DTMC II

Dr Ahmad Khonsari ECE Dept.

The University of Tehran

Calculation of stationary distribution finite chain

• brute-force (when there is limiting distribution)

is simply to take the transition matrix to a high power and then extract out any row

Solving global balance: via eigendecomposition

Note that the equation $\pi \cdot P = (\lambda = 1)\pi$ implies that the row vector π (if π is a column vector use π^T as a row vector) is a left eigenvector of P with eigenvalue equal to 1 (Recall xA= λ x where x is a row vector is definition of a left eigenvector, as opposed to the more standard right eigenvector Ax= λ x)

• Solving global balance : via a system of linear equations solution

solve the system of linear equations $\pi \cdot P = \pi$, $\Sigma_i \pi_i = 1$ (assume π is row vector)

These equations are known as the global balance equations, and this approach is introduced in Discrete Markov Chains: Finding the Stationary Distribution via solution of the global balance equations

Calculation of limiting Distribution (Pⁿ): finite chaingenerating function method (Kobayashi P429) $\pi_j(n+1)$, π^T

• We have $p_j(n+1) = \sum_{i \in S} p_i(n) P_{ij}(n)$ $j \in S$

- Or in Matrix form $p^{T}(n+1) = p^{T}(n)P$ where $p^{T}(n)$ is a row vector of dimension |S|
- If MC is homogeneous we find $p^{\mathrm{T}}(n+1) = p^{\mathrm{T}}(n-1)PP = ... = p^{\mathrm{T}}(0)P^{\mathrm{n}}$
- let g(z) denote the generating function of the vector sequence { p(n); n = 0, 1, 2, ...}, i.e.,

$$\boldsymbol{g}(z) = \sum_{n=0}^{\infty} \boldsymbol{p}(n) z^n$$

$$\boldsymbol{p}^{\mathrm{T}}(n+1) = \boldsymbol{p}^{\mathrm{T}}(n)\boldsymbol{P} \qquad \text{for all } n=0,1,\dots \qquad \boldsymbol{g}(z) = \sum_{n=0}^{\infty} \boldsymbol{p}(n)z^n$$
$$\boldsymbol{\Sigma}^{\infty} = \boldsymbol{p}^{\mathrm{T}}(n+1) = {}^n \boldsymbol{\Sigma}^{\infty} = \boldsymbol{p}^{\mathrm{T}}(n) = {}^n \boldsymbol{P} = {}^n \boldsymbol{P}$$

$$\sum_{n=0}^{\infty} \boldsymbol{p}^{\mathrm{T}}(n+1) z^{n} = \sum_{n=0}^{\infty} \boldsymbol{p}^{\mathrm{T}}(n) z^{n} \boldsymbol{P} = \boldsymbol{g}^{\mathrm{T}}(z) \boldsymbol{P}$$

• We have $g^{\mathrm{T}}(z)\mathbf{P} = \sum_{n=0}^{\infty} \mathbf{p}^{\mathrm{T}}(n+1)z^{n} = z^{-1}\sum_{n=0}^{\infty} \mathbf{p}^{\mathrm{T}}(n+1)z^{n+1} = z^{-1}\sum_{n=0}^{\infty} \mathbf{p}^{\mathrm{T}}(n+1)z^{n+1} + z^{-1}\mathbf{p}^{\mathrm{T}}(0)z^{0} - z^{-1}\mathbf{p}^{\mathrm{T}}(0)z^{0}$ $= z^{-1}g^{\mathrm{T}}(z) - z^{-1}\mathbf{p}^{\mathrm{T}}(0)$

• And thus

• From

and

٠

$$\boldsymbol{g}^{\mathrm{T}}(z) = \boldsymbol{p}^{\mathrm{T}}(0)[\boldsymbol{I} - \boldsymbol{P}z]^{-1}$$

• where *I* is the $M \times M$ identity matrix, where M = |S|.

