

More on Markov Chains

For discrete time ($T = \mathbb{Z}^+$) a (the) stationary dist'n $\underline{\pi}$ is any Pf satisfying $\underline{\pi}' = \underline{\pi}' P$.

Note - $\underline{\pi}$ is not unique

We know

- j is recurrent iff $P_{jj}(1) = \infty$ (ie $\sum_n P_{jj}^{(n)} = \infty$)

- j recurrent then it is positive recurrent iff

$$\pi_j = \lim_{\Delta \uparrow 1} (1-\Delta) P_{jj}(\Delta) > 0$$

$$\text{+ then } \pi_j = \frac{1}{E(T_{jj})}$$

Notice that

$$\frac{1 - F_{jj}(\Delta)}{1 - \Delta} = \frac{1 - E(\Delta^{T_{jj}})}{1 - \Delta} = \frac{1 - \sum_k \Delta^k P(T_{jj} = k)}{1 - \Delta}$$

$$= \left[\sum_k P(T_{jj} = k) - \sum_k \Delta^k P(T_{jj} = k) \right] / (1 - \Delta)$$

$$= \sum_k P(T_{jj} = k) \frac{1 - \Delta^k}{1 - \Delta}$$

$$= \sum_k P(T_{jj} = k) (1 + \Delta + \dots + \Delta^{k-1})$$

+ this increases with Δ + has lim as $\Delta \uparrow 1$ of $\sum_k P(T_{jj} = k) k$ $\overbrace{\hspace{10em}}^{E(T_{jj})}$

- process irreducible & positive recurrent
 $\Rightarrow \underline{\pi}$ is unique & > 0
 $\pi_i = \frac{1}{E(T_{ii})}$

If the process is irreducible with a countably ∞ state space then
 $\underline{\pi}$ exists \Rightarrow positive recurrent

If the process is irreducible & finite then
 all states are positive recurrent

Irreducible + positive recurrent + aperiodic
 $\Rightarrow \lim_{n \rightarrow \infty} P_{ij}^{(n)} = \pi_j > 0$
 & $\underline{\pi}$ is the unique stationary dist'n } proved

For irreducible + aperiodic + null recurrent we do not have a stationary distribution. (By applying renewal theory we in fact know $P_{ij}^{(n)} \rightarrow 0$.) From the coupling argument we still have
 $P_{ik}^{(n)} - P_{jk}^{(n)} \rightarrow 0 \quad \forall i, j, k \quad (*)$

We now use this to show $P_{ij}^{(n)} \rightarrow 0$.

Now suppose $p_{ij}(n)$ does not $\rightarrow 0$ as $n \rightarrow \infty$. For fixed $i+j$ \exists a subsequence of the n 's so that $p_{ij}(n_k^{(i)}) \rightarrow \geq 0$ ^{but not all 0} say α_j . α_j will not depend on i $\circ\circ$ of (*) but the subsequence might. Now let $\epsilon_i \downarrow 0$. For each i choose n_k so that $n_k \in \{n_k^{(i)}\}$ & $p_{ij}(n_k)$ is within ϵ_i of α_j . In addition choose the n_k 's to increase with i . We then have

$$p_{ij}(n_k) \rightarrow \alpha_j \quad \text{for each } j$$

& not all α_j are 0. (We have just employed a diagonalization argument.) Let F be a finite set of j states. Then

$$\sum_{j \in F} \alpha_j = \lim_{k \rightarrow \infty} \underbrace{\sum_{j \in F} p_{ij}(n_k)}_{\leq \sum_j p_{ij}(n_k) = 1} \leq 1$$

Hence $0 < \sum_j \alpha_j \leq 1$.

