

## Supplement

### Sup1

It can be obtained by calculation,  $\det(\lambda \mathbf{I} - \mathbf{T}_X) = (\lambda - 1)(\lambda - (2a_X - 1))$ , so the eigenvalues of  $\mathbf{T}_X$  are  $\lambda_1 = 1, \lambda_2 = 2a_X - 1$ . The right eigenvector corresponding to  $\lambda_1$  is  $\mathbf{r}_1 = \left(\frac{1}{\sqrt{2}}, \frac{1}{\sqrt{2}}\right)^T$ , the left eigenvector is  $\boldsymbol{\ell}_1^T = \left(\frac{1}{\sqrt{2}}, \frac{1}{\sqrt{2}}\right)$ . The right eigenvector corresponding to  $\lambda_2$  is  $\mathbf{r}_2 = \left(\frac{1}{\sqrt{2}}, -\frac{1}{\sqrt{2}}\right)^T$ , the left eigenvector is  $\boldsymbol{\ell}_2^T = \left(\frac{1}{\sqrt{2}}, -\frac{1}{\sqrt{2}}\right)$ . Since  $\mathbf{T}_X$  is a symmetric matrix, by the definition of simple matrix,  $\mathbf{T}_X$  is a simple matrix. So it can be obtained from the Spectral decomposition theorem that

$$\begin{aligned} \mathbf{T}_X^k &= \lambda_1^k \mathbf{r}_1 \boldsymbol{\ell}_1^T + \lambda_2^k \mathbf{r}_2 \boldsymbol{\ell}_2^T \\ &= \begin{pmatrix} \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} \end{pmatrix} \begin{pmatrix} \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \end{pmatrix} + (2a_X - 1)^k \begin{pmatrix} \frac{1}{\sqrt{2}} \\ -\frac{1}{\sqrt{2}} \end{pmatrix} \begin{pmatrix} \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} \end{pmatrix} \\ &= \begin{pmatrix} \frac{1}{2} + \frac{1}{2}(2a_X - 1)^k & \frac{1}{2} - \frac{1}{2}(2a_X - 1)^k \\ \frac{1}{2} - \frac{1}{2}(2a_X - 1)^k & \frac{1}{2} + \frac{1}{2}(2a_X - 1)^k \end{pmatrix} \\ &= \frac{1}{2} \begin{pmatrix} 1 + (2a_X - 1)^k & 1 - (2a_X - 1)^k \\ 1 - (2a_X - 1)^k & 1 + (2a_X - 1)^k \end{pmatrix}. \end{aligned} \quad (1)$$

When  $k \geq 1$ ,

$$\begin{aligned} P(d_1^X d_{k+1}^X = 1) &= P(d_{k+1}^X = 1 | d_1^X = 1) P(d_1^X = 1) + \\ &\quad P(d_{k+1}^X = -1 | d_1^X = -1) P(d_1^X = -1) \\ &= \frac{1}{2} (1 + (2a_X - 1)^k) \times \frac{1}{2} + \frac{1}{2} (1 - (2a_X - 1)^k) \times \frac{1}{2} \\ &= \frac{1}{2} (1 + (2a_X - 1)^k). \end{aligned} \quad (2)$$

$$\begin{aligned} P(d_1^X d_{k+1}^X = -1) &= P(d_{k+1}^X = -1 | d_1^X = 1) P(d_1^X = 1) + P(d_{k+1}^X = 1 | d_1^X = -1) P(d_1^X = -1) \\ &= \frac{1}{2} (1 - (2a_X - 1)^k) \times \frac{1}{2} + \frac{1}{2} (1 + (2a_X - 1)^k) \times \frac{1}{2} \\ &= \frac{1}{2} (1 - (2a_X - 1)^k). \end{aligned} \quad (3)$$

So,

$$\begin{aligned} \mathbb{E}(d_1^X d_{k+1}^X) &= \frac{1}{2} (1 + (2a_X - 1)^k) - \frac{1}{2} (1 - (2a_X - 1)^k) \\ &= (2a_X - 1)^k. \end{aligned} \quad (4)$$

In the same way, the stationary distribution of  $d_i^Y$  is  $P(d_i^Y = 1) = P(d_i^Y = -1) = 1/2$ ,  $\mathbb{E}((d_1^Y)^2) = 1$ ,  $\mathbb{E}(d_1^Y d_{k+1}^Y) = (2a_Y - 1)^k$ .

