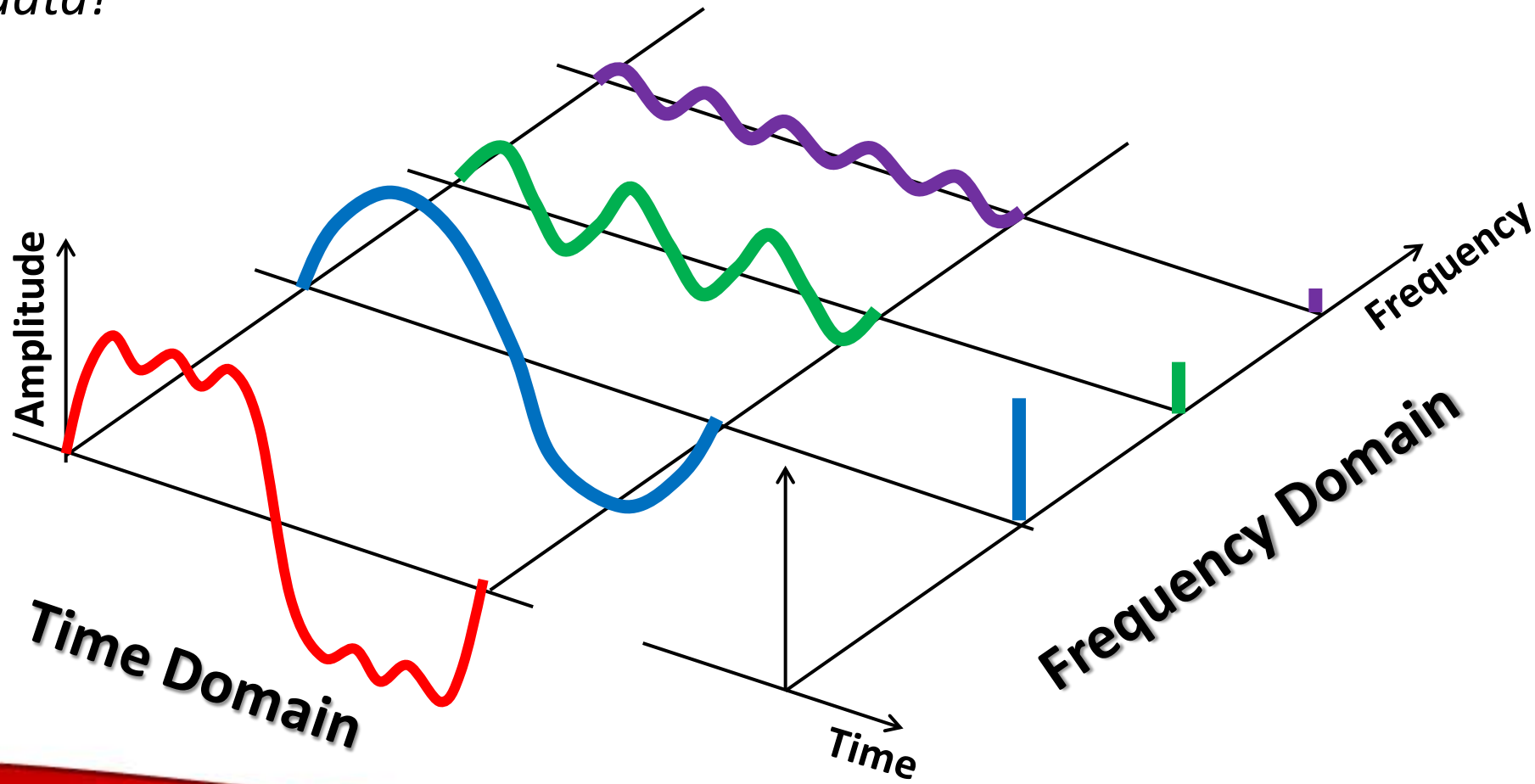


# Fourier Transforms

Modified from: <http://hometheaterhifi.com/technical/technical-reviews/up-sampling-aliasing-filtering-ringing-a-clarification-of-terminology/>

*Arguably the most important set of algorithm for analysis and manipulation of discrete data!*



