

# Discrete-Time Markov Chains



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# Outline

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- **What is a Markov chain?**
  - State space, Markov property
  - Stationary chains and transition matrix  $P$
- **Examples**
  - Random walk, Gambler's ruin, Ehrenfest model (and i.i.d. as a degenerate case)
- **Basic computations**
  - $n$ -step transition probabilities  $P^{(n)}$
  - Chapman–Kolmogorov equations and matrix powers  $P^{(n)} = P^n$
  - $\pi_n(j) = \mathbb{P}(X_n = j)$  and  $\pi_n = \pi_0 P^n$

# Definitions of DTMC

Consider a stochastic process  $\{X_n : n = 0, 1, 2, \dots\}$  taking values in a finite or countable set  $\mathcal{X}$ .

- $\mathcal{X}$  is called the **state space**
- If  $X_n = i$ ,  $i \in \mathcal{X}$ , we say the process is in state  $i$  at time  $n$
- Since  $\mathcal{X}$  is countable, we can label states by integers (e.g.  $\{0, 1, 2, \dots\}$ ,  $\mathbb{Z}$ , or  $\{0, \dots, n\}$  depending on the model).

## Definition

A stochastic process  $\{X_n : n = 0, 1, 2, \dots\}$  is called a **Markov chain** if it has the following property:

$$\begin{aligned} P(X_{n+1} = j | X_n = i, X_{n-1} = i_{n-1}, \dots, X_2 = i_2, X_1 = i_1, X_0 = i_0) \\ = P(X_{n+1} = j | X_n = i) \end{aligned}$$

for all states  $i_0, i_1, i_2, \dots, i_{n-1}, i, j \in \mathcal{X}$  and  $n \geq 0$ .

# Transition Probability Matrix

If  $P(X_{n+1} = j | X_n = i) = P_{ij}$  does not depend on  $n$ , then the process  $\{X_n\}$  is called a **stationary Markov chain**. From now on, we consider stationary Markov chains only.

$\{P_{ij}\}$  is called the **transition probabilities**.

$$\mathbb{P} = \begin{pmatrix} P_{00} & P_{01} & P_{02} & \cdots & P_{0j} & \cdots \\ P_{10} & P_{11} & P_{12} & \cdots & P_{1j} & \cdots \\ \vdots & \vdots & \vdots & \ddots & \vdots & \ddots \\ P_{i0} & P_{i1} & P_{i2} & \cdots & P_{ij} & \cdots \\ \vdots & \vdots & \vdots & \ddots & \vdots & \ddots \end{pmatrix}$$

is called the **transition probability matrix**.

Naturally, the transition probabilities  $\{P_{ij}\}$  satisfy:

- $P_{ij} \geq 0$  for all  $i, j$
- **Row sums are 1:**  $\sum_j P_{ij} = 1$  for all  $i$ .

In other words,  $\mathbb{P} \mathbf{1} = \mathbf{1}$ , where  $\mathbf{1} = (1, 1, \dots, 1, \dots)^T$ .

# Random Walk

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Consider the following random walk on integers

$$X_{n+1} = \begin{cases} X_n + 1 & \text{with prob } p \\ X_n - 1 & \text{with prob } 1 - p \end{cases}$$

This is a Markov chain because given  $X_n, X_{n-1}, X_{n-2}, \dots$ , the distribution of  $X_{n+1}$  depends only on  $X_n$  but not  $X_{n-1}, X_{n-2}, \dots$ .  
The state space is

$$\mathcal{X} = \{\dots, -3, -2, -1, 0, 1, 2, 3, \dots\} = \mathbb{Z} = \text{all integers}$$

The transition probability is

$$P_{ij} = \begin{cases} p & \text{if } j = i + 1 \\ 1 - p & \text{if } j = i - 1 \\ 0 & \text{otherwise} \end{cases}$$

# Gambler's Ruin

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In each round of a gambling game a player either wins \$1, with probability  $p$ , or loses \$1, with probability  $1 - p$ . The gambler starts with  $\$k$ . The game stops when the player either loses all their money, or gains a total of  $\$n$  ( $n > k$ ).

