

Week 1: Introduction – A lower bounded random walk

Consider a game in which players bet \$1 and win \$1 with probability p or lose their bets with probability $q = 1 - p$. The wealth of a player as a function of time is a stochastic process. Indeed, if the player's wealth at time t is $w(t)$ the wealth at time $t + 1$ is either $w(t) + 1$ (with probability p) or $w(t) - 1$ (with probability q). Formally, we can write this as

$$\begin{aligned}\mathbb{P}[w(t + 1) = w(t) + 1 \mid w(t)] &= p, \\ \mathbb{P}[w(t + 1) = w(t) - 1 \mid w(t)] &= q.\end{aligned}\tag{1}$$

The first equation is read “the probability of $w(t + 1)$ being equal to $w(t) + 1$ given $w(t)$ is p .” The expressions in (??) are true as long as $w(t) \neq 0$. When $w(t) = 0$ the gambler is ruined and $w(t + 1) = 0$. A rather sophisticated, but sometimes useful way of expressing this fact is to write

$$\mathbb{P}[w(t + 1) = 0 \mid w(t) = 0] = 1.\tag{2}$$

This stochastic process is sometimes called a *lower bounded random walk*. That is because the wealth can be interpreted as the position on a line and wealth variations as steps taken randomly to the left or to the right. The origin is home, in that if the walker reaches 0 it no longer moves. We saw in class that if $p > 1/2$, then $w(t)$ is likely to diverge, making this a good game to play. But in this exercise, we let p take any value.

A Simulation of a realization of the process. Write a function that takes as inputs the probability p , the initial wealth $w(0) = w_0$, and a maximum number of rounds T . The function must return a vector of length *at most* $T + 1$ containing the player's wealth history $w(0), \dots, w(T)$, computed according to the stochastic process described in (??) and (??). If the wealth is depleted at time $t < T$, i.e., if $w(t) = 0$ for some $t < T$, then the function should return a vector of length $t + 1$ with the player's wealth history up to time t , i.e., $w(0), \dots, w(t)$. The function must also return a boolean variable that distinguishes between runs that resulted in a broke player and those that did not. Show plots of simulated processes with $w_0 = 20$ and $T = 10^3$ for $p = 0.25$, $p = 0.5$, and $p = 0.75$.

B Probability of ruin. Fixing $p = 0.55$ and $w_0 = 10$, use your function from part ?? to estimate the probability $B(p, w_0)$ of eventually going broke (or the equivalent random walk eventually reaching home), i.e., the probability of having $w(t) = 0$ for some t . Because once the player's wealth reaches zero it stays zero forever, this probability can be written as the limit

$$B(p, w_0) = \lim_{t \rightarrow \infty} \mathbb{P}[w(t) = 0 \mid w(0) = w_0].\tag{3}$$

Strictly speaking, you would need to run the simulation forever to make sure the gambler does not run out of money eventually. However, you can truncate simulations at time $T = 100$ for this exercise. Doing this, you are computing the probability of reaching home between times 0 and T , which we assume is a good approximation for the probability of reaching home between times 0

and ∞ . Formally, we are assuming that $\mathbb{P}[w(100) = 0 \mid w(0) = w_0]$ is a good approximation for the limit in (??).

To estimate $\mathbb{P}[w(T) = 0 \mid w(0) = w_0]$, we must run the code from part ?? multiple times. Let the result of the n -th runs be the wealth path $w_n(t)$ and define the function $\mathbb{I}[w_n(T) = 0]$ to be equal to 1 if the wealth at time T is $w_n(T) = 0$ and 0 otherwise. The probability of the gambler going broke can then be estimated as

$$\hat{B}_N(p, w_0) = \frac{1}{N} \sum_{n=1}^N \mathbb{I}[w_n(T) = 0]. \quad (4)$$

The expression in (??) is just the average number of times the gambler was ruined across all experiments. The function $\mathbb{I}[w_n(T) = 0]$ is called the *indicator function* of the event $w_n(T) = 0$ because it “indicates” whether the event occurred by taking the value 1.

To compute $\hat{B}_N(p, w_0)$ you need to decide on a number of experiments N . The more experiments you run, the more accurate your estimation. Alas, the more you need to wait. Report your probability estimate and the number of experiments N used. Explain your criteria for selecting N .

C Probability of ruin as a function of initial wealth. We want to study the probability of the gambler going broke as a function of the initial wealth. Fix $p = 0.55$ and vary the initial wealth between $w_0 = 1$ and $w_0 = 20$. Show a plot of your probability estimates $\hat{B}_N(p, w_0)$ as a function of the initial wealth. The number of experiments N you use to compute probability estimates for different initial wealths need not be the same.

D Probability of ruin as a function of p . The goal is to understand the variation of the probability of ruin for different values of the probability p . Fix $w_0 = 10$ and vary p between 0.3 and 0.7—increments 0.02 should do. Show a plot of your probability estimates $\hat{B}_N(p, w_0)$ as a function of p . You should observe a fundamentally different behavior for $p < 1/2$ and $p > 1/2$. Comment on that.

E Time to ruin. Fix $p = 0.4$. With this value of p , the gambler’s wealth will eventually deplete independently of the initial wealth w_0 . This is something remarkable: despite the process being random, it is possible to say that $w(t)$ eventually becomes 0. This needs to be qualified though. Unlikely as it may be, there is a chance of winning all hands. Of course, the probability of this happening becomes smaller as the gambler plays more hands. What we can say about a lower bounded random walk is that with probability one, the wealth $w(t)$ approaches 0 as t grows. Formally, the limit $\lim_{t \rightarrow \infty} w(t)$ satisfies

$$\mathbb{P} \left[\lim_{t \rightarrow \infty} w(t) = 0 \right] = 1. \quad (5)$$

In words, we say that “ $w(t) \rightarrow 0$ as $t \rightarrow \infty$ *almost surely*”. Different wealth paths are possible, but almost all of them result in a broke gambler. If we think of probabilities as measuring the likelihood of an event, the measure of the event $w(t) \neq 0$ is asymptotically null. An important quantity here is therefore the time t at which $w(t) = 0$ for the first time, which we can write as

$$T_0 = \min_t \{w(t) = 0\}. \quad (6)$$

Using your function from part ??, estimate the probability distribution of T_0 and its average value.