# EE3731C: Signal Processing Methods

**Tutorial II-3** 



A one dimensional random walk starts at  $S_0 = 0$  and at each step moves by  $\pm 1$  with equal probability. To define this walk formally, take independent random variables  $Z_1$ ,  $Z_2$ ,..., each of which is 1 with probability 1/2 and -1 with probability 1/2, and set

$$S_n = Z_1 + Z_2 + ... + Z_n$$

This sequence  $\{S_n\}$  is called the simple random walk on the integers.

- (i) Find  $E[S_n]$  and  $E[S_n^2]$ .
- (ii) What is the root mean square displacement of the walk after *n* steps?

The PDF of 
$$Z_i$$
 is  $P(Z_i) = \begin{cases} 1/2 & \text{if } Z_i = 1 \\ 1/2 & \text{if } Z_i = -1 \end{cases}$ 

So 
$$E[Z_i] = \frac{1}{2} \times 1 + \frac{1}{2} \times (-1) = 0$$
  $S_n = Z_1 + Z_2 + \dots + Z_n$   $E[Z_i^2] = \frac{1}{2} \times (1)^2 + \frac{1}{2} \times (-1)^2 = 1$   $E[Z_i Z_j] = 0, \ \forall i \neq j$ 

$$E[S_n] = E[Z_1 + Z_2 + \dots + Z_n] = \sum_{i=1}^n E[Z_i] = 0$$

$$E[S_n^2] = E[(Z_1 + Z_2 + \dots + Z_n)^2] = \sum_{i=1}^n E[Z_i^2] = n$$

The root mean square displacement of the walk after n steps is  $\sqrt{n}$ 

Let w[n] denote a Gaussian white noise sequence with mean zero and variance 1. Determine the mean and autocorrelation functions of x[n] in the following cases.

(i) 
$$x[n] = 0.8 x[n-1] + 0.5 w[n]$$

(ii) 
$$x[n] = 0.4w[n] - 0.1w[n-1]$$

Hints: Since x[n] is generated by the same rule at all times, x[n] will be stationary. When x[n-1] is a function past inputs and past noise values, it is independent of w[n] because the noise is independent.

(i) 
$$x[n] = 0.8 x[n-1] + 0.5 w[n]$$

$$m_{x} = E\{0.8x[n-1]+0.5w[n]\}$$

$$= 0.8E\{x[n-1]\}+0.5E\{w[n]\}$$

$$= 0.8m_{x}$$

$$= 0.8m_{x}$$

$$E\{w[n]\}=0$$

$$E\{w[n]w[n+m]\}=\delta[m]$$

$$m_x = 0.8 m_x \Rightarrow m_x = 0$$

Since x[n-1] is a function past inputs and past noise values, and is therefore independent of w[n].

$$E\{x[n-1]w[n]\} = E\{x[n-1]\}E\{w[n]\} = 0$$

(i) 
$$x[n] = 0.8 x[n-1] + 0.5 w[n]$$

$$R_{x}[m] = E\{(0.8x[n-1]+0.5w[n])(0.8x[n+m-1]+0.5w[n+m])\}$$

$$= 0.64R_{x}[m]+0.25R_{w}[m]+0.4R_{wx}[m-1]+0.4R_{xw}[m+1]$$

$$0.36R_{x}[m] = 0.25R_{w}[m]+0.4R_{wx}[m-1]+0.4R_{xw}[m+1]$$

$$R_{x}[m] = \frac{25}{36}\delta[m]+\frac{10}{9}R_{wx}[m-1]+\frac{10}{9}R_{xw}[m+1]$$

$$E\{x[n-1]w[n]\} = E\{x[n-1]\}E\{w[n]\} = 0$$

$$R_{xw}[m] = 0 \quad \forall m > 0$$

$$R_{xw}[m] = 0 \quad \forall m < 0$$

$$R_{xw}[m] = R_{wx}[-m]$$

(i) 
$$x[n] = 0.8 x[n-1] + 0.5 w[n]$$

$$m = 0$$
:  $R_{wx}[0] = E\{w[n]x[n]\} = E\{w[n](0.8x[n-1]+0.5w[n])\} = 0.5$ 

$$m > 0: R_{wx}[m] = E\{w[n]x[n+m]\}$$
  
=  $E\{[w[n](0.8x[n+m-1]+0.5w[n+m])]\} = 0.8R_{wx}[m-1]$ 

$$R_{wx}[m] = \begin{cases} 0.5 \times (0.8)^m, m \ge 0 \\ 0, & \text{otherwise} \end{cases} \Rightarrow R_{wx}[m] = 0.5 \times (0.8)^m u[m]$$

$$R_{x}[m] = \frac{25}{36} \delta[m] + \frac{10}{9} R_{wx}[m-1] + \frac{10}{9} R_{xw}[m+1] \qquad R_{xw}[m] = R_{wx}[-m]$$

$$R_{x}[m] = \frac{25}{36} \delta[m] + \frac{5}{9} \times \left[ (0.8)^{(m-1)} u[m-1] + (0.8)^{-(m+1)} u[-(m+1)] \right] = \frac{25}{36} (0.8)^{|m|}$$

(ii) 
$$x[n] = 0.4w[n] - 0.1w[n-1]$$

$$m_x = E\{0.4w[n] - 0.1w[n-1]\} = 0.4E\{w[n]\} - 0.1E\{w[n-1]\} = 0$$

$$R_{x}[m] = E\{(0.4w[n] - 0.1w[n-1])(0.4w[n+m] - 0.1w[n+m-1])\}$$

$$= 0.16R_{w}[m] - 0.04R_{w}[m-1] - 0.04R_{w}[m+1] + 0.01R_{w}[m]$$

$$= 0.17R_{w}[m] - 0.04R_{w}[m-1] - 0.04R_{w}[m+1]$$

$$= 0.17\delta[m] - 0.04\delta[m-1] - 0.04\delta[m+1]$$

Let w[n] denote a Gaussian white noise sequence with mean zero and variance 1. Find the power spectrum of the random process x[n] in the following cases.

