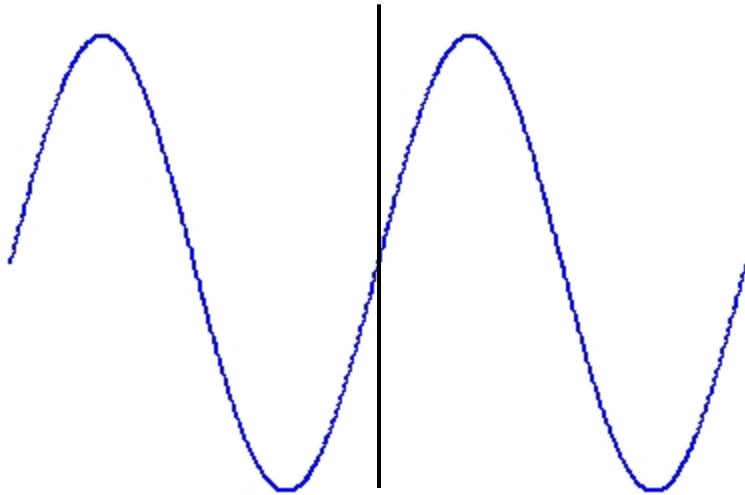
- Which pluck will contain only odd harmonics?
- Which pluck has stronger higher harmonics?

# Odd vs Even Harmonics and Symmetry

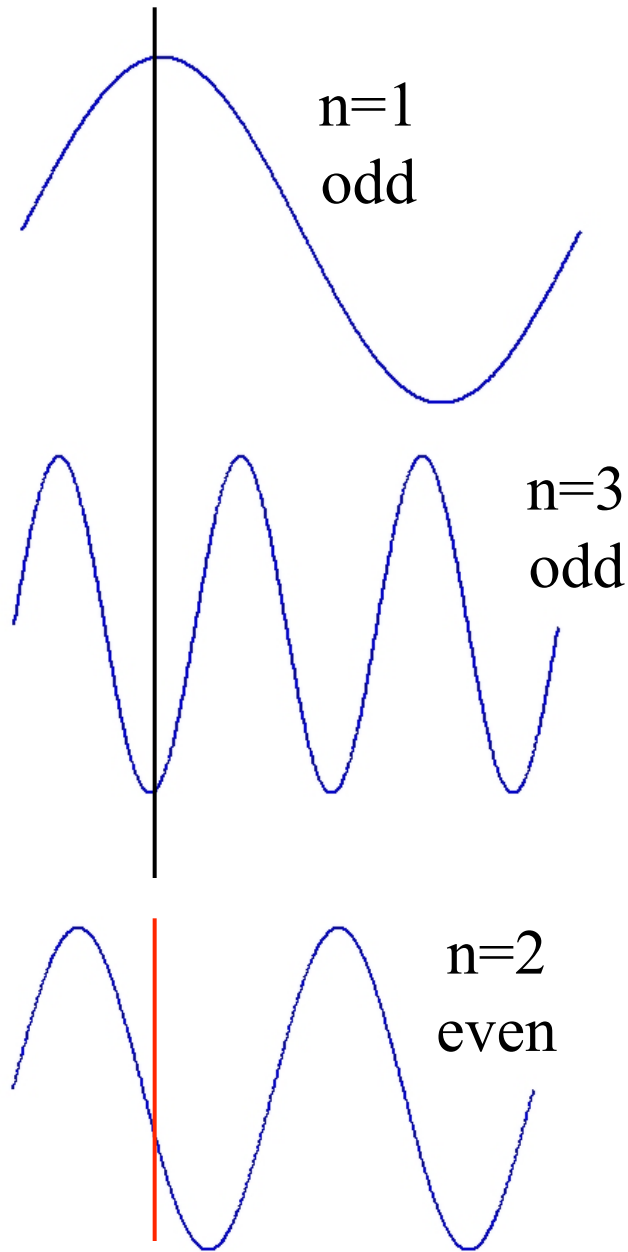


- Sines are Anti-symmetric about mid-point
- If you mirror around the middle you get the same shape but upside down

# More on Symmetry



- Sines are anti-symmetric
- Cosines are symmetric



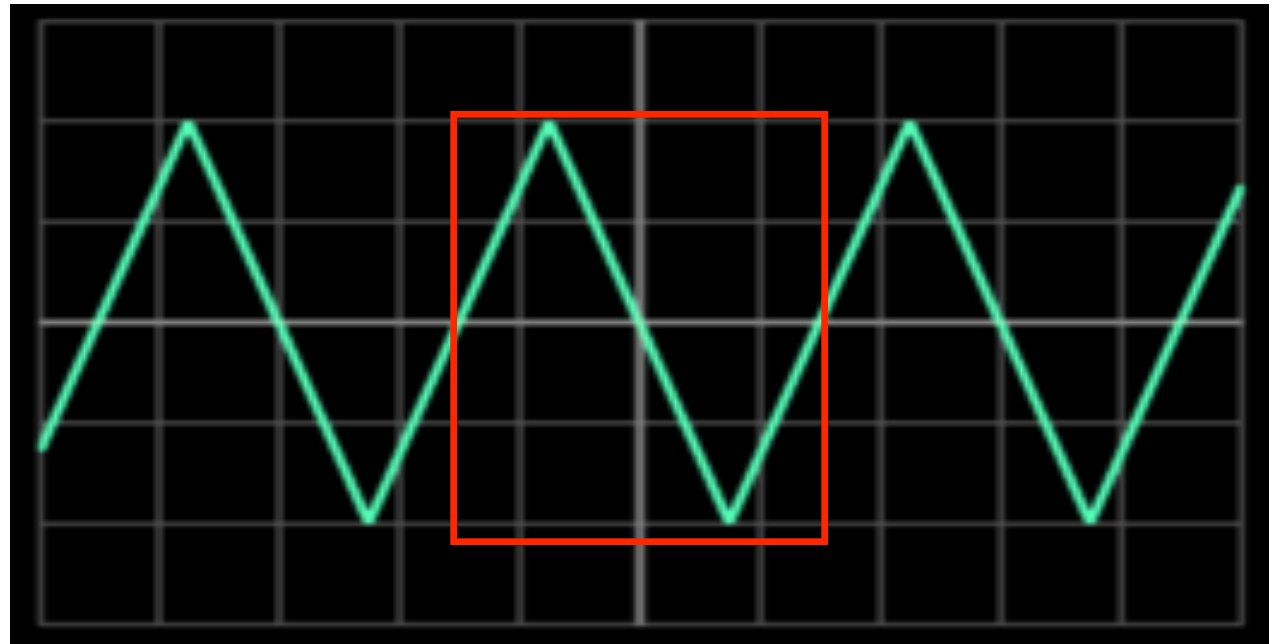
# Symmetry

- Additional symmetry of odd sines if you consider reflection at the black line.
- About this line, Odd harmonics are symmetric but even ones are anti-symmetric



# Symmetry of the triangle wave

Obeys same symmetry as the odd harmonics so cannot contain even harmonic components



# Odd Fourier components

Both triangle waves and square waves contain **odd** Fourier components.

