

1.

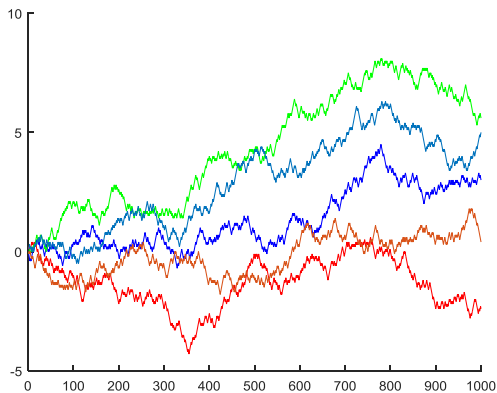
```
% problem 1
clear;
close all;
clc;

p = 0.5;
q = 1-p;
n = 1000;
trials = 10000;
s = 0.1;

x = cumsum((rand(n,trials)<p).*2*s-s); % generate 10000 random walks. each
column represent one random walk.

figure(1);
hold on;
plot(x((1:n),1), 'r');
plot(x((1:n),2), 'b');
plot(x((1:n),3), 'g');
plot(x((1:n),4));
plot(x((1:n),5));
hold off;

mean_sim = mean(x(n,(1:trials)))
mean_est = n*(p-q)*s
var_sim = var(x(n,(1:trials)))
var_est = 4*n*p*q*s^2
```



```
mean_sim = -0.0465
mean_est = 0
var_sim = 10.1783
var_est = 10.0000
```

2.

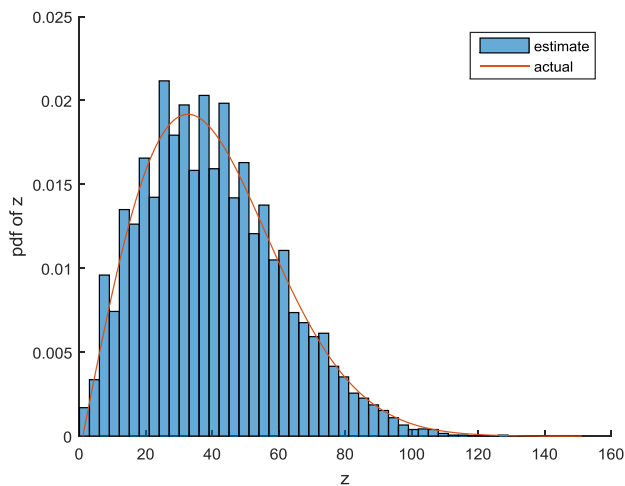
```
% problem 2
clear;
close all;
clc;

p = 0.5;
q = 1-p;
n = 1000; % the number of steps in a random walk
T = 0.5; % each time step length
t = n*T;
trials = 10000;
s = 1;
alpha = s^2/T;

x = cumsum((rand(n,trials)<p).*2*s-s); % generate 10000 random walks. each
column represent one random walk
y = cumsum((rand(n,trials)<p).*2*s-s);
x_est = x(n,1:trials); % end points of 10000 random walk
y_est = y(n,1:trials);
z_est = sqrt(x_est.^2+y_est.^2);

x_actual = 0:1:150;
z_actual = raylpdf(x_actual,sqrt(alpha*t)); % parameters: (x, std)

figure(1); hold on;
histogram(z_est,'normalization','pdf'); % estimated plot for z(rayleigh)
plot(z_actual);
legend('estimate','actual');
xlabel('z');
ylabel('pdf of z');
```



Analytical parts for 1,2

1. b) because each step in a random walk is likely to increase more than to decrease

c) because the distribution of random walks will spread more

2. By eq(10-52) on p. 446 when  $t \gg T$ . (Wiener process)

$$f_x(x, t) = \frac{1}{\sqrt{2\pi\alpha t}} e^{-x^2/2\alpha t} \quad f_y(y, t) = \frac{1}{\sqrt{2\pi\alpha t}} e^{-y^2/2\alpha t}$$

