

Homework 3

Chapter 5, #13

A drunkard executes a "random walk" in the following way :
each minute he takes a step north or south , with prob. $\frac{1}{2}$ each,
and his successive step directions are independent. His step
length is 50 cm. Use the CLT to approximate the probability
distribution of his location after 1 hour. Where is he most likely
to be ?

Solution Let $X_i = \begin{cases} +50 & \text{with prob. } \frac{1}{2} \\ -50 & \text{with prob. } \frac{1}{2} \end{cases}$

Then the location of the drunkard after 1h is

$$Y = \sum_{i=1}^n X_i, \text{ where } n = \underline{60} \text{ (number of steps)}$$

According to the CLT, $nY = \bar{X}_n \sim N(\mu, \frac{\sigma^2}{n})$,

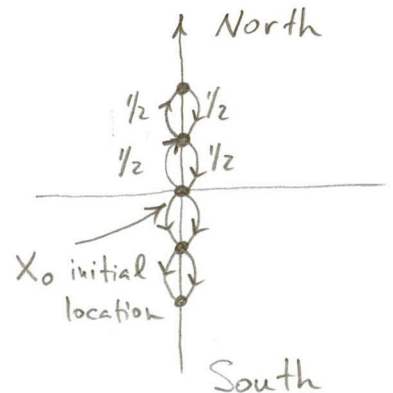
where $\mu = E[X_i] = 0$

$$\sigma^2 = V[X_i] = E[X_i^2] = 2500\frac{1}{2} + 2500\frac{1}{2} = 2500$$

$$\text{Thus, } nY \sim N(0, \frac{2500}{n}) \Rightarrow Y \sim N(0, n \cdot 2500) = \boxed{N(0, 15 \cdot 10^4)}$$

He is most likely to be at $\arg \max_x N(x | 0, 15 \cdot 10^4) = 0$.

Thus, he is most likely to be where he started his walk.



Chapter 5, #15

Suppose that you bet \$5 on each of a sequence of $n=50$ independent fair games. Use the CLT to approximate the probability that you will lose more than \$75.

Solution Let $X_i = \begin{cases} 5 & \text{w.p. } 1/2 \\ -5 & \text{w.p. } 1/2 \end{cases} \quad i=1, \dots, n$

The total payoff is $T = \sum_{i=1}^n X_i$

$$IP(T < -75) = IP(n \bar{X}_n < -75) = IP\left(\bar{X}_n < -\frac{75}{n}\right) \Leftrightarrow$$

$$\bar{X}_n \stackrel{\circ}{\sim} N\left(\mu, \frac{\sigma^2}{n}\right), \quad \mu = E[X_i] = 0$$

$$\sigma^2 = V[X_i] = 25 \cdot \frac{1}{2} + 25 \cdot \frac{1}{2} = 25$$

$$\Leftrightarrow IP\left(\frac{(\bar{X}_n - \mu)\sqrt{n}}{\sigma} < \frac{(-\frac{75}{n} - \mu)\sqrt{n}}{\sigma}\right) \approx \Phi\left(-\frac{75}{\sigma\sqrt{n}}\right) = \Phi(-2.121) \approx \underline{\underline{0.017}}$$

