

5.C.1. Spectral Properties

For a time independent \mathbf{Q} , we have

$$Q_{mn}(s) = P_{1|1}(m, s | n, s+1) = P_{1|1}(m, 0 | n, 1) = Q_{mn} \quad \forall s \quad (5.23)$$

(5.21) can be solved by iteration starting with

$$P_{1|1}(m, s | n, s) = \delta_{mn}$$

For the 1st time step (or iteration), (5.21) becomes

$$\begin{aligned} P_{1|1}(m, s | n, s+1) &= \sum_k P_{1|1}(m, s | k, s) P_{1|1}(k, s | n, s+1) \\ &= \sum_k \delta_{mk} P_{1|1}(k, s | n, s+1) \\ &= P_{1|1}(m, s | n, s+1) \\ &= Q_{mn} \end{aligned}$$

For the 2nd time step,

$$\begin{aligned} P_{1|1}(m, s-1 | n, s+1) &= \sum_k P_{1|1}(m, s-1 | k, s) P_{1|1}(k, s | n, s+1) \\ &= \sum_k P_{1|1}(m, s-1 | k, s) Q_{kn} \\ &= \sum_k Q_{mk} Q_{kn} \\ &= (\mathbf{Q}^2)_{mn} \end{aligned}$$

For the 3rd time step,

$$\begin{aligned} P_{1|1}(m, s-2 | n, s+1) &= \sum_k P_{1|1}(m, s-2 | k, s) P_{1|1}(k, s | n, s+1) \\ &= \sum_k (\mathbf{Q}^2)_{mk} Q_{kn} \\ &= (\mathbf{Q}^3)_{mn} \end{aligned}$$

In general,

$$P_{1|1}(m, s_0 | n, s) = (\mathbf{Q}^{s-s_0})_{mn} \quad (5.24)$$

Similarly, (5.20) gives, by iteration,

$$\begin{aligned} P(n, s+1) &= \sum_m P(m, s) Q_{mn} \\ &= \sum_{m,k} P(k, s-1) Q_{km} Q_{mn} = \sum_k P(k, s-1) (\mathbf{Q}^2)_{kn} \\ &\vdots \\ &= \sum_k P(k, 0) (\mathbf{Q}^{s+1})_{kn} \end{aligned} \quad (5.25)$$

As in quantum mechanics, the Dirac notation greatly clarifies the mathematics.

The state of the system at time s is represented by the **probability vector** $|p(s)\rangle$ in the sample space, which is spanned by the eigenvectors $|n\rangle$ of the stochastic variable operator Y with

$$\begin{aligned} Y |n\rangle &= y(n) |n\rangle & n &= 1, \dots, M \\ \langle n | &= |n\rangle^T & & \quad T = \text{transpose operator} \end{aligned}$$

Note: $|p(s)\rangle$ is analogous to the state vector $|\psi(t)\rangle$ in the Hilbert space of quantum mechanics except that it is real.

Probability $P(n, s)$ is then the projection (or component) of $|p(s)\rangle$ along $|n\rangle$:

$$P(n, s) = \langle p(s) | n \rangle = \langle n | p(s) \rangle \quad (a)$$

From (5.20), we see that the conditional probability $P_{1|1}$ is an operator that relates probability vectors of different times. Hence, we set

$$P_{1|1}(m, s_0 | n, s) = \langle m | P_{1|1}(s_0 | s) | n \rangle \quad (b)$$

so that (5.20) becomes

$$\begin{aligned} \langle p(s+1) | n \rangle &= \sum_{m=1}^M \langle p(s) | m \rangle \langle m | P_{1|1}(s | s+1) | n \rangle \\ &= \langle p(s) | P_{1|1}(s | s+1) | n \rangle \end{aligned} \quad (c)$$

where the completeness condition

$$\sum_{m=1}^M | m \rangle \langle m | = I \quad (d)$$

was used.