The gambler's successive fortunes form a Markov chain on  $\{0, 1, \dots, n\}$  with  $X_0 = k$  and transition matrix given by

$$P_{ij} = \begin{cases} p, & \text{if } j = i + 1, \quad 0 < i < n, \\ 1 - p, & \text{if } j = i - 1, \quad 0 < i < n, \\ 1, & \text{if } i = j = 0, \text{ or } i = j = n, \\ 0, & \text{otherwise.} \end{cases}$$

# Transition Matrix

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Here is the transition matrix with  $n = 6$  and  $p = \frac{1}{3}$ :

$$P = \begin{pmatrix} 1 & 0 & 0 & 0 & 0 & 0 & 0 \\ \frac{2}{3} & 0 & \frac{1}{3} & 0 & 0 & 0 & 0 \\ 0 & \frac{2}{3} & 0 & \frac{1}{3} & 0 & 0 & 0 \\ 0 & 0 & \frac{2}{3} & 0 & \frac{1}{3} & 0 & 0 \\ 0 & 0 & 0 & \frac{2}{3} & 0 & \frac{1}{3} & 0 \\ 0 & 0 & 0 & 0 & \frac{2}{3} & 0 & \frac{1}{3} \\ 0 & 0 & 0 & 0 & 0 & 0 & 1 \end{pmatrix}.$$

Gambler's ruin is an example of *simple random walk with absorbing boundaries*. Since  $P_{00} = P_{nn} = 1$ , when the chain reaches 0 or  $n$ , it stays there forever.

## Ehrenfest Diffusion Model

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Two containers  $A$  and  $B$ , containing a sum of  $K$  balls. At each stage, a ball is selected at random from the totality of  $K$  balls, and move to the other container. Let

$$X_0 = \# \text{ of balls in container } A \text{ in the beginning}$$

$$X_n = \# \text{ of balls in container } A \text{ after } n \text{ movements, } n = 1, 2, 3, \dots$$

$$\mathcal{X} = \{0, 1, 2, \dots, K\}$$

$$P_{ij} = \begin{cases} \frac{i}{K} & \text{if } j = i - 1 \\ \frac{K-i}{K} & \text{if } j = i + 1 \\ 0 & \text{otherwise} \end{cases}$$

## IID sequence

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An independent and identically distributed sequence of random variables is trivially a Markov chain. Assume that  $X_0, X_1, \dots$  is an i.i.d. sequence that takes values in  $\{1, \dots, k\}$  with

$$\mathbb{P}(X_n = j) = p_j, \quad \text{for } j = 1, \dots, k \text{ and } n \geq 0,$$

where

$$p_1 + \dots + p_k = 1.$$

By independence,

$$\mathbb{P}(X_1 = j \mid X_0 = i) = \mathbb{P}(X_1 = j) = p_j.$$

The transition matrix is

$$P = \begin{pmatrix} p_1 & p_2 & \cdots & p_k \\ p_1 & p_2 & \cdots & p_k \\ \vdots & \vdots & \ddots & \vdots \\ p_1 & p_2 & \cdots & p_k \end{pmatrix}.$$

# Joint Distribution of Random Variables in a Markov Chain

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Suppose  $\{X_n : n = 0, 1, 2, \dots\}$  is a stationary Markov chain with

- state space  $\mathcal{X}$  and
- transition probabilities  $\{P_{ij} : i, j \in \mathcal{X}\}$ .

Define  $\pi_0(i) = \mathbb{P}(X_0 = i)$ ,  $i \in \mathcal{X}$  to be the distribution of  $X_0$ .

What is the joint distribution of  $X_0, X_1, X_2$ ?

# Joint Distribution

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$$\begin{aligned}\mathbb{P}(X_0 = i_0, X_1 = i_1, X_2 = i_2) &= \mathbb{P}(X_0 = i_0)\mathbb{P}(X_1 = i_1 | X_0 = i_0)\mathbb{P}(X_2 = i_2 | X_1 = i_1, X_0 = i_0) \\ &= \mathbb{P}(X_0 = i_0)\mathbb{P}(X_1 = i_1 | X_0 = i_0)\mathbb{P}(X_2 = i_2 | X_1 = i_1) \quad (\text{Markov}) \\ &= \pi_0(i_0)P_{i_0i_1}P_{i_1i_2}\end{aligned}$$