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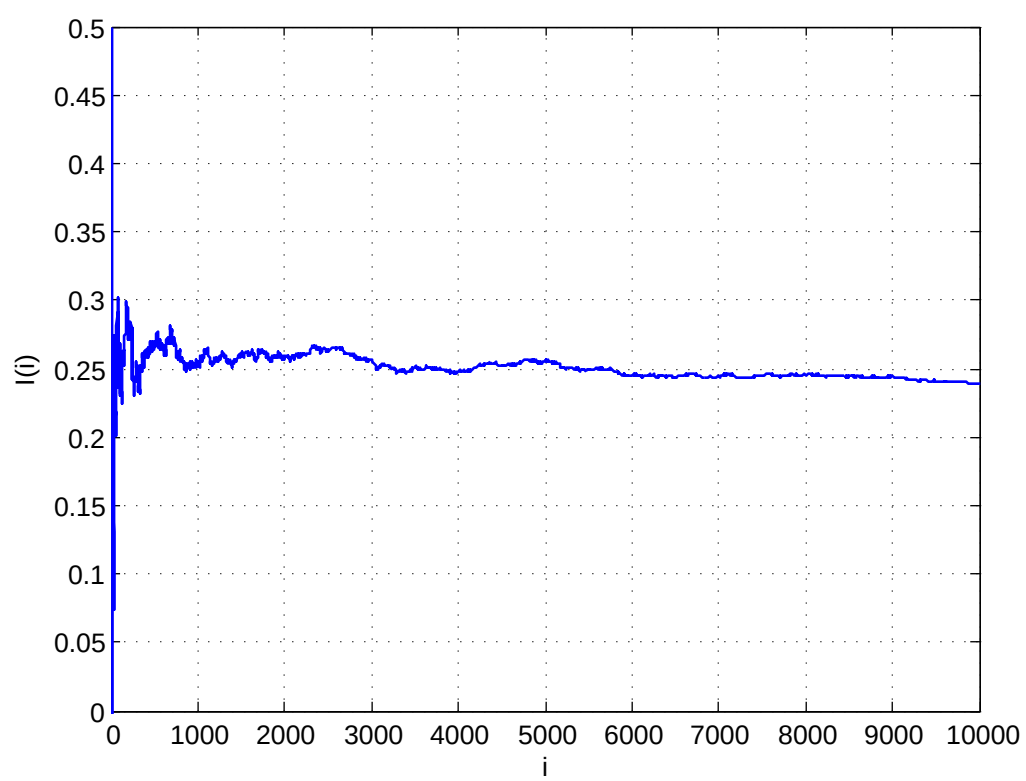
% Matlab source code.
% Problem #19, Chapter 5.
% Part (a): Use the Monte Carlo method with n=100 and n=1000 to
estimate
%  $\int_0^1 \cos(2\pi x) dx$ . Compare the estimates with the exact answer.
clear;
n1=100;           % number of Monte Carlo samples for the 1st estimate
n2=1000;          % number of Monte Carlo samples for the 2nd estimate
x1=rand(1,n1); % 1xn1 vector with components  $\sim U[0,1]$ 
x2=rand(1,n2); % 1xn2 vector with components  $\sim U[0,1]$ 
I_true=0;         % true value of the integral
I1=mean(cos(2*pi.*x1)); % 1st Monte Carlo estimate
I2=mean(cos(2*pi.*x2)); % 2nd Monte Carlo estimate

% Output: I1=-0.0255547, I2=-0.007258.
% The 2nd estimate is more accurate

% Part (b): Use Monte Carlo to evaluate  $\int_0^1 \cos(2\pi x^2) dx$ 
n=10^4;          % number of Monte Carlo samples
x=rand(1,n); % 1xn vector with components  $\sim U[0,1]$ 
for i=1:n
    I(i)=mean(cos(2*pi.*(x(1:i).^2)));
    % I(i) is the Monte Carlo estimate that is based on i samples
end

% Output: the integral is approximately 0.24
% See also the Figure that show convergence of I(i) to 0.24 as i
increases

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• Chapter 6

③ $\bar{X}_n = \frac{1}{n} \sum_{i=1}^n X_i$, $X_i \stackrel{iid}{\sim} \mathcal{N}(0,1)$, $n=16$

Find c such that $IP(|\bar{X}_n| < c) = \frac{1}{2}$

Solution : Th.1: $\bar{X}_n \sim \mathcal{N}(0, \frac{1}{n}) \Rightarrow Z = \sqrt{n} \bar{X}_n \sim \mathcal{N}(0,1)$

$$IP(|\bar{X}_n| < c) = IP(|\sqrt{n} \bar{X}_n| < \sqrt{n} \cdot c) = IP(|Z| < \sqrt{n} \cdot c)$$

$$= \Phi(\sqrt{n}c) - \Phi(-\sqrt{n}c) = \Phi(\sqrt{n}c) - (1 - \Phi(\sqrt{n}c)) =$$

$$= 2\Phi(\sqrt{n}c) - 1$$

Therefore : $IP(|\bar{X}_n| < c) = \frac{1}{2} \Leftrightarrow 2\Phi(\sqrt{n}c) - 1 = \frac{1}{2}$

$$\sqrt{n}c = \Phi^{-1}\left(\frac{3}{4}\right)$$

$$c = \frac{1}{4} \Phi^{-1}\left(\frac{3}{4}\right) \approx 0.17$$