Since (5.25a) is true for all $| n \rangle$, we have

$$\langle p(s+1) | = \langle p(s) | P_{1|1}(s | s+1) \quad (e)$$

(5.23) then becomes

$$Q_{mn} = \langle m | P_{1|1}(s | s+1) | n \rangle \quad (f)$$

so that \mathbf{Q} is just the matrix form of the operator $P_{1|1}(s | s+1)$. This justifies the use of \mathbf{Q} to denote both the operator $P_{1|1}(s | s+1)$ and its matrix representation. (f) thus becomes

$$Q_{mn} = \langle m | \mathbf{Q} | n \rangle \quad (g)$$

Similarly, (5.24) becomes

$$\begin{aligned} \langle m | P_{1|1}(s_0 | s) | n \rangle &= (\mathbf{Q}^{s-s_0})_{mn} \\ \rightarrow P_{1|1}(s_0 | s) &= P_{1|1}(s_0 | s_0+1) \dots P_{1|1}(s-1 | s) = \mathbf{Q}^{s-s_0} \end{aligned} \quad (h)$$

The eigenvalues of \mathbf{Q} are given by the solutions of the secular equation

$$\det | \mathbf{Q} - \lambda I | = 0 \quad (5.26)$$

where I is the $M \times M$ unit matrix. As can be seen from (f) & (h), \mathbf{Q} is a real but non-symmetric matrix in the basis $\{ | n \rangle \}$. Therefore, the left and right eigenvectors for the same eigenvalue are not equal. Let

$$\mathbf{Q} | \psi_i \rangle = \lambda_i | \psi_i \rangle \quad (i)$$

$$\langle \chi_i | \mathbf{Q} = \langle \chi_i | \lambda_i \quad (j)$$

Since \mathbf{Q} is always real, we define

$$\langle \chi_i | = | \psi_i \rangle^T$$

(j) is then equivalent to

$$\mathbf{Q}^T | \chi_i \rangle = \lambda_i | \chi_i \rangle \quad (k)$$

which is the preferred form when using computer software for calculation.

Caution: for a real but non-symmetric matrix, both its eigenvalues and eigenvectors can be complex.

With the help of the completeness condition and

$$\psi_i(n) = \langle n | \psi_i \rangle = \langle \psi_i | n \rangle$$

(i) can be written as

$$\begin{aligned} \sum_n \langle m | \mathbf{Q} | n \rangle \langle n | \psi_i \rangle &= \lambda_i \langle m | \psi_i \rangle \\ \rightarrow \sum_n Q_{mn} \psi_i(n) &= \lambda_i \psi_i(m) \end{aligned} \quad (5.28)$$

Similarly, (k) becomes

$$\begin{aligned} & \sum_n \langle \chi_i | n \rangle \langle n | \mathbf{Q} | m \rangle = \langle \chi_i | m \rangle \lambda_i \\ \rightarrow & \sum_n \chi_i(n) Q_{nm} = \chi_i(m) \lambda_i \end{aligned} \quad (5.27)$$

(i) & (j) give

$$\begin{aligned} & \langle \chi_i | \mathbf{Q} | \psi_j \rangle = \lambda_j \langle \chi_i | \psi_j \rangle \\ & \langle \chi_i | \mathbf{Q} | \psi_j \rangle = \lambda_i \langle \chi_i | \psi_j \rangle \\ \rightarrow & 0 = (\lambda_j - \lambda_i) \langle \chi_i | \psi_j \rangle \\ \therefore & \langle \chi_i | \psi_j \rangle = 0 \quad \text{for} \quad \lambda_i \neq \lambda_j \end{aligned}$$

i.e., left and right eigenvectors belonging to different eigenvalues are orthogonal.

For eigenvectors belonging to the same eigenvalue, we can use the Schmidt orthogonalization method to obtain a set of mutually orthogonal vectors.

