### EE512 Stochastic Processes Fall 2016

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NOTE: Durrett's book for Markov Chain.

### **Stopping time**

T is a stopping time wrt  $\{X_n\}_{n\geq 0}$  if  $\{T=n\}$  can be determined from  $X_0, X_1, \ldots, X_n$ 

### Strong markov property

$$P(X_{T+m} = z | X_Y = y, T = n) = P(X_m = z | X_0 = y) = p^m(z)$$

## Time of first return

State space of MC is finite.

For a state  $y \in S$ ,  $T_y = \min\{n \ge 1 : X_n = y\}$ 

since  $n \ge 1$  so  $X_0$  is not relevant here.

If  $X_n \neq y$  for any finite  $n \geq 1$ , then we say that  $T_y = \infty$ 

Is  $T_y$  a stopping time?

For  $I_{T_y=1}$  all we need to know is  $X_1 [= y, \neq y]$ 

If  $\{T_y = k\}$  then  $X_1, X_2, X_{k-1} \neq y$  and  $X_k = y$ 

$$P_y(T_y < \infty) = P(T_y < \infty | X_0 = y)$$

Time to return to y is finite given start  $X_0 = y$ .

$$P(T_y < \infty | X_0 = y) = \sum_{n \ge 1} P(T_y = n | X_0 = y) = \rho_{yy}$$

 $\rho_{yy}$ : Probability of 1st return to y happening in finite time given  $X_0 = y$ 

# Time of $2^{nd}$ return

$$\begin{split} T_y^2 &= \min\{n > T_1 : X_n = y | X_0 = y\} \\ P(T_y^2 < \infty) = ? \\ \{T_y^2 M \infty\} \subset \{T_y < \infty\} \\ \text{Thus, } P_y(T_y^2 < \infty) \leq \rho_{yy} \\ P(T_y^2 < \infty) = P(T_y^2 < \infty, T_y^1 < \infty) + P(T_y^2 < \infty, T_y^1 = \infty) = P(T_y^2 < \infty, T_y^1 < \infty) = P(Y_y^2 < \infty | T_y^1 < \infty) \\ \infty) P(T_y^1 < \infty) = \rho_{yy}^2 \\ \text{Inductively, } P(T_y^k < \infty) = \rho_{yy}^k \\ \text{Two cases:} \end{split}$$

- $\rho_{yy} = 1$  In this case we say that y is a RS[Recurrent State].  $P_y(T_y^1 < \infty) = 1$  so guaranteed to come back to  $y P_y(T_y^k < \infty) = 1$ . Number of times the MC visits given  $X_0 = y$  is infinite
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Let N(y) = Total number of visits to y from time 1 onwards  $\{N(y) = \infty\} = \bigcup_{k \ge 0} \{N(y) \ge k\}$  $P_y(N(y) = \infty) = P(\bigcup_{k \ge 0} \{N(y) \ge k\}) = \lim_{k \longrightarrow \infty} P_y(N(y) > k) = \lim_{k \longrightarrow \infty} P(T_y^k < \infty) = \lim_{k \longrightarrow \infty} \rho_{yy}^k = 1$ 

IF  $\rho_{yy} < 1$  then  $P(T'_y < \infty) < 1$  and  $P(T'_y = \infty) > 0$ 

States with  $\rho_{yy} < 1$  are called transient state

$$\begin{split} P(N(y) = \infty) &= P_y(\cap \{N(y) \ge k\} \\ &= \lim_{k \longrightarrow \infty} P(T_y^k < infty) \\ &= \rho_{yy}^k \\ &= 0 \end{split}$$

For recurrent states:  $N_y = \infty$  with probability 1. for transient states,  $N_y < \infty$  with probability 1.

### Example

 $\rho_{00} = 1$  so state 0 is recurrent[also absorbing]

In general any absorbing state is also recurrent.

Simimarly state 4 is also recurrant.

If we start at 1, we visit 1 only finitel maaaaaaaaa Recurring states do not have to be absorbing states.

Example 2 A-A : 0.5 A:B : 0.5 B-B : 1

Both 0, 1 are recurrent

$$P(T_0^1 = k) = \frac{1}{2^k}$$

 $P(T_0^1 < \infty) = 1$  Thus, 0 is a recurring state, but not an absorbing state.

### **Recurrent and Transient state**

$$\begin{split} \rho_{xy} &= P(T_y < \infty | X_0 = x) = P_x(T_y < \infty) \\ \text{if } \rho_{xy} &= 0 \implies P_x(T_y = \infty) = 1 \\ \text{E.g A-A} &= 1 \text{ B-A} = 0.6 \text{ B-C=0.4 C-C=1} \end{split}$$

$$\rho_{10} = 0.6$$

Suppose there exists a finite number m such that the m step transition probab from x to y is positivie

$$p^{m}(x,y) > 0$$
, then  $\rho_{xy} > 0$ 

Converse: Suppose  $\rho_{xy} > 0$  then there exists a finite number m such that  $p^m(x, y) > 0$ 

 $\rho_{xy} = P_x(T_y < \infty) = \sum_{m=1} P_x(T_y = m) > 0 \implies$  there exists at least one m = n such that  $P_x(T_y = n) > 0$ 

If  $\rho_{xy} > 0$  then we say that x, y communicate or  $x \longrightarrow y$ 

**Theorem** If  $\rho_{xy} > 0$  and  $\rho_{yx} < 1$  then x is a transient state.

Example:

 $\rho_{21}=1$ 

 $\rho_{12} = 0$ 

So 2 is transient