Now

$$\underbrace{\sum_{l \in F} p_{il}(n_k) p_{lj}}_{\rightarrow \sum_{l \in F} \alpha_l p_{lj}} \leq p_{ij}(n_k+1) = \underbrace{\sum_l p_{il} p_{lj}}_{\rightarrow \alpha_j}$$

$$\circ \circ \quad \sum_{l \in F} \alpha_l P_{lj} \leq \alpha_j$$

↳ so

$$\sum_l \alpha_l P_{lj} \leq \alpha_j, \quad \forall j$$

In fact equality must hold $\forall j$ for if we had $<$ for some j then

$$\sum_l \alpha_l = \sum_l \alpha_l \underbrace{\sum_j P_{lj}}_1 = \sum_{l,j} \alpha_l P_{lj} = \sum_{j,l} \alpha_l P_{lj} < \sum_j \alpha_j$$

which is impossible. $\circ \circ$

$$\sum_l \alpha_l P_{lj} = \alpha_j, \quad \forall j$$

Now set $\alpha = \sum_j \alpha_j$ & $\pi_j = \alpha_j / \alpha$. We then have π is a RF & $\pi' = \pi' P$. But this cannot be. It follows that

$$P_{ij}(n) \rightarrow 0, \quad \forall i, j$$

Irreducible + aperiodic + transient — obvious

that $P_{ij}(n) \rightarrow 0$

Theorem MC irreducible & aperiodic then

$$P_{ij}(n) \rightarrow \frac{1}{E(T_{ij})}$$

Assume irreducible + aperiodic

Let $i, j \in S$. We want to show $P_{ij}(n) > 0$
 $\forall n$ large enough.

Sol'n Since j is aperiodic $\exists m_1, \dots, m_r$ with
 $\gcd = 1$ & $P_{jj}(m_s) > 0$ for $1 \leq s \leq r$. Now $\exists M$
for $n \geq M$ we can write $n = \sum_{s=1}^r a_s m_s$
 \uparrow
 $\in \mathbb{Z}^+$

Now use the CKE to get

$$P_{ij}(n) \geq \prod_{s=1}^r P_{ij}(m_s)^{a_s} > 0$$

Hence $P_{ij}(n) > 0 \quad \forall n \geq M$.

Now since $i \leftrightarrow j$ choose k so that $P_{ij}^{(k)} > 0$.

Then $P_{ij}(n+k) \geq \underbrace{P_{ij}^{(k)}}_{>0} P_{ij}(n) > 0$ if $n \geq M$

Back to coupling

Let $\{X_n\} + \{Y_n\}$ be iid MC's with transition
matrix P which are irreducible + aperiodic. Set

$$\underline{Z}_n = (X_n, Y_n)$$

Let $i, j, k, l \in S$. Let $N(i, j) + N(k, l)$ be st

$$\begin{aligned} P_{ij}(n) &> 0, \quad \forall n \geq N(i, j) \\ P_{kl}(n) &> 0, \quad \forall n \geq N(k, l) \end{aligned}$$

Note that both $N(i,j)$ & $N(k,l)$ exist from the previous problem.

We now have $\forall n \geq \max\{N(i,j), N(k,l)\}$

$$P(\underline{Z}_n = (k,l) \mid \underline{Z}_0 = (i,j)) = p_{ik}^{(n)} p_{jl}^{(n)} > 0$$

so that $\{\underline{Z}_n\}$ is irreducible & aperiodic.

Remark Take $S = \{1,2\}$ & $P = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix}$.

Then $\{(1,1), (2,2)\}$ & $\{(1,2), (2,1)\}$ are closed sets of states for $\{\underline{Z}_n\}$.

