

Analysis Methods in Atmospheric and Oceanic Science

AOSC 652

Fourier Analysis

Week 9, Day 1

24 Oct 2016

AOSC 652: Analysis Methods in AOSC

Fourier Series and Spectral Analysis

Any function can be expressed as a sum of a series of sin and cosine curves:

$$f(t) = \sum_{n=1}^{\infty} a_n \sin(2\pi nt) + b_n \cos(2\pi nt) \quad (1)$$

$f(t)$ is a “signal” in the time domain,
 a_n and b_n are unknown coefficients, and
 n has units of 1/time corresponding to the various frequencies of the wave

AOSC 652: Analysis Methods in AOSC

Fourier Series and Spectral Analysis

Any function can be expressed as a sum of a series of sin and cosine curves:

$$f(t) = \sum_{n=1}^{\infty} a_n \sin(2\pi nt) + b_n \cos(2\pi nt) \quad (1)$$

$f(t)$ is a “signal” in the time domain,
 a_n and b_n are unknown coefficients, and
 n has units of 1/time corresponding to the various frequencies of the wave

Often, you’ll see the expression for the Fourier Transform written quite differently from the above equation

AOSC 652: Analysis Methods in AOSC

Fourier Series and Spectral Analysis

For example, equation 12.0.1 of Press *et al.* states:

$$f(t) = \int_{-\infty}^{\infty} H(f) e^{-2\pi i f t} df \quad (2)$$

AOSC 652: Analysis Methods in AOSC

Fourier Series and Spectral Analysis

For example, equation 12.0.1 of Press *et al.* states:

$$f(t) = \int_{-\infty}^{\infty} H(f) e^{-2\pi i f t} df \quad (2)$$

What is the relation between these two definitions ?

AOSC 652: Analysis Methods in AOSC

Fourier Series and Spectral Analysis

For example, equation 12.0.1 of Press *et al.* states:

$$f(t) = \int_{-\infty}^{\infty} H(f) e^{-2\pi i f t} df \quad (2)$$

What is the relation between these two definitions ?

$$e^{i\theta} = \cos \theta + i \sin \theta \quad (3)$$

AOSC 652: Analysis Methods in AOSC

Fourier Series and Spectral Analysis

For example, equation 12.0.1 of Press *et al.* states:

$$f(t) = \int_{-\infty}^{\infty} H(f) e^{-2\pi i f t} df \quad (2)$$

What is the relation between these two definitions ?

$$e^{i\theta} = \cos \theta + i \sin \theta \quad (3)$$

Therefore:

$$f(t) = \int_{-\infty}^{\infty} H(f) [\cos(-2\pi f t) + i \sin(-2\pi f t)] df \quad (4)$$

AOSC 652: Analysis Methods in AOSC

Fourier Series and Spectral Analysis

For example, equation 12.0.1 of Press *et al.* states:

$$f(t) = \int_{-\infty}^{\infty} H(f) e^{-2\pi i f t} df \quad (2)$$

What is the relation between these two definitions ?

$$e^{i\theta} = \cos \theta + i \sin \theta \quad (3)$$

Therefore:

$$f(t) = \int_{-\infty}^{\infty} H(f) [\cos(-2\pi f t) + i \sin(-2\pi f t)] df \quad (4)$$

Or:

$$f(t) = \int_{-\infty}^{\infty} H_{\text{REAL}}(f) \cos(2\pi f t) + i H_{\text{IMAGINARY}}(f) \sin(2\pi f t) df \quad (5)$$

AOSC 652: Analysis Methods in AOSC

Fourier Series and Spectral Analysis

For example, equation 12.0.1 of Press *et al.* states:

$$f(t) = \int_{-\infty}^{\infty} H(f) e^{-2\pi i f t} df \quad (2)$$

What is the relation between these two definitions ?

$$e^{i\theta} = \cos \theta + i \sin \theta \quad (3)$$

Therefore:

$$f(t) = \int_{-\infty}^{\infty} H(f) [\cos(-2\pi f t) + i \sin(-2\pi f t)] df \quad (4)$$

Or:

$$f(t) = \int_{-\infty}^{\infty} H_{\text{REAL}}(f) \cos(2\pi f t) + i H_{\text{IMAGINARY}}(f) \sin(2\pi f t) df \quad (5)$$

This expression used in Muller and MacDonald reading !



AOSC 652: Analysis Methods in AOSC

Fourier Series and Spectral Analysis

There are many ways to compute a_n and b_n in equation (1) :

$$f(t) = \sum_{n=1}^{\infty} a_n \sin(2\pi n t) + b_n \cos(2\pi n t) \quad (1)$$

or H_{REAL} and $H_{\text{IMAGINARY}}$ in equation (5)

$$f(t) = \int_{-\infty}^{\infty} H_{\text{REAL}}(f) \cos(2\pi f t) + i H_{\text{IMAGINARY}}(f) \sin(2\pi f t) df \quad (5)$$

AOSC 652: Analysis Methods in AOSC

Fourier Series and Spectral Analysis

Here, we will focus on a very simple treatment of (5), that given on page 52 of Muller and MacDonald:

$$f(t) = \int_{-\infty}^{\infty} H_{\text{REAL}}(f) \cos(2\pi f t) + i H_{\text{IMAGINARY}} \sin(2\pi f t) df \quad (5)$$

