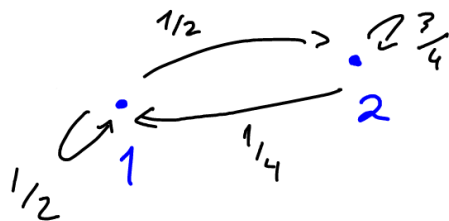


Exercise 1



1.1
$$P(T_1 = k | X_0 = 1) = P(X_0 = X_k = 1, X_1 = \dots = X_{k-1} = 2 | X_0 = 1)$$

(k=1)
$$= P(X_1 = 1 | X_0 = 1) = 1/2$$

(k > 1)
$$= P(X_1 = 2 | X_0 = 1) \times \prod_{j=2}^{k-1} P(X_j = 2 | X_{j-1} = 2) \times P(X_k = 1 | X_{k-1} = 2)$$

$$= \frac{1}{2} \times \left(\frac{3}{4}\right)^{k-2} \cdot \frac{1}{4}$$

Note that
$$P(T_1 < +\infty | X_0 = 1) = \sum_{k \geq 1} P(T_1 = k | X_0 = 1)$$

$$= \frac{1}{2} + \frac{1}{8} \cdot \sum_{k \geq 2} \left(\frac{3}{4}\right)^{k-2} = \frac{1}{2} + \frac{1}{8} \left(\frac{1}{1 - 3/4}\right)$$

$$= \frac{1}{2} + \frac{1}{8} \cdot 4 = 1. \quad \left(\text{This is because 1 is a rec. state}\right)$$

This describes the distribution of T_1 .

1.2 Find $\mu = (\mu_1, \mu_2)$ s.t. $\mu Q = \mu$ and $\mu \mathbf{1}^T = 1$

That is an eigenvector of Q^T for the eigenvalue 1.

$$\text{Ker}(Q^T - I) = \text{Ker} \begin{pmatrix} -1/2 & 1/4 \\ 1/2 & -1/4 \end{pmatrix} = \left\langle \begin{pmatrix} 1 \\ 2 \end{pmatrix} \right\rangle$$

Thus $\mu = \left(\frac{1}{3}, \frac{2}{3}\right)$.

$$E[T_1] = \frac{1}{2} \times 1 + \sum_{k \geq 2} k \cdot \frac{1}{8} \left(\frac{3}{4}\right)^{k-2} = \frac{1}{2} + \frac{1}{8} \cdot \left(\frac{4}{3}\right) \cdot \sum_{k \geq 2} k \cdot \left(\frac{3}{4}\right)^{k-1}$$

$$= \frac{1}{2} + \frac{1}{6} f\left(\frac{3}{4}\right) \quad \text{where} \quad f(x) := \sum_{k \geq 2} k \cdot x^{k-1} = g'(x)$$

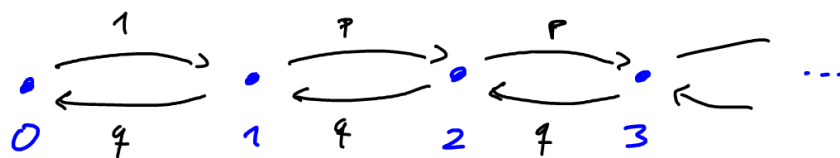
$$g(x) := \sum_{k \geq 2} x^k = \frac{x^2}{1-x}$$

Therefore,
$$f(x) = \frac{d}{dx} g(x) = \frac{2x(1-x) + x^2}{(1-x)^2} = \frac{2x - x^2}{(1-x)^2}$$

And
$$E[T_1] = \frac{1}{2} + \frac{1}{6} \frac{2 \cdot \frac{3}{4} - \left(\frac{3}{4}\right)^2}{\left(1 - \frac{3}{4}\right)^2} = \frac{1}{2} + \frac{1}{6} \frac{(24 - 9)/16}{1/16} = \frac{1}{2} + 4 - \frac{3}{2} = 3$$

Note that $\mu(1)^{-1} = 3 = \mathbb{E}[T_1]$ \square

Exercise 2



$$q = 1-p \in (0,1).$$

$$2.1 \quad \mu_i = \begin{cases} (p/q)^i, & \text{if } i \geq 1 \\ p, & \text{if } i = 0 \end{cases}$$

$$\begin{aligned} \text{Then } (\mu \cdot Q)_0 &= \mu_1 Q_{1,0} = \frac{p}{q} \cdot q = p = \mu_0 \\ (\mu \cdot Q)_1 &= \mu_0 Q_{0,1} + \mu_2 Q_{2,1} = p \cdot 1 + \left(\frac{p}{q}\right)^2 \cdot q \\ &= \frac{pq + p^2}{q} = \frac{p}{q} \underbrace{(p+q)}_{=1} = \mu_1 \end{aligned}$$

$$\begin{aligned} j \geq 2; \quad (\mu \cdot Q)_j &= \mu_{j-1} Q_{j-1,j} + \mu_{j+1} Q_{j+1,j} \\ &= \left(\frac{p}{q}\right)^{j-1} \cdot p + \left(\frac{p}{q}\right)^{j+1} q = \left(\frac{p}{q}\right)^j \underbrace{\left[\frac{q}{p} \cdot p + \frac{p}{q} \cdot q \right]}_{=1} \\ &= \mu_j \end{aligned}$$

Thus $\mu = \mu Q$ is a stationary measure.

$$2.2 \quad \mu(\mathbb{Z}_{>0}) = p + \sum_{i \geq 1} \left(\frac{p}{q}\right)^i = p + \frac{p/q}{1-p/q} < +\infty$$

$$p/q < 1 \text{ for } p < 0.5$$

Therefore, by Theorem 16.3, because this is an irreducible MC, and has a finite measure, all states are recurrent.

