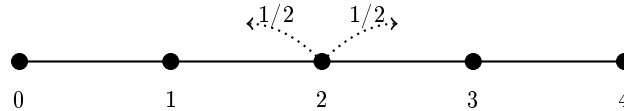


## Lecture 20

## Math 20 Fall 2014, Dartmouth College

A *simple random walk* on a graph is a random process  $\{X(0), X(1), X(2), \dots\}$  in which the nodes of the graph form the state space, with some starting state  $X(0)$  and for each  $X(t)$ ,  $X(t+1)$  can be any vertex adjacent to vertex  $X(t)$  with equal probability. In other words, imagine a token that starts out on some node in a graph and at each time interval moves to a random neighbor of that node. On a path as pictured, if you're on one of the nodes at ends you only have one way to go. If you're on a middle node, move either right or left with equal probability.

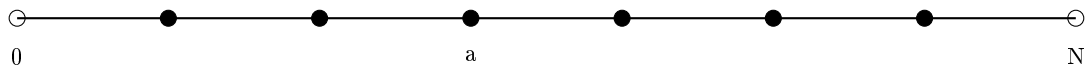


The walk is not "simple" if the probabilities of moving to different neighbors are not equal.

## 1 Gambler's Ruin

Alice has  $\$a$  and Bob has  $\$b$ , with  $a + b = N$ . They will flip a coin. If it comes up heads, Alice gives Bob  $\$1$ . If it comes up tails, Bob gives Alice  $\$1$ . They will repeat this until someone has  $\$N$ . What is the probability that Alice wins?

The amount of money Alice has can be represented as a simple random walk on a path with nodes from 0 to  $N$ , starting at  $a$ . What is the chance that the walk reaches  $N$  before it reaches 0?



A good way to evaluate this, is to keep  $N$  fixed and find the probability for various  $a$ s. Let  $P(x)$  be the probability that the walk that starts at node  $x$  reaches  $N$  before it reaches 0.

$$\begin{aligned} P(0) &= 0 \\ P(N) &= 1 \\ \text{For } 0 < x < N, P(x) &= P(x-1)/2 + P(x+1)/2. \end{aligned}$$

This last relation comes from the fact that this probability must be the same as the expected probability after the first move. And after first move we're at  $x-1$  or  $x+1$  each with probability  $1/2$ . So, if  $0 < x < N$ :

$$\begin{aligned} 2P(x) &= P(x-1) + P(x+1) \\ P(x) - P(x-1) &= P(x+1) - P(x) = \dots = P(1) - P(0) \end{aligned}$$

So the difference between each probability and the next is the same, the function must be linear! Since the difference is  $P(1)$ , we get:

$$\begin{aligned} P(x) &= xP(1) \\ P(N) &= 1 = NP(1) \\ P(1) &= 1/N \\ P(x) &= x/N. \end{aligned}$$

So a simple random walk on a path with nodes numbered from 0 to  $N$ , starting at  $x$  hits  $N$  before it hits 0 with probability  $x/N$ . What if Alice and Bob have the same amount of money to start with? What property of the graph does that correspond to?

## 2 Time to Absorption

In the Gambler's Ruin, 0 and  $N$  are called *absorbing states*, since once you hit one of them, you stay there and the game ends. What is the *time to absorption*, i.e. the expected amount of moves that the game will last?

Let  $T(x)$  be the expected time to absorption of a simple random walk on a path with nodes numbered from 0 to  $N$ , starting at a node  $x$ . Then:

$$\begin{aligned} T(0) &= 0 \\ T(N) &= 0 \\ \text{For } 0 < x < N, T(x) &= 1 + T(x-1)/2 + T(x+1)/2. \end{aligned}$$

The function  $T$  is unique, we prove it below.

*Claim:* There is only one function  $f$  on  $\{0, 1, 2, \dots, N\}$  such that  $f(0) = f(N) = 0$  and for any  $0 < x < N$ ,

$$f(x) = 1 + \frac{1}{2}f(x-1) + \frac{1}{2}f(x+1).$$

*Proof:* Suppose that there are TWO such functions,  $f$  and  $g$ . We will show that  $f(x) - g(x) = 0$ , so must in fact be the same function. Consider the function  $h(x) = f(x) - g(x)$ . We have:

$$\begin{aligned} h(0) &= 0 \\ h(N) &= 0 \\ \text{For } 0 < x < N, h(x) &= \frac{1}{2}h(x-1) + \frac{1}{2}h(x+1). \end{aligned}$$