For a discrete time series represented by variable $signal(time)$:

i.e., there exists $signal(1), signal(2), \dots, signal(N)$

and corresponding time elements $time(1), time(2), \dots, time(N)$

the values of H_{REAL} and $H_{\text{IMAGINARY}}$ at frequency f are given by:

$$H_{\text{REAL}} = 0$$

$$H_{\text{IMAGINARY}} = 0$$

do j=1 to N by 1

$$H_{\text{REAL}} = H_{\text{REAL}} + signal(j) * \cos(2\pi f time(j))$$

$$H_{\text{IMAGINARY}} = H_{\text{IMAGINARY}} + signal(j) * \sin(2\pi f time(j))$$

enddo

AOSC 652: Analysis Methods in AOSC

Fourier Series and Spectral Analysis

For a *discrete* time series represented by variable $signal(time)$:

i.e., there exists $signal(1), signal(2), \dots, signal(N)$

and corresponding time elements $time(1), time(2), \dots, time(N)$

the values of H_{REAL} and $H_{\text{IMAGINARY}}$ at frequency f are given by:

$$H_{\text{REAL}} = 0$$

$$H_{\text{IMAGINARY}} = 0$$

do $j=1$ to N by 1

$$H_{\text{REAL}} = H_{\text{REAL}} + signal(j) * \cos(2\pi f time(j))$$

$$H_{\text{IMAGINARY}} = H_{\text{IMAGINARY}} + signal(j) * \sin(2\pi f time(j))$$

enddo

Simply put:

- if the data oscillate in phase with a sin wave of given frequency f , the various values of $H_{\text{IMAGINARY}}$ will reinforce and sum together, leading to a large “summed” value of $H_{\text{IMAGINARY}}$
- if the data are not in phase with a sin wave of frequency f , various values of $H_{\text{IMAGINARY}}$ will tend to cancel and the “summed” value of $H_{\text{IMAGINARY}}$ will be small.

AOSC 652: Analysis Methods in AOSC

Fourier Series and Spectral Analysis

For a *discrete* time series represented by variable $signal(time)$:

i.e., there exists $signal(1), signal(2), \dots, signal(N)$

and corresponding time elements $time(1), time(2), \dots, time(N)$

the values of H_{REAL} and $H_{\text{IMAGINARY}}$ at frequency f are given by:

$$H_{\text{REAL}} = 0$$

$$H_{\text{IMAGINARY}} = 0$$

do $j=1$ to N by 1

$$H_{\text{REAL}} = H_{\text{REAL}} + signal(j) * \cos(2\pi f time(j))$$

$$H_{\text{IMAGINARY}} = H_{\text{IMAGINARY}} + signal(j) * \sin(2\pi f time(j))$$

enddo

Assuming time series is equally spaced at Δt and ranges from $t=0$ to $t=t_{\text{MAX}}$

- $f_{\text{LOW}} = 1 / t_{\text{MAX}}$ \Leftrightarrow lowest frequency of interest
- $f_{\text{NYQUIST}} = 1 / 2 \Delta t$ \Leftrightarrow highest frequency of interest

AOSC 652: Analysis Methods in AOSC

Fourier Series and Spectral Analysis

This is a good time to look at the equation, and get a sense of what the Fourier transform does. It takes your data, multiplies it by a sine wave, and then sums the results. If the data oscillate in phase with the sine wave, so that they are positive together and negative together, then all the terms in the sum are positive and the Fourier amplitude is large. If they drift into phase and then out of phase, then half of the values in the product will be positive and half will be negative, and the sum will be close to zero. You can also see that the sum is not particularly sensitive to sharp changes in the data (e.g. sudden terminations); it is more sensitive to the bulk behaviour of the data, e.g. are most of the data points positive when the sine wave is positive?

Page 52, Muller and MacDonald

AOSC 652: Analysis Methods in AOSC

Fourier Series and Spectral Analysis

The total power of a signal is the same whether we compute it in the time or frequency domain (Press *et al.*, page 492):

$$\textit{Total Power} \equiv \int_{-\infty}^{\infty} [\textit{signal}(\textit{time})]^2 d(\textit{time}) = \int_{-\infty}^{\infty} [H(f)]^2 df \quad (6)$$

AOSC 652: Analysis Methods in AOSC

Fourier Series and Spectral Analysis

The total power of a signal is the same whether we compute it in the time or frequency domain (Press *et al.*, page 492):

$$\text{Total Power} \equiv \int_{-\infty}^{\infty} [\text{signal}(\text{time})]^2 d(\text{time}) = \int_{-\infty}^{\infty} [H(f)]^2 df \quad (6)$$

Often one wants to know “how much power” is contained in the frequency interval between f and $f + df$. In such circumstances, we usually do not distinguish between positive and negative values of f , but rather regard f as varying between 0 (no frequency, or D.C.) and infinity:

$$\text{Power}(f) \equiv [H(f)]^2 + [H(-f)]^2 \quad 0 \leq f \leq \infty \quad (7)$$

AOSC 652: Analysis Methods in AOSC

Fourier Series and Spectral Analysis

The total power of a signal is the same whether we compute it in the time or frequency domain (Press *et al.*, page 492):

$$\text{Total Power} \equiv \int_{-\infty}^{\infty} [\text{signal}(\text{time})]^2 d(\text{time}) = \int_{-\infty}^{\infty} [H(f)]^2 df \quad (6)$$

Often one wants to know “how much power” is contained in the frequency interval between f and $f + df$. In such circumstances, we usually do not distinguish between positive and negative values of f , but rather regard f as varying between 0 (no frequency, or D.C.) and infinity:

$$\text{Power} (f) \equiv [H(f)]^2 + [H(-f)]^2 \quad 0 \leq f \leq \infty \quad (7)$$

