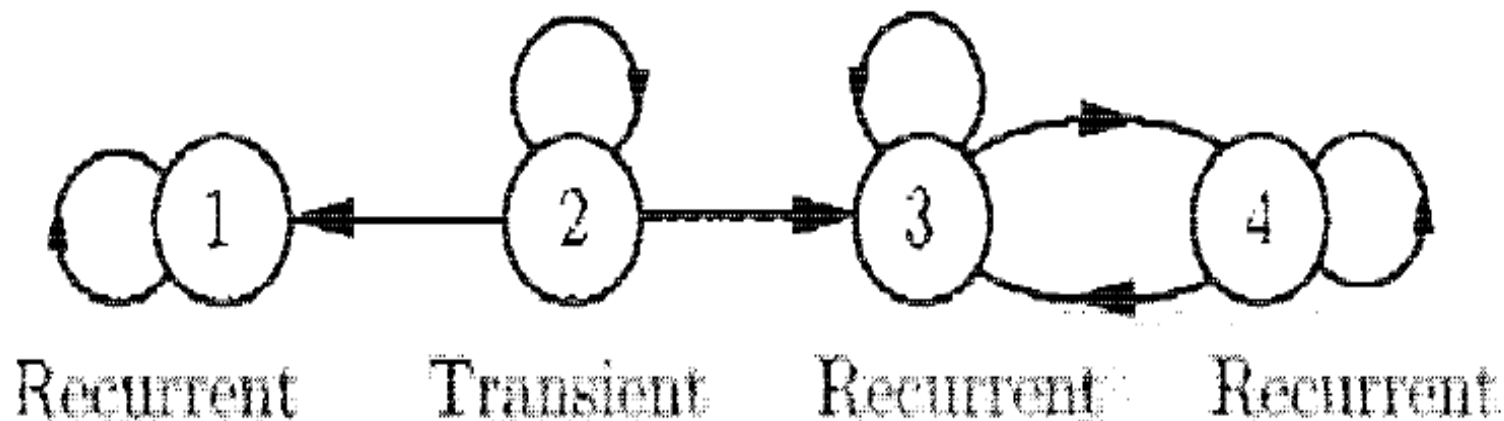


# More on Transience and Recurrence

## Recall: Recurrent and Transient states

- ▶ We say that a state  $i$  is recurrent if  $F_{ii} = P(\text{ ever returning to } i \text{ having started in } i ) = 1$ .
- ▶  $F_{ii}$  is not easy to calculate. (We will see this today)
- ▶ If a state is not recurrent, it is transient.
- ▶ For a transient state  $i$ ,  $F_{ii} < 1$ .
- ▶ If  $i \leftrightarrow j$  and  $i$  is recurrent, then  $j$  is recurrent.



# First passage probabilities

- ▶ Consider a DTMC with state space denoted by  $\mathcal{S}$ .
- ▶  $f_{ij}^n := P(X_n = j, X_k \neq j \text{ for } 1 \leq k \leq n-1 | X_0 = i)$ . ( $f_{ij}^0 = 0$ ).
- ▶  $\{f_{ij}^n, n \geq 0, i, j \in \mathcal{S}\}$  is called the first passage probabilities
- ▶ For a fixed  $n$ ,  $f^n$  denotes the matrix  $[[f_{ij}^n]]$ .
- ▶ For a fixed  $ij$  pair,  $\{f_{ij}^n, n \geq 0\}$  represents a probability mass function on  $\mathbb{Z}_+$ .
- ▶ This mass function can be degenerate which means it can have a point mass at infinity.
- ▶ Define:  $F_{ij} = \sum_{n=1}^{\infty} f_{ij}^n$ .

# First passage probabilities

- ▶ Define:  $F_{ij} = \sum_{n=1}^{\infty} f_{ij}^n$ .
- ▶  $F_{ij}$  has the interpretation of the probability of ever visiting state  $j$  when starting from state  $i$ .
- ▶ If  $F_{ij} = p < 1$ , then there is a finite probability  $1 - p$  with which you may not ever reach  $j$  when starting from state  $i$ .
- ▶ In this case  $\{f_{ij}^n, n \geq 0\}$  is not a proper mass function since it does not sum to 1.
- ▶ If  $F_{ij} = 1$ , then the from  $i$  you can certainly reach  $j$ .
- ▶ In this case  $\{f_{ij}^n, n \geq 0\}$  is a proper mass function.
- ▶ Let  $T_{ij}$  denote the first passage time from  $i$  to  $j$ .
- ▶  $T_{ij}$  has the probability mass function  $\{f_{ij}^n, n \geq 0\}$ . In other words,  $P(T_{ij} = k) = f_{ij}^k$ .

