Random Walks and Markov Chains

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Random Walks





More Random Walks





More Random Walks





Section 1

Markov Chains



Definitions

- A stochastic process is a sequence of random variables $[X_0, X_1, X_2, X_3, \ldots]$ where $X_i \in \mathbb{Z}$.
- If $X_t = i$, we say the process has value *i* at time *t*.
- A *Markov Chain* is a random process with the Markov property:

$$\mathbb{P}(X_{t+1} = i_{t+1} \mid X_t = i_t, X_{t-1} = i_{t-1}, \dots, X_0 = i_0)$$
$$= \mathbb{P}(X_{t+1} = i_{t+1} \mid X_t = i_t)$$



Transition Probabilities

• By the Markov property, the transition probabilities between any two states can be expressed as a single value:

$$p_{ij} = \mathbb{P}(X_{t+1} = j \mid X_t = i)$$

Theorem

A random process $(X_t)_{t\geq 0}$ is Markov if and only if $\forall i_0, \ldots, i_n$:

$$\mathbb{P}(X_1 = i_1, \dots, X_n = i_n \mid X_0 = i_0) = p_{i_0 i_1} p_{i_1 i_2} \dots p_{i_{n-1} i_n}$$



Transition Matrix

The *transition matrix* of a Markov Chain contains all transition probabilities:

p_{00}	p_{01}	p_{02})
p_{10}	p_{11}	p_{12}	
p_{20}	p_{21}	p_{22}	
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Example: Random Walk on Number Line

Consider a symmetric random walk on the number line:





Recurrence and Transience

• We say state i is *recurrent* if, starting at i,

 $\mathbb{P}(\text{chain visits } i \text{ infinitely many times}) = 1$

• We say state i is *transient* if, starting at i,

 $\mathbb{P}(\text{chain visits } i \text{ infinitely many times}) = 0$

https://illinois.zoom.us/j/84778016612?pwd=b0FDVWwxdDlMMjR4TzNqamo0ZGZF Ex. In the chain $\begin{pmatrix} 0 & 1 \\ 0 & 1 \end{pmatrix}$, state 0 is transient, state 1 is recurrent.

Recurrence and Transience

Theorem For a Markov Chain starting at i,

1. $\mathbb{P}(\text{chain eventually returns to } i) = 1 \Leftrightarrow$

i is recurrent & $\mathbb{E}[$ number of visits to $i] = \infty$,

2. $\mathbb{P}(\text{chain eventually returns to } i) < 1 \Leftrightarrow$

i is transient & $\mathbb{E}[$ number of visits to *i* $] < \infty$.

Goal: prove (0,0) is a recurrent state in a random walk on \mathbb{Z}^2 , and (0,0,0) is a transient state in a random walk on \mathbb{Z}^3 .



Section 2

A Proof



Recurrence of 2-D Walk

Want to show: $\mathbb{E}[$ number of visits to $(0,0)] = \infty$.

Proof

- 1. Note that $\mathbb{E}[\text{number of visits to } (0,0)] = \sum_{t=0}^{\infty} p_{00}^{(t)};$
- Since returning to the origin can only be accomplished with an even number of steps, we will only consider the even terms.
 Then, p^{2t}₀₀ = ¹/₄^t ∑^t_{i=0} (2t)!/(i!i!(t-i)!(t-i)!) = ¹/_{4^{2t}} (^{2t})/_t ∑^t_{i=0} n!n!/(i!i!(t-i)!(t-i)!) =
- 3. Then, $p_{00}^{2t} = \frac{1}{4^t} \sum_{i=0}^{t} \frac{(2t)!}{i!i!(t-i)!(t-i)!} = \frac{1}{4^{2t}} {2t \choose t} \sum_{i=0}^{t} \frac{n!n!}{i!i!(t-i)!(t-i)!} = \frac{1}{4^{2t}} {2t \choose t} \sum_{i=0}^{t} \frac{(t)!}{i!i!(t-i)!(t-i)!} = \frac{1}{4^{2t}} \sum_{i=0}^{t} \frac{(t)!}{i!i!(t-i)!} = \frac{1}{4^{2t}} \sum_{i=0}^{t} \frac{(t)!}{i!(t-i)!} = \frac{1}{4^{2t}$
- 4. Since $\sum p_{00}^{(t)} \sim \sum \frac{1}{\pi t}$, we may conclude that both series diverge, hence (0,0) is recurrent.



Transience of 3-D Walk

Want to show: $\mathbb{E}[$ number of visits to $(0,0,0)] < \infty$.

Proof

- 1. Note that $\mathbb{E}[\text{number of visits to } (0,0,0)] = \sum_{t=0}^{\infty} p_{00}^{(t)};$
- 2. Since returning to the origin can only be accomplished with an even number of steps, we will only consider the even terms.

3. Then,
$$p_{00}^{2t} = \frac{1}{6^{2t}} \sum_{i+j+k=t} \frac{(2t)!}{(i!j!k!)^2} = \frac{1}{6^{2t}} {2t \choose t} \sum_{i=0}^t {t \choose i,j,k}^2 \leq \frac{1}{6^{2t}} {2n \choose n} \sum_{i=0}^t {n \choose n,3,n/3,n/3}^2 = \frac{1}{2^{2t}} {2n \choose n} {n \choose n/3,n/3,n/3} \frac{1}{3^t} \sim \frac{1}{2} \left(\frac{3}{\pi t}\right)^{3/2}$$

4. Since $\sum p_{00}^{(t)} \leq \sum \frac{1}{2} \left(\frac{3}{\pi t}\right)^{3/2}$, we may conclude that both series converge, hence (0,0,0) is recurrent.



Questions?



Acknowledgement

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A drunk man will find his way home, but a drunk bird may get lost forever.

— Shizuo Kakutani



Bibliography I



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