• Assume

$$\boldsymbol{P} = \begin{bmatrix} 0 & 1 & 0 \\ \frac{1}{4} & \frac{1}{4} & \frac{1}{2} \\ 0 & \frac{1}{2} & \frac{1}{2} \end{bmatrix}$$

• Then

$$P^{2} = \begin{bmatrix} 0 & 1 & 0 \\ \frac{1}{4} & \frac{1}{4} & \frac{1}{2} \\ 0 & \frac{1}{2} & \frac{1}{2} \end{bmatrix} \cdot \begin{bmatrix} 0 & 1 & 0 \\ \frac{1}{4} & \frac{1}{4} & \frac{1}{2} \\ 0 & \frac{1}{2} & \frac{1}{2} \end{bmatrix} = \begin{bmatrix} \frac{1}{4} & \frac{1}{4} & \frac{1}{2} \\ \frac{1}{8} & \frac{3}{8} & \frac{1}{2} \end{bmatrix} \cdot \begin{bmatrix} 0 & 1 & 0 \\ \frac{1}{4} & \frac{1}{4} & \frac{1}{2} \\ 0 & \frac{1}{2} & \frac{1}{2} \end{bmatrix} = \begin{bmatrix} \frac{1}{16} & \frac{9}{16} & \frac{3}{8} \\ \frac{9}{64} & \frac{25}{64} & \frac{15}{32} \\ \frac{3}{32} & \frac{15}{32} & \frac{7}{16} \end{bmatrix} \cdot \begin{bmatrix} 1 & \frac{9}{16} & \frac{4}{9} & \frac{4}{9} \\ \frac{1}{9} & \frac{4}{9} & \frac{4}{9} \\ \frac{1}{9} & \frac{4}{9} & \frac{4}{9} \end{bmatrix} \cdot \begin{bmatrix} 1 & \frac{9}{16} & \frac{1}{8} \\ \frac{9}{64} & \frac{25}{64} & \frac{15}{32} \\ \frac{3}{32} & \frac{15}{32} & \frac{7}{16} \end{bmatrix} \cdot \lim_{n \to \infty} P^{n} = P^{\infty} = \begin{bmatrix} \frac{1}{9} & \frac{4}{9} & \frac{4}{9} \\ \frac{1}{9} & \frac{4}{9} & \frac{4}{9} \\ \frac{1}{9} & \frac{4}{9} & \frac{4}{9} \end{bmatrix} \cdot \begin{bmatrix} 1 & \frac{9}{16} & \frac{1}{8} \\ \frac{9}{64} & \frac{25}{64} & \frac{15}{32} \\ \frac{3}{32} & \frac{15}{32} & \frac{7}{16} \end{bmatrix} \cdot \lim_{n \to \infty} P^{n} = P^{\infty} = \begin{bmatrix} \frac{1}{9} & \frac{4}{9} & \frac{4}{9} \\ \frac{1}{9} & \frac{4}{9} & \frac{4}{9} \end{bmatrix} \cdot \begin{bmatrix} 1 & \frac{4}{9} & \frac{4}{9} \\ \frac{1}{9} & \frac{4}{9} & \frac{4}{9} \end{bmatrix} \cdot \begin{bmatrix} 1 & \frac{4}{9} & \frac{4}{9} \\ \frac{1}{9} & \frac{4}{9} & \frac{4}{9} \end{bmatrix} = \begin{pmatrix} \frac{1}{9} & \frac{4}{9} & \frac{4}{9} \\ \frac{1}{9} & \frac{4}{9} & \frac{4}{9} \end{bmatrix}$$

5

• the inverse of a matrix **A** can be expressed as

$$A^{-1} = \frac{\operatorname{adj}(A)}{\operatorname{det}|A|},$$

- where adj (A) is the *adjugate* (called *classical adjoint*) matrix of A .
- Applying this result to

$$\boldsymbol{g}^{\mathrm{T}}(z) = \boldsymbol{p}^{\mathrm{T}}(0)[\boldsymbol{I} - \boldsymbol{P}z]^{-1}$$

• we have

$$[\boldsymbol{I} - \boldsymbol{P}\boldsymbol{z}]^{-1} = \frac{\boldsymbol{B}(\boldsymbol{z})}{\Delta(\boldsymbol{z})}$$



$$\mathbf{A} = egin{bmatrix} a_{11} & a_{12} & a_{13} \ a_{21} & a_{22} & a_{23} \ a_{31} & a_{32} & a_{33} \end{bmatrix} \qquad ext{adj}(\mathbf{A}) = \mathbf{C}^{\mathsf{T}} = egin{bmatrix} + & a_{22} & a_{23} \ a_{32} & a_{33} & a_{33} \ - & a_{32} & a_{33} \ - & a_{33} \ - & a_{33} & a_{33} \ - & a_{$$

$$egin{array}{c|c} a_{im} & a_{in} \ a_{jm} & a_{jn} \end{array} = \det egin{bmatrix} a_{im} & a_{in} \ a_{jm} & a_{jn} \end{bmatrix}$$