**Sup2** It can be obtained by calculation,  $\det(\lambda \mathbf{I} - \mathbf{T}_X) = (\lambda - 1)(\lambda - (b_X + c_X - 2d_X))(\lambda - (b_X - c_X))$ , so the eigenvalues of  $\mathbf{T}_X$  are  $\lambda_1 = 1, \lambda_2 = b_X + c_X - 2d_X, \lambda_3 = b_X - c_X$ . The right eigenvector corresponding to  $\lambda_1$  is  $\mathbf{r}_1 = (1, 1, 1)^T$ , the left eigenvector is  $\boldsymbol{\ell}_1^T = \left(\frac{d_X}{1-b_X-c_X+2d_X}, \frac{1-b_X-c_X}{1-b_X-c_X+2d_X}, \frac{d_X}{1-b_X-c_X+2d_X}\right)$ . The right eigenvector corresponding to  $\lambda_2$  is  $\mathbf{r}_2 = \left(1, -\frac{2d_X}{1-b_X-c_X}, 1\right)^T$ , the left eigenvector is  $\boldsymbol{\ell}_2^T = \left(\frac{1-b_X-c_X}{2(1-b_X-c_X+2d_X)}, -\frac{1-b_X-c_X}{1-b_X-c_X+2d_X}, \frac{1-b_X-c_X}{2(1-b_X-c_X+2d_X)}\right)$ . The right eigenvector corresponding to  $\lambda_3$  is  $\mathbf{r}_3 = (1, 0, -1)^T$ , the left eigenvector is  $\boldsymbol{\ell}_3^T = \left(\frac{1}{2}, 0, -\frac{1}{2}\right)$ . Since the elements of the transition probability matrix are all non-negative values,  $0 < 1 - b_X - c_X < 1, 0 < 1 - 2d_X < 1$ , thus  $0 < 1 - b_X - c_X + 2d_X < 2$ .

It can be obtained by simple calculations, the right (left) eigenvectors of  $\mathbf{T}_X$  are linearly independent, and the algebraic multiplicity of each eigenvalue is the same as the geometric multiplicity, so  $\mathbf{T}_X$  is a simple matrix. It can be obtained from the Spectral decomposition theorem that

$$\begin{aligned}
T_X^k &= \lambda_1^k r_1 \ell_1^T + \lambda_2^k r_2 \ell_2^T + \lambda_3^k r_3 \ell_3^T \\
&= \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} \begin{pmatrix} \frac{d_X}{1-b_X-c_X+2d_X} & \frac{1-b_X-c_X}{1-b_X-c_X+2d_X} & \frac{d_X}{1-b_X-c_X+2d_X} \end{pmatrix} + (b_X + c_X - 2d_X)^k \begin{pmatrix} 1 \\ -\frac{2d_X}{1-b_X-c_X} \\ 1 \end{pmatrix} \times \\
&\quad \begin{pmatrix} \frac{1-b_X-c_X}{2(1-b_X-c_X+2d_X)} & -\frac{1-b_X-c_X}{1-b_X-c_X+2d_X} & \frac{1-b_X-c_X}{2(1-b_X-c_X+2d_X)} \end{pmatrix} + (b_X - c_X)^k \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix} \begin{pmatrix} \frac{1}{2} & 0 & -\frac{1}{2} \end{pmatrix} \\
&= \begin{pmatrix} T_{-1,-1}^{X,k} & T_{-1,0}^{X,k} & T_{-1,1}^{X,k} \\ T_{0,-1}^{X,k} & T_{0,0}^{X,k} & T_{0,1}^{X,k} \\ T_{1,-1}^{X,k} & T_{1,0}^{X,k} & T_{1,1}^{X,k} \end{pmatrix}. \tag{5}
\end{aligned}$$

