

1. Basics

GA Let  $(\Omega, \mathcal{F}, P)$  be a probability space,  $E = \{1, 2, \dots, l\}$ ,  $l \in \mathbb{N}$ , a finite set (or  $E$  a countable set), and  $(X_n)_{n \in \mathbb{N}_0}$  a sequence of random variables  $X_i: \Omega \rightarrow E$ ,  $i \in \mathbb{N}_0$ .

1.1 Def  $(X_n)_{n \in \mathbb{N}_0}$  is called discrete-time stochastic process with state space  $E$ .

If  $X_k = i$ ,  $i \in E$ ,  $k \in \mathbb{N}_0$ , the process is said to be in state  $i$  at time  $k$ .

For  $\omega \in \Omega$ , the  $E$ -valued sequence  $(X_0(\omega), X_1(\omega), X_2(\omega), \dots)$  is called realization (or trajectory or sample path) associated with  $\omega$ .

1.2 Remark

1. Recall the notion of distribution  $\mu: E \rightarrow \mathbb{R}$ ,  $i \mapsto P(X=i)$  for random variable  $X: \Omega \rightarrow E$ .

Notation:  $P(X=i) = P(\{X=i\}) = P(\{\omega \in \Omega \mid X(\omega)=i\})$

1.3 Def

A discrete-time stochastic process  $(X_n)_{n \in \mathbb{N}_0}$  is called Markov chain if the Markov property

$$(1) \quad P(X_n = i_n \mid X_{n-1} = i_{n-1}, \dots, X_0 = i_0) = P(X_n = i_n \mid X_{n-1} = i_{n-1})$$

holds f.a.  $n \in \mathbb{N}$ ,  $i_0, \dots, i_n \in E$  (assuming both sides of equ. (1) are defined).

If  $p_{ij}(n) := P(X_n = j \mid X_{n-1} = i)$  does not depend on  $n$ , then the MC is called (time) homogeneous.

1.4 Remark

The conditional probability  $P(A|B)$  is not defined, if  $P(B) = 0$ .

In the following consider statements containing cond. prob. only when defined, and only consider homogeneous MC.

1.5 Def

For a MC  $(X_n)$ , the matrix  $P = (p_{ij})_{i,j \in E} \in [0,1]^{l \times l}$  with  $p_{ij} = P(X_1 = j \mid X_0 = i)$  is called transition matrix of  $(X_n)$ .

A function  $\alpha: E \rightarrow [0,1]$  with  $\sum_{i=1}^l \alpha(i) = 1$  and  $\alpha(i) = P(X_0 = i)$  is called initial distribution of  $(X_n)$ .

### 1.6 Remark

$p_{ij}$  is the one-step transition probability from state  $i$  to state  $j$ .

It follows that  $p_{ij} \geq 0$  f.a.  $i, j \in E$  and  $\sum_{j=1}^l p_{ij} = 1$ , i.e.  $P$  is a stochastic matrix.

### 1.7 Theorem

The following statements are equivalent:

- (1)  $(X_n)$  is a MC,
- (2) if ex.  $P = (p_{ij})_{i, j \in E} \in [0, 1]^{l \times l}$  with  $P(X_n = i_n | X_{n-1} = i_{n-1}, \dots, X_0 = i_0) = p_{i_{n-1} i_n}$  f.a.  $n \in \mathbb{N}$ ,  $i_0, \dots, i_n \in E$ .
- (3) if ex.  $P = (p_{ij})_{i, j \in E} \in [0, 1]^{l \times l}$ ,  $\alpha \in [0, 1]^l$  with  $P(X_0 = i_0, \dots, X_n = i_n) = \alpha_{i_0} p_{i_0 i_1} \dots p_{i_{n-1} i_n}$  f.a.  $n \in \mathbb{N}_0$ ,  $i_0, \dots, i_n \in E$ .

### 1.8 Remarks

A MC can be defined by giving a state space, a corresponding stochastic matrix and an initial distribution.

If no initial distribution is specified, the transition matrix describes a family of MCs.

### 1.9 Ex

Describe repression of a gene:

- $E = \{0, 1\}$ ,  $X_n = 0$  - gene free at time  $n$
- $X_n = 1$  - gene repressed at time  $n$

- Assumptions:
- 1. gene free at time  $n$ , then gene repressed at time  $n+1$  with prob.  $p \geq 0$
  - 2. gene repressed at time  $n$ , then gene free at time  $n+1$  with prob.  $q \geq 0$

$$\rightarrow P = \begin{pmatrix} 1-p & p \\ q & 1-q \end{pmatrix}$$

1.10 Def Given the trans. matrix  $P \in [0, 1]^{l \times l}$  of a MC  $(X_n)$  with state space  $E$ , the directed labeled graph  $G$  with vertex set  $E$  and edges  $(i, j, P_{ij})$ ,  $i, j \in E$ ,  $P_{ij} \neq 0$  is called transition graph of  $(X_n)$ .

### 1.11 Ex

Random walk on  $E = \mathbb{N}$ .

choose  $p_k \in (0, 1)$  for all  $k \in \mathbb{N}$ .



## 2. Canonical representation

Goal: A representation of a MC amenable to simulation.

### 2.1 Theorem

Let  $(Z_k)_{k \in \mathbb{N}}$  be a sequence of independent and identically distributed (i.i.d.) random variables  $Z_i: \Omega \rightarrow D$  (with  $(D, \mathcal{D})$  some measurable space), and let  $X_0: \Omega \rightarrow E$  be a random variable independent of  $Z_1, Z_2, \dots$

Consider a fct  $f: E \times D \rightarrow E$ .

Then the recurrence equation

$$(*) \quad X_{k+1} = f(X_k, Z_{k+1})$$

defines a homogeneous MC  $(X_k)_{k \in \mathbb{N}_0}$  on state space  $E$ .

### 2.2 Remark

For the transition probabilities holds  $p_{ij} = P(X_1 = j | X_0 = i) = P(f(i, Z_1) = j)$

### 2.3 Ex Random walk on $\mathbb{Z}$

$X_0: \Omega \rightarrow \mathbb{Z}$  independent of i.i.d. random variables  $Z_1, Z_2, \dots: \Omega \rightarrow \{-1, 1\}$  with  $P(Z_k = 1) = q$ ,  $P(Z_k = -1) = 1 - q$  for some  $q \in (0, 1)$ .

MC  $(X_n)$  defined by  $X_{n+1} = f(X_n, Z_{n+1}) = X_n + Z_{n+1}$ ,  $f: \mathbb{Z} \times \{-1, 1\} \rightarrow \mathbb{Z}$

### 2.4 Remark

1. Given  $(X_n)$ ,  $P = (p_{ij})_{i, j \in E}$  one can always define

•  $(Z_k)$  sequence of i.i.d random variables uniformly distributed on  $[0, 1]$

•  $f: E \times [0, 1] \rightarrow E$  with  $f(i, z) = k$  if  $\sum_{j=1}^{k-1} p_{ij} < z \leq \sum_{j=1}^k p_{ij}$

(i.e.  $f$  not adopting value  $k$  if  $\sum_{j=1}^{k-1} p_{ij} = \sum_{j=1}^k p_{ij}$ , i.e.  $p_{ik} = 0$ ).

2. Simulation of  $(X_n)$ : obtain realization  $(x_0, x_1, x_2, \dots)$

• choose  $x_0 \in E$  randomly (acc. to distribution of  $X_0$ )

• choose  $y_1, y_2, \dots \in D$  according to  $Z_1$  (used representatively for all  $Z_k$ )

• define  $x_{k+1} = f(x_k, y_{k+1})$