

# MATH 285: Stochastic Processes

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Today: Hitting times. First step analysis.  
Stopping times

- Homework 1 is due on Friday, January 14, 11:59 PM

## Expected hitting times

Let  $(X_n)_{n \geq 0}$  be a Markov chain with transition probabilities  $p(i,j)$  and state space  $S$ .

Notation:  $P_i[Y] = P[Y | X_0 = i]$ ,  $E_i[Y] = E[Y | X_0 = i]$

Let  $A \subset S$ ,  $\tau_A := \min\{n \geq 0 : X_n \in A\}$

Q: How long (on average) does it take to reach  $A$ ?

Compute  $E_i[\tau_A] =$

By definition,  $E_i[Y] = \sum_{k=1}^{\infty} k P[Y=k | X_0=i]$  ( $y \in \{0, 1, 2, \dots\}$ )

First step analysis (conditioning on the first step)

$g(i) = E_i[\tau_A] =$

## Expected hitting times

If  $i \in A$ , then  $g(i) = 0$ . Suppose  $i \notin A$ .

Then

$$\begin{aligned} \mathbb{P}[\tau_A = k \mid X_1 = j, X_0 = i] &= \mathbb{P}[X_0 \notin A, X_1 \notin A, \dots, X_{k-1} \notin A, X_k \in A \mid X_1 = j, X_0 = i] \\ &= \mathbb{P}[X_0 \notin A, X_1 \notin A, \dots, X_{k-2} \notin A, X_{k-1} \in A \mid X_0 = j] \\ &= \mathbb{P}[\tau_A = k-1 \mid X_0 = j] \end{aligned}$$

Compute the expectation

$$g(i) = \sum_{j \in S} \mathbb{E}[\tau_A \mid X_1 = j, X_0 = i] \mathbb{P}[X_1 = j \mid X_0 = i]$$

=

=

## Expected hitting times

Conclusion:

$$\begin{cases} g(i) = 1 + \sum_{j \in S} p(i,j) g(j) & \text{if } i \notin A \\ g(i) = 0 & \text{if } i \in A \end{cases}$$

Example 3.2 On average how many times do we need to toss a coin to get two consecutive heads?

Denote by  $X_n$  the number of consecutive heads after  $n^{\text{th}}$  toss.

$$X_n \in \{0, 1, 2\}, \quad P = \begin{bmatrix} 0 & 1 & 2 \\ 0 & \frac{1}{2} & \frac{1}{2} & 0 \\ 1 & \frac{1}{2} & 0 & \frac{1}{2} \\ 2 & 0 & 0 & 1 \end{bmatrix}$$

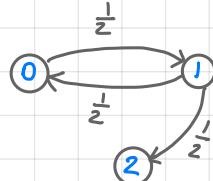
$$g(2) = 0 \quad g(1) =$$

$$g(0) =$$

$$g(0) =$$

$$g(1) = \quad g(0) =$$

Starting from state 0 it takes on average 6 tosses to reach state 2.



## Stopping Times

Def 3.3 Let  $(X_n)_{n \geq 0}$  be a discrete time stochastic process.

A stopping time is a

such that for each  $n$  the event  $\{T = n\}$  depends only on

Examples  $T_1 = \min\{n \geq 0 : X_n = i\}$  is a stopping time

$$\{T_1 = n\} =$$

$T_2 = \max\{n \geq 0 : X_n = i\}$  is not a stopping time

$$\{T_2 = n\} =$$

Recall Markov property: If  $(X_n)$  is  $\text{Markov}(\lambda, P)$ , then

conditional on  $X_m = l$ , the process  $(X_{m+n})_{n \in \mathbb{N}}$  is  $\text{Markov}(S_l, P)$  independent of  $X_0, X_1, \dots, X_m$

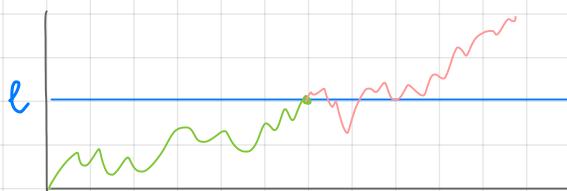
## Strong Markov property

Proposition 3.6 Let  $(X_n)$  be a time-homogeneous Markov chain with state space  $S$  and transition probabilities  $p(i,j)$ .

Let  $T$  be a stopping time,  $\ell \in S$  and  $\mathbb{P}[X_T = \ell] > 0$ .

Then, conditional on  $X_T = \ell$ ,  $(X_{T+n})_{n \geq 0}$  is a time-homogeneous

independent of  $X_0, \dots, X_T$ . In other words, if  $A$  is an event that depends only on  $X_0, X_1, \dots, X_T$  and  $\mathbb{P}[A \cap \{X_T = \ell\}] > 0$  then for all  $n \geq 0$  and all  $i_0, i_1, \dots, i_n \in S$

$$\mathbb{P}[X_{T+1} = i_1, X_{T+2} = i_2, \dots, X_{T+n} = i_n | A \cap \{X_T = \ell\}] =$$


Proof. Use the partition  $\{\{T=m\}\}_{m=0}^{\infty}$  and apply Markov property (see the notes)

## Classification of states : recurrence and transience

Let  $(X_n)$  be a Markov chain with state space  $S$ .