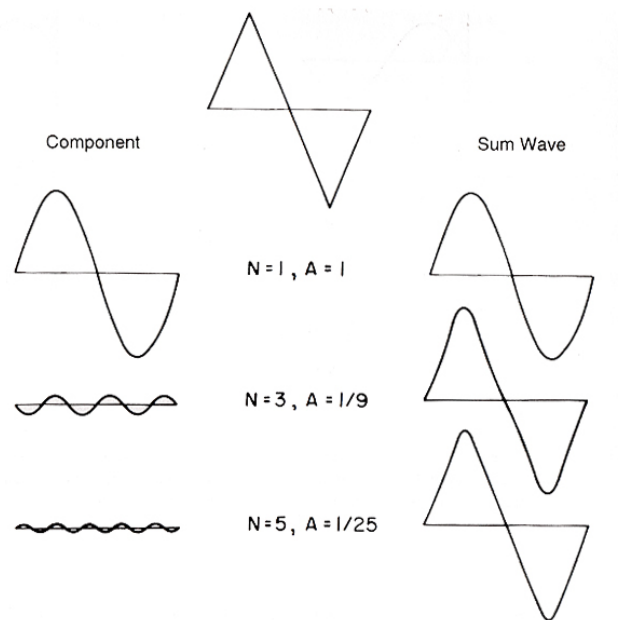
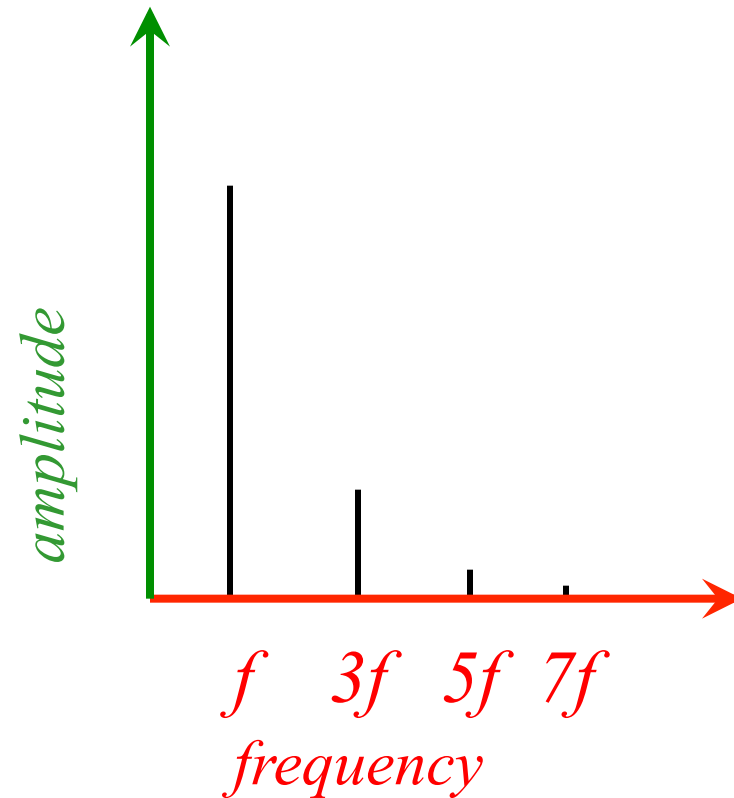
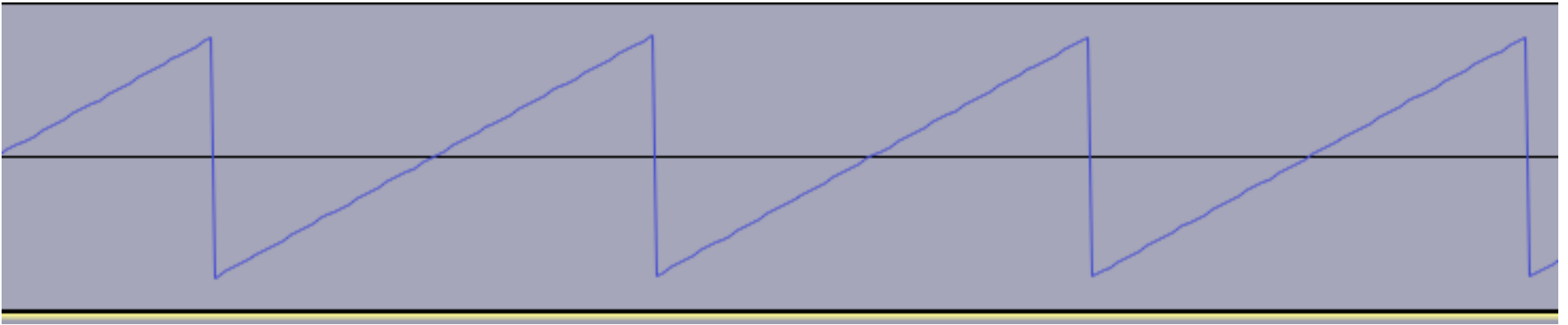


Figure 4-4 Fourier synthesis of a triangular wave. At the left are the successive harmonics; at the right are the sum waves including each successive harmonic. The graph at the top is the wave being synthesized.



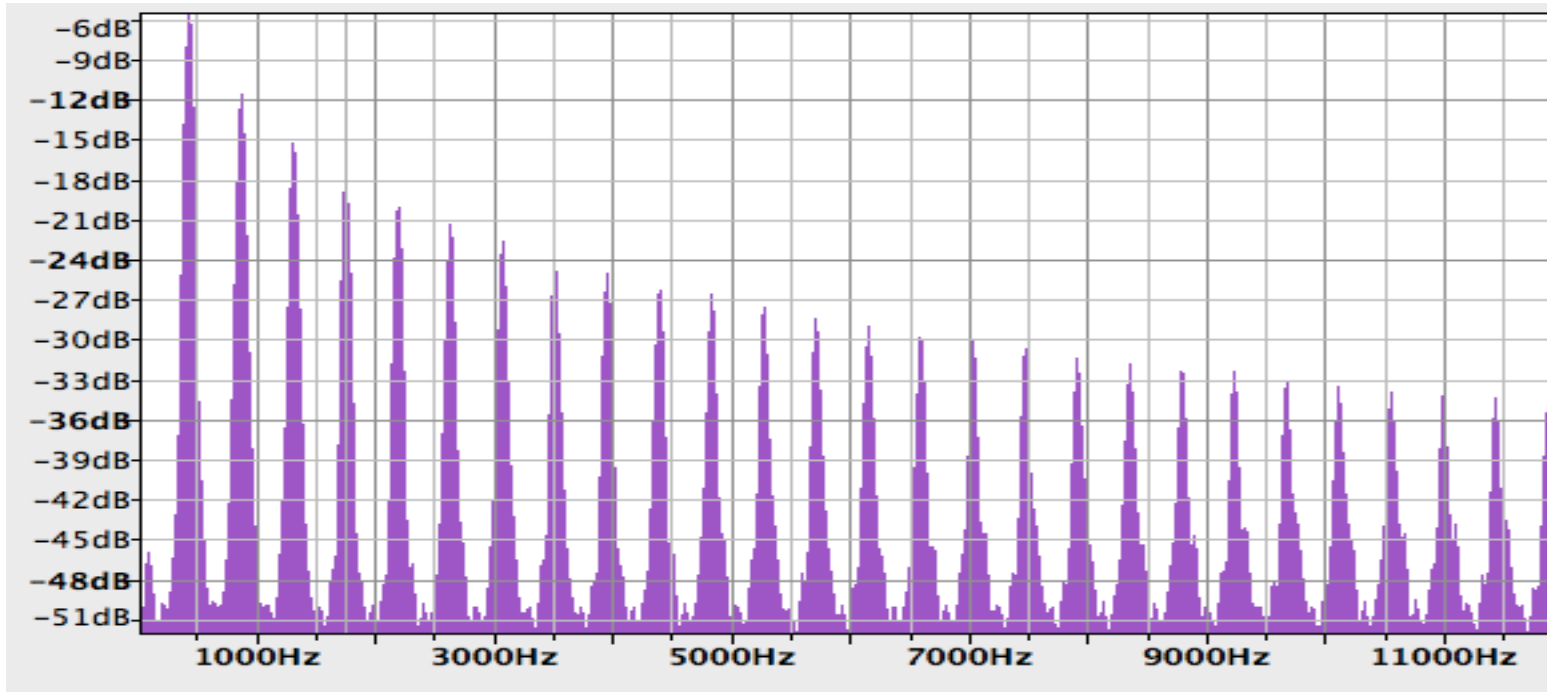
# Sawtooth



- What overtones are present in this wave?  
Use its symmetry to guess the answer.

# Spectrum of sawtooth

All integer harmonics are present. The additional symmetry about the  $\frac{1}{4}$  wave that both triangle and square wave have is not present in the sawtooth.