- Where $B(z) = adj (I - Pz) = \begin{bmatrix} 1 - \frac{3z}{4} - \frac{z^2}{8} & z\left(1 - \frac{z}{2}\right) & \frac{z^2}{2} \\ \frac{z}{4}\left(1 - \frac{z}{2}\right) & 1 - \frac{z}{2} & \frac{z}{2} \\ \frac{z^2}{8} & \frac{z}{2} & 1 - \frac{z}{4} - \frac{z^2}{4} \end{bmatrix}$
- And

$$\Delta(z) \triangleq \det[I - Pz] = (1 - z)(1 + \frac{z}{4} - \frac{z^2}{8})$$

• Hence $\begin{bmatrix} 1 - \frac{3z}{z} - \frac{z^2}{z} \\ z & z \end{bmatrix} = \begin{bmatrix} z^2 \\ z & z \end{bmatrix}$

$$[I - Pz]^{-1} = \frac{1}{\Delta(z)} \begin{bmatrix} 1 - \frac{3z}{4} - \frac{z}{8} & z\left(1 - \frac{z}{2}\right) & \frac{z}{2} \\ \frac{z}{4}\left(1 - \frac{z}{2}\right) & 1 - \frac{z}{2} & \frac{z}{2} \\ \frac{z^2}{8} & \frac{z}{2} & 1 - \frac{z}{4} - \frac{z^2}{4} \end{bmatrix}$$

Then by substituting the obtained expression and $p^{T}(0) = (p_0(0) \ p_1(0) \ p_2(0))$

• Into

$$\boldsymbol{g}^{\mathrm{T}}(z) = \boldsymbol{p}^{\mathrm{T}}(0)[\boldsymbol{I} - \boldsymbol{P}z]^{-1} = (p_0(0) \ p_1(0) \ p_2(0)) \ \frac{1}{\Delta(z)} \begin{bmatrix} 1 - \frac{3z}{4} - \frac{z^2}{8} & z\left(1 - \frac{z}{2}\right) & \frac{z^2}{2} \\ \frac{z}{4}\left(1 - \frac{z}{2}\right) & 1 - \frac{z}{2} & \frac{z}{2} \\ \frac{z^2}{8} & \frac{z}{2} & 1 - \frac{z}{4} - \frac{z^2}{4} \end{bmatrix}$$
• We obtain

 $\boldsymbol{g}^{\mathrm{T}}(z) = (g_0(z) \quad g_1(z) \quad g_2(z))$

where

$$g_{0}(z) = \frac{1}{\Delta(z)} \left[p_{0}(0) \left(1 - \frac{3z}{4} - \frac{z^{2}}{8} \right) + p_{1}(0) \frac{z}{4} \left(1 - \frac{z}{2} \right) + p_{2}(0) \frac{z^{2}}{2} \right]$$
$$g_{1}(z) = \frac{1}{\Delta(z)} \left[p_{0}(0) z \left(1 - \frac{z}{2} \right) + p_{1}(0) \left(1 - \frac{z}{2} \right) + p_{2}(0) \frac{z}{2} \right]$$
$$g_{2}(z) = \frac{1}{\Delta(z)} \left[p_{0}(0) \frac{z^{2}}{8} + p_{1}(0) \frac{z}{2} + p_{2}(0) \left(1 - \frac{z}{4} - \frac{z^{2}}{4} \right) \right]$$

9

- Inverting the PGF, we find $\{p_i(n); i = 0, 1, 2, n = 0, 1, 2, ...\}$.
- The limiting probabilities $\lim_{n\to\infty} p_i(n)$ can be obtained, however, without going through the inversion of the PGF. By applying the <u>final</u> <u>value theorem</u>, (see next 2 slides) we have

$$\lim_{n \to \infty} p_0(n) = \lim_{z \to 1} (1 - z) g_0(z) = \frac{1}{9}$$

$$\lim_{n \to \infty} p_1(n) = \lim_{z \to 1} (1-z)g_1(z) = \frac{4}{9}$$

$$\lim_{n \to \infty} p_2(n) = \lim_{z \to 1} (1-z)g_2(z) = \frac{4}{9}$$

- Problem 9.7 (Kobayashi P236):
- Generating function of a sequence. Let F(z) be the generating function of a sequence or vector { f_k ; k = 0, 1, 2, ...} defined by
- Find { f_k ; k = 0, 1, 2, ...} for the following F(z): $F(z) = \sum_{k=0}^{\infty} f_k z^k$