Def 4.1 A state  $i \in S$  is called recurrent if

A state  $i \in S$  is called transient if

### Remark

Let  $T_{i,k} = \text{time } X_n \text{ (starting from } i\text{) visits state } i \text{ } k^{\text{th}} \text{ time}$

$$T_{i,1} = 0, T_{i,k+1} =$$

Then, for  $k \geq 2$ ,  $T_{i,k}$  are stopping times. Indeed,

$$\{T_{i,2} = m\} =$$

$$\{T_{i,k} = m\} = \bigsqcup_{l=k-2}^{m-1} \{T_{i,k-1} = l, T_{i,k} = m\} = \bigsqcup_{l=k-2}^{m-1} \{T_{i,k-1} = l, X_{l+1} \neq i, \dots, X_{m-1} \neq i, X_m = i\}$$

↑ depends on  $X_0, \dots, X_l$

# Classification of states : recurrence and transience

Denote  $T_i := T_{i,2} =$

$r_i :=$

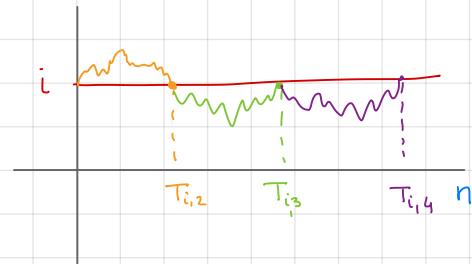
## Theorem 4.2

Let  $i \in S$ . Then

(1)  $i$  is recurrent  $\Leftrightarrow$   $\Leftrightarrow$

(2)  $i$  is transient  $\Leftrightarrow$   $\Leftrightarrow$

and



Proof. Step 1: By the Strong Markov property

$$\mathbb{P}_i[T_{i,k+1} < \infty | T_{i,k} < \infty] =$$

$$\mathbb{P}_i[T_{i,k+1} < \infty] =$$

Step 2: Denote  $N_i :=$   $\leftarrow \# \text{ times } (X_n) \text{ visits state } i$

$\forall k \geq 1, \{N_i \geq k\} =$   $, \text{ so } \mathbb{P}_i[N_i \geq k] = \mathbb{P}_i[T_{i,k} < \infty] =$

## Classification of states : recurrence and transience

Thus  $\mathbb{E}_i[N_i] = \sum_{k=1}^{\infty} \mathbb{P}_i[N_i \geq k] =$

$$\mathbb{E}_i[N_i] = \mathbb{E}_i\left[\sum_{n=0}^{\infty} \mathbb{1}_{X_n=i}\right] = \sum_{n=0}^{\infty} \mathbb{P}_i[X_n=i] =$$

Since  $r_i \in [0, 1]$ ,  $\sum_{\ell=0}^{\infty} r_i^\ell = \infty \Leftrightarrow$  ,  $\sum_{\ell=0}^{\infty} r_i^\ell < \infty \Leftrightarrow$

Step 3:  $r_i = 1 \Leftrightarrow \forall k \quad \mathbb{P}_i[N_i \geq k] = 1$ , i.e.,  $i$  is

Step 4:  $r_i < 1 \Leftrightarrow \mathbb{P}_i[N_i \geq k] = r_i^k \rightarrow 0, k \rightarrow \infty$ ,

so  $\mathbb{P}_i[N_i = \infty] = 0$ , i.e.,  $i$  is

$$\sum_{n=0}^{\infty} p_n(i, i) = \sum_{\ell=0}^{\infty} r_i^\ell =$$



## Recurrence and transience of RW

### Example 4.5

Let  $(X_n)$  be a random walk on  $\mathbb{Z}$ ,  $p(i,j) = \begin{cases} p, & j=i+1 \\ 1-p, & j=i-1 \\ 0, & \text{otherwise} \end{cases}$

Fix  $i \in \mathbb{Z}$ . Is  $i$  recurrent or transient?

Use the  $\sum_{n=0}^{\infty} p_n(i,i)$  criterion.

Notice that  $p_n(i,i) = 0$  if  $n$  is odd

Goal: compute  $\sum_{n=0}^{\infty} p_{2n}(i,i)$



$$p_{2n}(i,i) = \quad (\text{trivial for } p=0 \text{ or } p=1)$$

Case 1:  $p \in (0,1)$ ,  $p \neq \frac{1}{2}$ . Then  $p(1-p) < \frac{1}{4}$

$$\sum_{n=0}^{\infty} p_{2n}(i,i) = \sum_{n=0}^{\infty} \binom{2n}{n} (p(1-p))^n$$

$$\binom{2n}{n} < 4^n$$

$\Rightarrow$  all states are

## Recurrence and transience of RW

Case 2:  $p = \frac{1}{2}$

$$\binom{2n}{n} = \frac{(2n)!}{n! n!} \quad \leftarrow \text{use Stierling's approximation}$$

$$n! \sim \sqrt{2\pi n} \cdot \frac{n^n}{e^n}$$

$$\binom{2n}{n} \sim$$

$$\sum_{n=0}^{\infty} p_n(i,i) =$$

$\Rightarrow$  all states are