Obviously $\{\underline{Z}_n\}$ is reducible. The reason this happens is that $\{X_n\}$ is periodic (it is irreducible)

Once we have $\{\underline{Z}_n\}$ irreducible & aperiodic we follow the coupling proof to get (*) which then leads to $p_{ij}^{(n)} \rightarrow 0 = \frac{1}{E(T_{ij})} \leftarrow \infty$

Sums of rv's

Let X_0, X_1, \dots be rv's + set

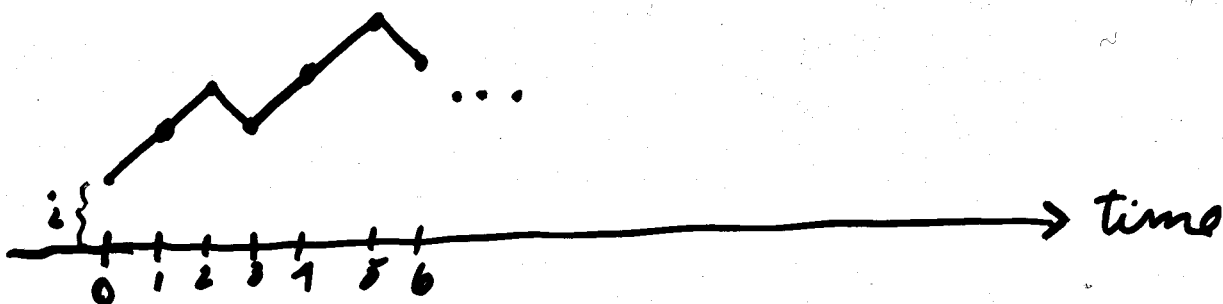
$$S_t = X_0 + X_1 + \dots + X_t$$

The process $\{S_t; t=0, 1, \dots\}$ is called a random walk. Typically, X_1, X_2, \dots are ind/ iid while $S_0 = X_0$ is treated as a constant & is the starting point of the walk. In the independent case $\{S_t\}$ is a Markov process & if we add the id assumption it is time homogeneous.

The simple random walk on \mathbb{Z}

Here, X_1, X_2, \dots are iid taking values ± 1 with probabilities $p = P(X_i = 1)$ + $q = P(X_i = -1)$ with $0 < p < 1$.

Set the starting point $S_0 = i$. A typical sample path would look like



We have seen that this Markov Chain is irreducible and is recurrent in the symmetric case ($p=q=1/2$), however the process is null recurrent in the sense that average return times are infinite. When $p \neq q$ then the process is transient. Set

$Y = \sum_{t=1}^{\infty} I_{A_t}$, where $A_t = \{S_t = i\}$. In the recurrent case $Y \stackrel{a.s.}{=} +\infty$. In the transient case Y is geometric with

$E(Y) < \infty$ (the i is to remind us that we are conditioning on the initial state. We will drop it when this point is clear.). So

$$P(Y = \infty) = 1 \Leftrightarrow E(Y) = \sum P(A_t) = \infty \left\{ \begin{array}{l} \Leftrightarrow p=q=1/2 \\ \Leftrightarrow EX_1 = 0 \end{array} \right\}$$

Notice $\{Y = \infty\} = \{A_t \text{ i.o.}\}$ so that we have

$$P(A_t \text{ i.o.}) = 1 \Leftrightarrow \sum P(A_t) = \infty$$

This reminds us of the Borel Cantelli Lemma & 0-1 Law. Certainly $\sum P(A_t) < \infty \Rightarrow P(A_t \text{ i.o.}) = 0$ but note the A_t 's are not typically independent. It is tempting to approach the problem via tail events, but is $\{A_t \text{ i.o.}\}$ one of them?

For random walks with general iid steps X_k the recurrence problem is more complicated (see Breiman's text Probability, section 3.7). We will only consider integer steps X_k . Without loss of generality we take $S_0 = X_0 = 0$ so that we are then interested in the event $\{S_m = 0 \text{ i.o.}\}$. We have

Theorem $E(X_k) = 0 \Rightarrow P(S_m = 0 \text{ i.o.}) = 1$

Proof For any state j let $T \geq 1$ be the first time $S_m = j$ and set

$$N_j = \sum_{m=1}^{\infty} I_{\{S_m = j\}}$$

Then $E(N_j) = E[E(N_j | T)] = \sum_{t=1}^{\infty} \left[\sum_{m \geq t} P(S_m = j | T=t) \right] P(T=t)$