Since  $z(t) = \sqrt{x^2(t) + y^2(t)}$ , by eq(6-70) on p. 190

$$f_z(z) = \frac{z}{\sigma^2} e^{-z^2/2\sigma^2} U(z)$$

In this case the variances of  $f_x(x, t)$  and  $f_y(y, t)$  are  $\alpha t$  instead of  $\sigma^2$

$$\text{Thus } f_z(z, t) = \frac{z}{\alpha t} e^{-z^2/2\alpha t} U(z) \quad S^2 = \alpha T \Rightarrow \alpha = \frac{S^2}{T}$$

3.

$$3. \quad H_1(s) = \frac{1}{LCs^2 + RCs + 1} \quad H_1(w) = \frac{1}{-LCw^2 + RCjw + 1}$$

$V(t)$  is output,  $n_e(t)$  is input

by eq. (9-149), (10-73), (10-74)

$$S_v(w) = S_{n_e}(w) |H(w)|^2 = \frac{2kTB}{(1-w^2LC)^2 + w^2R^2C^2} \leftarrow$$

$$|H(w)|^2 = H(w) H^*(w) = \frac{1}{\{ (1-w^2LC) + RCjw \} \{ (1-w^2LC) - RCjw \}} = \frac{1}{(1-w^2LC)^2 + w^2R^2C^2}$$

$$Z(s) = \frac{Ls + R}{LCs^2 + RCs + 1} \quad Z(jw) = \frac{(R + Ljw)(1 - w^2LC) - RCjw}{(-LCw^2 + RCjw + 1)(1 - w^2LC) - RCjw}$$

$$Z(jw) = \frac{R - w^2LCR - R^2Cjw + Ljw - w^3L^2Cj + R^2Lw^2}{(1 - w^2LC)^2 + w^2R^2C^2}$$

$$\operatorname{Re} Z(jw) = \frac{R}{(1 - w^2LC)^2 + w^2R^2C^2}$$

Thus,  $S_v(w) = 2kT \cdot \operatorname{Re} Z(jw)$  Verified Nyquist Theorem (10-75)

$$H_2(s) = \frac{1}{R + Ls} \quad H_2(w) = \frac{1}{R + Ljw} \quad |H_2(w)|^2 = \frac{1}{R^2 + L^2w^2}$$

$$S_i(w) = S_{n_e}(w) |H_2(jw)|^2 = \frac{2kTB}{R^2 + L^2w^2}$$

Short circuit admittance

$$Y(s) = \frac{1}{R + Ls} \quad Y(jw) = \frac{1}{(R + Ljw)(R - Ljw)} = \frac{R - Ljw}{R^2 + L^2w^2}$$

$S_i(w) = 2kT \operatorname{Re} Y(jw)$  Verified Nyquist (10-78)

4.

$$4. \quad p(z_n=1)=p, \quad p(z_n=-1)=q=1-p \quad S=1$$

$$X_n = \sum_{i=1}^n z_i$$

$$a) \quad X_4 = z_1 + z_2 + z_3 + z_4 = -2 \quad p(X_4=-2) = p q^3$$

$$b) \quad E[X_n] = n(p-q) \quad \text{by eq (10-55)}$$

$$c) \quad \text{Var } X(n) = 4npq$$

$$d) \quad R_{XX}(n, m) = E[X_n X_m] = \sum_{i=1}^n \sum_{j=1}^m E[z_i z_j]$$

$$= \sum_{\substack{i=1 \\ i=j}}^n \sum_{j=1}^m E[z_i z_j] + \sum_{\substack{i=1 \\ i \neq j}}^n \sum_{\substack{j=1 \\ j \neq i}}^m E[z_i z_j] = \sum_{i=1}^{\min(n, m)} E[z_i^2] + \sum_{\substack{i=1 \\ i \neq j}}^n \sum_{j=1}^m E[z_i] E[z_j]$$

$$E[z_i^2] = p(1)^2 + q(-1)^2 = p+q = 1$$

$$= \underbrace{\min(n, m) \times 1}_{\substack{\# \text{ of the occurrence} \\ i=j}} + \underbrace{(nm - \min(n, m))}_{\substack{\# \text{ of the occurrence} \\ i \neq j}} (p-q)^2$$

e) no, because the mean of  $X_n$  changes as  $n$  changes