Therefore, assuming all eigenvectors are properly normalized, we can always write

$$\begin{aligned} & \langle \chi_i | \psi_j \rangle = \delta_{ij} \quad (5.30) \\ \rightarrow & \sum_{n=1}^M \langle \chi_i | n \rangle \langle n | \psi_j \rangle = \delta_{ij} \\ & \sum_{n=1}^M \chi_i(n) \psi_j(n) = \delta_{ij} \end{aligned} \quad (5.30a)$$

Consider the expansion

$$\begin{aligned} & \langle p | = \sum_{i=1}^M \alpha_i \langle \chi_i | \quad (m) \\ \rightarrow & \langle p | \psi_j \rangle = \sum_{i=1}^M \alpha_i \langle \chi_i | \psi_j \rangle = \sum_{i=1}^M \alpha_i \delta_{ij} = \alpha_j \end{aligned}$$

Putting this back into (m) gives

$$\langle p | = \sum_{i=1}^M \langle p | \psi_i \rangle \langle \chi_i |$$

Since this is true for arbitrary $\langle p |$, we must have

$$\sum_{i=1}^M | \psi_i \rangle \langle \chi_i | = I \quad (5.31)$$

which is the **completeness condition** for the eigenvectors.

Note that what (5.31) means is that these possibly complex eigenvectors span an M -D real space that is also spanned by the M real vectors $| n \rangle$. To be consistent, eigenvalues and eigenvectors, if complex, must appear in conjugate pairs. This is so since \mathbf{Q} is real.

In terms of components with respect to the basis $| n \rangle$, we have

$$\begin{aligned} & \sum_{i=1}^M \langle m | \psi_i \rangle \langle \chi_i | n \rangle = \langle m | n \rangle \\ \rightarrow & \sum_{i=1}^M \psi_i(m) \chi_i(n) = \delta_{mn} \end{aligned} \quad (5.31a)$$

Taking the absolute value of (5.28) gives

$$\left| \sum_{n=1}^M Q_{mn} \psi_i(n) \right| = |\lambda_i| |\psi_i(m)|$$

Since all Q_{mn} are real,

$$|\lambda_i| |\psi_i(m)| \leq \sum_n Q_{mn} |\psi_i(n)| \tag{5.32}$$

Let

$$|\psi_i(n)| \leq C \quad \forall n$$

then

$$\sum_{n=1}^M Q_{mn} |\psi_i(n)| \leq C \sum_{n=1}^M Q_{mn} = C \tag{5.33}$$

where the discrete version of (5.13)

$$\sum_{n=1}^M P_{1|1}(m, s | n, s+1) = \sum_{n=1}^M Q_{mn} = 1 \tag{n}$$

was used.

Let

$$|\psi_i(n_0)| = C$$

then setting $m = n_0$ in (5.32) gives

$$|\lambda_i| C \leq C$$

$$\rightarrow |\lambda_i| \leq 1 \quad \forall i \tag{5.34}$$

Using (n), we see that (5.28) can always be satisfied by

$$\lambda_1 = 1 \quad \text{with} \quad \psi_1(n) = c \quad \forall n \tag{o}$$

where c is a normalization constant and we have set arbitrarily $i = 1$.

In other words, every Q has an eigenvalue of 1 with left eigenvector given by (o).

For the right eigenvector, setting $\lambda_1 = 1$ in (5.27) gives

$$\chi_1(m) = \sum_{n=1}^M \chi_1(n) Q_{nm} \tag{5.35}$$

Setting

$$\chi_1(m) = \langle p(s) | m \rangle = P(m, s)$$

(5.35) becomes

$$\begin{aligned} \langle p(s) | m \rangle &= \sum_{n=1}^M \langle p(s) | n \rangle \langle n | P_{1|1}(s | s+1) | m \rangle \\ &= \langle p(s+1) | m \rangle \quad [(5.20) \text{ used. }] \end{aligned}$$

$$\rightarrow |p(s)\rangle = |p(s+1)\rangle$$

χ_1 is therefore a stationary probability vector. We denote this as

$$\chi_1(n) = P_{ST}(n) \tag{p}$$

The constant c in (o) can be determined from the normalization

$$\langle \chi_1 | \psi_1 \rangle = 1$$

$$\rightarrow \sum_{n=1}^M \chi_1(n) \psi_1(n) = c \sum_{n=1}^M \chi_1(n) = 1$$

Using the sum rule for probabilities

$$\sum_{n=1}^M \chi_1(n) = \sum_{n=1}^M P_{ST}(n) = 1$$

we have

$$c = 1 \quad \rightarrow \quad \psi_1(n) = 1 \quad \forall n \quad (\text{o}')$$