(since $m < T \Rightarrow S_m \neq j$)

$$= \sum_{t=1}^{\infty} P(T=t) \left[1 + \sum_{m=t+1}^{\infty} P(S_m = j | T=t) \right]$$

$$= \sum_{t=1}^{\infty} P(T=t) [1 + E(N_0)]$$

$$= P(T < \infty) [1 + E(N_0)]$$

$$\leq 1 + E(N_0)$$

Now take $\epsilon > 0$. For integer $M > 0$ we have

$$\sum_{m=1}^{\infty} P(|S_m| \leq M) = \sum_{m=1}^{\infty} \sum_{j=-M}^M P(S_m = j)$$

$$= \sum_{j=-M}^M \sum_{m=1}^{\infty} P(S_m = j) = \sum_{j=-M}^M E(N_j)$$

$$\leq (2M+1) [1 + E(N_0)]$$

By the SLLN

$$\frac{S_m}{m} \xrightarrow{a.s.} 0 \Rightarrow \frac{|S_m|}{m} \xrightarrow{a.s.} 0 \Rightarrow \frac{|S_m|}{m} \rightarrow 0$$

so that $P(|S_m| \leq \epsilon m) \rightarrow 1$. Now take m_0 large enough so that $m \geq m_0$ implies $P(|S_m| \leq \epsilon m) \geq \frac{1}{2}$ say.

In particular, for M large enough we have

$$P(|S_m| \leq M) \geq \frac{1}{2} \quad \text{for } m_0 \leq m \leq \frac{M}{\epsilon}$$

so that

$$1 + E(N_0) \geq \frac{1}{2M+1} \sum_{m=1}^{\infty} P(|S_m| \leq M)$$

$$\geq \frac{1}{2M+1} \sum_{m_0 \leq m \leq \frac{M}{\epsilon}} P(|S_m| \leq M)$$

$$\geq \frac{1}{2M+1} \sum_{m_0 \leq m \leq \frac{M}{\epsilon}} \frac{1}{2}$$

$$\geq \frac{1}{(2M+1)2} \left(\frac{M}{\epsilon} - m_0 \right)$$

$$\rightarrow \frac{1}{4\epsilon} \quad \text{as } M \rightarrow \infty$$

It then follows (since $\epsilon > 0$ is arbitrary) that $E(N_0) = \infty$ qed

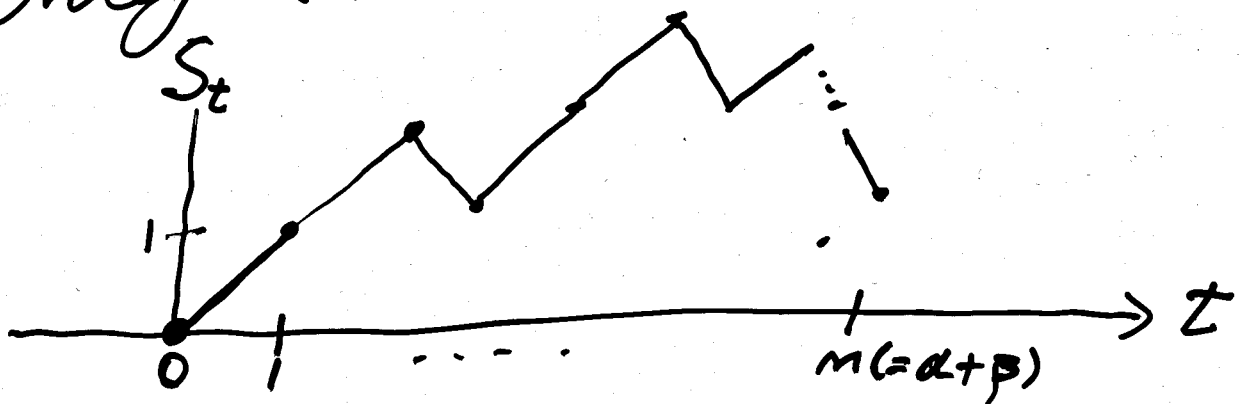
The Ballot Problem (≈ 1850)