In general,

$$\begin{aligned}\mathbb{P}(X_0 = i_0, X_1 = i_1, X_2 = i_2, \dots, X_{n-1} = i_{n-1}, X_n = i_n) &= \pi_0(i_0)P_{i_0i_1}P_{i_1i_2} \dots P_{i_{n-1}i_n}\end{aligned}$$

## **$n$ -Step Transition Probabilities**

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Suppose  $\{X_n\}$  is a stationary Markov chain with state space  $\mathcal{X}$ .  
Define the  $n$ -step transition probabilities

$$P_{ij}^{(n)} = \mathbb{P}(X_{n+k} = j \mid X_k = i) \quad \text{for } i, j \in \mathcal{X} \text{ and } n, k = 0, 1, 2, \dots$$

How to calculate  $P_{ij}^{(n)}$ ?

# Chapman-Kolmogorov Equations

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Suppose  $\{X_n\}$  is a stationary Markov chain with state space  $\mathcal{X}$ . Define the  $n$ -step transition probabilities

$$P_{ij}^{(n)} = \mathbb{P}(X_{n+k} = j | X_k = i) \quad \text{for } i, j \in \mathcal{X} \text{ and } n, k = 0, 1, 2, \dots$$

Then for all  $m, n \geq 1$ ,

$$P_{ij}^{(m+n)} = \sum_{k \in \mathcal{X}} P_{ik}^{(m)} P_{kj}^{(n)}$$

# Proof

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$$\begin{aligned} P_{ij}^{(m+n)} &= \mathbb{P}(X_{m+n} = j | X_0 = i) \\ &= \sum_{k \in \mathcal{X}} \mathbb{P}(X_{m+n} = j, X_m = k | X_0 = i) \\ &= \sum_{k \in \mathcal{X}} \mathbb{P}(X_m = k | X_0 = i) \mathbb{P}(X_{m+n} = j | X_m = k, X_0 = i) \\ &= \sum_{k \in \mathcal{X}} \mathbb{P}(X_m = k | X_0 = i) \mathbb{P}(X_{m+n} = j | X_m = k) \quad (\text{Markov}) \\ &= \sum_{k \in \mathcal{X}} P_{ik}^{(m)} P_{kj}^{(n)} \end{aligned}$$

# Chapman-Kolmogorov Equation in Matrix Notation

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For  $n = 1, 2, 3, \dots$ , let

$$P^{(n)} = \begin{pmatrix} P_{00}^{(n)} & P_{01}^{(n)} & P_{02}^{(n)} & \cdots & P_{0j}^{(n)} & \cdots \\ P_{10}^{(n)} & P_{11}^{(n)} & P_{12}^{(n)} & \cdots & P_{1j}^{(n)} & \cdots \\ \vdots & \vdots & \vdots & \ddots & \vdots & \ddots \\ P_{i0}^{(n)} & P_{i1}^{(n)} & P_{i2}^{(n)} & \cdots & P_{ij}^{(n)} & \cdots \\ \vdots & \vdots & \vdots & \ddots & \vdots & \ddots \end{pmatrix}$$

be the *n-step transition probability matrix*.

The Chapman-Kolmogorov equation just asserts that

$$P^{(m+n)} = P^{(m)} \times P^{(n)}$$

Note  $P^{(1)} = P$ ,  $\Rightarrow P^{(2)} = P^{(1)} \times P^{(1)} = P \times P = P^2$ .

By induction,

$$P^{(n)} = P^{(n-1)} \times P^{(1)} = P^{n-1} \times P = P^n$$

## Distribution of $X_n$

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Define  $\pi_n(i) = \mathbb{P}(X_n = i)$ ,  $i \in \mathcal{X}$  to be the marginal distribution of  $X_n$ ,  $n = 1, 2, \dots$ . Then again by the law of total probabilities,

$$\begin{aligned}\pi_n(j) &= \mathbb{P}(X_n = j) \\ &= \sum_{k \in \mathcal{X}} \mathbb{P}(X_0 = k) \mathbb{P}(X_n = j | X_0 = k) \\ &= \sum_{k \in \mathcal{X}} \pi_0(k) P_{kj}^{(n)}\end{aligned}\tag{1.1}$$

Suppose the state space  $\mathcal{X}$  is  $\{0, 1, 2, \dots\}$ .