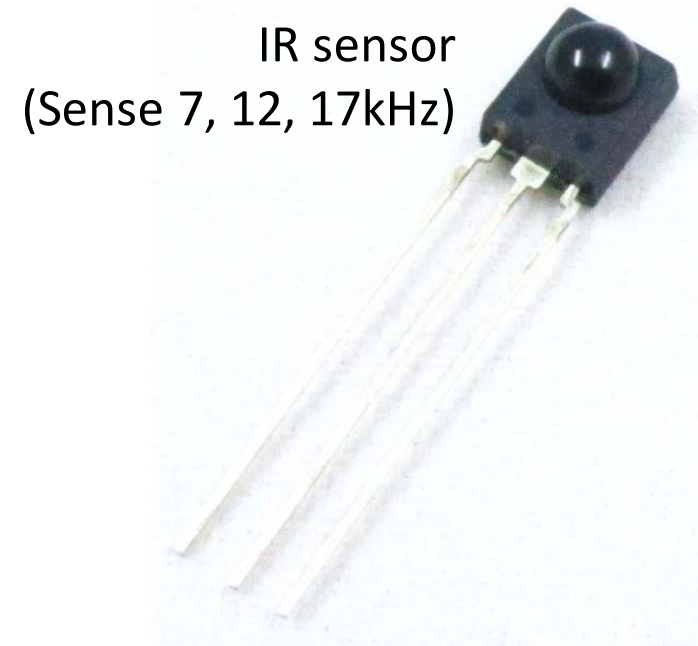
# Fourier Transforms

Why should *you* care *now*?

- Because you need to analyze noisy sensor signals!
- Analog filters
  - Bulky and require accurate (expensive) components
  - Low relative Q-values
- Discrete filters
  - Require memory, processing time, and an ADC
  - High relative Q-values and are versatile



Microphone  
(Sense 660Hz)



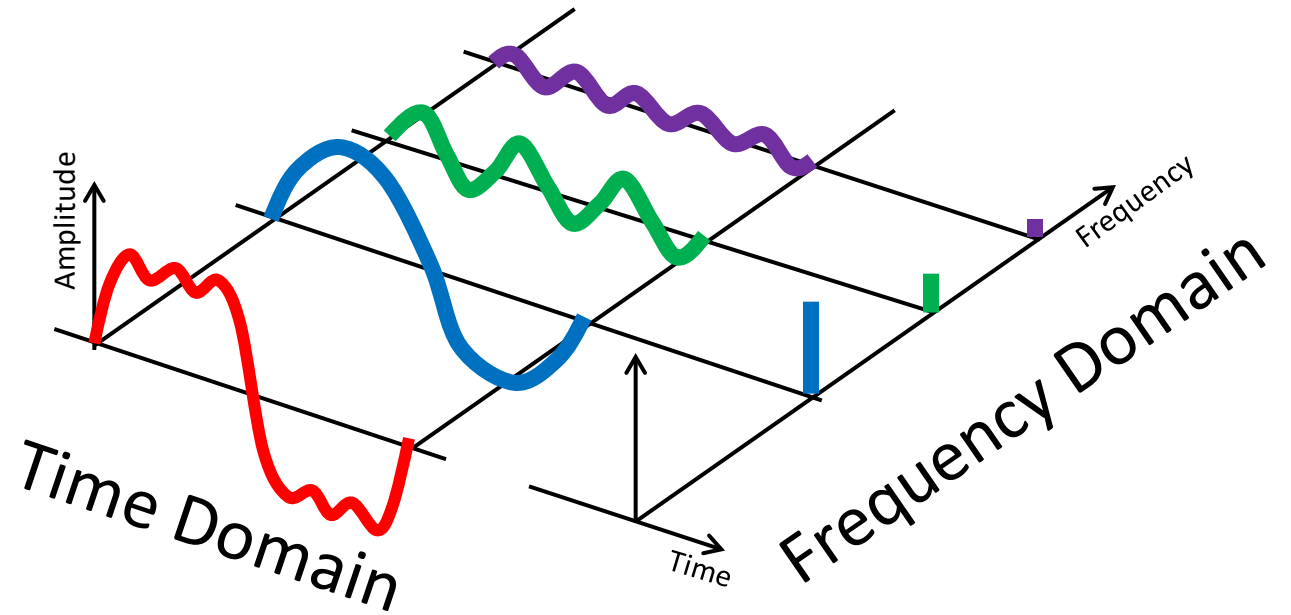
IR sensor  
(Sense 7, 12, 17kHz)