2 candidates in an election $A + B$.
Suppose A wins. Let $\alpha = \#$ of
votes for A & $\beta = \#$ for B .
 $P(A \text{ was always ahead of } B)$

Assign $+1$ to a vote for A &
 -1 to a vote for B .
As the votes come in you have
 $0, X_1, X_2, \dots, X_n$ ($n = \alpha + \beta$)

Let $S_t = 0 + X_1 + \dots + X_t$ then

$\{S_t\}$ is a random walk on
the integers which starts at 0 .



The solution to the problem is to calculate $P(S_t > 0 \text{ for } 0 < t \leq m)$ when starting from 0. In order to solve this consider our simple random walk $\{S_t\}$ starting from a say (ie $S_0 = a$). Suppose we wish to calculate $P(S_m = b)$. Consider

$$a \rightarrow s_1 \rightarrow s_2 \rightarrow \dots \rightarrow s_m = b$$

m arrows

Let $r = \#$ of steps to the right (ie $\#$ of $+1$'s)
 $l = \#$ of steps to the left

Then $r + l = m$ & $r - l = b - a$ if we are to end up at b at time m . Hence

$$r = \frac{m + b - a}{2}, \quad l = \frac{m - b + a}{2}$$

so that

$$P(S_m = b) = \binom{m}{\frac{m + b - a}{2}} p^{\frac{m + b - a}{2}} q^{\frac{m - b + a}{2}}$$

req'd \rightarrow (\uparrow $\#$ of paths from a to b with $\frac{m + b - a}{2}$ steps to the right)

Suppose now that both $a, b > 0$. Set

$$N_m(a, b) = \# \text{ of paths from } a \text{ to } b$$

$$* N_m^0(a, b) = \# \text{ of paths from } a \text{ to } b \text{ which hit the time axis at least once}$$

Every path from a to b which crosses the t -axis at least once has a reflection about the t -axis which is a path from $-a$ to b . The reverse is also true so that

$$N_m^0(a, b) = N_m(-a, b)$$

We have just employed the reflection principle.

We are now in a position to solve the ballot problem. So, let $b > 0$ & $a = 0$. The # of paths from $a = 0$ to b which do not revisit the t -axis is

$$\frac{b}{m} N_m(0, b)$$

To see this note that every such path first goes to 1 (more specifically $(1, 1)$). Hence the number of desired paths is

$$N_{m-1}(1, b) - N_{m-1}^0(1, b) = N_{m-1}(1, b) - N_{m-1}(-1, b)$$

Now, $N_m(a, b) = \binom{m}{\frac{m+b-a}{2}}$ so that the req'd # is

$$\binom{m-1}{\frac{m-1+b-1}{2}} - \binom{m-1}{\frac{m-1+b+1}{2}}$$

$$= \binom{m-1}{\frac{m+b}{2} - 1} - \binom{m-1}{\frac{m+b}{2}} = \binom{m-1}{k-1} - \binom{m-1}{k}, \quad k = \frac{m+b}{2}$$

$$= \frac{(m-1)!}{(k-1)!(m-k)!} - \frac{(m-1)!}{k!(m-k-1)!}$$

$$= \frac{k m (m-1)!}{k m (k-1)!(m-k)!} - \frac{m(m-1)!(m-k)}{m k!(m-k-1)(m-k)}$$

$$= \frac{k m!}{m k!(m-k)!} - \frac{(m-k)}{m} \frac{m!}{k!(m-k)!}$$

$$= \binom{m}{k} \frac{k - (m-k)}{m} = \frac{2k-m}{m} \binom{m}{k}$$

$$= \frac{m+b-m}{m} \binom{m}{\frac{m+b}{2}} = \frac{b}{m} N_m(0, b)$$

For the ballot problem $m = \alpha + \beta$, $b = \alpha - \beta$ ($\forall a=0$)
so that