On the other hand, $(l) \mid \psi_j \rangle$ gives

$$\begin{aligned} \langle \chi_i \mid \mathbf{Q} \mid \psi_j \rangle &= \lambda_i \langle \chi_i \mid \psi_j \rangle \\ &= \lambda_i \delta_{ij} \end{aligned} \quad [(5.30) \text{ used. }] \quad (\text{q})$$

$$\rightarrow \quad \sum_{i=1}^M \langle \chi_i \mid \mathbf{Q} \mid \psi_j \rangle = \lambda_j$$

which implies

$$\mathbf{Q} = \sum_{i=1}^M \lambda_i \mid \psi_i \rangle \langle \chi_i \mid \quad (\text{r})$$

Proof: (r) gives

$$\begin{aligned} \langle \chi_i \mid \mathbf{Q} \mid \psi_j \rangle &= \sum_{k=1}^M \lambda_k \langle \chi_i \mid \psi_k \rangle \langle \chi_k \mid \psi_j \rangle \\ &= \sum_{k=1}^M \lambda_k \delta_{ik} \delta_{kj} = \lambda_i \delta_{ij} \end{aligned}$$

which is (p).

$\langle m \mid (q) \mid n \rangle$ gives

$$\begin{aligned} Q_{mn} &= \sum_{i=1}^M \lambda_i \langle m \mid \psi_i \rangle \langle \chi_i \mid n \rangle \\ &= \sum_{i=1}^M \lambda_i \psi_i(m) \chi_i(n) \quad (5.36) \\ &= P_{1|1}(m, s \mid n, s+1) \quad \forall s \quad [\text{see (5.23). }] \end{aligned}$$

From (5.24), we have

$$\begin{aligned} P_{1|1}(m, s \mid n, s+2) &= (\mathbf{Q}^2)_{mn} \\ &= \sum_{i,j=1}^M \lambda_i \lambda_j \langle m \mid \psi_i \rangle \langle \chi_i \mid \psi_j \rangle \langle \chi_j \mid n \rangle \\ &= \sum_{i=1}^M \lambda_i^2 \langle m \mid \psi_i \rangle \langle \chi_i \mid n \rangle \\ &= \sum_{i=1}^M \lambda_i^2 \psi_i(m) \chi_i(n) \end{aligned}$$

By induction, we have

$$P_{1|1}(m, s_0 \mid n, s) = (\mathbf{Q}^{s-s_0})_{mn} = \sum_{i=1}^M \lambda_i^{s-s_0} \psi_i(m) \chi_i(n) \quad (5.37)$$

which, together with

$$P(n, s) = \chi_1(n)$$

provide the general solution to the Markov chain.

\mathbf{Q} is called **regular** if

$$\mathbf{Q}^N \neq \mathbf{0} \quad \forall N$$

Since $|\lambda_i| \leq 1$ with $\lambda_1 = 1$, we write (5.37) as

$$P_{1|1}(m, s_0 \mid n, s) = \psi_1(m) \chi_1(n) + \sum_{i=2}^M \lambda_i^{s-s_0} \psi_i(m) \chi_i(n)$$

$$= P_{ST}(n) + \sum_{i=2}^M \lambda_i^{s-s_0} \psi_i(m) \chi_i(n) \tag{5.38}$$

where (o), (o') & (p) were used.

Now,

$$|\lambda_i| < 1 \rightarrow \lim_{s \rightarrow \infty} |\lambda_i|^{s-s_0} = 0 \quad \forall i \geq 2$$

$$\therefore \lim_{s \rightarrow \infty} \lambda_i^{s-s_0} = 0 \quad \forall i \geq 2$$

