

1060-710

# Mathematical and Statistical Methods for Astrophysics

## Problem Set 5

Assigned 2009 October 15

Due 2009 October 22

**Show your work on all problems!** Be sure to give credit to any collaborators, or outside sources used in solving the problems. Note that if using an outside source to do a calculation, you should use it as a reference for the method, and actually carry out the calculation yourself; it's not sufficient to quote the results of a calculation contained in an outside source.

## 1 Colored Noise

This problem should be done in your favorite numerical analysis environment. Please turn in a printout of the final set of commands you used as well as the plots themselves.

Consider a  $T = 16$  sec of data sampled at  $\frac{1}{\delta t} = 1024$  Hz, so that  $N = 16384$ .

- Generate an  $N$ -point series  $\{x_j\}$  of random samples drawn from a distribution with zero mean and unit variance, i.e.,  $\langle x_j \rangle = 0$  and  $\langle x_j x_\ell \rangle = \delta_{j\ell}$ . (This can be done in matplotlib with `randn(N)`.) Plot  $x_j$  versus  $t_j - t_0 = j\delta t$  for  $t_j - t_0$  from 0 to 16 sec.
- Take the discrete Fourier transform  $\hat{x}_k$  and plot  $|\hat{x}_k|$  versus  $f_k = k\delta f = k/T$  for  $f_k$  from  $-512$  Hz to  $512$  Hz.
- Calculate the average

$$\frac{1}{N} \sum_{k=-N/2}^{N/2-1} |\hat{x}_k|^2 \quad (1.1)$$

How does it compare to your theoretical expectations for  $\langle |\hat{x}_k|^2 \rangle$ ?

- Construct  $\hat{y}_k = \hat{x}_k e^{-f_k^2/2\sigma_f^2}$ , where  $\sigma_f = 16$  Hz, for  $k \in [-N/2, N/2 - 1]$  and plot  $|\hat{y}_k|$  versus  $f_k$  for  $f_k$  from  $-512$  Hz to  $512$  Hz.
- Take the inverse discrete Fourier Transform to get  $y_j$  and plot  $y_j$  versus  $t_j - t_0$  for  $t_j - t_0$  from 0 to 16 sec.

## 2 Power Spectral Density

Consider a single random variable  $\psi$  which is equally likely to fall anywhere between 0 and  $2\pi$ , so that expectation values of random variables whose only randomness comes from  $\psi$  can be calculated as

$$\langle F(\psi) \rangle = \frac{1}{2\pi} \int_0^{2\pi} F(\psi) d\psi . \quad (2.1)$$

Let  $x(t) = A \cos(2\pi f_0 t + \psi)$  for some fixed  $A$  and  $f_0$ .

- Find  $\langle x(t) \rangle$  and  $\langle x(t)x(t') \rangle$  and show that  $x(t)$  is wide-sense stationary.
- Find the power spectral density  $P_x(f)$ .

## 3 Averages and Expectation Values

Let  $\{x_k | k = 0 \dots N - 1\}$  be a set of  $N$  uncorrelated random variables all drawn from the same distribution with (possibly unknown) mean  $\mu$  and variance  $\sigma^2$  so that the expectation values are

$$\langle x_k \rangle = \mu \quad (3.1a)$$

$$\langle (x_k - \mu)(x_\ell - \mu) \rangle = \delta_{k\ell} \sigma^2 \quad (3.1b)$$

- Consider the average of the  $N$  instantiations

$$\bar{x} = \frac{1}{N} \sum_{k=0}^{N-1} x_k \quad (3.2)$$

and show that its expectation value  $\langle \bar{x} \rangle$  is equal to  $\mu$ . (This is pretty easy to show if you remember that the expectation value is a linear operation, so that  $\langle \alpha + \beta \rangle = \langle \alpha \rangle + \langle \beta \rangle$ .) This means that  $\bar{x}$  is an *unbiased estimator* of the mean of the underlying distribution, even though it's constructed from a finite number of samples from that distribution.

- Calculate the expected variance of  $\bar{x}$ , i.e.,

$$\langle (\bar{x} - \langle \bar{x} \rangle)^2 \rangle = \langle (\bar{x} - \mu)^2 \rangle = \text{what?} \quad (3.3)$$

- Suppose we know the exact value of  $\mu$  (via some sort of physical principle or something); we could use

$$\overline{(x - \mu)^2} = \frac{1}{N} \sum_{k=0}^{N-1} (x_k - \mu)^2 \quad (3.4)$$

as an estimator of the underlying variance  $\sigma^2$ . Show that this is an unbiased estimator, i.e., that its expectation value is indeed  $\sigma^2$ :

$$\langle \overline{(x - \mu)^2} \rangle = \sigma^2 \quad (3.5)$$

- d) Now suppose the true mean  $\mu = \langle x \rangle$  is not known, and must also be estimated from the same  $N$  data points. We can consider

$$\overline{(x - \bar{x})^2} = \frac{1}{N} \sum_{k=0}^{N-1} (x_k - \bar{x})^2 \quad (3.6)$$

as a potential estimator of  $\sigma^2$ . Calculate its expectation value

$$\left\langle \overline{(x - \bar{x})^2} \right\rangle, \quad (3.7)$$

keeping in mind that both  $x_k$  and  $\bar{x}$  are random variables. This will *not* be equal to  $\sigma^2$  which means (3.6) is a *biased* estimator.

- e) Modify (3.6) to produce an unbiased estimator of  $\sigma$  which can be calculated from only the samples  $\{x_k\}$ . (This can't include the unknown actual value of  $\mu$  nor any expectation values, only averages calculated from the actual samples  $\{x_k\}$ .)