# Fast Fourier Transforms

*frequency = The sum of contributions to this frequency from each point in time*

$$F(x) = \sum_{n=0}^{N-1} f(n) \cdot e^{-i2\pi xn/N}$$

$$f(n) = \frac{1}{N} \sum_{n=0}^{N-1} F(x) \cdot e^{i2\pi kn/N}$$

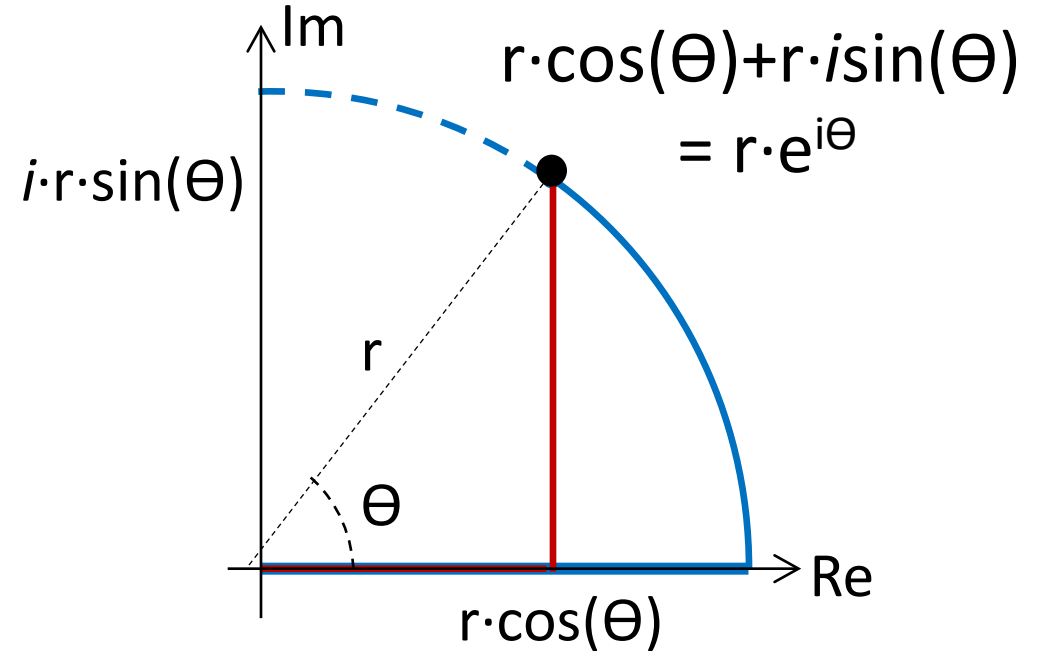


*Time point = The sum of contributions to this time point from each frequency*

# Describing a signal in terms of cycles

$$F(x) = \sum_{n=0}^{N-1} f(n) \cdot e^{-i2\pi xn/N}$$

$$f(n) = \frac{1}{N} \sum_{n=0}^{N-1} F(x) \cdot e^{i2\pi kn/N}$$

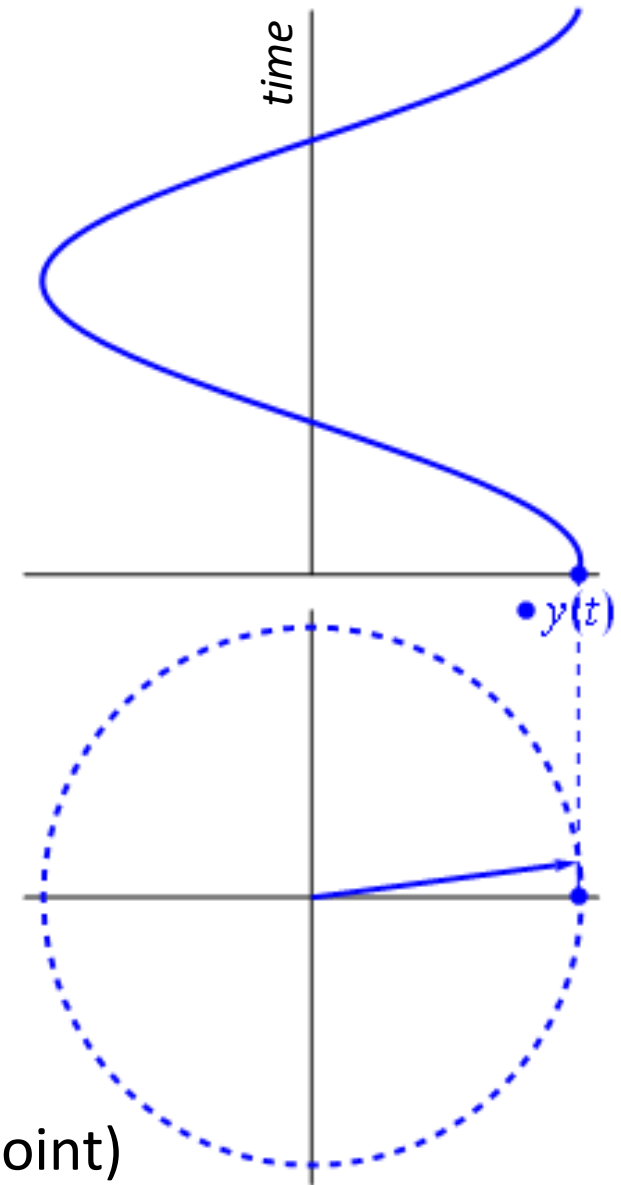


# Describing a signal in terms of cycles

$$F(x) = \sum_{n=0}^{N-1} f(n) \cdot e^{-i2\pi xn/N}$$

$$f(n) = \frac{1}{N} \sum_{k=0}^{N-1} F(x) \cdot e^{i2\pi kn/N}$$

- Frequency (speed)
- Amplitude (radius)
- Phase Angle (starting point)