and (5.38) gives

$$\lim_{s \rightarrow \infty} P_{1|1}(m, s_0 | n, s) = P_{ST}(n) \tag{5.39}$$

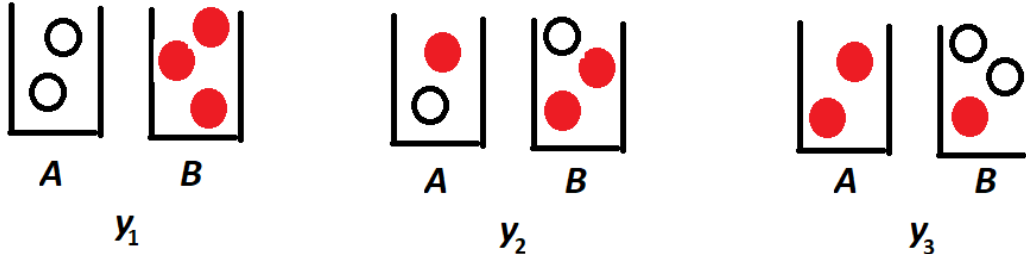
Similarly, (5.25) gives

$$\begin{aligned} \lim_{s \rightarrow \infty} P(n, s) &= \sum_{m=1}^M P(m, 0) P_{ST}(n) \\ &= P_{ST}(n) \quad [(5.6) \text{ used. }] \end{aligned} \tag{5.40}$$

Markov chains with regular Q are **ergodic**, which means every state is eventually reachable starting from any state. This is no longer true if Q is block diagonal. In which case, multiple long-time states exist.

Exercise 5.1.

Consider 2 pots, A & B , with 3 red & 2 white balls distributed between them so that A always has 2 & B has 3 balls. There are 3 different configurations as shown in the figure below.



At each time step, transition between configurations is attained by picking up randomly 1 ball each from A & B and interchanging them.

- (a) Compute the transition matrix Q and the conditional probability $P_{1|1}(m, s_0 | n, s)$.
- (b) If initially $P(1, 0) = 1$, $P(2, 0) = 0$ and $P(3, 0) = 0$, compute the probabilities $P(n, s)$ for $n = 1, 2, 3$ at time s .
- (c) Assume the realization $y(n) = n$. Compute the first moment, $\langle y(s) \rangle$, and the autocorrelation function, $\langle y(0) y(s) \rangle$, for the same initial conditions as (b).

Answer (a)

Starting with $Y(s) = y_1$, one can only reach $Y(s + 1) = y_2$. Therefore

$$Q_{11} = 0 \quad Q_{12} = 1 \quad Q_{13} = 0$$

Starting with $Y(s) = y_2$, one can reach $Y(s + 1) = y_1$ by exchanging a red ball in A with a white ball in

B with a probability $\frac{1}{2} \times \frac{1}{3} = \frac{1}{6}$. Therefore $Q_{21} = \frac{1}{6}$.

There are two ways to remain in y_2 . The first is to pick up a white ball from both pots with probability

$$\frac{1}{2} \times \frac{1}{3} = \frac{1}{6}. \text{ The second is to pick a red ball from both pots with probability } \frac{1}{2} \times \frac{2}{3} = \frac{1}{3}. \text{ Therefore,}$$

$$Q_{22} = \frac{1}{6} + \frac{1}{3} = \frac{1}{2}.$$

In order to change to y_3 , one must pick a white ball from A and a red ball from B with probability

$$\frac{1}{2} \times \frac{2}{3} = \frac{1}{3}. \text{ Therefore } Q_{23} = \frac{1}{3}.$$

Starting with $Y(s) = y_3$, there is no way to reach $Y(s+1) = y_1$. Therefore $Q_{31} = 0$.

One can reach $Y(s+1) = y_2$ by exchanging a red ball in A with a white ball in B with a probability

$$1 \times \frac{2}{3} = \frac{2}{3}. \text{ Therefore } Q_{32} = \frac{2}{3}.$$

In order to remain in y_3 , one must pick a red ball from both A and B with probability $1 \times \frac{1}{3} = \frac{1}{3}$. There-

fore $Q_{33} = \frac{1}{3}$.