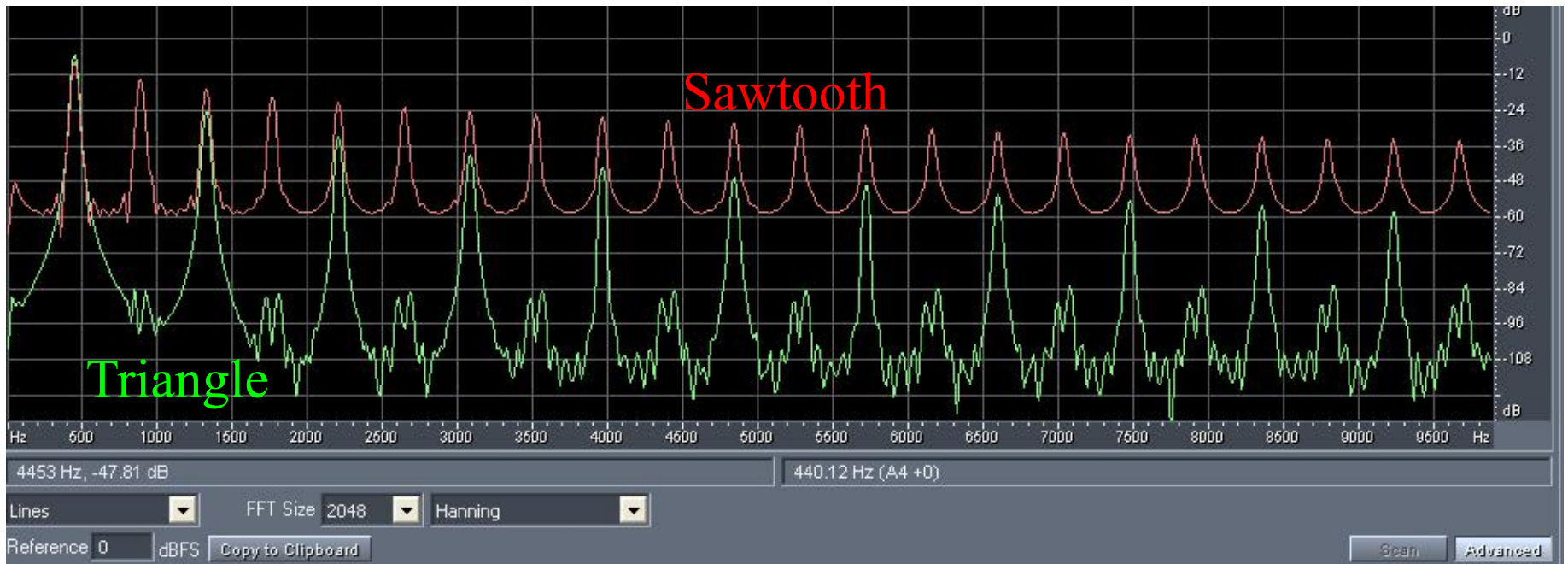


# Generated tones

Order of 440Hz tones:



Sine, Triangle, Sawtooth, Square, Rectangular  
with 10%/90%



# Symmetry as a compositional element

- From Larry Solomon’s “Symmetry as a compositional element” -- last phrase of Bartok’s Music for Strings, Percussion and Celesta, movement I

The image shows a musical score for a string quartet, specifically the last phrase of Bartok's Music for Strings, Percussion and Celesta, movement I. The score is written in 6/8 time and begins with a *ppp* dynamic marking. The melody is characterized by reflection symmetry in tones, with an axis of symmetry centered on the letter A. The score is divided into two halves by a vertical line, with a curved line above it indicating symmetry. A speaker icon is visible below the score.

- Reflection symmetry in tones --- axis of symmetry is an A
- microcosmos vol 6 141 Free variations

A1

Sehr mäßig ♩. = ca 40

Anton Webern, Op. 27

The image displays four systems of musical notation for a piano piece. Each system consists of a treble clef staff and a bass clef staff. The notation includes notes, rests, and various musical symbols. Annotations include circled letters and numbers: P<sub>0</sub>, R<sub>0</sub>, I<sub>2</sub>, and R<sub>0</sub>. Dynamic markings such as *pp*, *p*, *f*, *dim.*, and *rit.* are present. A double-headed arrow with a vertical line through it is used to indicate reflection in time. Measure numbers 1 through 18 are marked throughout the score.

Reflection in time

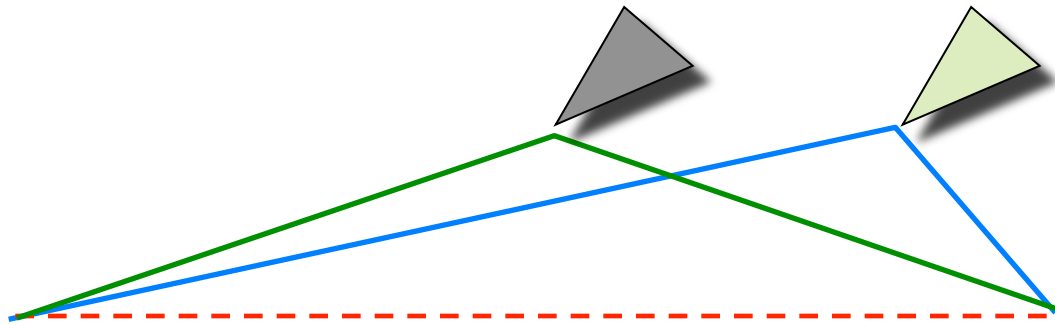
Axis of symmetry is a time



(Example from Larry Solomon) Anton Webern, Opus 27

# Predicting the spectrum of a plucked string

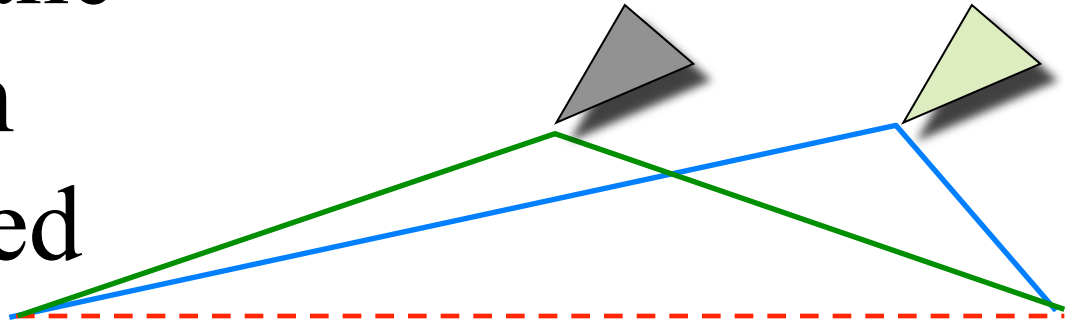
- Can one predict the amplitude of each mode (overtone/harmonic?) following plucking?



- Using the procedure to measure the Fourier coefficients it is possible to predict the amplitude of each harmonic tone.



# Predicting the spectrum of a plucked string



- You know the shape just before it is plucked.
- You know that each mode moves at its own frequency
- The shape when released  $f(x, t = 0)$
- We rewrite this as

$$f(x, t = 0) = \sum_{m=0}^{\infty} A_m \sin \frac{2\pi m x}{2L}$$

# Predicting the motion of a plucked string (continued)

Each harmonic has its own frequency of oscillation, the  $m$ -th harmonic moves at a frequency  $f_m = mf_0$  or  $m$  times that of the fundamental mode.

$$f(x, t = 0) = \sum_{m=0}^{\infty} A_m \sin \frac{2\pi mx}{2L} \quad \text{initial condition}$$

$$f(x, t) = \sum_{m=0}^{\infty} A_m \sin \frac{2\pi mx}{2L} \cos 2\pi mf_0 t$$

# Moving string in general

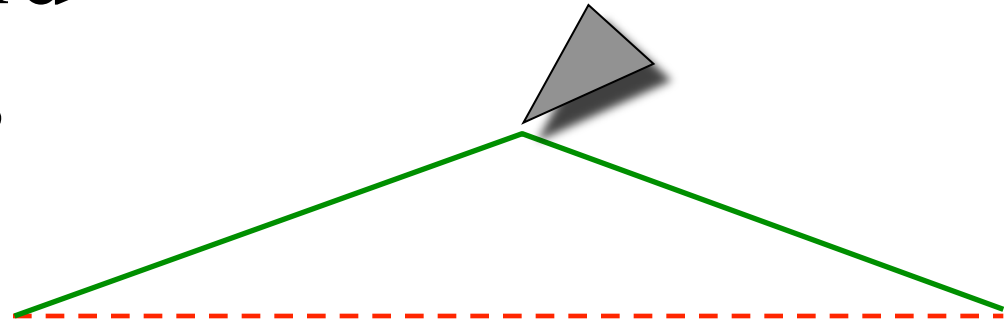
$$f(x, t) = \sum_{m=0}^{\infty} A_m \sin \frac{2\pi m x}{2L} \cos 2\pi m f_0 t$$

Does this make sense? Some checks:

Are left and right boundaries fixed?

Is the string not moving at  $t=0$ ?