If $\text{signal}(\text{time})$ is a real function, the two terms above can be shown to be equal and:

$$\text{Power} (f) = 2 [H(f)]^2 \quad 0 \leq f \leq \infty \quad (8)$$

AOSC 652: Analysis Methods in AOSC

Fourier Series and Spectral Analysis

The total power of a signal is the same whether we compute it in the time or frequency domain (Press *et al.*, page 492):

$$\text{Total Power} \equiv \int_{-\infty}^{\infty} [\text{signal}(\text{time})]^2 d(\text{time}) = \int_{-\infty}^{\infty} [H(f)]^2 df \quad (6)$$

Often one wants to know “how much power” is contained in the frequency interval between f and $f + df$. In such circumstances, we usually do not distinguish between positive and negative values of f , but rather regard f as varying between 0 (no frequency, or D.C.) and infinity:

$$\text{Power}(f) \equiv [H(f)]^2 + [H(-f)]^2 \quad 0 \leq f \leq \infty \quad (7)$$

If $\text{signal}(\text{time})$ is a real function, the two terms above can be shown to be equal and:

$$\text{Power}(f) = 2 [H(f)]^2 \quad 0 \leq f \leq \infty \quad (8)$$

Press *et al.* (pages 492 & 493) discuss this factor of 2 and how it is sometimes not used. We'll work with normalized power spectrum, such that maximum value is by definition unity, so we will also not consider this factor of 2.

AOSC 652: Analysis Methods in AOSC

Fourier Series and Spectral Analysis

Returning to our heuristic code:

For a *discrete* time series represented by variable $signal(time)$:

i.e., there exists $signal(1), signal(2), \dots, signal(N)$

and corresponding time elements $time(1), time(2), \dots, time(N)$

the values of H_{REAL} and $H_{\text{IMAGINARY}}$ at frequency f are given by:

$$H_{\text{REAL}} = 0$$

$$H_{\text{IMAGINARY}} = 0$$

do j=1 to N by 1

$$H_{\text{REAL}} = H_{\text{REAL}} + signal(j) * \cos(2\pi f time(j))$$

$$H_{\text{IMAGINARY}} = H_{\text{IMAGINARY}} + signal(j) * \sin(2\pi f time(j))$$

enddo

$$\underline{Power = H_{\text{REAL}}^2 + H_{\text{IMAGINARY}}^2} \quad (\text{see page 52, Muller \& MacDonald})$$

AOSC 652: Analysis Methods in AOSC

Fourier Series and Spectral Analysis

Let's conduct some power spectrum analyses:

Copy files `~rjs/aosc652/week_09/tone*.dat`
& `~rjs/aosc652/week_09/fourier_analysis.f`

to your work area.

File `tone01.dat` contains a *signal vs time*, for *time = 0 to 0.25 sec*

First, using either `hppltd`, Python, or IDL produce a plot of signal vs time:

AOSC 652: Analysis Methods in AOSC

Fourier Series and Spectral Analysis

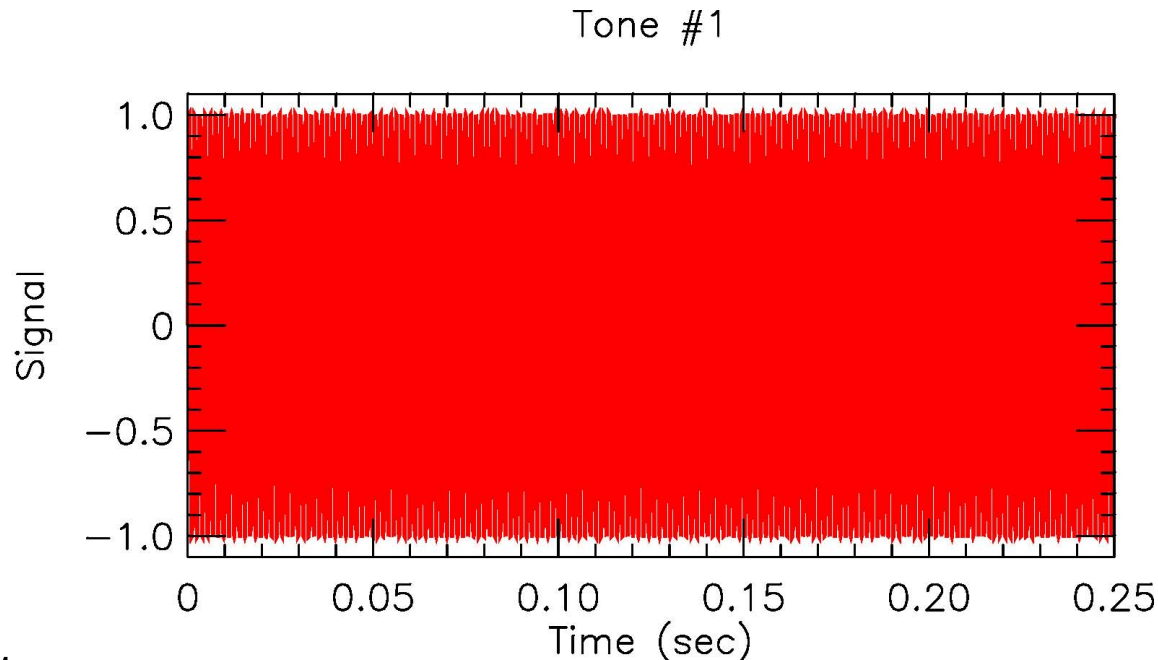
Let's conduct some power spectrum analyses:

Copy files `~rjs/aosc652/week_09/tone*.dat`
& `~rjs/aosc652/week_09/fourier_analysis.f`

to your work area.