Thus,

$$Q = \begin{pmatrix} 0 & 1 & 0 \\ \frac{1}{6} & \frac{1}{2} & \frac{1}{3} \\ 0 & \frac{2}{3} & \frac{1}{3} \end{pmatrix} \quad Q^2 = \begin{pmatrix} \frac{1}{6} & \frac{1}{2} & \frac{1}{3} \\ \frac{1}{12} & \frac{23}{36} & \frac{5}{18} \\ \frac{1}{9} & \frac{5}{9} & \frac{1}{3} \end{pmatrix}$$

Using the *Mathematica* code in the section Code, we get

$$\begin{aligned} \lambda_1 = 1 & \quad | \psi_1 \rangle = \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} & \quad \langle \chi_1 | = \left(\frac{1}{3} \quad 2 \quad 1 \right) \\ \lambda_2 = -\frac{1}{3} & \quad | \psi_2 \rangle = \begin{pmatrix} 3 \\ -1 \\ 1 \end{pmatrix} & \quad \langle \chi_2 | = (1 \quad -2 \quad 1) \\ \lambda_3 = \frac{1}{6} & \quad | \psi_3 \rangle = \begin{pmatrix} -\frac{3}{2} \\ -\frac{1}{4} \\ 1 \end{pmatrix} & \quad \langle \chi_3 | = \left(-\frac{1}{2} \quad -\frac{1}{2} \quad 1 \right) \end{aligned}$$

The eigenvectors above are mutually orthogonal not normalized. Keeping $| \psi_i \rangle$ as is, the normalized $\langle \chi_i |$ are easily found to be

$$\langle \chi_1 | = \frac{3}{10} \left(\frac{1}{3} \quad 2 \quad 1 \right) \quad \langle \chi_2 | = \frac{1}{6} (1 \quad -2 \quad 1) \quad \langle \chi_3 | = \frac{8}{15} \left(-\frac{1}{2} \quad -\frac{1}{2} \quad 1 \right)$$

The conditional probability matrix is therefore

$$\begin{aligned} P_{1|1}(s_0 | s) &= \sum_{i=1}^3 \lambda_i^{s-s_0} | \psi_i \rangle \langle \chi_i | \\ &= \begin{pmatrix} \frac{1}{10} & \frac{3}{5} & \frac{3}{10} \\ \frac{1}{10} & \frac{3}{5} & \frac{3}{10} \\ \frac{1}{10} & \frac{3}{5} & \frac{3}{10} \end{pmatrix} + \left(-\frac{1}{3} \right)^{s-s_0} \begin{pmatrix} \frac{1}{2} & -1 & \frac{1}{2} \\ -\frac{1}{6} & \frac{1}{3} & -\frac{1}{6} \\ \frac{1}{6} & -\frac{1}{3} & \frac{1}{6} \end{pmatrix} + \left(\frac{1}{6} \right)^{s-s_0} \begin{pmatrix} \frac{2}{5} & \frac{2}{5} & -\frac{4}{5} \\ \frac{1}{15} & \frac{1}{15} & -\frac{2}{15} \\ -\frac{4}{15} & -\frac{4}{15} & \frac{8}{15} \end{pmatrix} \end{aligned}$$

where $| \psi_i \rangle \langle \chi_i |$ was calculated as "rIProd" in the section Code.

Answer (b)

The initial condition can be written as

$$\langle p(0) | = (1 \ 0 \ 0)$$

Using

$$\langle p(s) | = \langle p(0) | P_{1|1}(0 | s)$$

we have [see Code]

$$\langle p(s) | = \left(\frac{1}{10} + \frac{1}{2} \left(-\frac{1}{3} \right)^s + \frac{1}{5} \times 2^{1-s} \times 3^{-s}, \frac{3}{5} + (-1)^{1+s} 3^{-s} + \frac{1}{5} \times 2^{1-s} \times 3^{-s}, \frac{3}{10} + \frac{1}{2} \left(-\frac{1}{3} \right)^s - \frac{1}{5} \times 2^{2-s} \times 3^{-s} \right)$$

Note that

$$\begin{aligned} P(n, s) &= \langle p(s) | n \rangle \\ &= \sum_m \langle p(0) | m \rangle \langle m | P_{1|1}(0 | s) | n \rangle \\ &= \sum_m P(m, 0) \langle m | P_{1|1}(0 | s) | n \rangle \\ &= \langle 1 | P_{1|1}(0 | s) | n \rangle \end{aligned}$$

Answer (c)