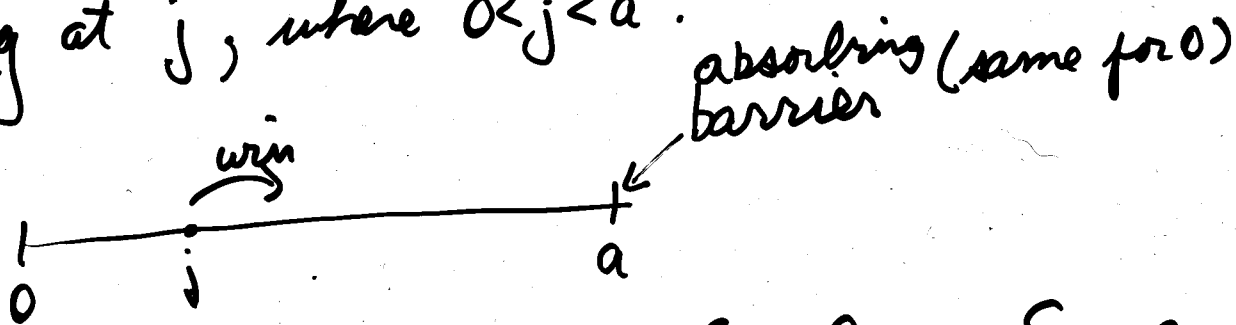
$$P(A \text{ is always ahead}) = \frac{\frac{b}{m} N_m(0, b)}{N_m(0, b)} = \frac{b}{m}$$

$$= \frac{\alpha - \beta}{\alpha + \beta}$$

m

The Gambler's Ruin Problem

We consider a simple random walk $\{S_t\}$ starting at j , where $0 < j < a$.



The process stops if either $S_t = 0$ or $S_t = a$. This Markov Chain has $a+1$ states with both 0 and a being absorbing states (they are of course recurrent) and $\{1, \dots, a-1\}$ being a transient class. There are 3 classes to this chain $\{0\}$, $\{1, \dots, a-1\}$, $\{a\}$. The only nonzero transition probabilities (1-step) are

$$P_{i,i+1} = p + P_{i,i-1} = q, \quad 0 < i < a$$

Set $p_j(t) = P(S_t = 0 | S_0 = j)$. This is also denoted by $p_{j0}(t)$. We have (recall the backward equations) for $t > 0$

$$p_j(t) = p p_{j+1}(t-1) + q p_{j-1}(t-1), \quad 0 < j < a$$

$$p_0(t) = p_0(t-1), \quad p_a(t) = p_a(t-1)$$

with boundary conditions

$$p_j(0) = \begin{cases} 1, & j=0 \\ 0, & j>0 \end{cases}$$

"Clearly" $p_j(t)$ increases with t & since it is bounded we have a limit p_j called the probability of ultimate ruin. If we let $t \rightarrow \infty$ in the above equations we obtain

$$p_j = p p_{j+1} + q p_{j-1} \quad (0 < j < a)$$

with boundary conditions $p_0 = 1$, $p_a = 0$. This difference equation is easily solved yielding

$$p_j = \frac{(q/p)^j - (q/p)^a}{1 - (q/p)^a}, \quad p \neq q$$

$$= \frac{a-j}{a}, \quad p = q = \frac{1}{2}$$

In the extreme case $a = \infty$ (playing against an infinitely rich opponent!) this reduces to

$$p_j = (q/p)^j, \quad p > q$$

$$= 1, \quad p \leq q$$

These results can also be obtained via martingale methods (later).