(a)
$$F(z) = \frac{1}{1 - \alpha z};$$

(b) $F(z) = \frac{1}{(1 - \alpha z)^2};$
(c) $F(z) = \frac{\alpha z}{(1 - \alpha z)^2};$

- Final value theorem. (Problem 9.12.)
- Refer to previous slide (problem 9.7) and show that

$$\lim_{z \to 1} (1-z)F(z) = \lim_{k \to \infty} f_k$$

- An alternative method to evaluate the state probability vector *p(n)* is to use the eigenvalues and eigenvectors of the TPM *P*. (spectral expansion method, since the set of eigenvalues of a matrix is also called its spectrum.)
- This method is similar to the generalized Fourier series expansion or Karhunen– Loève expansion method.
- In this case, however, **P** is not a symmetric matrix.

Calculation of limiting Distribution (Pⁿ): finite Chain-Spectral expansion method (Papoulis 4th ed page 706)

• Let λ_i be the *i*th **eigenvalue** and u_i be the associated **right-eigenvector** of the Markov TPM $Pu_i = \lambda_i u_i, i \in S = \{0, 1, 2, ..., M - 1\},\$

where M = |S| is the number of states and u_i is a column vector, making its transpose u_i^T a row vector.

- We assume that all eigenvalues are distinct; i.e., there is no multiplicity of any of the eigenvalues.
- Let us form an $M \times M$ matrix **U** by $U = [u_0 \ u_1 \ \cdots \ u_{M-1}]$ and a diagonal matrix

$$\Lambda = \begin{bmatrix} \lambda_0 & 0 & \dots & 0 \\ 0 & \lambda_1 & \dots & 0 \\ & \dots & & \\ 0 & 0 & \dots & \lambda_{M-1} \end{bmatrix}$$

• The TPM can be expanded as $P = U \land U^{-1} = U \land V$ Where $V = U^{-1}$

By multiplying **V** on the left of **P** = $U \land U^{-1} = U \land V$ we have **VP** = $\land V$

• Defining a set of row vectors
$$\boldsymbol{v}_i^T$$
 by $\mathbf{V} = \begin{bmatrix} \boldsymbol{v}_0^T \\ \boldsymbol{v}_1^T \\ \dots \\ \boldsymbol{v}_{M-1}^T \end{bmatrix}$

- we find from $VP = \Lambda V$ $v_i^\top P = \lambda_i v_i^\top, i \in S$.
- Therefore, \boldsymbol{v}_i^T is the **left-eigenvector** associated with the eigenvalue λ_i .

 We have VU = I, which implies that v_i and u_j are biorthonormal; i.e.,

$$\boldsymbol{v}_i^{\top}\boldsymbol{u}_j = \delta_{ij}, i, j \in \mathcal{S}.$$

- And $P^2 = U \land U^{-1} U \land U^{-1} = U \land {}^2 U^{-1}$
- By repeating the same procedure (n 1) times, we have

$$P^n = U\Lambda^n U^{-1} = \sum_{i \in \mathcal{S}} \lambda_i^n u_i v_i^{\top} = \sum_{i \in \mathcal{S}} \lambda_i^n E_i,$$

• where the matrices are the **projection matrices** $E_i = u_i v_i^{\top}, i \in S$,

By taking the transpose of VU = UV= I and expanding

$$I = (UV)^{\top} = \sum_{i \in S} v_i u_i^{\top} = \sum_{i \in S} E_i^{\top},$$

- We find $\sum_{i \in S} E_i = \sum_{i \in S} E_i^{\top} = I,$
- which corresponds to the case n = 0 in the expansion formula

$$P^n = U\Lambda^n U^{-1} = \sum_{i \in \mathcal{S}} \lambda_i^n u_i v_i^{\top} = \sum_{i \in \mathcal{S}} \lambda_i^n E_i,$$

 we may write the *n*-step transition probability from state *i* to state *j* as

$$P_{ij}^{(n)} = \sum_{k \in \mathcal{S}} \lambda_k^n u_{ki} v_{kj}, i, j \in \mathcal{S}, n = 0, 1, 2, \dots$$