According to (5.7), the 1st moment is

$$\begin{aligned} \langle y(s) \rangle &= \sum_{n=1}^3 y(n) P(n, s) \\ &= \sum_{n=1}^3 n \langle p(s) | n \rangle \\ &= \frac{11}{5} - \frac{6^{1-s}}{5} \quad [\text{ see Code. }] \end{aligned}$$

For the autocorrelation,

$$\begin{aligned} \langle y(0) y(s) \rangle &= \sum_{m, n=1}^3 y(m) y(n) P_2(m, 0; n, s) \\ &= \sum_{m, n=1}^3 y(m) y(n) P(m, 0) P_{1|1}(m, 0 | n, s) \\ &= \frac{11}{5} - \frac{6^{1-s}}{5} \quad [\text{ see Code. }] \\ &= \langle y(s) \rangle \end{aligned}$$

This coincidence is due to the special initial conditions & realization we used:

$$\begin{aligned} &\sum_{m, n=1}^3 y(m) y(n) P(m, 0) P_{1|1}(m, 0 | n, s) \\ &= \sum_{n=1}^3 y(n) P_{1|1}(1, 0 | n, s) \\ &= \sum_{n=1}^3 n \langle 1 | P_{1|1}(0 | s) | n \rangle \\ &= \sum_{n=1}^3 n P(n, s) \\ &= \langle y(s) \rangle \end{aligned}$$

Code

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Q = 
$$\begin{pmatrix} 0 & 1 & 0 \\ 1 & 1 & 1 \\ 0 & 2 & 1 \\ & 3 & 3 \end{pmatrix};$$


(* eigenvalues & right eigenvectors *)
{λ, rev} = Eigensystem[Q]

{{1, - $\frac{1}{3}$ ,  $\frac{1}{6}$ }, {{1, 1, 1}, {3, -1, 1}, {- $\frac{3}{2}$ , - $\frac{1}{4}$ , 1}}}

(* Checking *)
Table[Q.rev[[i]] == λ[[i]] rev[[i]], {i, 3}]
{True, True, True}

(* eigenvalues & left eigenvectors *)
{λ, lev} = Eigensystem[Q^T]

{{1, - $\frac{1}{3}$ ,  $\frac{1}{6}$ }, {{ $\frac{1}{3}$ , 2, 1}, {1, -2, 1}, {- $\frac{1}{2}$ , - $\frac{1}{2}$ , 1}}}

(* Checking *)
Table[lev[[i]].Q == λ[[i]] lev[[i]], {i, 3}]
{True, True, True}

(* normalization constants *)
nconst = MapThread[Dot, {lev, rev}]

{ $\frac{10}{3}$ , 6,  $\frac{15}{8}$ }

(* Normalized left eigenvectors; right eigenvectors unchanged *)
lev = lev/nconst

{{ $\frac{1}{10}$ ,  $\frac{3}{5}$ ,  $\frac{3}{10}$ }, { $\frac{1}{6}$ , - $\frac{1}{3}$ ,  $\frac{1}{6}$ }, {- $\frac{4}{15}$ , - $\frac{4}{15}$ ,  $\frac{8}{15}$ }}

(* Orthonormality:  $\langle \chi_i | \psi_j \rangle = \delta_{ij}$  *)
Table[lev[[i]].rev[[j]], {i, 3}, {j, 3}] // MatrixForm


$$\begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}$$


(*  $|\psi_i\rangle\langle \chi_i|$  *)
r1Prod = Table[Outer[Times, rev[[i]], lev[[i]], {i, 3}];
MatrixForm/@r1Prod


$$\left\{ \begin{pmatrix} \frac{1}{10} & \frac{3}{5} & \frac{3}{10} \\ \frac{1}{10} & \frac{3}{5} & \frac{3}{10} \\ \frac{1}{10} & \frac{3}{5} & \frac{3}{10} \end{pmatrix}, \begin{pmatrix} \frac{1}{2} & -1 & \frac{1}{2} \\ -\frac{1}{6} & \frac{1}{3} & -\frac{1}{6} \\ \frac{1}{6} & -\frac{1}{3} & \frac{1}{6} \end{pmatrix}, \begin{pmatrix} \frac{2}{5} & \frac{2}{5} & -\frac{4}{5} \\ \frac{1}{15} & \frac{1}{15} & -\frac{2}{15} \\ -\frac{4}{15} & -\frac{4}{15} & \frac{8}{15} \end{pmatrix} \right\}$$


