

**Discussion 08**

Spring 2023

**1. Moving Books Around**

You have  $N$  books labelled  $1, \dots, N$  on your shelf. At each time step, you pick a book  $i$  with probability  $\frac{1}{N}$ , place it on the left of all others on the shelf, then repeat this process, each step independent of any other step. Construct a suitable Markov chain which takes values in the set of all  $N!$  permutations of the books.

- a. Find the transition probabilities of the Markov chain.
- b. Find its stationary distribution.

*Hint:* You can guess the stationary distribution before computing it.

**Solution:**

- a. The state space consists of all  $N!$  permutations on  $N$  books. The transition probabilities are then

$$P((\sigma_1, \dots, \sigma_{i-1}, \sigma_i, \sigma_{i+1}, \dots, \sigma_N), (\sigma_i, \sigma_1, \dots, \sigma_{i-1}, \sigma_{i+1}, \sigma_N)) = \frac{1}{N}$$

for  $i = 1, \dots, N$ , and 0 otherwise.

- b. By symmetry, every state  $\sigma \in S_N$  should have the same stationary probability,

$$\pi(\sigma) = \frac{1}{N!}.$$

We can verify that this probability distribution satisfies the balance equations. Let  $\sigma^{(1)} = (\sigma_1, \dots, \sigma_{i-1}, \sigma_i, \dots, \sigma_N)$  be a permutation, and for  $i = 2, \dots, n$ , let  $\sigma^{(i)}$  be the permutation with  $\sigma_1$  in the  $i$ th position,  $(\sigma_2, \dots, \sigma_{i-1}, \sigma_1, \sigma_i, \dots, \sigma_N)$ . With this notation,

$$\pi(\sigma^{(1)}) = \sum_{i=1}^N \pi(\sigma^{(i)}) P(\sigma^{(i)}, \sigma^{(1)}) = \sum_{i=1}^N \frac{1}{N!} \cdot \frac{1}{N} = \frac{1}{N!}.$$

## 2. Markov Chain Practice

Consider a Markov chain with three states 0, 1, 2, and suppose its transition probabilities are  $P(0, 1) = P(0, 2) = \frac{1}{2}$ ,  $P(1, 0) = P(1, 1) = \frac{1}{2}$ ,  $P(2, 0) = \frac{2}{3}$ , and  $P(2, 2) = \frac{1}{3}$ .

- Classify the states in the chain. Is this chain periodic or aperiodic?
- In the long run, what fraction of time does the chain spend in state 1?
- Suppose that  $X_0$  is chosen according to the steady-state or stationary distribution. What is  $\mathbb{P}(X_0 = 0 \mid X_2 = 2)$ ?

### Solution:

- The Markov chain is one recurrent, aperiodic class.
- By solving  $\pi P = \pi$ , we have

$$\pi = \frac{1}{11} \begin{bmatrix} 4 & 4 & 3 \end{bmatrix}.$$

Thus  $\pi(1) = 4/11$ .

- By the definition of conditional probability,

$$\mathbb{P}(X_0 = 0 \mid X_2 = 2) = \frac{\mathbb{P}(X_0 = 0, X_2 = 2)}{\mathbb{P}(X_2 = 2)} = \frac{\mathbb{P}(X_0 = 0, X_1 = 2, X_2 = 2)}{\mathbb{P}(X_2 = 2)}.$$

Note that we used the fact that the only possible two-step path from  $X_0 = 0$  to  $X_2 = 2$  in this chain is  $0 \rightarrow 2 \rightarrow 2$ . Now,  $\mathbb{P}(X_2 = 2) = \mathbb{P}(X_0 = 2)$  because  $X_0$  is chosen according to the stationary distribution  $\pi$ , so

$$\frac{\mathbb{P}(X_0 = 0, X_1 = 2, X_2 = 2)}{\mathbb{P}(X_2 = 2)} = \frac{\pi(0) \cdot (1/2) \cdot (1/3)}{\pi(2)} = \frac{2}{9}.$$

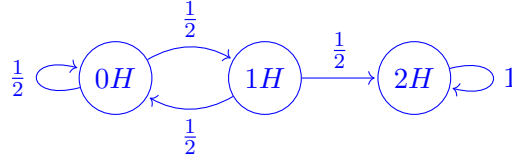
### 3. Hitting Time with Coins

Consider a sequence of fair coin flips.

- What is the expected number of flips until we first see two heads in a row?
- What is the expected number of flips until we see a head followed immediately by a tail?

**Solution:**

- We can create a Markov chain to compute the expected hitting time.  $2H$  represents all sequences with  $HH$  as a subsequence,  $1H$  all sequences that end in  $H$  but do not contain  $HH$ , and  $0H$  all other sequences, including the initial empty sequence.

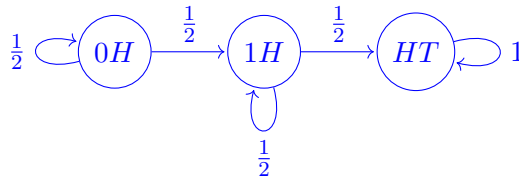


From here, we can set up our hitting-time equations, letting  $\beta(i)$  denote the expected number of flips until two consecutive heads, given that we are in state  $i$  right now:

$$\begin{aligned}
 \beta(0H) &= 1 + \mathbb{P}(H) \cdot \beta(1H) + \mathbb{P}(T) \cdot \beta(0H) \\
 &= 1 + \frac{1}{2}\beta(1H) + \frac{1}{2}\beta(0H) \\
 \beta(1H) &= 1 + \mathbb{P}(H) \cdot \beta(2H) + \mathbb{P}(T) \cdot \beta(0H) \\
 &= 1 + \frac{1}{2}\beta(2H) + \frac{1}{2}\beta(0H) \\
 \beta(2H) &= 0.
 \end{aligned}$$

Solving this system of equations gives us  $\beta(1H) = 4$  and  $\beta(0H) = 6$ . Thus, it takes 6 flips on average until we first see two heads in a row.

- This part has a slightly different setup: if we flip heads after we just flipped a head, we do not need to reset to the initial state.



Letting  $\beta(i)$  be the expected number of flips until we see  $HT$ , we have the equations

$$\begin{aligned}
 \beta(0H) &= 1 + \mathbb{P}(H) \cdot \beta(1H) + \mathbb{P}(T) \cdot \beta(0H) \\
 &= 1 + \frac{1}{2}\beta(1H) + \frac{1}{2}\beta(0H) \\
 \beta(1H) &= 1 + \mathbb{P}(H) \cdot \beta(HT) + \mathbb{P}(T) \cdot \beta(1H) \\
 &= 1 + \frac{1}{2}\beta(HT) + \frac{1}{2}\beta(1H) \\
 \beta(HT) &= 0.
 \end{aligned}$$

Solving this system gives  $\beta(1H) = 2$  and  $\beta(0H) = 4$ .