# Sum of forward + backwards travelling waves



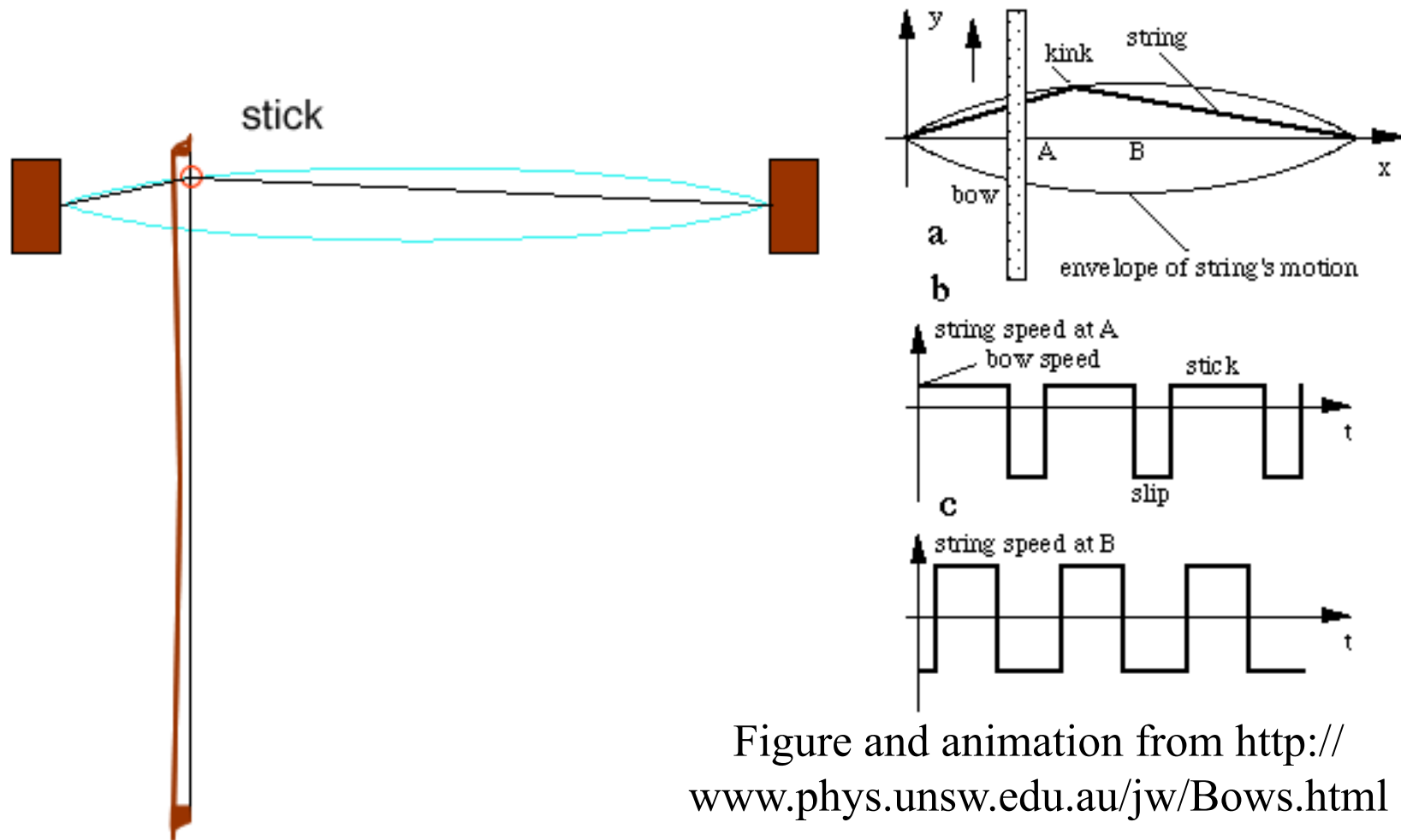
Initial condition given above, and the velocity everywhere is zero.

This is equal to the sum of two traveling waves



Shape of wave form can be predicted at future times by considering each traveling wave and how it reflects off of the boundaries

# Violin and stick slip motion



# Iphone films

- <http://www.wired.com/gadgetlab/2011/07/iphones-rolling-shutter-captures-amazing-slo-mo-guitar-string-vibrations/>

Each line scanned at a different time

The “rolling shutter”

Between 24 and 30fps.

1280 x 720 pixels

At fastest 0.033s per frame If I

divide by 1/1000 then ~ 30

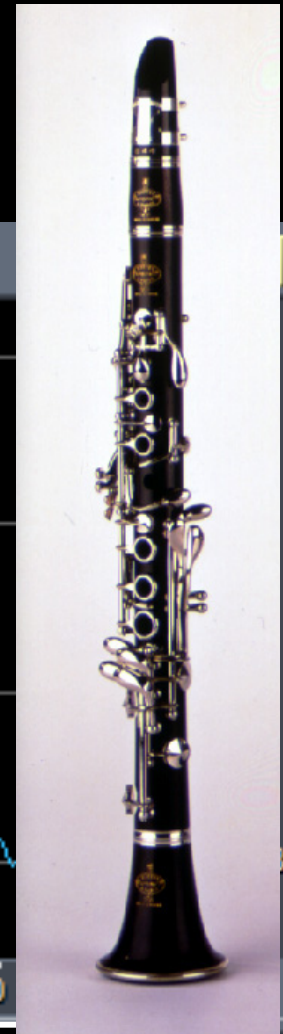
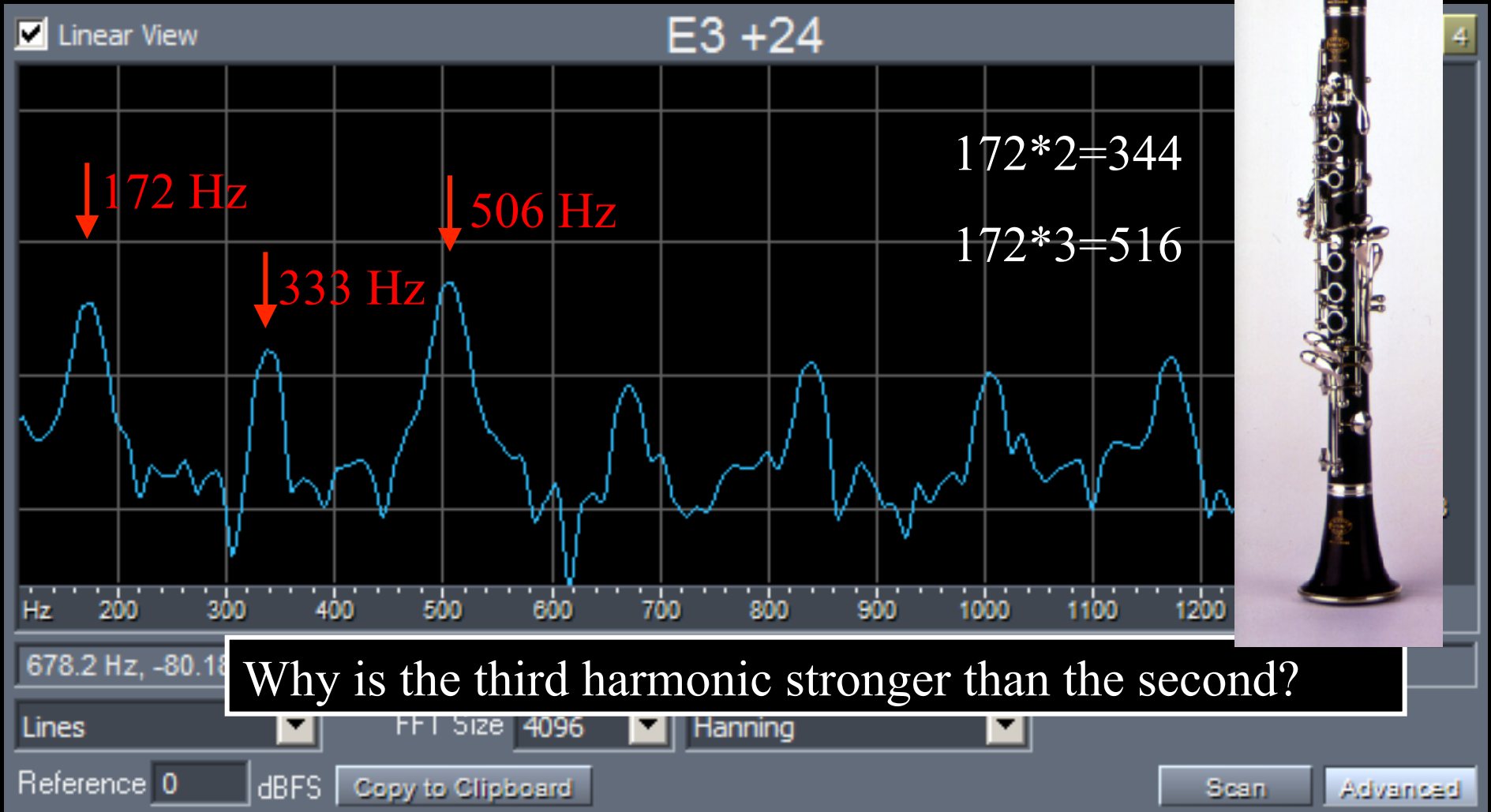
microseconds delay between lines