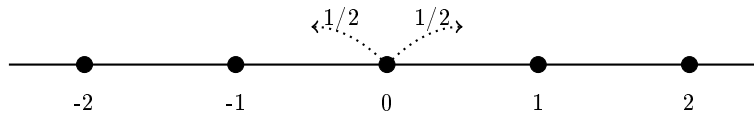
Now we need to show that this implies  $h(x) = 0$  for all nodes  $x$ . Suppose that  $0 < a$  is the smallest node such that  $h(a) \neq 0$ . Since  $h(a) = \frac{1}{2}h(a-1) + \frac{1}{2}h(a+1) = 0 + \frac{1}{2}h(a+1)$ ,  $h(a+1)$  has to be even further from 0. Similarly,  $h(a+2)$  has to be even further from 0 than that, and so on, until we reach  $h(N) \neq 0$ , which contradicts the assumptions. We conclude therefore that such situation is impossible and  $f(x) = g(x)$  for all nodes  $x$ .  $\square$

Since the function  $T$  is unique, if we guess one we know it's the only one that fits. Looking at  $N = 2$ ,  $N = 3$  we get an impression that function could be  $x(N-x)$ . Let's try it.

$$\begin{aligned} RHS &= 1 + (x-1)(N-x+1)/2 + (x+1)(N-x-1)/2 \\ &= 1 + (x-1)(N-x)/2 + (x+1)(N-x)/2 + (x-1)/2 - (x+1)/2 \\ &= 1 + (x-1-x+1)(N-x)/2 + (x-1-x-1)/2 = x(N-x) = LHS. \end{aligned}$$

The expected time to absorption of this walk is  $x(N-x)$ .

## 3 Walk on the Integers



Take a simple random walk on the integers, the line stretches out to infinity on both sides. If you start at 0, what is the probability distribution of your position after  $k$  steps?

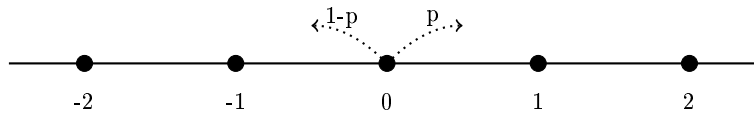
If  $k$  is even:

$$m(i) = \begin{cases} \frac{1}{2^k} \binom{k}{(k+i)/2} & \text{if } i \text{ is even and } -k \leq i \leq k \\ 0 & \text{if } i \text{ otherwise.} \end{cases}$$

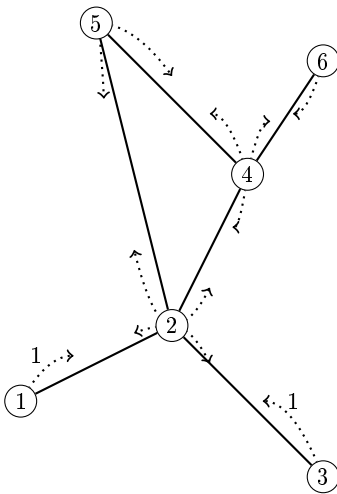
If  $k$  is odd:

$$m(i) = \begin{cases} \frac{1}{2^k} \binom{k}{(k+i)/2} & \text{if } i \text{ is odd and } -k \leq i \leq k \\ 0 & \text{if } i \text{ otherwise.} \end{cases}$$

What if the walk is not simple, but instead wherever you are the probability of a step to the right is  $p$  and a step to the left is  $q$ ?



#### 4 Arbitrary Networks



Random Walks are an example of a *Markov chain*. We will look at Markov Chains more closely next week, but here is the gist of it: Markov chains are random processes with a set state space, where the distribution of where it goes next only depends on the current position, not on the entire history. *Markov chains* are often presented in the form of *transition matrix*. If  $M$  the transition matrix for a Markov chain, entry  $T_{ij}$ , i.e. the one in row  $i$  and column  $j$  is the probability that if we are in state  $i$ , we will be in state  $j$  on the next step. For the simple random walk on the graph pictured above, the transition matrix is:

$$T = \begin{bmatrix} 0 & 1 & 0 & 0 & 0 & 0 \\ \frac{1}{4} & 0 & \frac{1}{4} & \frac{1}{4} & \frac{1}{4} & 0 \\ 0 & 1 & 0 & 0 & 0 & 0 \\ 0 & \frac{1}{3} & 0 & 0 & \frac{1}{3} & \frac{1}{3} \\ 0 & \frac{1}{2} & 0 & \frac{1}{2} & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 & 0 \end{bmatrix}$$