# Describing a signal in terms of cycles

$$F(x) = \sum_{n=0}^{N-1} f(n) \cdot e^{-i2\pi xn/N}$$

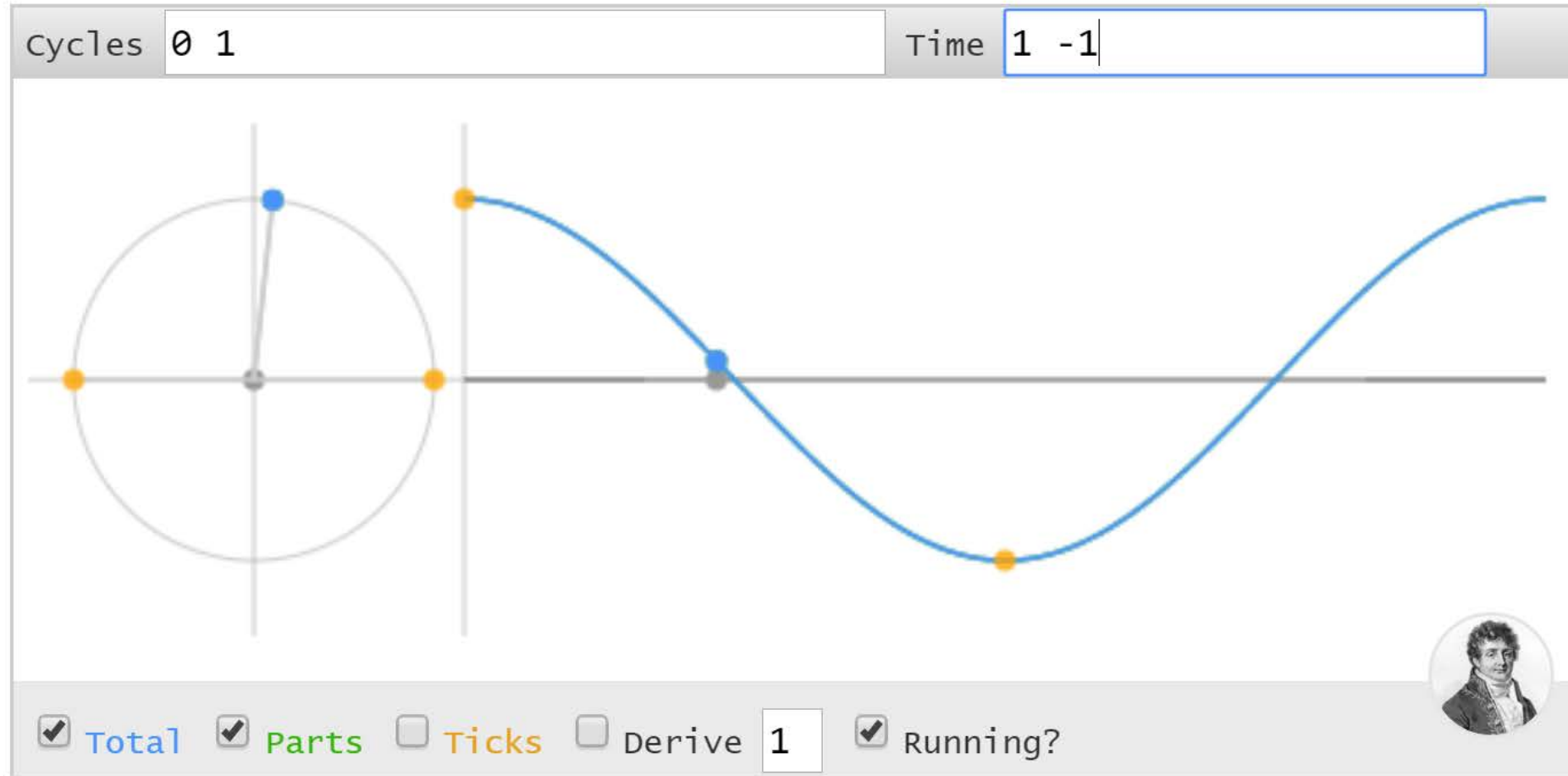
To find the **energy** at a **particular frequency**,  
**spin your signal** around a circle  
at that frequency,  
and average a bunch of points along that  
path.