# Guitar string

- Length of string,  $L$ , is about a meter, frequency of lowest string is 82Hz,  $P=0.012s$
- Speed on the string  
 $v/(2L) = f \rightarrow v = 2Lf \sim 160 \text{ m/s}$
- The delay between lines is 30 microseconds corresponding to a distance of  $160\text{m/s} \times 30 \text{ microseconds} = 5e-3\text{m} = 0.5\text{cm}$
- Number of lines to get there and back travel times  $0.012/33e-6=400$  (half the picture) as expected
- Maybe could do this calculation more efficiently by considering what fraction of wavelength fits in view of camera, giving phase information

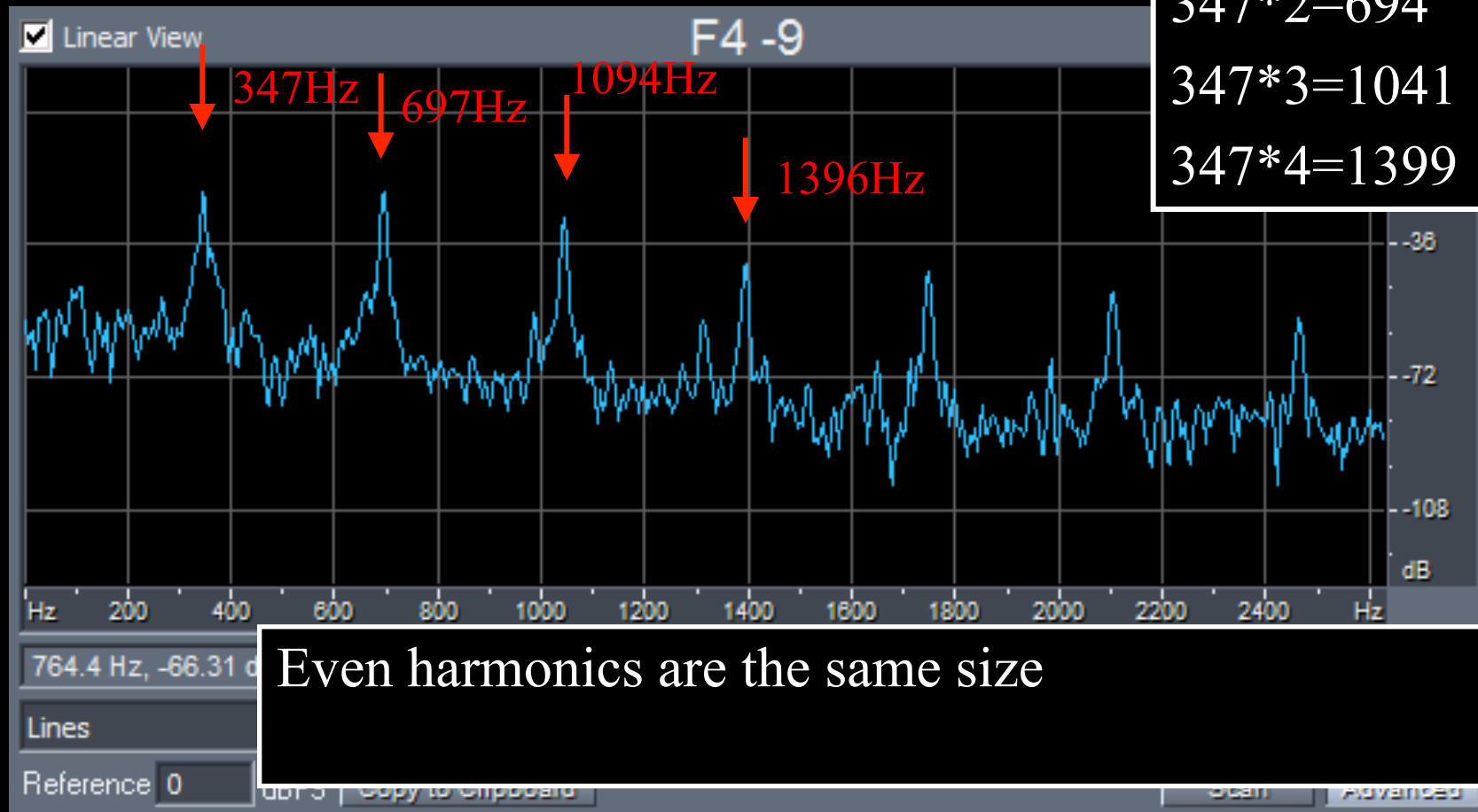
# Clarinet spectrum



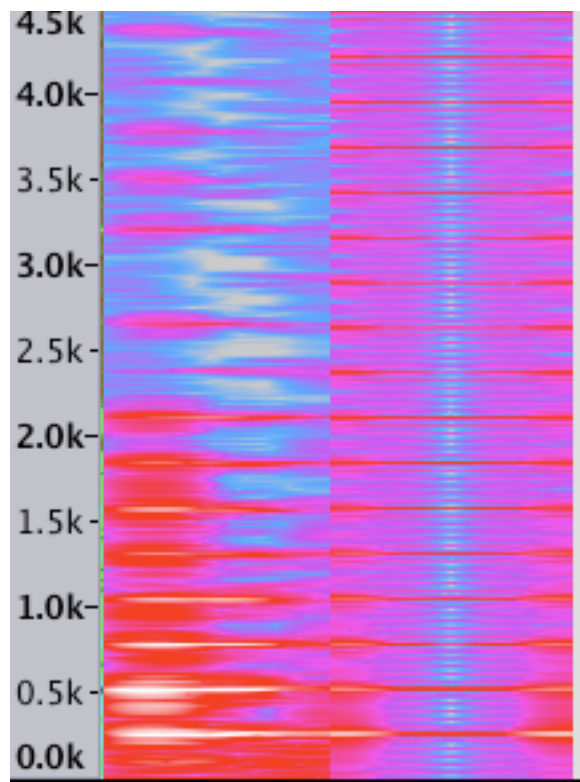
Why is the third harmonic stronger than the second?



# Piano spectrum



# Piano spectrum



- C4 piano on left, sawtooth at same frequency on right.
- High overtones are higher in piano.
- Why?

# Are these frequency shifts important?

Butler (example 2.4).



a) Piano playing C4

b) Piano playing C4 but the partials have been lowered by digital processing so that their frequencies are exact integer multiples of the fundamental.

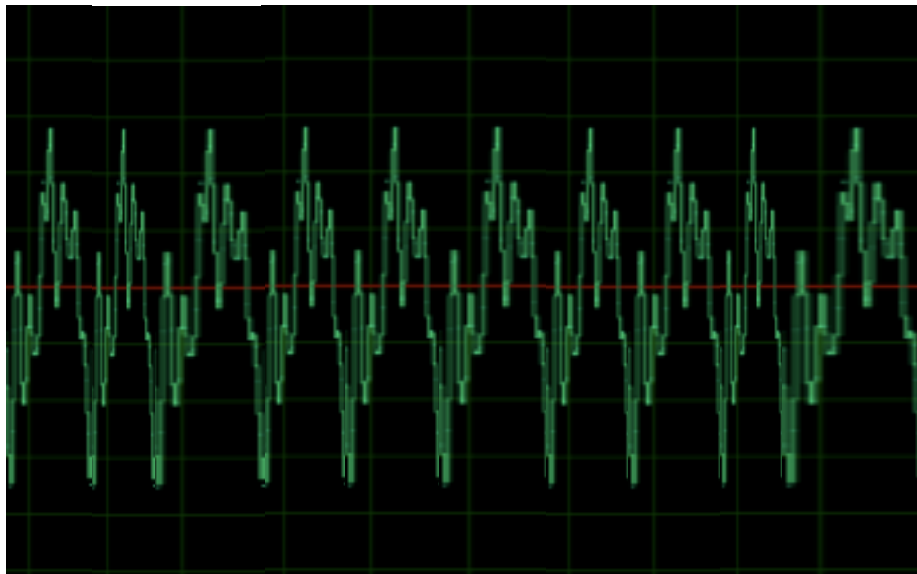
Pair of tones repeated 3 times.

# Synthesized voicing

- Voice and many instruments make a nearly periodic signal
- Overtones are all integer multiples of each other
- Frequencies are fixed
- However if a tone is synthesized to have exact integer overtones and fixed frequencies it sounds electronic
- How do you synthesize more realistic tones?

# Irregularities are important

- Slight frequency shifts
- Slight timing differences in the periodic waveform



Timing differences from turbulence in throat and other sources.

If there is no irregularity then the tones are unnatural and dull.

# Synthesized singer

Cookdemo70

- a. No vibrato
- b. Random and periodic vibrato and singer scooping slightly upward at beginning of each note



# Nearly Periodic Waveforms

- Voice, guitar, flute, horn, didgeridu, piano: all have ladder spectrum

Why nearly periodic signals?