If we write the marginal distribution of  $X_n$  as a row vector

$$\pi_n = (\pi_n(0), \pi_n(1), \pi_n(2), \dots),$$

then equation (1.1) is equivalent to

$$\pi_n = \pi_0 P^{(n)} = \pi_0 P^n$$

# Ehrenfest Model with 4 Balls

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$$\mathbb{P} = \begin{pmatrix} & 0 & 1 & 2 & 3 & 4 \\ 0 & 0 & 4/4 & 0 & 0 & 0 \\ 1 & 1/4 & 0 & 3/4 & 0 & 0 \\ 2 & 0 & 2/4 & 0 & 2/4 & 0 \\ 3 & 0 & 0 & 3/4 & 0 & 1/4 \\ 4 & 0 & 0 & 0 & 4/4 & 0 \end{pmatrix}$$

**Q1** Find  $P_{4,2}^{(4)} = \mathbb{P}(X_4 = 2 | X_0 = 4)$ .

**Q2** Find  $P_{4,2}^{(10)} = \mathbb{P}(X_{10} = 2 | X_0 = 4)$ .

**Q3** Given  $\mathbb{P}(X_0 = i) = 1/5$  for  $i = 0, 1, 2, 3, 4$ , find  $\mathbb{P}(X_4 = 2)$

**Q4** Find  $\mathbb{P}(X_{10} = 2, X_k \geq 2, \text{ for } 1 \leq k \leq 9 | X_0 = 4)$

## Example: Ehrenfest Model, 4 Balls (Cont'd)

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To find  $P_{4,2}^{(10)}$  for Q4, it's awful lots of work to compute  $P^{10} \dots$

There are ways to save some work. By the C-K equation,

$$P_{4,2}^{(10)} = \underbrace{P_{4,0}^{(5)} P_{0,2}^{(5)}}_{=0} + P_{4,1}^{(5)} P_{1,2}^{(5)} + \underbrace{P_{4,2}^{(5)} P_{2,2}^{(5)}}_{=0} + P_{4,3}^{(5)} P_{3,2}^{(5)} + \underbrace{P_{4,4}^{(5)} P_{4,2}^{(5)}}_{=0}$$

because it's impossible to move between even states in odd number of moves.

We just need to find  $P_{4,1}^{(5)}$ ,  $P_{4,3}^{(5)}$ ,  $P_{1,2}^{(5)}$ , and  $P_{3,2}^{(5)}$

## Example: Ehrenfest Model, 4 Balls (Cont'd)

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$$P^5 = P^2 \times P^3$$

$$= 2 \begin{pmatrix} 0 & 1 & 2 & 3 & 4 \\ 1/4 & 0 & 3/4 & 0 & 0 \\ 0 & 5/8 & 0 & 3/8 & 0 \\ 1/8 & 0 & 3/4 & 0 & 1/8 \\ 0 & 3/8 & 0 & 5/8 & 0 \\ 0 & 0 & 3/4 & 0 & 1/4 \end{pmatrix} \times 2 \begin{pmatrix} 0 & 1 & 2 & 3 & 4 \\ 0 & 5/8 & 0 & 3/8 & 0 \\ 5/32 & 0 & 3/4 & 0 & 3/32 \\ 0 & 1/2 & 0 & 1/2 & 0 \\ 3/32 & 0 & 3/4 & 0 & 5/32 \\ 0 & 3/8 & 0 & 5/8 & 0 \end{pmatrix}$$

$$= 2 \begin{pmatrix} 0 & 1 & 2 & 3 & 4 \\ 0 & 0 & 0 & 0 & 0 \\ 3/4 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \\ 0 & 15/32 & 0 & 17/32 & 0 \end{pmatrix}$$

$$P_{4,2}^{(10)} = P_{4,1}^{(5)} P_{1,2}^{(5)} + P_{4,3}^{(5)} P_{3,2}^{(5)} = \frac{15}{32} \times \frac{3}{4} + \frac{17}{32} \times \frac{3}{4} = \frac{3}{4}.$$

