

Addendum on some Basic Results in Gambling Theory

Consider the martingale as well as Markov chain $\{M_n\}_{n=0}^\infty$ given by $M_n = m_0 + \sum_{i=1}^n X_i$ where $\{X_i\}_{i=1}^\infty$ are iid. (independent identically distributed) random variables with $\mathbf{P}\{X_i = -1\} = \mathbf{P}\{X_i = 1\} = 1/2$ (a so called Rademacher distribution): This is the simplest possible (non-trivial) example of a fair gambling process with initial fortune m_0 .

For the so called n -step transition probability $p_{ij}^{(n)} = \mathbf{P}\{M_n = j | M_0 = i\}$ we then have $p_{jj}^{(n)} = 0$ for $n = 2k + 1$ odd and

$$p_{jj}^{(n)} = \binom{2k}{k} (1/2)^k (1/2)^k \quad \text{for } n = 2k \text{ even}$$

[=the probability that a Bin($2k, 1/2$)-distributed random variable is equal to k (=the probability that half of the n steps go upwards and the other half of them downwards)].

It is a theorem in Markov theory that a state j is recurrent (that is, it is certain to get back to the state eventually if starting in it) if $\sum_{n=1}^\infty p_{jj}^{(n)} = \infty$, while j is transient (that is, it is not certain to get back to the state if starting in it) if $\sum_{n=1}^\infty p_{jj}^{(n)} < \infty^1$.

Now, according to Stirling's formula we have $k! \sim \sqrt{2\pi k} k^k e^{-k}$ as $k \rightarrow \infty$ (where \sim means "behaves asymptotically like"). Hence

$$\binom{2k}{k} (1/2)^k (1/2)^k = \frac{(2k)! (1/2)^{2k}}{(k!)^2} \sim \frac{\sqrt{4\pi k} (2k)^{2k} e^{-2k} (1/2)^{2k}}{(\sqrt{2\pi k} k^k e^{-k})^2} = \frac{1}{\sqrt{\pi k}} \quad \text{as } k \rightarrow \infty,$$

so that $\sum_{n=1}^\infty p_{jj}^{(n)} = \sum_{k=1}^\infty p_{jj}^{(2k)} = \infty$: Consequently, $\{M_n\}_{n=0}^\infty$ is recurrent!

Now consider the probability $\mathbf{P}\{M_n = j \text{ for some } n \geq 1\} = \mathbf{P}\{M_n = j \text{ for some } n \geq 1 | M_0 = m_0\}$ to reach a state j eventually with initial fortune m_0 . If we write $j = m_0 + N$ for a suitable integer N we see that $\mathbf{P}\{M_N = j\} = (1/2)^{|N|}$. As the chain is recurrent we have after each (certain!) return of the chain to the initial state the probability $(1/2)^{|N|}$ to reach j in N steps. And by repeating this experiment add infinum (at each return to m_0) we see that we must have $\mathbf{P}\{M_n = j \text{ for some } n \geq 1\} = 1$.

To prove that the expected value of the time considered in the previous paragraph (found to be finite with probability 1) it takes the chain to reach the state j when

¹Proof: Let $P_j(z) = \sum_{n=0}^\infty p_{jj}^{(n)} z^n$ and $F_j(z) = \sum_{n=0}^\infty f_{jj}^{(n)} z^n$ for $|z| < 1$ with $f_{jj}^{(n)} = \mathbf{P}\{M_n = j, M_{n-1} \neq j, \dots, M_1 \neq j | M_0 = j\}$ and $f_{jj}^{(0)} = 0$. Then $p_{jj}^{(n)} = \sum_{k=0}^n f_{jj}^{(k)} p_{jj}^{(n-k)}$ (a discrete convolution) for $n \geq 1$, so that $P_j(z) = 1 + \sum_{n=1}^\infty p_{jj}^{(n)} z^n = 1 + \sum_{n=1}^\infty \sum_{k=0}^n f_{jj}^{(k)} p_{jj}^{(n-k)} z^n = 1 + \sum_{n=0}^\infty \sum_{k=0}^n f_{jj}^{(k)} p_{jj}^{(n-k)} z^n = 1 + \sum_{k=0}^\infty (\sum_{n=k}^\infty p_{jj}^{(n-k)} z^{n-k}) f_{jj}^{(k)} z^k = 1 + P_j(z) F_j(z)$, so that $\mathbf{P}\{M_n = j \text{ for some } n \geq 1 | M_0 = j\} = \sum_{n=1}^\infty f_{jj}^{(n)} = \lim_{z \uparrow 1} F_j(z) = \lim_{z \uparrow 1} (P_j(z) - 1) / P_j(z) = 1$ (=recurrence) if $\sum_{n=1}^\infty p_{jj}^{(n)} = \lim_{z \uparrow 1} P_j(z) = \infty$ while the same probability is strictly less than 1 (=transience) if $\sum_{n=1}^\infty p_{jj}^{(n)} = \lim_{z \uparrow 1} P_j(z) < \infty$.

starting at the state m_0 is always infinite we may without loss of generality consider the special case when $m_0 = 0$ and $j = 1$, as it is clear that if the expectation is infinite in that special case, then it will be infinite for all other cases as well (as the chain then has a longer way to travel to its goal). Now consider the modified chain $\{\bar{M}_n\}_{n=0}^\infty$ with state space E , initial probability $\mathbf{p}(0)$ and transition probability matrix P given by

$$E = [\dots -2 \ -1 \ 0 \ 1], \quad \mathbf{p}(0) = [\dots 0 \ 0 \ 0 \ 1] \quad \text{and} \quad P = \begin{bmatrix} \dots & \vdots & \vdots & \vdots & \vdots \\ \dots & 0 & 1/2 & 0 & 0 \\ \dots & 1/2 & 0 & 1/2 & 0 \\ \dots & 0 & 1/2 & 0 & 1/2 \\ \dots & 0 & 0 & 1 & 0 \end{bmatrix},$$

respectively. Clearly, it is enough to show that we have $\mathbf{E}\{T_1\} = \infty$ for the recurrence time $T_1 = \min\{n \geq 1 : \bar{M}_n = 1\}$ as the expectation we are interested in for the original chain must be $\mathbf{E}\{T_1\} - 1$. This in turn we can establish by means of proving that the chain doesn't have a stationary distribution, as in the presence of a stationary distribution $\pi = [\dots -\pi_{-2} \ \pi_{-1} \ \pi_0 \ \pi_1]$ we must have $\mathbf{E}\{T_1\} = 1/\pi_1$ where all elements of π (and in particular π_1) must be strictly positive as a consequence of the fact that all states communicate with each other. However, the equation system $\pi = \pi P$ to find the stationary distribution (if it exists) spells out as

$$(1/2) \pi_{n-1} + (1/2) \pi_{n+1} = \pi_n \quad \text{for } n < 0, \quad (1/2) \pi_{-1} + \pi_1 = \pi_0 \quad \text{and} \quad (1/2) \pi_0 = \pi_1,$$

which in turn has unique solution $\pi = [\dots \ \pi_0 \ \pi_0 \ \pi_0 \ (1/2)\pi_0]$. This however can never be a probability distribution (as it cannot sum up to 1), and we are done!

Så de så!²

²Remark of the type setter.