- Stringed instruments. Modes of vibration have frequencies that are integer multiples of a fundamental tone. All modes are excited by plucking. Harmonics are modes.
- Wind instruments. Mode frequencies are close to integer multiples of a fundamental. Excitation builds on one mode. Excitation (mouth) is nearly periodic. Resulting sound contains harmonics. The harmonics may not be modes. Sometimes other modes can be seen in the sound spectrum that are not harmonics.
- Voice. Excitation is nearly periodic. Tract resonances give formants, but not key toward driving sound. Emerging sound since nearly periodic contains harmonics.

Not all musical sounds are nearly periodic in nature

# Some history

Images and information from [http://physics.kenyon.edu/EarlyApparatus/Rudolf\\_Koenig\\_Apparatus/Helmholtz\\_Resonator/Helmholtz\\_Resonator.html](http://physics.kenyon.edu/EarlyApparatus/Rudolf_Koenig_Apparatus/Helmholtz_Resonator/Helmholtz_Resonator.html)



Earliest sound spectra taken by Helmholtz ~1860 who used glass spheres or cylinders, each with a difference size and hole diameter setting its resonant frequency. The opposite side would have a slender opening that could be held in the ear. The enclosed volume of air acts as a spring connected to the mass of the slug of air, and vibrates in an adiabatic fashion at a frequency dependent on the density and volume of the air, its molecular composition, and the mass of the slug of air in the neck.



- Sets of these were built and ordered by universities to allow spectra of sounds to be measured in the lab
- This very large set of twenty two Helmholtz resonators is in the Garland Collection of Classic Physics Apparatus at Vanderbilt University. These were bought by Chancellor Garland to outfit the Vanderbilt physics department for the opening of the university in 1875. Garland had previously gone to visit Koenig in Paris to discuss his order. in 1889 a set of nineteen resonators cost 170 francs.



# Tunable resonators



- a cylindrical resonator permits the volume of the resonator to be changed by sliding the tubes in and out. The notes (and hence the resonant frequencies) are engraved on the side of the apparatus. This is one of a number of tunable Helmholtz resonators at the University of Vermont.

# Tunable resonators

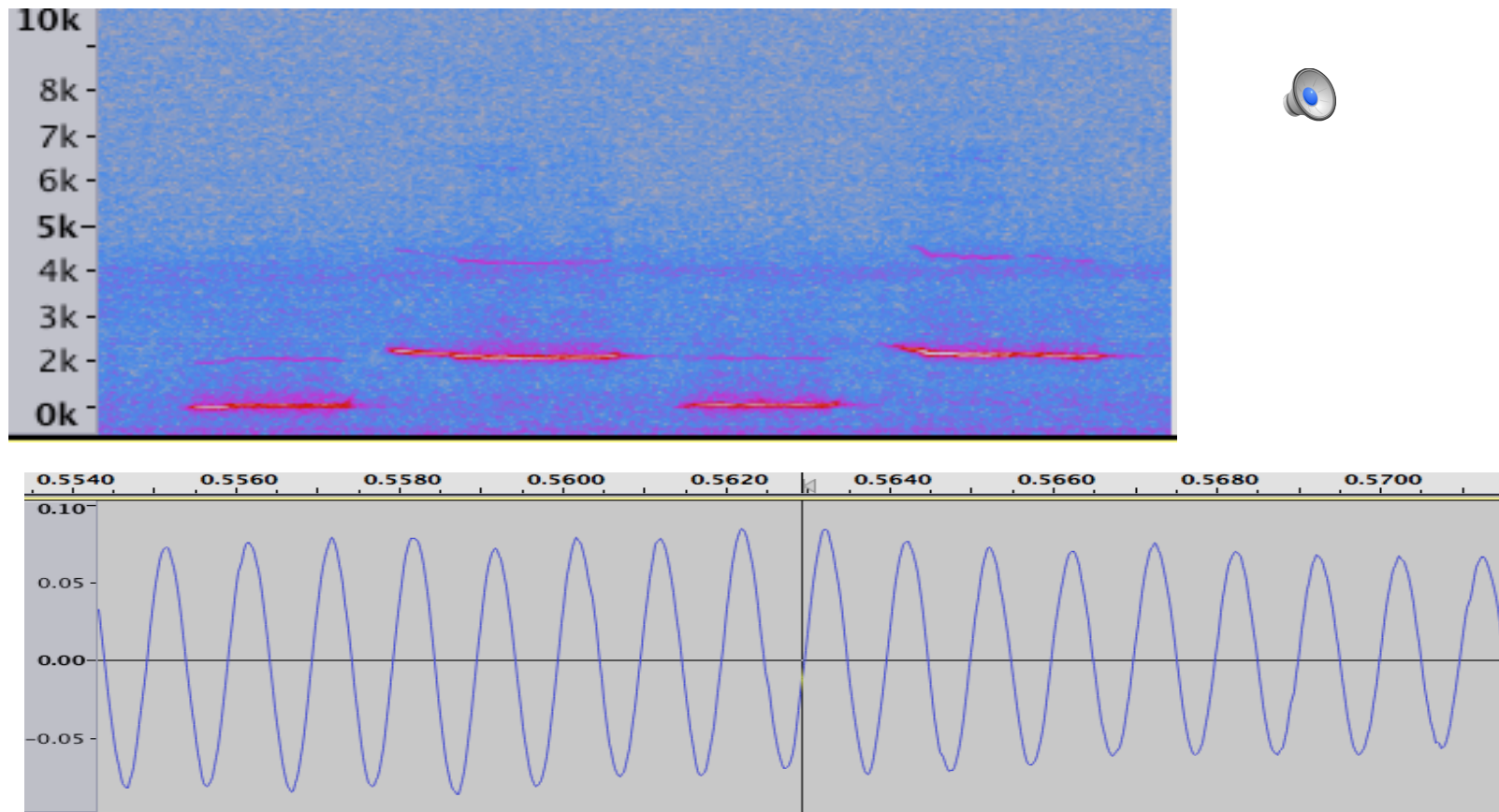


Ocarinas and  
whistling

Unlike with flutes  
the pitch is not set  
by the effective  
length of the  
instrument

# Whistle

- To do : film a whistle of across an octave



# Fourier analysis in 1890



- "Manometric Flame Analyser for the timbre of sounds, with 14 universal resonators --- originally 650 francs" (\$130). The adjustable Helmholtz resonators are tuned to the fundamental frequency of the sound to be analyzed, plus its harmonics. The holes on the other side of the resonators are connected by the rubber tubes to manometric flame capsules, and the variation in the height of the flames observed in the rotating mirror. The variation is proportional to the strength of the Fourier component of the sound.
- The picture at the left, below, shows the manometric capsules and the jets where the flames are produced. Note the black background to make the flames more visible.
- BTW nice display at U Toronto!

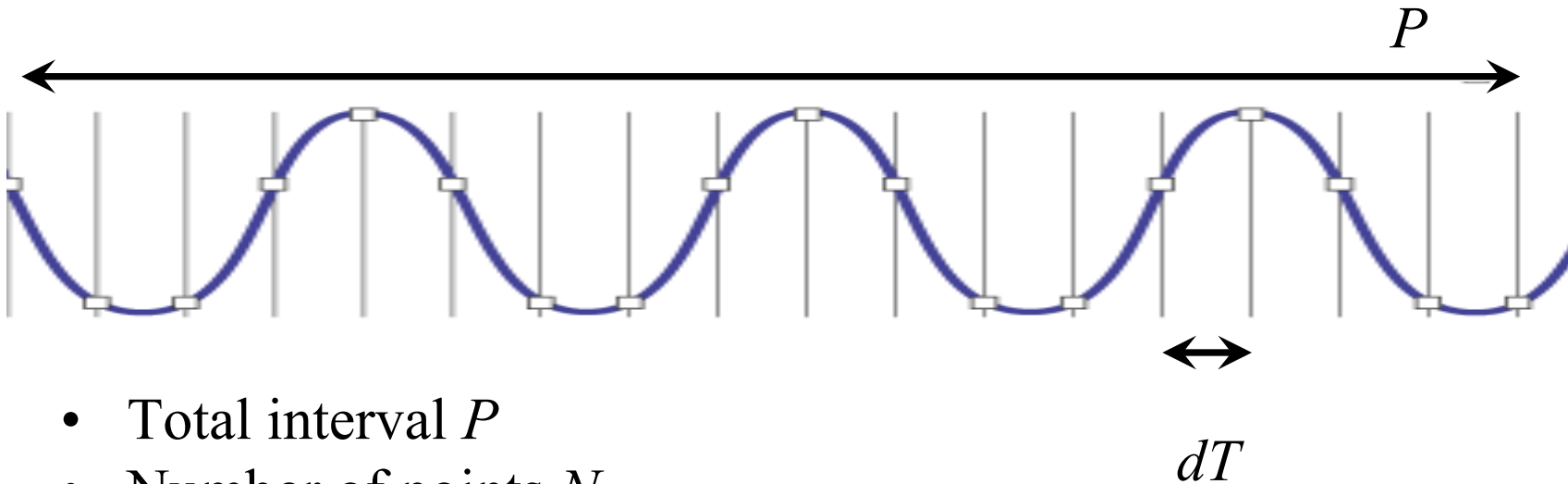
# Now how is the frequency analysis computed?

