

Let $f_i = P(\exists n > 0 \text{ so that } X_n = i | X_0 = i)$
 Recurrent stat: $f_i = 1$
 Transient state: $f_i < 1$
 i Recurrent stat \Leftrightarrow expectation of number returns to i is ∞ .
 Let $I_n = \begin{cases} 1 & X_n = i \\ 0 & \text{else} \end{cases}$
 $E[\sum_{n=1}^{\infty} I_n | X_0 = i] = \sum_{n=1}^{\infty} E[I_n | X_0 = i] = \sum_{n=1}^{\infty} P(X_n = i | X_0 = i) = \sum_{n=1}^{\infty} f_i^n$
Proposition 4.1
 State i is recurrent if $\sum_{n=1}^{\infty} P_{ii}^n = \infty$
 - - - transient if $\sum_{n=1}^{\infty} P_{ii}^n < \infty$
Remark 4.3.1 In a finite state Markov chain all states cannot be transient. If they were all states are visited only a finite number of times which is impossible since there finite number of states.

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Corollary 4.2
 state i recurrent
 state i communicates with $j \Rightarrow j$ is recurrent
 state $j, j \leftrightarrow i$
Proof. ~~Propose~~ \exists k and m so that $P_{ij}^k > 0, P_{ji}^m > 0$
 Then $P_{jj}^{m+k} \geq P_{ji}^m P_{ij}^k$
 probability to go from j to j in $m+k$ steps \geq probability to go from j to j in $m+k$ steps but via state i
 Hence $\sum_{n=1}^{\infty} P_{jj}^n \geq P_{ji}^m P_{ij}^k \sum_{n=1}^{\infty} P_{ii}^n = \infty$
 $P_{jj}^n (\sum_{i=1}^k P_{ji}^i) P_{ij}^k \rightarrow \sum_{n=1}^{\infty} P_{jj}^n = \infty$ i.e. j recurrent
Remark 4.3.2 Transience is a class property. If i is transient and communicates with j must be transient, since if it were recurrent also i would be recurrent.

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Remark 4.3.3 A finite irreducible chain has one class. The states cannot be transient, therefore the states must be recurrent.
Example states $\{0, 1, 2, 3\}$

$$P = \begin{matrix} & \begin{matrix} 0 & 1 & 2 & 3 \end{matrix} \\ \begin{matrix} 0 \\ 1 \\ 2 \\ 3 \end{matrix} & \begin{bmatrix} 0 & 0 & 1/2 & 1/2 \\ 1 & 0 & 0 & 0 \\ 2 & 0 & 1 & 0 \\ 3 & 0 & 0 & 0 \end{bmatrix} \end{matrix}$$

 Then all state communicates, e.g. $0 \rightarrow 2 \rightarrow 1 \rightarrow 0$
Example, states $\{0, 1, 2, 3, 4\}$


$$P = \begin{matrix} & \begin{matrix} 0 & 1 & 2 & 3 & 4 \end{matrix} \\ \begin{matrix} 0 \\ 1 \\ 2 \\ 3 \\ 4 \end{matrix} & \begin{bmatrix} 1/2 & 1/2 & 0 & 0 & 0 \\ 1/2 & 1/2 & 0 & 0 & 0 \\ 0 & 0 & 1/2 & 1/2 & 0 \\ 0 & 0 & 1/2 & 1/2 & 0 \\ 1/4 & 1/4 & 0 & 0 & 1/2 \end{bmatrix} \end{matrix}$$