$$f(n) = \frac{1}{N} \sum_{k=0}^{N-1} F(x_k) \cdot e^{i2\pi kn/N}$$

Stuart Riffle, Blogaday

# Describing a signal in terms of cycles

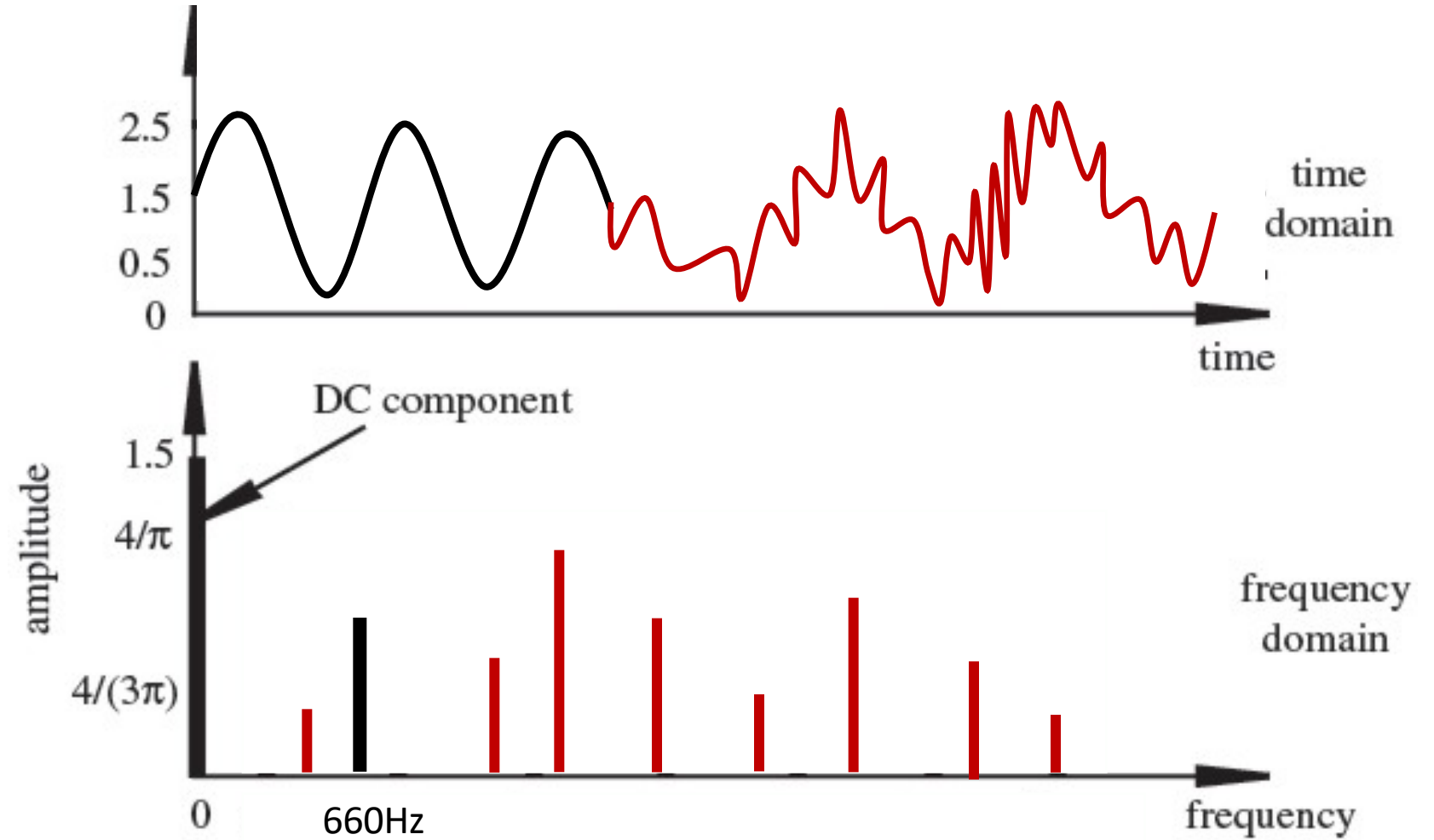
<https://betterexplained.com/articles/an-interactive-guide-to-the-fourier-transform/>