File `tone01.dat` contains a *signal vs time*, for *time = 0 to 0.25 sec*

First, using either `hplotd`, Python, or IDL produce a plot of signal vs time:



AOSC 652: Analysis Methods in AOSC

Fourier Series and Spectral Analysis

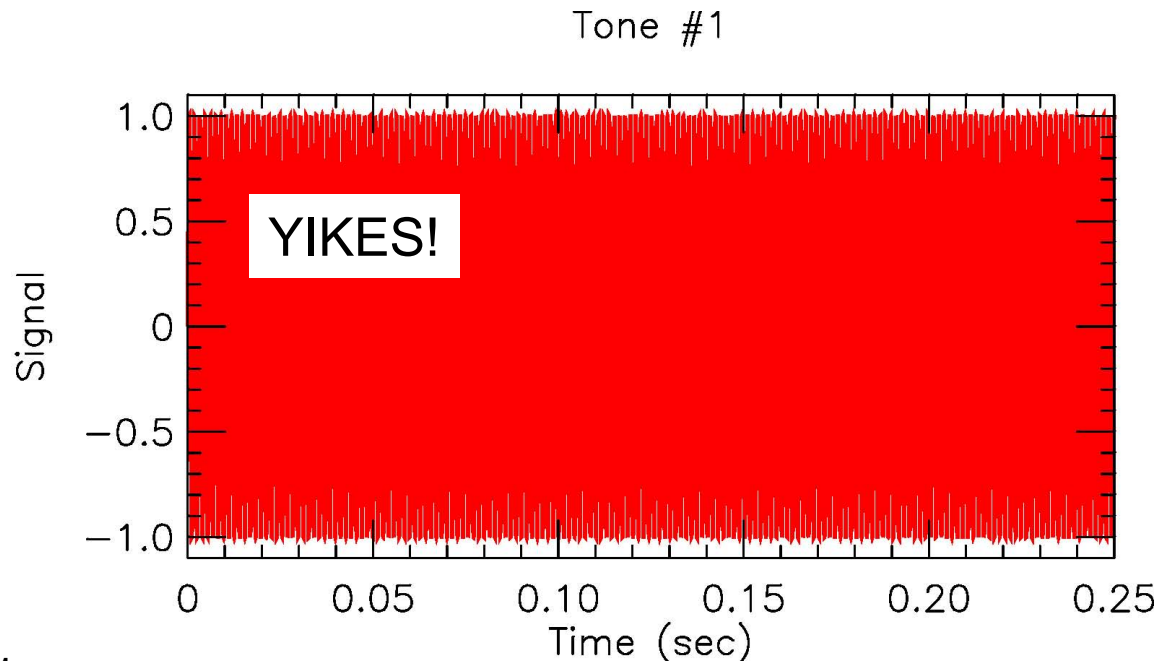
Let's conduct some power spectrum analyses:

Copy files `~rjs/aosc652/week_09/tones*.dat`
& `~rjs/aosc652/week_09/fourier_analysis.f`

to your work area.

File `tone01.dat` contains a *signal vs time*, for *time = 0 to 0.25 sec*

First, using either `hplotd`, Python, or IDL produce a plot of signal vs time:



AOSC 652: Analysis Methods in AOSC

Fourier Series and Spectral Analysis

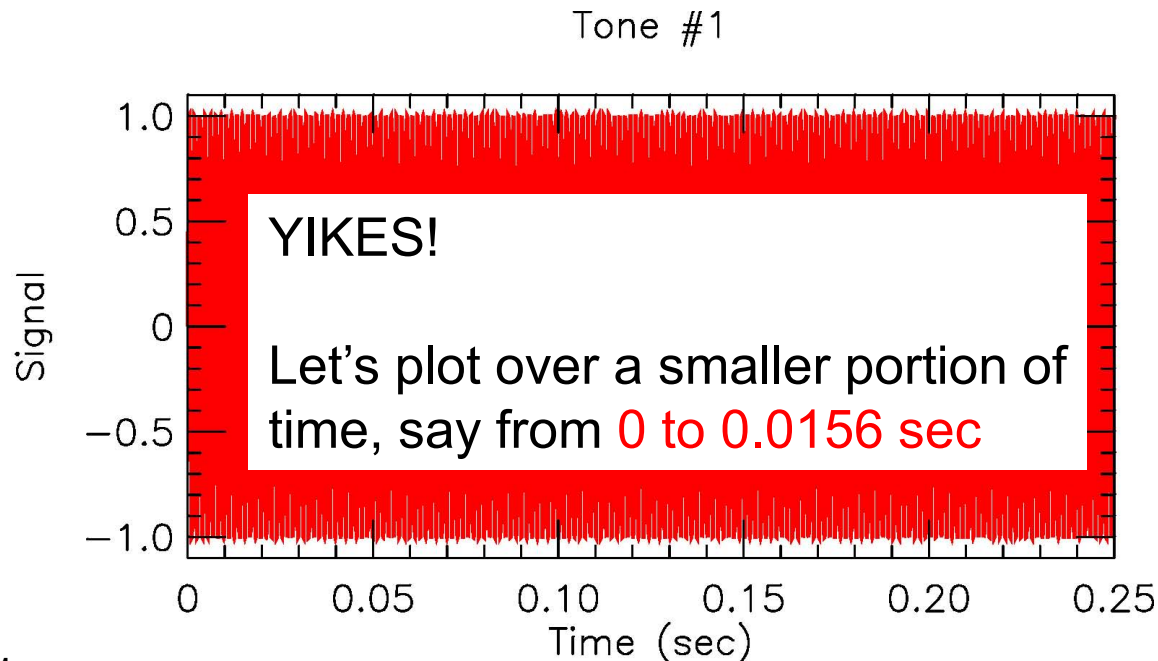
Let's conduct some power spectrum analyses:

Copy files `~rjs/aosc652/week_09/tones*.dat`
& `~rjs/aosc652/week_09/fourier_analysis.f`

to your work area.