(i) 
$$x[n] = 0.8 x[n-1] + 0.5 w[n]$$

(ii) 
$$x[n] = 0.4w[n] - 0.1w[n-1]$$

#### Two alternative solutions:

- Find the autocorrelation function and take its Fourier transform.
- Find the transfer function H(z) and use the following relationship to obtain the power spectrum.  $S_v(e^{j\omega}) = |H(e^{j\omega})|^2 S_v(e^{j\omega})$

(i) 
$$x[n] = 0.8 x[n-1] + 0.5 w[n] S_w(e^{j\omega}) = 1$$

Take Z transform of both sides, we get:

$$X(z) = 0.8X(z)z^{-1} + 0.5W(z)$$

$$X(z) - 0.8X(z)z^{-1} = 0.5W(z)$$

$$H(z) = \frac{X(z)}{W(z)} = \frac{0.5}{1 - 0.8z^{-1}} \qquad |H(e^{j\omega})|^2 = H(e^{j\omega})H(e^{-j\omega})$$

$$= |H(e^{j\omega})|^2 S_w(e^{j\omega}) = |H(e^{j\omega})|^2 = \frac{0.5}{(1 - 0.8z^{-j\omega})} \frac{0.5}{(1 - 0.8z^{-j\omega})}$$

$$S_{x}(e^{j\omega}) = |H(e^{j\omega})|^{2} S_{w}(e^{j\omega}) = |H(e^{j\omega})|^{2} = \frac{0.5}{(1 - 0.8e^{-j\omega})(1 - 0.8e^{j\omega})}$$

$$S_{x}(e^{j\omega}) = \frac{0.25}{(1 - 0.8e^{-j\omega})(1 - 0.8e^{j\omega})} = \frac{0.25}{1.64 - 1.6\cos(\omega)}$$

(ii) 
$$x[n] = 0.4w[n] - 0.1w[n-1]$$

Take Z transform of both sides, we get:

$$X(z) = 0.4W(z) - 0.1W(z)z^{-1}$$

$$X(z) = (0.4 - 0.1z^{-1})W(z)$$

$$H(z) = \frac{X(z)}{W(z)} = 0.4 - 0.1z^{-1}$$

$$S_{w}(e^{j\omega}) = 1$$

$$S_{x}(e^{j\omega}) = |H(e^{j\omega})|^{2} S_{w}(e^{j\omega}) = |H(e^{j\omega})|^{2} = (0.4 - 0.1e^{-j\omega})(0.4 - 0.1e^{+j\omega})$$

$$S_x(e^{j\omega}) = 0.17 - 0.04e^{-j\omega} - 0.04e^{+j\omega} = 0.17 - 0.08\cos(\omega)$$

Let a WSS random process x[n] be the input to an LTI system with known impulse response h[n]. Suppose that the input x[n] is not observable, but we can measure the output y[n].

i.Propose an estimate of the power spectrum of y[n] (i.e., describe the processing you would apply to the observed output y[n] to estimate the power spectrum  $S_y(e^{j\omega})$ ). Hint: Consider using the DFT to estimate the "power" of the output at each frequency.

ii. How would you use the estimate of  $S_y(e^{j\omega})$  to estimate the power spectrum of the unobservable input x[n]?

i. Propose an estimate of the power spectrum of y[n].

Suppose y[n] is of length N

$$S_{y}(e^{j\omega}) = \frac{1}{N} \left| \sum_{n=0}^{N-1} y[n] e^{-j\omega n} \right|^{2}$$

Averaged Periodogram:

- 1) Divide y[n] into K non-overlapping segments of length M.
- 2) For each segment, compute its *M*-point DFT to obtain its periodogram.  $S_{y}(e^{j\omega_{k}}) = \frac{1}{M} \left| \sum_{n=0}^{M-1} y[n] e^{-j\omega_{k}n} \right|^{2}, \ \omega_{k} = \frac{2\pi k}{M}$
- 3) Average the periodograms over all *K* segments to obtain the averaged periodogram.

ii. How would you use the estimate of  $S_y(e^{j\omega})$  to estimate the power spectrum of the unobservable input x[n]?

$$S_y(f) = |H(f)|^2 S_x(f)$$

$$S_{y}(e^{j\omega}) = |H(e^{j\omega})|^{2} S_{x}(e^{j\omega})$$

Since the impulse response h[n] is known, we can compute

$$\left|H(e^{j\omega})^{2}\right| = \left|\sum_{n=-\infty}^{\infty} h[n]e^{-j\omega n}\right|^{2}$$

A zero mean noise process x[n] has the following autocorrelation function

$$R_{x}(m) = \left(\frac{1}{2}\right)^{|m|}$$

We process x[n] with a linear time-invariant system that satisfies the following difference equation:

$$y[n] = x[n+1] + x[n-1]$$

What is the autocorrelation function of y[n]?

$$y[n] = x[n+1] + x[n-1]$$
  $R_x(m) = \left(\frac{1}{2}\right)^{|m|}$ 

### Autocorrelation function of y[n]:

$$R_{y}[m] = E\{y[n]y[n+m]\}$$

$$= E\{(x[n+1]+x[n-1])(x[n+m+1]+x[n+m-1])\}$$

$$= 2R_{x}[m]+R_{x}[m-2]+R_{x}[m+2]$$

Alternatively, 
$$R_y[m] = R_x[m] * h[m] * h[-m]$$
  
$$h[m] = \delta[m+1] + \delta[m-1]$$