- The fast Fourier transform (FFT) is a discrete Fourier transform algorithm which reduces the number of computations needed for  $N$  points from  $2N^2$  to  $2N\log_2 N$  computations
- Discrete: works on data points rather than a function.
- A nice, space efficient algorithm exists for the number of points  $N$  equal to a power of 2.
- When you do a frequency analysis in Adobe Audition one of the parameters you can choose is  $N$  (and you will notice that the menu only allows powers of 2).

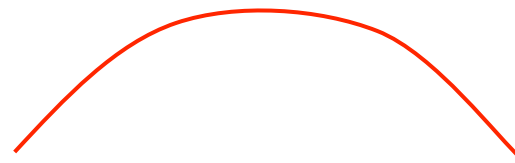
# The FFT algorithm

- A nice, memory efficient algorithm exists if the number of points is a power of 2
- Each component can be written as a sum of components from a transform of the interval divided in half.
- It maybe makes sense that the number of steps depends on  $\log N$

# Taking an FFT

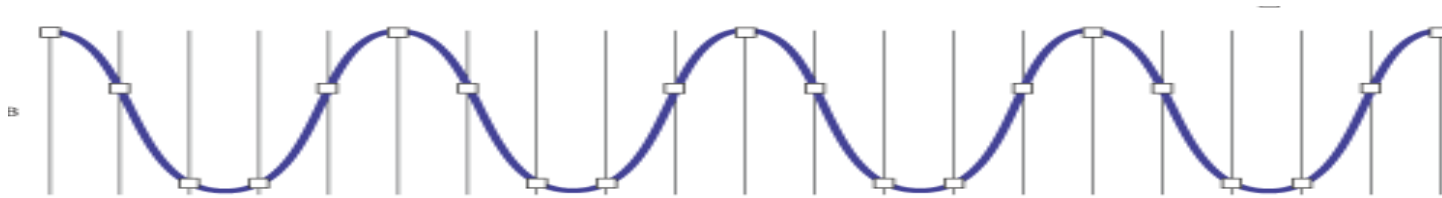


- Total interval  $P$
- Number of points  $N$
- Sampling  $dt$
- $P=N*dt$
- Windowing function – entire interval is multiplied by a function





# Output of FFT



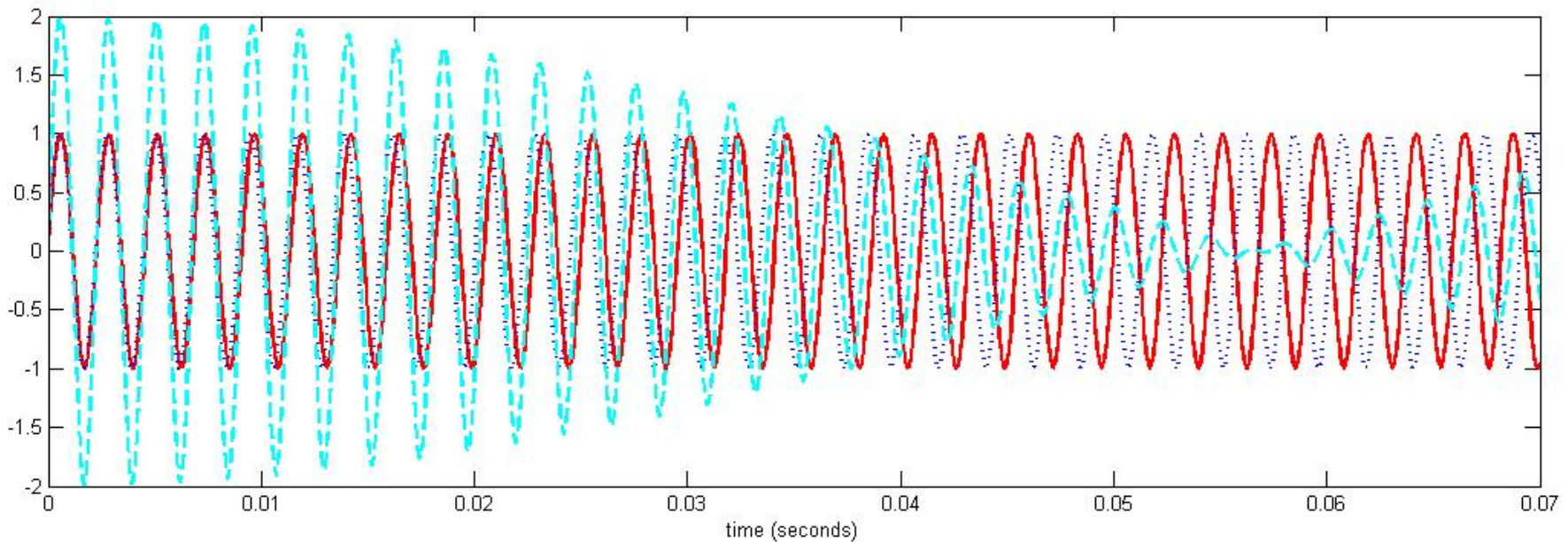
- Frequencies are computed at frequencies
- $f, 2f, 3f, 4f, \dots, Nf$  where  $1/f=P$  is the length of the interval used to compute the FFT and  $N$  is the number of points
- Difference between frequencies measured is set by the length of the whole interval  $P$ .
- If  $P$  (or number of points  $N$ ) is too small then precision of FFT is less.

# Accuracy of FFT

- To get better frequency measurements you need a larger interval to measure in
- You can't make extremely fine frequency measurements over extremely small time intervals
- Similar to a Heisenberg uncertainty relation