- The state probability vector at step *n* is given from
- as

$$p^{\top}(n) = p^{\top}(0) \boldsymbol{P}^n,$$

$$p^{\top}(n) = p^{\top}(0)P^n = \sum_{k \in \mathcal{S}} \lambda_k^n p^{\top}(0)u_k v_k^{\top}.$$

• Thus, the probability that the Markov chain is in state *i* at time *n* is given by

$$p_i(n) = \sum_{k \in \mathcal{S}} \lambda_k^n \left(p^\top(0) u_k \right) v_{ki}, i \in \mathcal{S}.$$

• The marginal PMF $p(n) = (p_0(n) \ p_1(n) \dots \ p_i(n) \dots p_{|S|}(n))$

- Assume $P = \begin{bmatrix} 0 & 1 & 0 \\ \frac{1}{4} & \frac{1}{4} & \frac{1}{2} \\ 0 & \frac{1}{3} & \frac{1}{3} \end{bmatrix}.$
- The characteristic equation that determines the eigenvalues is given by
 - $\det |\boldsymbol{P} \lambda \boldsymbol{I}| = 0,$

• And rearranged as

$$\det |\boldsymbol{I} - \lambda^{-1}\boldsymbol{P}| = \Delta(\lambda^{-1}) = 0,$$

- where $\Delta(z)$ was defined earlier.
- Since we find that there are three roots for $\Delta(z) = 0$,
- $z_0 = 1$, $z_1 = -2$, and $z_2 = 4$, we readily find the three eigenvalues of
- **P**: $\lambda_0 = z_0^{-1} = 1$, $\lambda_1 = z_1^{-1} = -1/2$, $\lambda_2 = z_2^{-1} = 1/4$
- And the corresponding right-eigenvectors are

$$u_0 = \begin{bmatrix} 1\\1\\1 \end{bmatrix}, u_1 = \begin{bmatrix} 4\\-2\\1 \end{bmatrix}, \text{ and } u_2 = \begin{bmatrix} 4\\1\\-2 \end{bmatrix}.$$

• Thus we find

$$U = [u_0 u_1 u_2] = \begin{bmatrix} 1 & 4 & 4 \\ 1 & -2 & 1 \\ 1 & 1 & -2 \end{bmatrix}$$

• And its inverse

$$V = U^{-1} = \frac{1}{9} \begin{bmatrix} 1 & 4 & 4 \\ 1 & -2 & 1 \\ 1 & 1 & -2 \end{bmatrix} = \frac{1}{9} U,$$

• https://www.wolframalpha.com/calculators/eigenvalue-calculator

• From which we find

$$= U^{-1} = \frac{1}{9} \begin{bmatrix} 1 & 4 & 4 \\ 1 & -2 & 1 \\ 1 & 1 & -2 \end{bmatrix} = \frac{1}{9}U,$$

23

V

$$v_0^{\top} = \frac{1}{9}(1, 4, 4), v_1^{\top} = \frac{1}{9}(1, -2, 1), v_2^{\top} = \frac{1}{9}(1, 1, -2)$$

• The projection matrices are

$$u_{0} = \begin{bmatrix} 1\\1\\1\\1 \end{bmatrix}, u_{1} = \begin{bmatrix} 4\\-2\\1 \end{bmatrix}, \text{ and } u_{2} = \begin{bmatrix} 4\\1\\-2\\1 \end{bmatrix}.$$

$$E_{0} = u_{0}v_{0}^{\top} = \frac{1}{9}\begin{bmatrix} 1 & 4 & 4\\1 & 4 & 4\end{bmatrix},$$

$$\begin{bmatrix} 1\\1\\1\\1 \end{bmatrix} \begin{bmatrix} 1 & 4 & 4 \end{bmatrix}$$

$$\begin{bmatrix} 1\\1\\1\\1 \end{bmatrix} \begin{bmatrix} 1 & 4 & 4 \end{bmatrix}$$

$$E_{1} = u_{1}v_{1}^{\top} = \frac{1}{9}\begin{bmatrix} 4 & -8 & 4\\-2 & 4 & -2\\1 & -2 & 1 \end{bmatrix},$$

$$\begin{bmatrix} 4\\-2\\1\\1 \end{bmatrix} \begin{bmatrix} 1 & -2 & 1 \end{bmatrix}$$

$$\begin{bmatrix} 4\\1\\-2\end{bmatrix} \begin{bmatrix} 1 & 1 & -2 \end{bmatrix}$$

$$E_{2} = u_{2}v_{2}^{\top} = \frac{1}{9}\begin{bmatrix} 4 & 4 & -8\\1 & 1 & -2\\2 & 2 & -4 \end{bmatrix}.$$