## Example: Ehrenfest Model, 4 Balls (Cont'd)

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$$\pi_0 = \left( \frac{1}{5}, \frac{1}{5}, \frac{1}{5}, \frac{1}{5}, \frac{1}{5} \right).$$

$$\pi_4 = \pi_0 P^4 = \left( \frac{1}{5}, \frac{1}{5}, \frac{1}{5}, \frac{1}{5}, \frac{1}{5} \right) \begin{pmatrix} 5/32 & 0 & 3/4 & 0 & 3/32 \\ 0 & 17/32 & 0 & 15/32 & 0 \\ 1/8 & 0 & 3/4 & 0 & 1/8 \\ 0 & 15/32 & 0 & 17/32 & 0 \\ 3/32 & 0 & 3/4 & 0 & 5/32 \end{pmatrix}$$

$$\pi_4(2) = \left( \frac{1}{5}, \frac{1}{5}, \frac{1}{5}, \frac{1}{5}, \frac{1}{5} \right) \begin{pmatrix} 3/4 \\ 0 \\ 3/4 \\ 0 \\ 3/4 \end{pmatrix}$$

$$= \frac{1}{5} \cdot \frac{3}{4} + \frac{1}{5} \cdot 0 + \frac{1}{5} \cdot \frac{3}{4} + \frac{1}{5} \cdot 0 + \frac{1}{5} \cdot \frac{3}{4} = \frac{9}{20}$$

## Example: Ehrenfest Model, 4 Balls (Cont'd)

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Q6: Find  $\mathbb{P}(X_{10} = 2, X_k \geq 2, \text{ for } 1 \leq k \leq 9 | X_0 = 4)$ .

**Tip:** Create another process  $\{W_n, n = 0, 1, 2, \dots\}$  with an absorbing state  $A$

$$W_n = \begin{cases} X_n & \text{if } X_k \geq 2 \text{ for all } k = 0, 1, 2, \dots, n \\ A & \text{if } X_k < 2 \text{ for some } k \leq n \end{cases}$$

What is the state space of  $\{W_n\}$ ?

Is  $\{W_n\}$  a Markov chain?

$$W_{n+1} = \begin{cases} A & \text{if } W_n = A \\ W_n + 1 & \text{with prob. } \frac{4-W_n}{4} \text{ if } W_n \neq A \\ W_n - 1 & \text{with prob. } \frac{W_n}{4} \text{ if } W_n = 3 \text{ or } 4 \\ A & \text{with prob. } \frac{W_n}{4} \text{ if } W_n = 2 \end{cases}$$

## Example: Ehrenfest Model, 4 Balls (Cont'd)

---

Q6: Find  $\mathbb{P}(X_{10} = 2, X_k \geq 2, \text{ for } 1 \leq k \leq 9 | X_0 = 4)$ .

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What is the state space of  $\{W_n\}$ ?  $\{A, 2, 3, 4\}$

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## Example: Ehrenfest Model, 4 Balls (Cont'd)

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Q6: Find  $\mathbb{P}(X_{10} = 2, X_k \geq 2, \text{ for } 1 \leq k \leq 9 | X_0 = 4)$ .

**Tip:** Create another process  $\{W_n, n = 0, 1, 2, \dots\}$  with an absorbing state  $A$

$$W_n = \begin{cases} X_n & \text{if } X_k \geq 2 \text{ for all } k = 0, 1, 2, \dots, n \\ A & \text{if } X_k < 2 \text{ for some } k \leq n \end{cases}$$

What is the state space of  $\{W_n\}$ ?  $\{A, 2, 3, 4\}$

Is  $\{W_n\}$  a Markov chain?

$$W_{n+1} = \begin{cases} A & \text{if } W_n = A \\ W_n + 1 & \text{with prob. } \frac{4-W_n}{4} \text{ if } W_n \neq A \\ W_n - 1 & \text{with prob. } \frac{W_n}{4} \text{ if } W_n = 3 \text{ or } 4 \\ A & \text{with prob. } \frac{W_n}{4} \text{ if } W_n = 2 \end{cases}$$

Yes,  $\{W_n\}$  is a Markov chain.