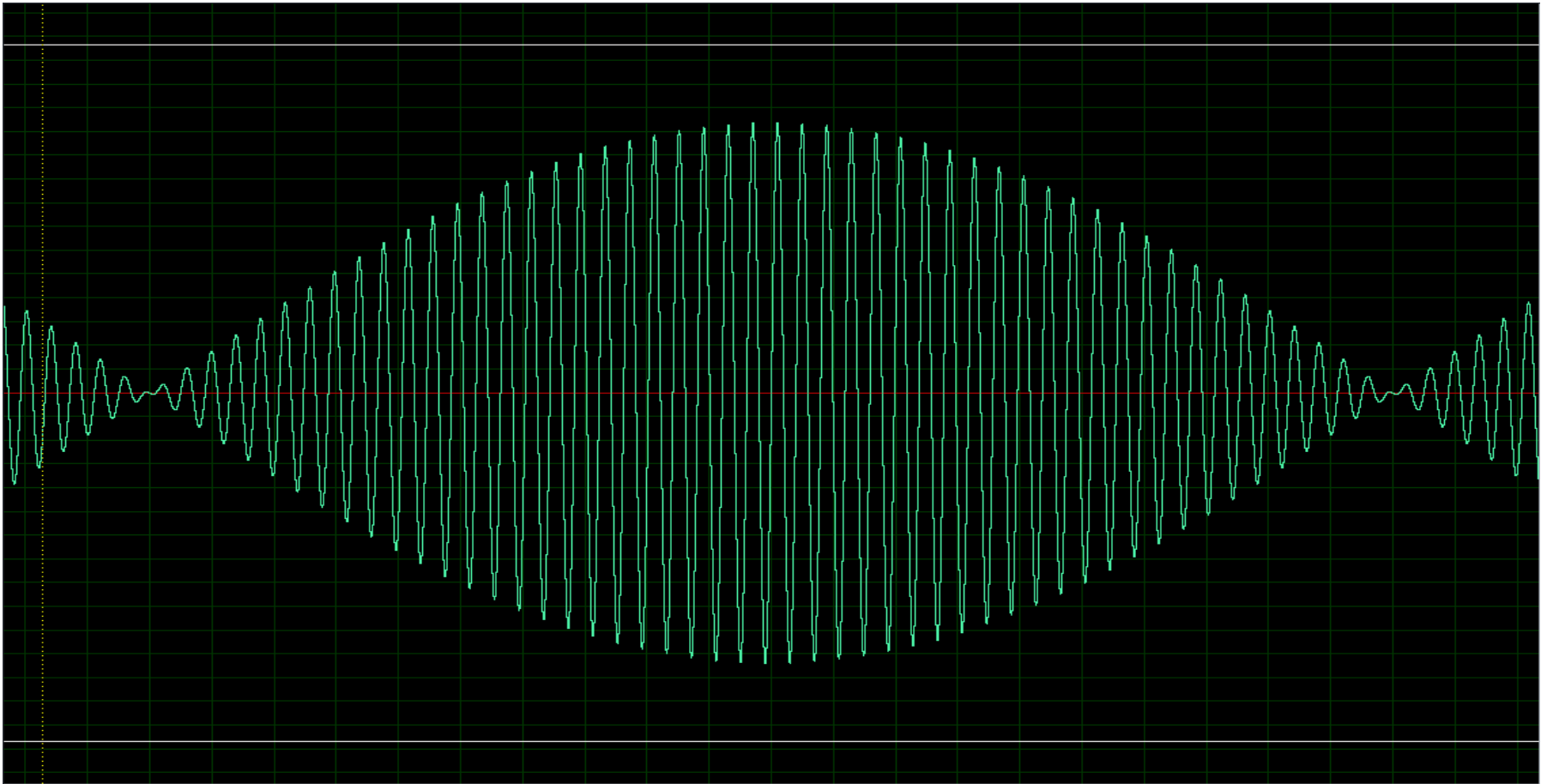
$$\Delta P \Delta f \sim 1$$

# Sum of two sine waves with frequencies very close together



Frequency  $f$  and  $1.02f$  and their sum

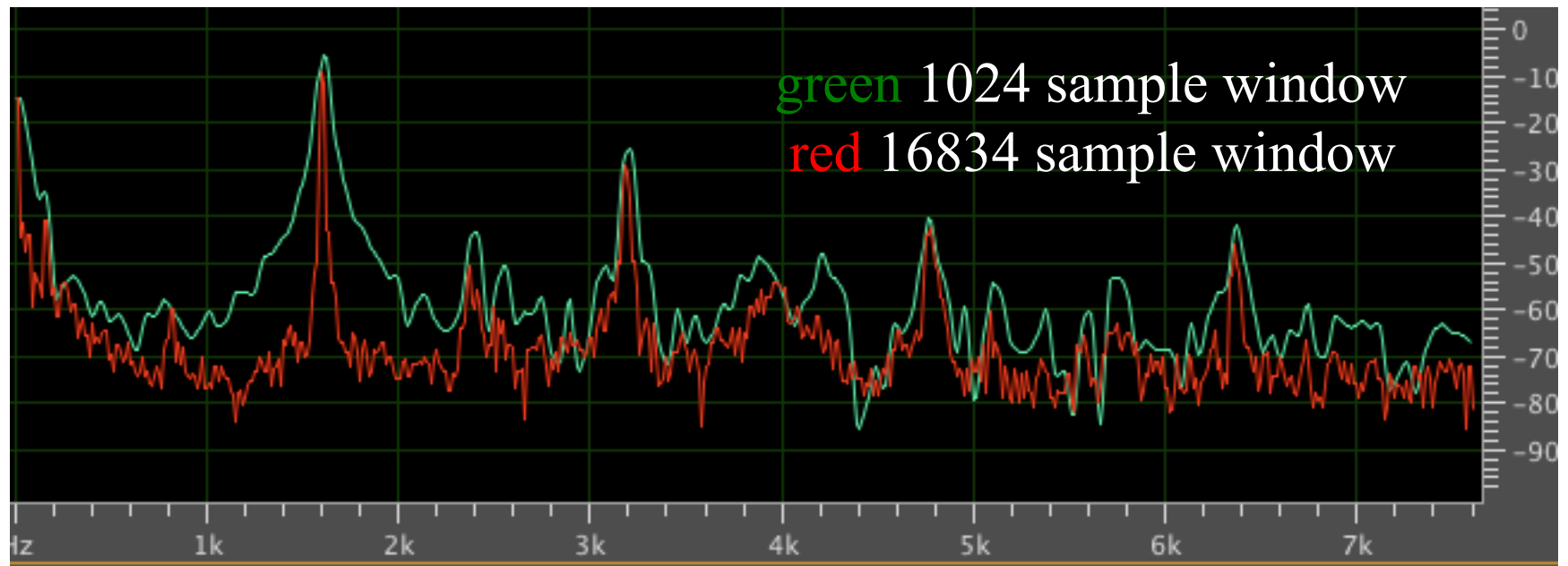
The closer the two frequencies, the longer it takes until they start to cancel



If I measure a fixed frequency over a small window then I don't know whether I have a single frequency or a sum of nearby frequencies. The longer the window I measure a pure sine wave, the more exactly I know the frequency of the sine wave.

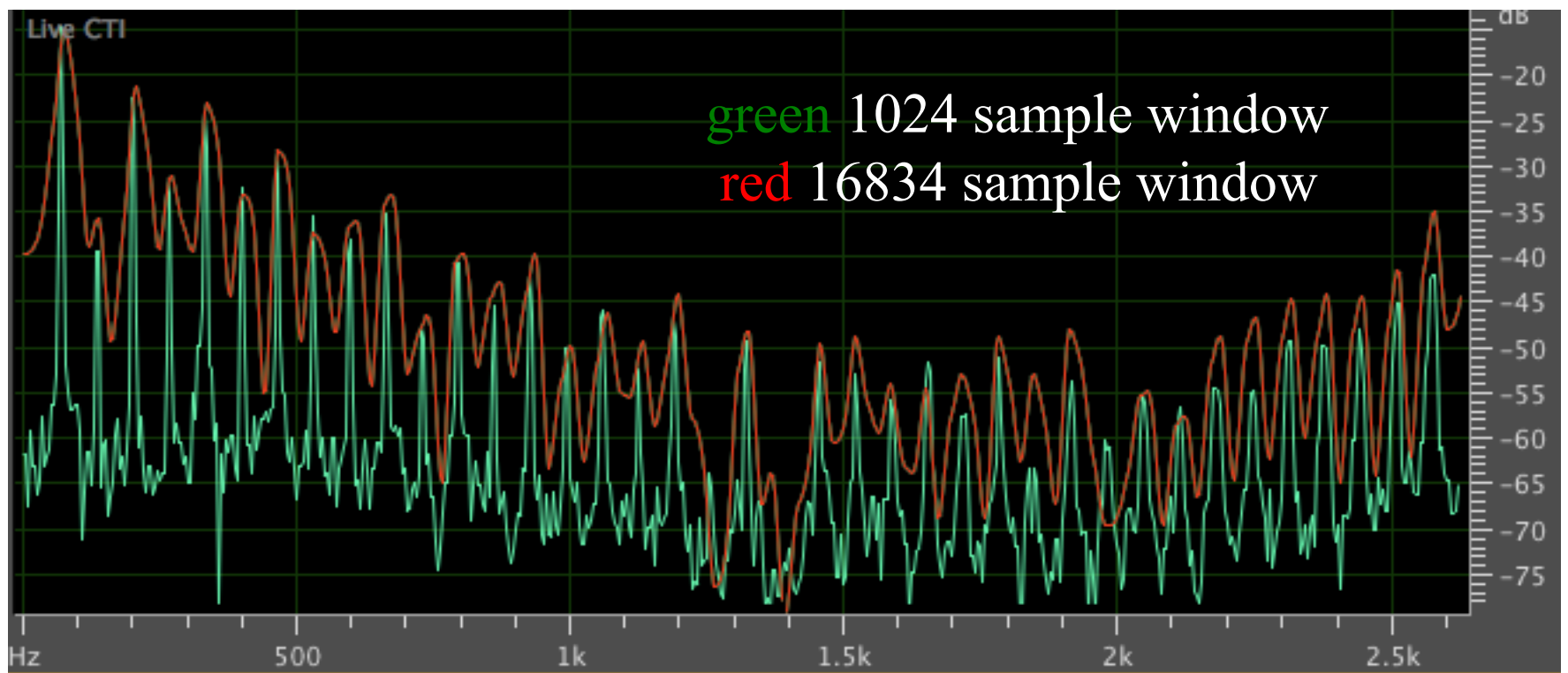
# Effect of window length on FFT precision

- Demo in Audition or Audacity different FFT lengths and windows on a sine wave



piccolo sound

# Window length and precision



digi low frequency sound

# Effect of Window function on FFT



$n=2408$  on digi sound

# Terminology

- Fourier decomposition
- Spectrum
- Spectral analysis
- Sampling rate
- Phase
- FFT (Fast Fourier Transform)



# Good/Bad physics -- Animusic



# Good/Bad Physics

- Donald Duck in Mathemagic land

# Recommended Reading

- Berg and Stork Chap 4