• The n-step TPM is calculated as

 $P^n = \sum_{i=0}^{n} \lambda_i^n E_i$

 $\lambda_0 = z_0^{-1} = 1$, $\lambda_1 = z_1^{-1} = -1/2$, $\lambda_2 = z_2^{-1} = 1/4$

$$E_{0} = u_{0}v_{0}^{\top} = \frac{1}{9} \begin{bmatrix} 1 & 4 & 4 \\ 1 & 4 & 4 \\ 1 & 4 & 4 \end{bmatrix},$$

$$E_{1} = u_{1}v_{1}^{\top} = \frac{1}{9} \begin{bmatrix} 4 & -8 & 4 \\ -2 & 4 & -2 \\ 1 & -2 & 1 \end{bmatrix},$$

$$E_{2} = u_{2}v_{2}^{\top} = \frac{1}{9} \begin{bmatrix} 4 & 4 & -8 \\ 1 & 1 & -2 \\ 2 & 2 & -4 \end{bmatrix}.$$

$$E_{2} = u_{2}v_{2}^{\top} = \frac{1}{9} \begin{bmatrix} 4 & 4 & -8 \\ 1 & 1 & -2 \\ 2 & 2 & -4 \end{bmatrix}.$$

- Assume the MC is initially at state 1;
- i.e., **p**^T(0) = (1, 0, 0). Then,

$$P^{n} = \sum_{i=0}^{2} \lambda_{i}^{n} E_{i}$$

$$= \frac{1}{9} \begin{bmatrix} 1 & 4 & 4 \\ 1 & 4 & 4 \\ 1 & 4 & 4 \end{bmatrix} + \frac{1}{9} \left(-\frac{1}{2} \right)^{n} \begin{bmatrix} 4 & -8 & 4 \\ -2 & 4 & -2 \\ 1 & -2 & 1 \end{bmatrix} + \frac{1}{9} \left(\frac{1}{4} \right)^{n} \begin{bmatrix} 4 & 4 & -8 \\ 1 & 1 & -2 \\ 2 & 2 & -4 \end{bmatrix}.$$

$$\begin{split} p^{\top}(n) &= p^{\top}(0) P^n = \frac{1}{9} (1, 4, 4) + \frac{1}{9} \left(-\frac{1}{2} \right)^n (4, -8, 4) + \frac{1}{9} \left(\frac{1}{4} \right)^n (4, 4, -8) \\ &= (p_0^{(n)} p_1^{(n)} p_2^{(n)}), \end{split}$$

- Assume the MC is initially at state 1;
- i.e., **p**^T(0) = (1, 0, 0). Then,

$$p^{\top}(n) = p^{\top}(0)P^{n} = \frac{1}{9}(1, 4, 4) + \frac{1}{9}\left(-\frac{1}{2}\right)^{n}(4, -8, 4) + \frac{1}{9}\left(\frac{1}{4}\right)^{n}(4, 4, -8)$$
$$= (p_{0}^{(n)}p_{1}^{(n)}p_{2}^{(n)}),$$

• where

$$\begin{split} p_0^{(n)} &= \frac{1}{9} + \frac{4}{9} \left(-\frac{1}{2} \right)^n + \frac{4}{9} \left(\frac{1}{4} \right)^n, \\ p_1^{(n)} &= \frac{4}{9} - \frac{8}{9} \left(-\frac{1}{2} \right)^n + \frac{4}{9} \left(\frac{1}{4} \right)^n, \\ p_2^{(n)} &= \frac{4}{9} + \frac{4}{9} \left(-\frac{1}{2} \right)^n - \frac{8}{9} \left(\frac{1}{4} \right)^n. \end{split}$$

- We can also verify that $p_0^{(n)} + p_1^{(n)} + p_2^{(n)} = 1$ for all *n* as expected, because the *n*-step TPM **P**ⁿ is a stochastic matrix.
- We also find that, in the limit $_{n \to \infty}$, $p_0^{(n)} \to 1/9$, $p_1^{(n)} \to 4/9$, $p_2^{(n)} \to 4/9$
- These steady-state probabilities are independent of the initial probability **p**(0).
- The left eigenvector v_0 , associated with the eigenvalue $\lambda_0 = 1$, determines the steady-state solution.