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(* Completeness: $\sum_{i=1}^3 |\psi_i\rangle \langle \chi_i| = \mathbf{I}$ *)

$\sum_{i=1}^3 \text{r1Prod}[[i]] // \text{MatrixForm}$

$$\begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}$$

(* $P_{1|1}(s_0|s)$ *)

$P11[s_0, s] := \text{Sum}[\text{ev}[[i]]^{s-s_0} \text{r1Prod}[[i]], \{i, 3\}]$

$P11[0, s] // \text{MatrixForm}$

$$\begin{pmatrix} \frac{1}{10} + \frac{1}{2} \left(-\frac{1}{3}\right)^s + \frac{1}{5} \times 2^{1-s} \times 3^{-s} & \frac{3}{5} + (-1)^{1+s} 3^{-s} + \frac{1}{5} \times 2^{1-s} \times 3^{-s} & \frac{3}{10} + \frac{1}{2} \left(-\frac{1}{3}\right)^s - \frac{1}{5} \times 2^{2-s} \times 3^{-s} \\ \frac{1}{10} - \frac{1}{2} (-1)^s 3^{-1-s} + \frac{1}{5} \times 2^{-s} \times 3^{-1-s} & \frac{3}{5} + (-1)^s 3^{-1-s} + \frac{1}{5} \times 2^{-s} \times 3^{-1-s} & \frac{3}{10} - \frac{1}{2} (-1)^s 3^{-1-s} - \frac{1}{5} \times 2^{1-s} \times 3^{-s} \\ \frac{1}{10} + \frac{1}{2} (-1)^s 3^{-1-s} - \frac{1}{5} \times 2^{2-s} \times 3^{-1-s} & \frac{3}{5} + \left(-\frac{1}{3}\right)^{1+s} - \frac{1}{5} \times 2^{2-s} \times 3^{-1-s} & \frac{3}{10} + \frac{1}{2} (-1)^s 3^{-1-s} + \frac{1}{5} \times 2^{3-s} \times 3^{-s} \end{pmatrix}$$

(* Q *)

$P11[s, s+1] // \text{MatrixForm}$

$$\begin{pmatrix} 0 & 1 & 0 \\ \frac{1}{6} & \frac{1}{2} & \frac{1}{3} \\ 0 & \frac{2}{3} & \frac{1}{3} \end{pmatrix}$$

(* Q^2 *)

$P11[s, s+2] // \text{MatrixForm}$

$$\begin{pmatrix} \frac{1}{6} & \frac{1}{2} & \frac{1}{3} \\ \frac{1}{12} & \frac{23}{36} & \frac{5}{18} \\ \frac{1}{9} & \frac{5}{9} & \frac{1}{3} \end{pmatrix}$$

(* Answer to (b) *)

$p_0 = \{1, 0, 0\};$

$ps = p_0 \cdot P11[0, s]$

$$\left\{ \frac{1}{10} + \frac{1}{2} \left(-\frac{1}{3}\right)^s + \frac{1}{5} \times 2^{1-s} \times 3^{-s}, \frac{3}{5} + (-1)^{1+s} 3^{-s} + \frac{1}{5} \times 2^{1-s} \times 3^{-s}, \frac{3}{10} + \frac{1}{2} \left(-\frac{1}{3}\right)^s - \frac{1}{5} \times 2^{2-s} \times 3^{-s} \right\}$$

(* Answer to (c) *)

(* 1st moment *)

$yb = \sum_{n=1}^3 n ps[[n]] // \text{Simplify}$

$$\frac{11}{5} - \frac{6^{1-s}}{5}$$

(* autocorrelation function *)

$yyb = \sum_{m=1}^3 \sum_{n=1}^3 m n p_0[[m]] P11[0, s] [[m, n]] // \text{Simplify}$

$$\frac{11}{5} - \frac{6^{1-s}}{5}$$