Where

$$\begin{aligned}
T_{-1,-1}^{X,k} &= \frac{d_X}{1-b_X-c_X+2d_X} + \frac{(1-b_X-c_X)(b_X+c_X-2d_X)^k}{2(1-b_X-c_X+2d_X)} + \frac{1}{2}(b_X-c_X)^k, \\
T_{-1,0}^{X,k} &= \frac{1-b_X-c_X}{1-b_X-c_X+2d_X} - \frac{(1-b_X-c_X)(b_X+c_X-2d_X)^k}{1-b_X-c_X+2d_X}, \\
T_{-1,1}^{X,k} &= \frac{d_X}{1-b_X-c_X+2d_X} + \frac{(1-b_X-c_X)(b_X+c_X-2d_X)^k}{2(1-b_X-c_X+2d_X)} - \frac{1}{2}(b_X-c_X)^k, \\
T_{0,-1}^{X,k} &= \frac{d_X}{1-b_X-c_X+2d_X} - \frac{d_X(b_X+c_X-2d_X)^k}{1-b_X-c_X+2d_X}, \\
T_{0,0}^{X,k} &= \frac{1-b_X-c_X}{1-b_X-c_X+2d_X} + \frac{2d_X(b_X+c_X-2d_X)^k}{1-b_X-c_X+2d_X}, \\
T_{0,1}^{X,k} &= \frac{d_X}{1-b_X-c_X+2d_X} - \frac{d_X(b_X+c_X-2d_X)^k}{1-b_X-c_X+2d_X}, \\
T_{1,-1}^{X,k} &= \frac{d_X}{1-b_X-c_X+2d_X} + \frac{(1-b_X-c_X)(b_X+c_X-2d_X)^k}{2(1-b_X-c_X+2d_X)} - \frac{1}{2}(b_X-c_X)^k, \\
T_{1,0}^{X,k} &= \frac{1-b_X-c_X}{1-b_X-c_X+2d_X} - \frac{(1-b_X-c_X)(b_X+c_X-2d_X)^k}{1-b_X-c_X+2d_X}, \\
T_{1,1}^{X,k} &= \frac{d_X}{1-b_X-c_X+2d_X} + \frac{(1-b_X-c_X)(b_X+c_X-2d_X)^k}{2(1-b_X-c_X+2d_X)} + \frac{1}{2}(b_X-c_X)^k. \tag{6}
\end{aligned}$$

Let  $k \rightarrow \infty$ , the stationary distribution of  $d_i^X$  can be obtained that

$$\begin{aligned}
\varphi_X &= \left( \frac{d_X}{1-b_X-c_X+2d_X}, \frac{1-b_X-c_X}{1-b_X-c_X+2d_X}, \frac{d_X}{1-b_X-c_X+2d_X} \right) \\
&= (\varphi_{-1}^X, \varphi_0^X, \varphi_1^X). \tag{7}
\end{aligned}$$

So,

$$\begin{aligned}
P(d_1^X d_{k+1}^X = 1) &= P(d_{k+1}^X = 1 | d_1^X = 1) P(d_1^X = 1) \\
&\quad + P(d_{k+1}^X = -1 | d_1^X = -1) P(d_1^X = -1) \\
&= \varphi_1^X T_{1,1}^{X,k} + \varphi_{-1}^X T_{-1,-1}^{X,k}. \tag{8}
\end{aligned}$$

The same can be obtained,

$$\begin{aligned}
P(d_1^X d_{k+1}^X = -1) &= P(d_{k+1}^X = -1 | d_1^X = 1) P(d_1^X = 1) \\
&\quad + P(d_{k+1}^X = 1 | d_1^X = -1) P(d_1^X = -1) \\
&= \varphi_1^X T_{1,-1}^{X,k} + \varphi_{-1}^X T_{-1,1}^{X,k}. \tag{9}
\end{aligned}$$

Thus,

$$\mathbb{E}(d_1^X d_{k+1}^X) = \varphi_1^X T_{1,1}^{X,k} + \varphi_{-1}^X T_{-1,-1}^{X,k} - \varphi_1^X T_{1,-1}^{X,k} - \varphi_{-1}^X T_{-1,1}^{X,k}. \tag{10}$$

In addition, it can be obtained by the stable distribution that

$$\mathbb{E}((d_1^X)^2) = \varphi_{-1}^X + \varphi_1^X. \tag{11}$$

Similarly,

$$\begin{aligned}
\varphi_Y &= (\varphi_{-1}^Y, \varphi_0^Y, \varphi_1^Y) \\
&= \left( \frac{d_Y}{1 - b_Y - c_Y + 2d_Y}, \frac{1 - b_Y - c_Y}{1 - b_Y - c_Y + 2d_Y}, \frac{d_Y}{1 - b_Y - c_Y + 2d_Y} \right). \\
\mathbb{E}((d_1^Y)^2) &= \varphi_{-1}^Y + \varphi_1^Y. \\
\mathbb{E}(d_1^Y d_{k+1}^Y) &= \varphi_1^Y T_{1,1}^{Yk} + \varphi_{-1}^Y T_{-1,-1}^{Yk} - \varphi_1^Y T_{1,-1}^{Yk} - \varphi_{-1}^Y T_{-1,1}^{Yk}.
\end{aligned}$$