# First recurrence probabilities

- ▶ Define:  $F_{ii} = \sum_{n=1}^{\infty} f_{ii}^n$ .
- ▶  $f_{ii}^n$  : probability of starting in  $i$  and returning to state  $i$  for the first time at time  $n$
- ▶  $F_{ii}$  has the interpretation of the probability of ever returning to state  $i$ .
- ▶ If  $F_{ii} = p < 1$ , then there is a finite probability  $1 - p$  with which you may not return to state  $i$ .
- ▶ If  $F_{ii} = 1$ , then from  $i$  you can certainly return to  $i$ .
- ▶ For any  $i \in \mathcal{M}$ , the first return time  $T_{ii}$  has the probability mass function  $\{f_{ii}^n, n \geq 0\}$ .

# Mean passage and recurrence times

- ▶ Let  $\mu_{ij}$  be the mean (first passage) time from state  $i$  to  $j$ .
- ▶  $\mu_{ij} = \sum_{n=1}^{\infty} n f_{ij}^n$
- ▶ Let  $\mu_{ii}$  be the mean recurrence time at state  $i$ .
- ▶  $\mu_{ii} = \sum_{n=1}^{\infty} n f_{ii}^n$
- ▶ All the above definitions have an equivalent counterpart in a CTMC.
- ▶ For eg:  $f_{ij}^t$  has a natural interpretation. We won't go further into this.

# Transient and recurrent states

- ▶ Suppose for a state  $i$  we have  $F_{ii} = 1$ . Then we say that state  $i$  is recurrent.
- ▶ Once in  $i$ , you are certain to come back to  $i$ .
- ▶ If  $\mu_{ii} = \infty$ , it is called null recurrent. The chain is bound to return to state  $i$ , but possibly after an infinite time.
- ▶  $\mu_{ii} < \infty$ , it is called positive recurrent.
- ▶ If all states of the Markov chain are (null / positive) recurrent, it is called as a (null / positive) recurrent Markov chain.
- ▶ Null recurrence is possible in infinite state space models.
- ▶ State  $i$  is transient if  $F_{ii} < 1$ .
- ▶ You may not return back to  $i$  with a finite probability.

# Transient and recurrent states

- ▶ Consider a DTMC and consider  $X_0 = i$ .
- ▶ Let us count the number of times the chain is in state  $i$ .
- ▶ Let  $I_n$  denote an indicator variable which is 1 if  $X_n = i$  and 0 if  $X_n \neq i$ .
- ▶  $\sum_{n=1}^{\infty} I_n$  counts the number of times state  $i$  was visited.
- ▶  $E[I_n] = P(X_n = i | X_0 = i) = p_{ii}^{(n)}$ .
- ▶ The mean total number of visits to state  $i$  is given by  $\sum_{n=1}^{\infty} p_{ii}^{(n)}$
- ▶ Convergence or divergence of this sum also defines transient or recurrent states.



# Recurrent criteria

- ▶ The mean total number of visits to state  $i$  is given by  $\sum_{n=1}^{\infty} p_{ii}^n$
- ▶ Suppose the chain visits state  $i$  only exactly  $n$  times.
- ▶ The  $P(\text{exactly } n \text{ visits to } i) = F_{ii}^n(1 - F_{ii})$ .
- ▶ For a recurrent state,  $F_{ii} = 1$ . Hence  $P(\text{exactly } n \text{ visits to } i) = 0$ .
- ▶  $P(\text{exactly infinite visits to } i) = 1$ .
- ▶ Mean total number of visits is also infinite and hence  $\sum_{n=1}^{\infty} p_{ii}^n$  diverges.

# Transient state criteria

- ▶ The mean total number of visits to state  $i$  is given by  $\sum_{n=1}^{\infty} p_{ii}^n$
- ▶ Suppose the chain visits state  $i$  only exactly  $n$  times.
- ▶ The  $P(\text{exactly } n \text{ visits to } i) = F_{ii}^n(1 - F_{ii})$ .
- ▶ For transient state  $i$ ,  $F_{ii} < 1$ .
- ▶ The  $P(\text{exactly } n \text{ visits to } i) = F_{ii}^n(1 - F_{ii})$ . Compare this with geometric random variable.
- ▶ Mean total number of visits to state  $i$  is  $\frac{F_{ii}}{1-F_{ii}}$  which is finite.
- ▶ Hence for transient state  $\sum_{n=1}^{\infty} p_{ii}^n$  must converge.