Exercise 3

$$H_0 = \inf_n \{X_n = 0\}, \quad H_1 = \inf_n \{X_n = 1\}$$

$$\phi(s) := \mathbb{E}[s^{H_0} \mid X_0 = 1]$$

3.1 The strong Markov property says that $(\{X_{n+H_1}\}_{n \geq 0}, X_0 = 2)$ has the same distribution as $(\{X_n\}_{n \geq 0}, X_0 = 1)$.

Further, $(\{X_{n+H_1}\}_{n \geq 0}, X_0 = 2) \perp\!\!\!\perp \{X_0, \dots, X_{H_1}\}$.

It follows that $\mathbb{P}_1(H_0=k) = \mathbb{P}_2(H_1 + \tilde{H}_0 = H_1+k) \quad \forall k$
 and that $\tilde{H}_0 \perp H_1$.

Also $(\{X_n - 1\}_{n=0}^{H_1}, X_0=2) \sim (\{X_n\}_{n=0}^{H_0}, X_0=1)$ because
 the Markov chains have the same transition matrix and initial dist.
 under the identification $k \mapsto k-1$.

$$\text{Thus } \mathbb{P}_2(H_1=k) = \mathbb{P}_1(H_0=k).$$

It follows that $\mathbb{E}_2[S^{H_0}] = \mathbb{E}_2[S^{\tilde{H}_0 + H_1}] = \mathbb{E}_2[S^{\tilde{H}_0}] \mathbb{E}_2[S^{H_1}] = \phi(s)^2$
 $\mathbb{E}_1[S^{H_0} | X_1=2] = \mathbb{P}_1(X_2=2) \cdot \mathbb{E}_1[S^{H_0} | X_1=2] = p \cdot \mathbb{E}_2[S^{H_0+1}] = p \cdot s \cdot \phi(s)^2$

3.2

$$\mathbb{E}[S^{H_0} | X_0=1] = p \mathbb{E}[S^{H_0} | X_1=2] + q \mathbb{E}[S^{H_0} | X_1=0]$$

$$= p \cdot s \phi(s)^2 + q \cdot s \quad \text{as desired}$$

$$\phi^2 ps - \phi + qs = 0 \Rightarrow \phi(s) = \frac{1 \pm \sqrt{1 - 4pq s^2}}{2ps} \quad (**)$$

From $\sqrt{1-x} = 1 - \frac{x}{2} + O(x^2)$, for $s \rightarrow 0$ we get
 $\phi(s) = \frac{1 \pm (1 - 2pq s^2 + O(s^4))}{2ps}$

Now $\phi(0) = \mathbb{P}(H_0=0) \leq 1$ so for s close to zero
 we have $\phi(s) = \frac{1 - \sqrt{1 - 4pq s^2}}{2ps}$

Let us now study when do the positive and negative roots of $**$
 coincide. This happens when $1 - 4pq s^2 = 0$, or for $s^2 = \frac{1}{4pq}$

Now $pq \leq \left(\frac{p+q}{2}\right)^2 = \frac{1}{4}$ so $\frac{1}{4pq} \geq 1$. $(**)$

The positive and the negative root of $(**)$ are distinct for $1 - 4pq s^2 \neq 0$, so for $s \in [0, 1)$ they are always distinct according to $(*)$.

They also vary continuously wrt s , so

$$\phi(s) = \frac{1 - \sqrt{1 - 4pq s^2}}{2ps}$$

3.3 From $\sqrt{1-x} = 1 - \frac{x}{2} - \frac{x^2}{8} + O(x^3)$ we have

$$\phi(s) = \frac{2pq s^2 + 2p^2 q^2 s^4 + O(s^6)}{2ps} = qs + pq s^3 + O(s^5)$$

It follows that $P(H_0 = 3 \mid X_0 = 1) = p \cdot q$

3.4 $\phi(s) = \frac{1 - \sqrt{1 - 4pq s^2}}{2ps}$, so:

$$\lim_{s \rightarrow 1^-} \phi(s) = \frac{1 - \sqrt{1 - 4pq}}{2p} = \frac{1 - \sqrt{1 - 4p(1-p)}}{2p} = \frac{1 - \sqrt{(1-2p)^2}}{2p} = \begin{cases} 1, & \text{if } p \leq 1/2 \\ \frac{-2p+2}{2p}, & \text{if } p > 1/2 \end{cases}$$

$$= \begin{cases} 1, & \text{if } p \leq 1/2 \\ \frac{q}{p}, & \text{if } p > 1/2 \end{cases} = P(H_0 < +\infty \mid X_0 = 1)$$

3.5 Because $\frac{d}{ds}(1 - \sqrt{1 - 4pq s^2}) = \frac{8pq s}{2\sqrt{1 - 4pq s^2}} = \frac{4pq s}{\sqrt{1 - 4pq s^2}}$, we have

$$\phi'(s) = \frac{2ps \cdot \frac{4pq s}{\sqrt{1 - 4pq s^2}} - 2p(1 - \sqrt{1 - 4pq s^2})}{4p^2 s^2} = \frac{2q}{\sqrt{1 - 4pq s^2}} - \frac{1 - \sqrt{1 - 4pq s^2}}{2ps^2}$$

Thus, $\lim_{s \rightarrow 1^-} \phi'(s) = \frac{2q}{|1-2p|} - \frac{1 - |1-2p|}{2p} = \frac{2q}{1-2p} - 1$ $p < 0.5$

$$= \frac{2q - 1 + 2p}{1-2p} = \frac{1}{1-2p} \quad \square$$