## Example: Ehrenfest Model, 4 Balls (Cont'd)

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What is the transition probability of  $\{W_n\}$ ?

$$P_W = \begin{matrix} & \begin{matrix} A & 2 & 3 & 4 \end{matrix} \\ \begin{matrix} A \\ 2 \\ 3 \\ 4 \end{matrix} & \begin{pmatrix} 1 & 0 & 0 & 0 \\ 2/4 & 0 & 2/4 & 0 \\ 0 & 3/4 & 0 & 1/4 \\ 0 & 0 & 1 & 0 \end{pmatrix} \end{matrix}$$

Observe that  $\mathbb{P}_{W,i,j}$  equals the transition prob. of the original process  $\mathbb{P}_{i,j}$  for  $i, j \neq A$ .

$$P = \begin{matrix} & \begin{matrix} 0 & 1 & 2 & 3 & 4 \end{matrix} \\ \begin{matrix} 0 \\ 1 \\ 2 \\ 3 \\ 4 \end{matrix} & \begin{pmatrix} 0 & 4/4 & 0 & 0 & 0 \\ 1/4 & 0 & 3/4 & 0 & 0 \\ 0 & 2/4 & 0 & 2/4 & 0 \\ 0 & 0 & 3/4 & 0 & 1/4 \\ 0 & 0 & 0 & 1 & 0 \end{pmatrix} \end{matrix}$$

## Example: Ehrenfest Model, 4 Balls (Cont'd)

How does  $\{W_n\}$  helps us to solve Q6?

Observe that  $\mathbb{P}(X_{10} = 2, X_k \geq 2, \text{ for } 1 \leq k \leq 9 | X_0 = 4)$

$$= \mathbb{P}(W_{10} = 2 | W_0 = 4) = P_{W,4,2}^{(10)}$$

It's still an awful lot of work to compute  $P_{W,4,2}^{(10)}$ .

By the same way we calculate  $P_{4,2}^{(10)}$ , using C-K equation, we know

$$\mathbb{P}_{W,4,2}^{(10)} = \mathbb{P}_{W,4,A}^{(5)} \underbrace{\mathbb{P}_{W,A,2}^{(5)}}_{=0} + \underbrace{\mathbb{P}_{W,4,2}^{(5)} \mathbb{P}_{W,2,2}^{(5)}}_{=0} + \mathbb{P}_{W,4,3}^{(5)} \mathbb{P}_{W,3,2}^{(5)} + \underbrace{\mathbb{P}_{W,4,4}^{(5)} \mathbb{P}_{W,4,2}^{(5)}}_{=0}$$

in which

- $\mathbb{P}_{W,A,2}^{(5)} = 0$  because  $\{W_n\}$  will never leave  $A$ .
- $\mathbb{P}_{W,4,2}^{(5)} = \mathbb{P}_{W,4,4}^{(5)} = 0$  because  $\{W_n\}$  can never get from 4 to an even numbered state in odd numbers of steps.

Just need to find  $\mathbb{P}_{W,4,3}^{(5)}$  and  $\mathbb{P}_{W,3,2}^{(5)}$ .

## Example: Ehrenfest Model, 4 Balls (Cont'd)

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$$P_W^{(2)} = \begin{matrix} & \begin{matrix} A & 2 & 3 & 4 \end{matrix} \\ \begin{matrix} A \\ 2 \\ 3 \\ 4 \end{matrix} & \begin{pmatrix} 1 & 0 & 0 & 0 \\ 1/2 & 3/8 & 0 & 1/8 \\ 3/8 & 0 & 5/8 & 0 \\ 0 & 3/4 & 0 & 1/4 \end{pmatrix}, \quad P_W^{(3)} = \begin{matrix} & \begin{matrix} A & 2 & 3 & 4 \end{matrix} \\ \begin{matrix} A \\ 2 \\ 3 \\ 4 \end{matrix} & \begin{pmatrix} 1 & 0 & 0 & 0 \\ 11/16 & 0 & 5/16 & 0 \\ 3/8 & 15/32 & 0 & 5/32 \\ 3/8 & 0 & 5/8 & 0 \end{pmatrix} \end{matrix}$$

$$P_W^{(5)} = P_W^{(2)} \times P_W^{(3)} = \begin{matrix} & \begin{matrix} A & 2 & 3 & 4 \end{matrix} \\ \begin{matrix} A \\ 2 \\ 3 \\ 4 \end{matrix} & \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 75/256 & & \\ 75/256 & 0 & 25/64 & \end{pmatrix} \end{matrix}$$

So

$$\mathbb{P}_{W,4,2}^{(10)} = \mathbb{P}_{W,4,3}^{(5)} \mathbb{P}_{W,3,2}^{(5)} = \frac{25}{64} \times \frac{75}{256} = \frac{1875}{16384}.$$