# Lab 2

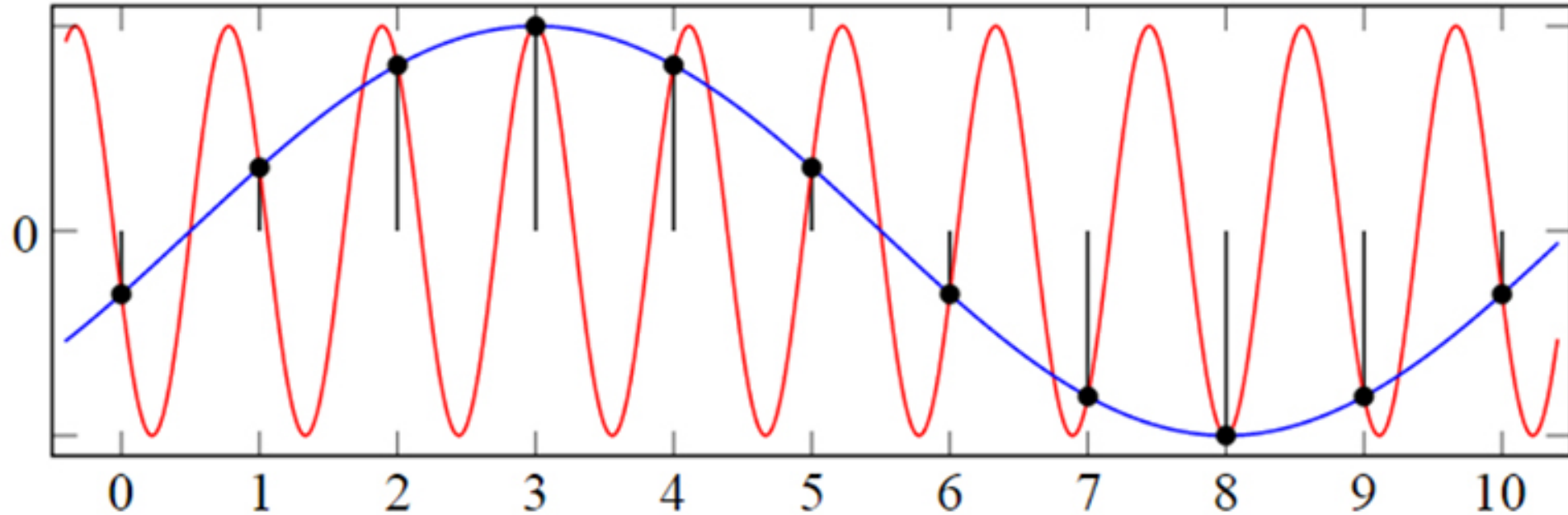


Detect start signal





# Discretized Signals and Aliasing

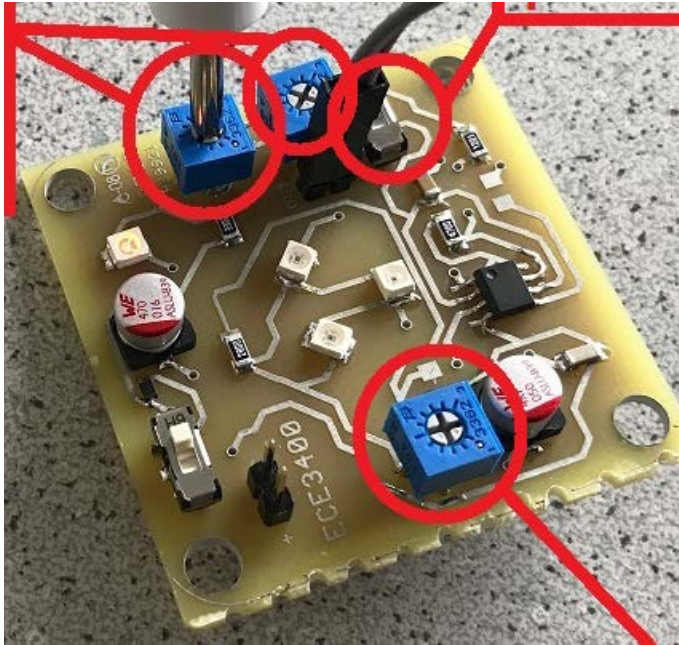


## *Nyquist Theorem*

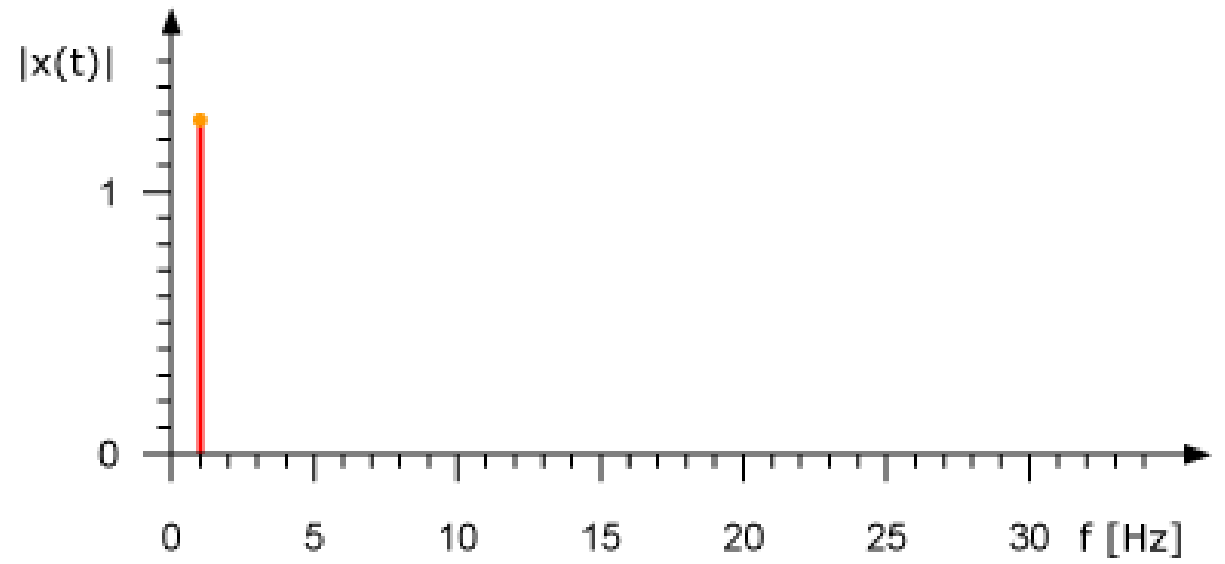
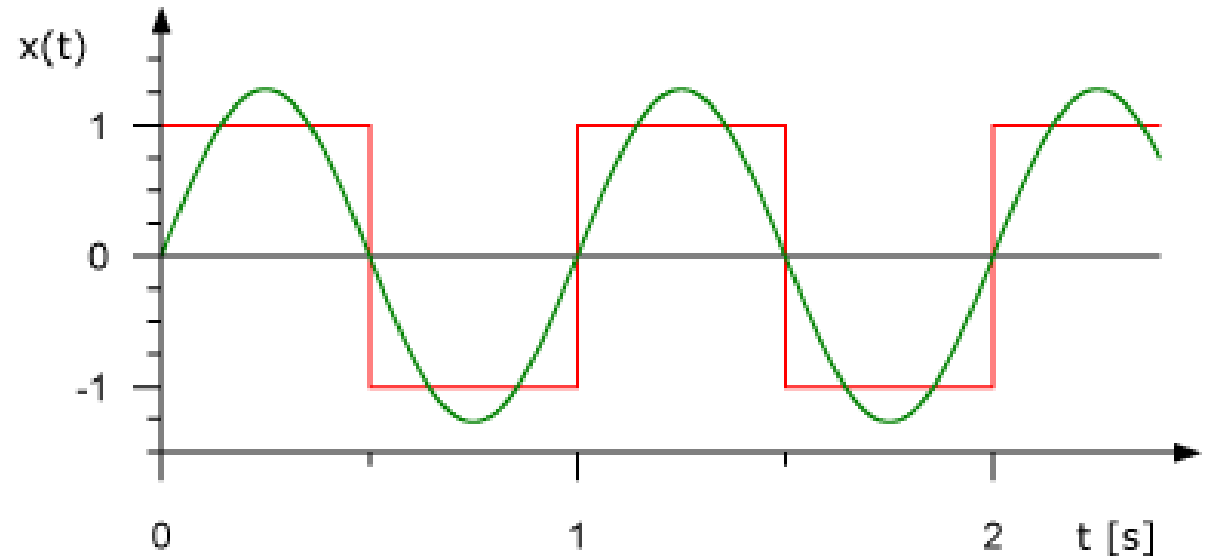
The sampling rate must be at least twice the highest frequency component of the signal:

$$f_{sample} = 2 \cdot f_{max}$$

# Lab 2



Treasures  
(7kHz, 12kHz, 17kHz)



# Open Music Labs FFT Library Functions

## `fft_run()`

Main FFT functional call

No input variables, no return variables

Assumes already re-ordered data is available in memory

Data is stored in array of size N, called `fft_input[]`

The array contains 2 16-bit values per FFT datapoint

Even positions are for real values, odd for imaginary

Every two positions corresponds to one bin =>  $N/2$  FFT bins

# Open Music Labs FFT Library Functions

Before calling `fft_run()`:

Decide `N` based on the number of FFT bins you want.

Fill `fft_input[]` with datapoints from the ADC. Since the datapoints are real, put them in the even positions, and fill the odd positions with 0's.

# Open Music Labs FFT Library Functions

After calling `fft_run()`:

Use one of the provided functions to process the FFT output and obtain the output array.

`fft_mag_log()` returns the output in sequential order of FFT frequency bins!

Check the value of the relevant bin (which bin # would you need to check for a 660Hz or 7kHz frequency?)

*(do the prelab!)*

*Go Build Robots!*