 classes $\{0, 1, 2, 3\}$ recurrent, $\{4\}$ transient

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Example, random walk
 state space is $\{0, \pm 1, \pm 2, \dots\}$
 $P_{i, i+1} = p = 1 - P_{i, i-1}$
 $0 < p < 1$
 $p = 1/2$ symmetric random walk
 Is the chain recurrent or transient?
 Remark $P_{nn}^{2n} = 0, n = 1, 2, \dots$
 $P_{00}^{2n} = P(X_{2n} = 0 | X_0 = 0) = P(\text{upward and downward moves } | X_0 = 0)$
 $= \binom{2n}{n} p^n (1-p)^n = \binom{2n}{n} (1-p)^n$
 Stirling's formula: $n! \sim n^n e^{-n} \sqrt{2\pi n}$
 when $a_n \sim b_n \Leftrightarrow \frac{a_n}{b_n} \rightarrow 1, n \rightarrow \infty$
 $P_{00}^{2n} \sim \frac{(2n)^{2n+1/2}}{n^{2n+1/2} e^{2n}} \cdot \frac{e^{-2n}}{\sqrt{2\pi}} [4p(1-p)]^n$
 $= \frac{2^{2n+1/2}}{\sqrt{\pi}} \cdot \frac{1}{\sqrt{2\pi}} [4p(1-p)]^n = \frac{[4p(1-p)]^n}{\sqrt{\pi}}$
 But $4p(1-p) \leq 1$ so $\sum_{n=1}^{\infty} \frac{[4p(1-p)]^n}{\sqrt{\pi}} = \begin{cases} \infty & \text{when } p = 1/2 \\ < \infty & \text{when } p \neq 1/2 \end{cases}$

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Example 2 dimension 
 Symmetric random walk in two dimension is recurrent.
 In three dimension the symmetric random walk is not recurrent.
4.4 Long-run proportions and limiting probabilities
 X_n Markov chain
 Let $f_{ij} = P(X_n = j \text{ some } n | X_0 = i)$
Proposition 4.3 If i is recurrent and $i \leftrightarrow j$ then $f_{ij} = 1$
Proof $P_{ij}^n > 0$ some n since $i \leftrightarrow j$. Fix
 Define 0-1 variable: success if $X_n = j$ before if $X_n = i$
 $P(\text{success}) = P_{ij}^n$
 Let t_j be the first time a flow n_j that X_n enters i .
 Define a 0-1 variable: success if $X_{n+t_j} = j$ before if $X_{n+t_j} = i$
 $P(\text{success}) = P_{ij}^n$
 Let t_i be the first time after t_n that X_n enters i and define a 0-1 variable: success if $X_{n+t_i} = i$ before if $X_{n+t_i} = j$
 $P(\text{success}) = P_{ij}^n$
 The 0-1 variables are independent since the starts over after visit to i . Hence a Bernoulli process

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Hence number of failures before first success is geometric distributed, with $p = P_{ij}^n$ and a success will therefore occur with probability 1.
 Assume that j recurrent
 $N_j = \min\{n > 0, X_n = j\}$ and $m_j = E[N_j | X_0 = j]$
Definition: A recurrent state j is positive recurrent if $m_j < \infty$ and it is null recurrent if $m_j = \infty$
 Suppose that X_n is irreducible and recurrent.
 Want to show that $1/m_j$ is the long-run proportion of the time X_n spends in j , denoted by π_j
Proposition 4.4 If the Markov chain is irreducible and recurrent then $1/m_j = \pi_j$ whatever the initial state is.
Proof Suppose $X_0 = i$.
 Define $T_i = \min\{n > 0, X_n = j\}$
 $T_i + T_j = \min\{n > 0, X_n = j\}$
 $T_i, T_i + T_j = \min\{n > 0, T_i, X_n = j\}$
 $T_i, T_i + T_j$ is the length between visits to j
 By proposition 4.3 T_j finite by Markov property $T_i, T_i + T_j = \dots$ i.e. $E[T_i] m_j = 1$

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The n 'th transition into j occurs at

$$T_1 + \dots + T_n$$

$$\text{Hence } \pi_j = \frac{n}{\lim_{n \rightarrow \infty} \sum_{i=1}^n T_i} = \frac{1}{\lim_{n \rightarrow \infty} \left[\frac{T_1}{n} + \frac{T_2 + \dots + T_n}{n} \right]}$$

But $\frac{T_1}{n} \rightarrow 0$ and by the strong law of large numbers

$$\frac{T_2 + \dots + T_n}{n} = \frac{n-1}{n} \left(\frac{T_2 + \dots + T_n}{n-1} \right) \rightarrow m_j^*$$

and hence $\pi_j = 1/m_j^*$ \square

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