

## 5.5

Let  $q = \begin{bmatrix} 0.6 \\ 0.1 \\ 0.3 \end{bmatrix}$  and  $A = \begin{bmatrix} 0.9 & 0 & 0.1 \\ 0 & 1 & 0 \\ 0.2 & 0 & 0.8 \end{bmatrix}$ . We define an HMM model  $\lambda = (q, A, B)$  where  $B$  defines arbitrary distributions for states 1, 2 and 3.

### 5.5.a

The state transition matrix is square, therefore the HMM is infinite. When the absorbing state number 2 is reached, the model keeps on generating samples therefore state 2 is not an exit state.

### 5.5.b

The eigenvalues of  $A^T$  are found by solving for  $x \in \mathbb{R}$ :  $\det(A^T - xI_3) = |A^T - xI_3| = 0$

$$\begin{aligned} |A^T - xI_3| = 0 &\Leftrightarrow \begin{vmatrix} 0.9-x & 0 & 0.2 \\ 0 & 1-x & 0 \\ 0.1 & 0 & 0.8-x \end{vmatrix} = 0 \\ &\Leftrightarrow (0.9-x)(1-x)(0.8-x) - (1-x)0.1 \cdot 0.2 = 0 \\ &\Leftrightarrow (1-x)[(0.9-x)(0.8-x) - 0.02] = 0 \\ &\Leftrightarrow (1-x)(0.7 - 1.7x + x^2) = 0 \quad (\text{note: } 1 \text{ is a root!}) \\ &\Leftrightarrow (1-x)(1-x)(0.7-x) = 0 \end{aligned}$$

Therefore the eigenvalues are 1 with algebraic multiplicity 2 and 0.7 with algebraic multiplicity 1.

The  $L_2$ -normalized eigenvectors associated with the eigenvalue 1 belong to the normalized eigenspace:  $E_1 = \{x \in \mathbb{R} \mid A^T x = x, \|x\|_2^2 = 1\}$ . Let us characterize  $E_1$ .

$$x = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} \in E_1 \Leftrightarrow \begin{cases} A^T x = x \\ \|x\|_2^2 = 1 \end{cases} \Leftrightarrow \begin{cases} 9x_1 + 2x_3 = 10x_1 \\ 10x_2 = 10x_2 \\ x_1 + 8x_3 = 10x_3 \\ x_1^2 + x_2^2 + x_3^2 = 1 \end{cases} \Leftrightarrow \begin{cases} x_1 = 2x_3 \\ x_2 = x_2 \\ x_1^2 + x_2^2 + x_3^2 = 1 \end{cases}$$

Since we have one degree of freedom in the system of equations above we can fix  $x_2 = 0$  for a vector  $p_1 \in E_1$  and  $x_2 = 1$  for a vector  $p_2 \in E_1$ .

Therefore we have  $p_1 = \begin{bmatrix} \frac{2}{\sqrt{5}} \\ 0 \\ \frac{1}{\sqrt{5}} \end{bmatrix}$  and  $p_2 = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}$ , two  $L_2$ -normalized eigenvectors associated with the eigenvalue 1.

Let us denote  $p_3$  a  $L_2$ -normalized eigenvector associated with the eigenvalue 0.7.  $p_3$  belongs to the normalized eigenspace:  $E_2 = \{x \in \mathbb{R} \mid A^T x = 0.7x, \|x\|_2^2 = 1\}$ . Let us characterize  $E_2$ .

$$x = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} \in E_2 \Leftrightarrow \begin{cases} A^T x = 0.7x \\ \|x\|_2^2 = 1 \end{cases} \Leftrightarrow \begin{cases} 9x_1 + 2x_3 = 7x_1 \\ 10x_2 = 7x_2 \\ x_1 + 8x_3 = 7x_3 \\ x_1^2 + x_2^2 + x_3^2 = 1 \end{cases} \Leftrightarrow \begin{cases} x_1 + x_3 = 0 \\ x_2 = 0 \\ x_1^2 + x_2^2 + x_3^2 = 1 \end{cases}$$

$$\Leftrightarrow \begin{cases} x_1 = \frac{1}{\sqrt{2}} \\ x_2 = 0 \\ x_3 = -\frac{1}{\sqrt{2}} \end{cases}$$

Therefore  $p_3 = \begin{bmatrix} \frac{1}{\sqrt{2}} \\ 0 \\ -\frac{1}{\sqrt{2}} \end{bmatrix}$  is the  $L_2$ -normalized eigenvector associated with the eigenvalue 0.7.

### 5.5.c

The HMM is stationary because  $A^T$  has eigenvalues equal to 1.

The stationary state probabilities are the linear combinations of  $(p_1, p_2)$  with positive components which sum to 1, *i.e* the vectors  $p_s \in \mathbb{R}^3 \quad \exists \alpha \in [0, 1] \quad p_s = \alpha p_1 + (1 - \alpha)p_2$ .

### 5.5.d

According to the transition matrix:

$$\forall t \in \mathbb{N}^+ \quad \forall n \in \mathbb{N}_*^+ \quad p[S_{t+n} = 1 \mid S_t = 2] = 0 \neq p[S_{t+n} = 1 \mid S_t = 1] > 0.$$

In words, the probability to reach state 1 at time  $t+n$  is different whether we are in state 1 or in state 2 at time  $t$ .

Therefore the HMM is not ergodic. Another argument is that the HMM will not converge towards one unique stationary state probability since we showed that there are infinitely many.