File `tone01.dat` contains a *signal vs time*, for *time = 0 to 0.25 sec*

First, using either `hplotd`, Python, or IDL produce a plot of signal vs time:



AOSC 652: Analysis Methods in AOSC

Fourier Series and Spectral Analysis

Let's conduct some power spectrum analyses:

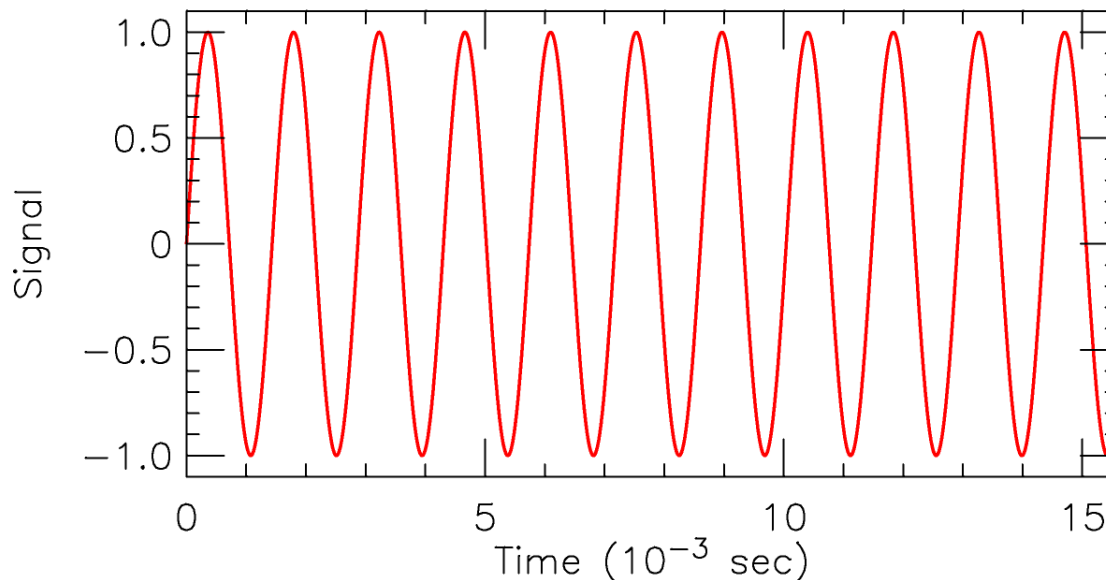
Copy files `~rjs/aosc652/week_09/tone*.dat`
& `~rjs/aosc652/week_09/fourier_analysis.f`

to you work area.

File `tone01.dat` contains a *signal vs time*, for *time = 0 to 0.25 sec*

First, using either `hplotd`, Python, or IDL produce a plot of signal vs time:

Tone #1



AOSC 652: Analysis Methods in AOSC

Fourier Series and Spectral Analysis

Compile program `fourier_analysis.f` and [compute power spectrum](#) of `tone01.dat`, from frequencies of [600 to 1300 Hz](#), every [1 Hz](#)

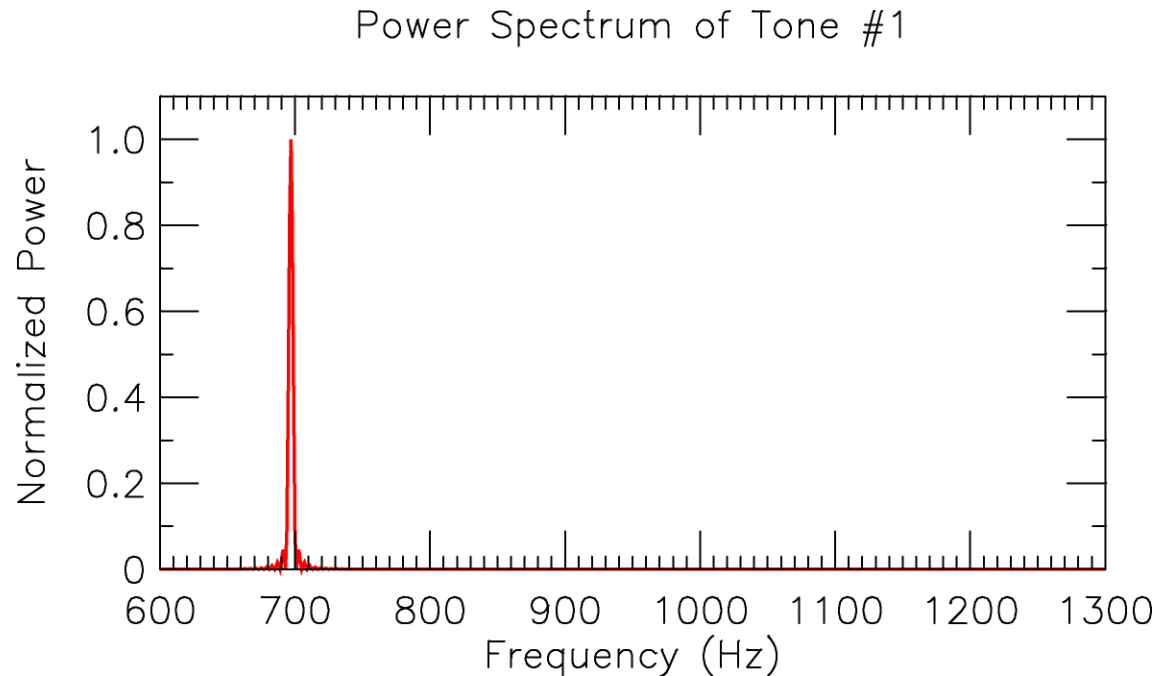
Plot the result from 600 to 1300 Hz:

AOSC 652: Analysis Methods in AOSC

Fourier Series and Spectral Analysis

Compile program `fourier_analysis.f` and [compute power spectrum](#) of `tone01.dat`, from frequencies of [600 to 1300 Hz](#), every 1 Hz

Plot the result from 600 to 1300 Hz:

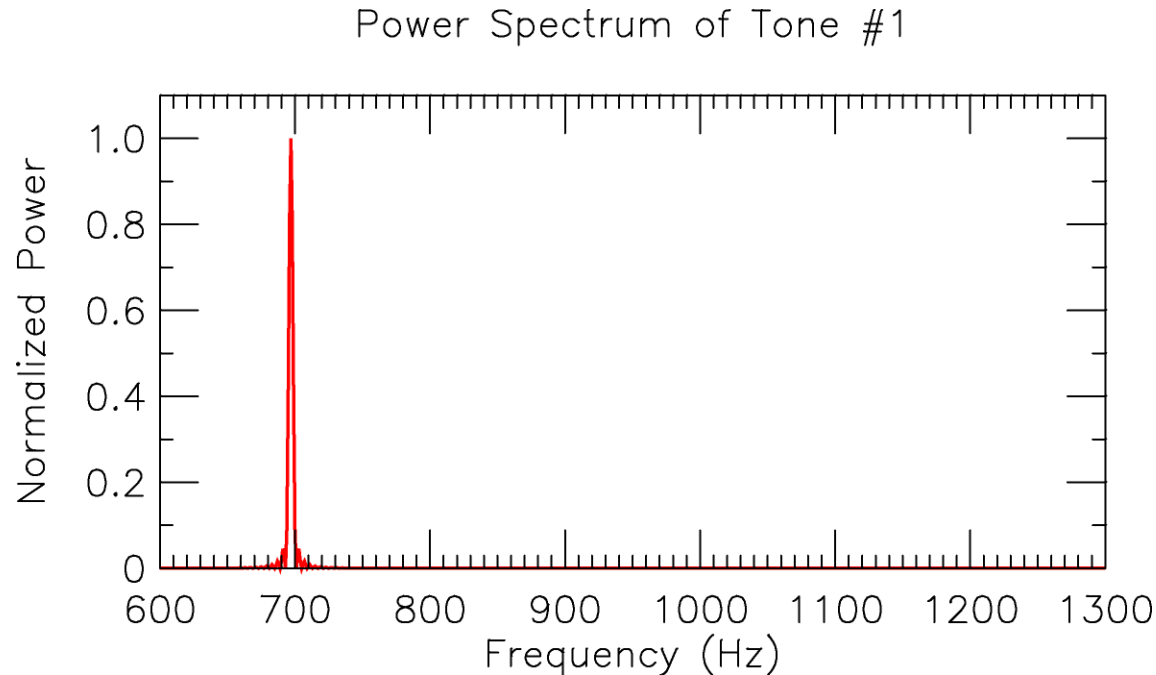


AOSC 652: Analysis Methods in AOSC

Fourier Series and Spectral Analysis

Compile program `fourier_analysis.f` and [compute power spectrum](#) of `tone01.dat`, from frequencies of [600 to 1300 Hz](#), every 1 Hz

Plot the result from 600 to 1300 Hz:



tone01.dat generated with FORTRAN program `~rjs/aosc652/week_09/tone01_gen.f`

AOSC 652: Analysis Methods in AOSC

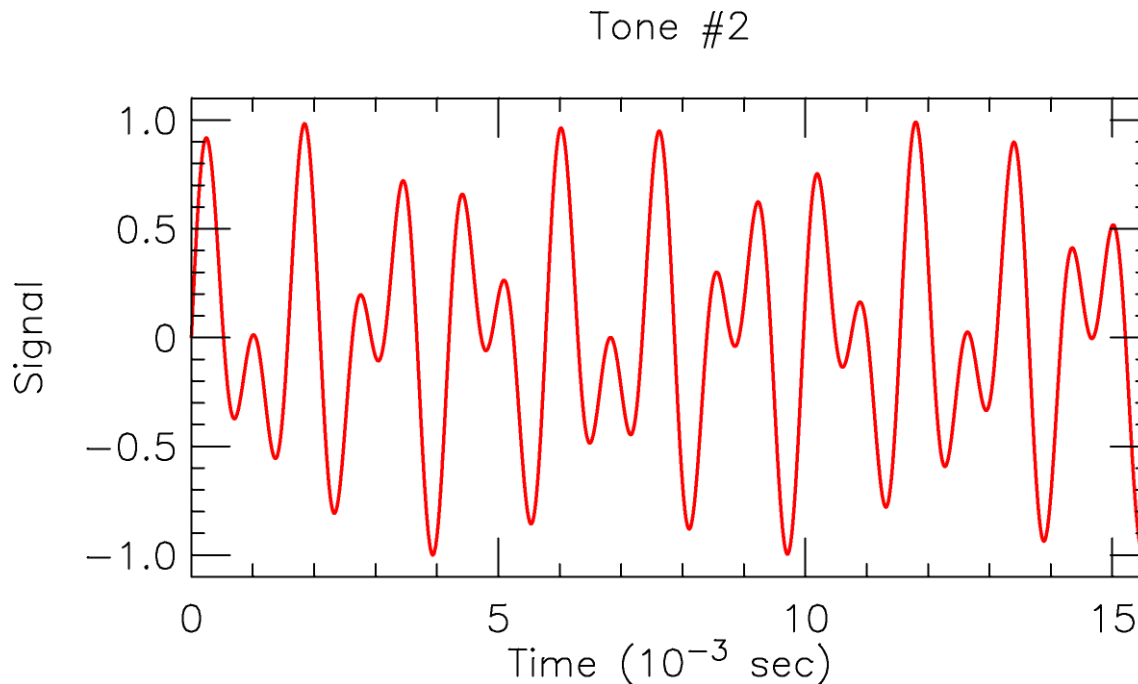
Fourier Series and Spectral Analysis

Plot tone02.dat, from 0 to 0.0156 sec :

AOSC 652: Analysis Methods in AOSC

Fourier Series and Spectral Analysis

Plot tone02.dat, from 0 to 0.0156 sec :

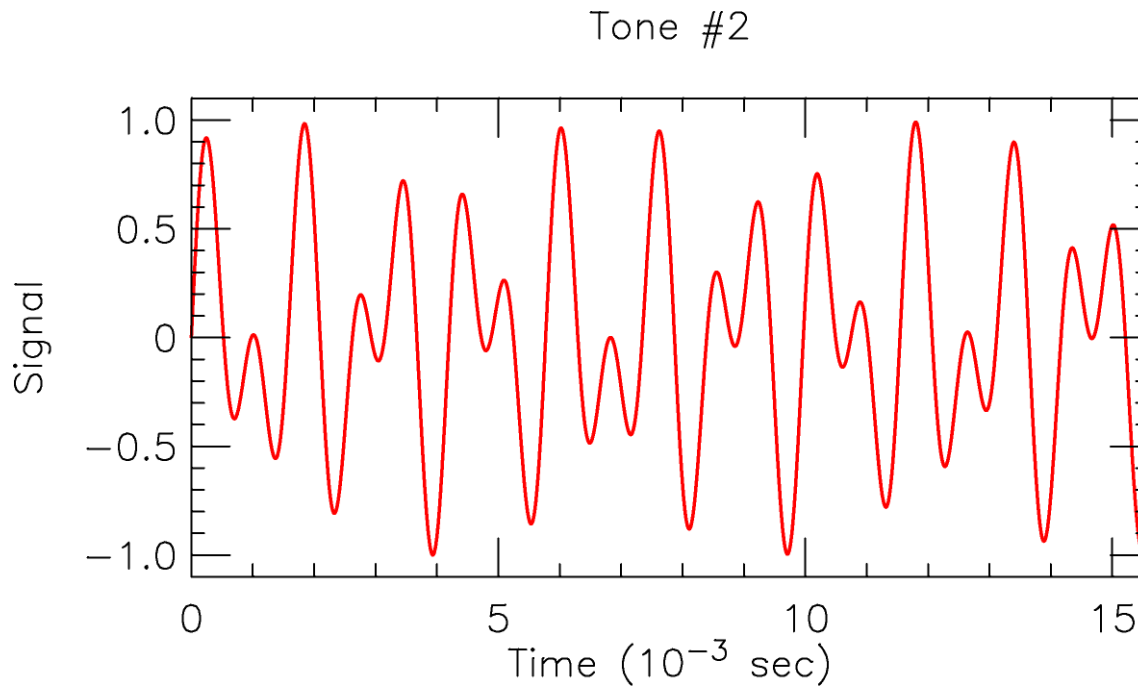


tone02.dat generated with FORTRAN program `~rjs/aosc652/week_09/tone02_gen.f`

AOSC 652: Analysis Methods in AOSC

Fourier Series and Spectral Analysis

Plot tone02.dat, from 0 to 0.0156 sec :



Does this look familiar?

If so, what is it?

AOSC 652: Analysis Methods in AOSC

Fourier Series and Spectral Analysis

Compute power spectrum of tone02.dat, from frequencies of 600 to 1300 Hz, every 1 Hz

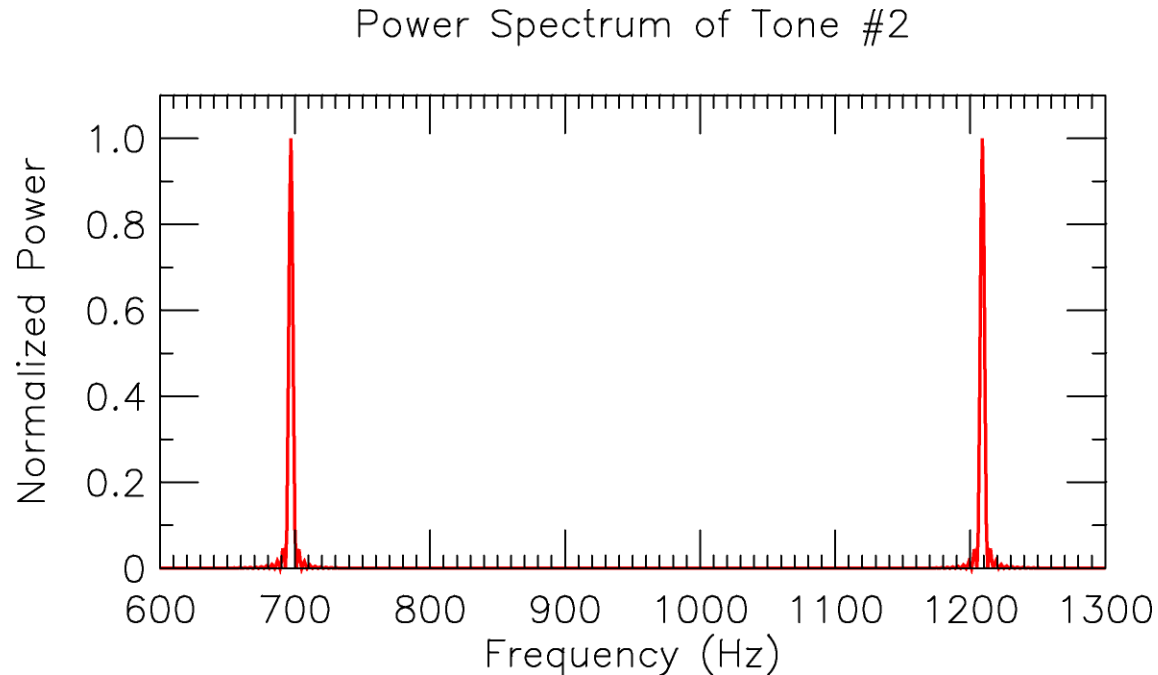
Plot the result from 600 to 1300 Hz:

AOSC 652: Analysis Methods in AOSC

Fourier Series and Spectral Analysis

Compute power spectrum of tone02.dat, from frequencies of 600 to 1300 Hz, every 1 Hz

Plot the result from 600 to 1300 Hz:

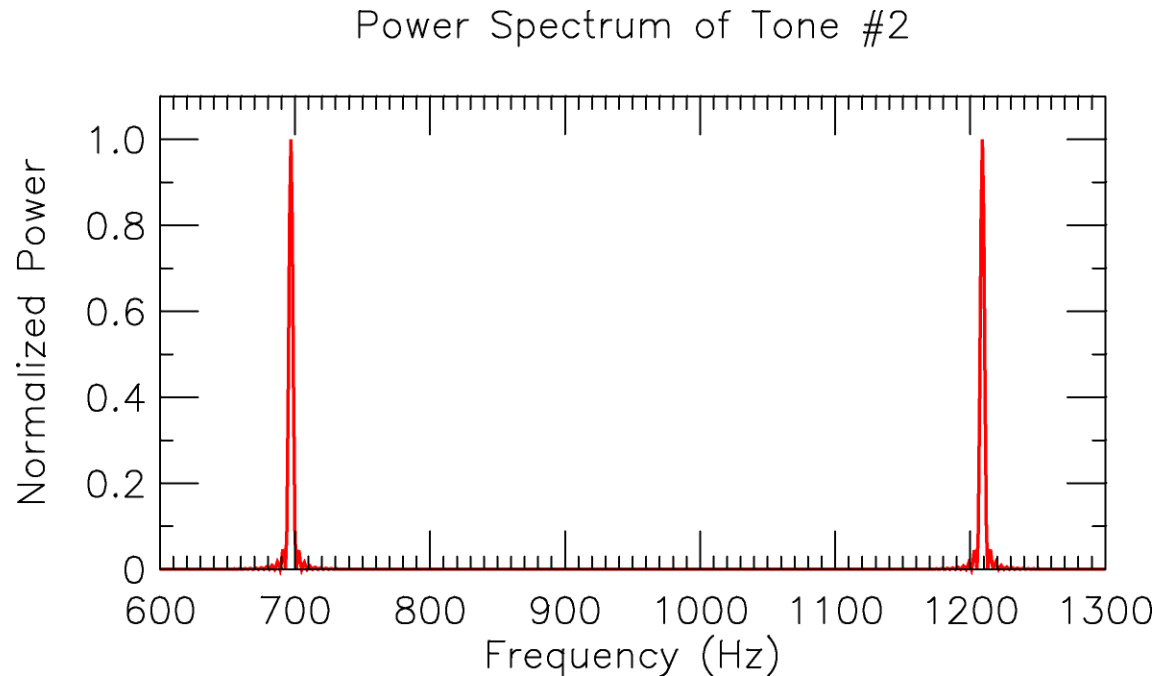


AOSC 652: Analysis Methods in AOSC

Fourier Series and Spectral Analysis

Compute power spectrum of tone02.dat, from frequencies of 600 to 1300 Hz, every 1 Hz

Plot the result from 600 to 1300 Hz:



tone02.dat generated with FORTRAN program `~rjs/aosc652/week_09/tone02_gen.f`

AOSC 652: Analysis Methods in AOSC

Fourier Series and Spectral Analysis

Mini-assignment for Wed:

Solar activity/sunspots (time scale of decades)

Copy file `~rjs/aosc652/week_09/sunspot_number_monthly.dat` to your work area.

This file contains *Sunspot Number vs Time*, from 1749 to present

Using program `fourier_analysis.f`, find the Power Spectrum of this time series, for time periods from **7 to 15 years** every **0.1 years** and:

- a) display the results
- b) answer a few questions about the results

Note: Power Spectrum must be found using the instructor provided FORTRAN program. Plots can be produced using either `hpptld`, MATLAB, Python, or IDL

Mini-assignment due **start of class on Wed**: no exceptions

AOSC 652: Analysis Methods in AOSC

Fourier Series and Spectral Analysis

Readings for Wed:

Everyone: pages 63 to 73 of Muller and MacDonald

Team Python: Chapter 18 of DeCaria

Team IDL: Chapter 25 of IDL

We'll all meet here at the start of class on Wed;
we'll split about 20 to 30 mins into class