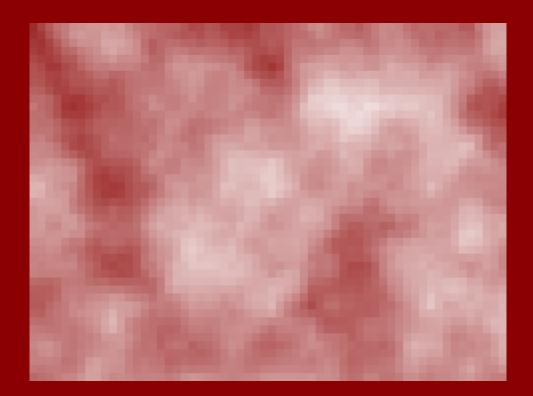
# Lecture notes on Markov Processes

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# Chapter 0: Manuel's notes

### Warning

These are unofficial lecture notes written by a student. They are messy, will almost surely contain errors, typos and misunderstandings and may not be kept up to date! I do however try my best and use these notes to prepare for my exams. Feel free to email me any corrections to mh@mssh.dev or s6mlhinz@uni-bonn.de. Happy learning!

## General Information

• Ecampus: Ecampus link

Basis: Basis linkWebsite: None

• Time slot(s): Tuesday 12-14 and Thursday 12-14

• Exams: Oral

• Exercises: To be handed in each Friday until noon

There are tutorials in the second week (14.10.24 and 16.10.24)!

Start of lecture 01 (08.10.2024)

# Chapter 1: Continuous time Markov chains

Two motivating examples:

# 1.1 Markov chain, transition function, infinitesimal description

We have a **countable state space** S and define

- Ω = {right-continuous functions ω : ℝ<sub>+</sub> →
   S with finitely many jumps in any finite time intervals}
- $\sigma$ -algebra  $\mathcal{F} := \sigma(\{\omega \to \omega(t) \text{ which is measurable } \forall t \geq 0\})$
- time shift operator:  $(\theta_s \omega)(t) := \omega(s+t)$

**Definition 1.1.** Denote  $X(t,\omega) := \omega(t)$ . Assume we have a collection of probability measures  $\{\mathbb{P}^x, x \in \mathcal{S}\}$  on  $(\Omega, \mathcal{F})$ , a right-continuous filtration  $\{\mathcal{F}_t, t \geq 0\}$  on  $(\Omega, \mathcal{F})$ . X is a continuous time Markov chain (MC) if

- (a) X is adapted to  $\{\mathcal{F}_t, t \geq 0\}$
- (b) Initial condition:  $\mathbb{P}^x(X(0) = x) = 1$
- (c) Markov property:  $\forall x \in \mathcal{S}, Y$  measurable function on  $\Omega, s \geq 0$ :

$$\mathbb{E}^x (Y \circ \theta_s \mid \mathcal{F}_s) = \mathbb{E}^{X(s)}(Y), \mathbb{P}^x \ a.s.$$

**Example.**  $Y = \max_{0 \le t \le 1} X(t) \to Y \circ \theta_s = \max_{s \le t \le s+1} X(t)$ 

**<u>Here:</u>** Time homogenous processes!

**Definition 1.2.** A transition function is function  $p_t(x,y), x,y \in \mathcal{S}, t \geq 0$  s.t.

- (a) **Positivity:**  $p_t(x,y) \ge 0$
- (b) Normalized:  $\sum_{y \in \mathcal{S}} p_t(x, y) = 1$
- (c) Continuity:  $\lim_{t\downarrow 0} p_t(x,x) = p_0(x,x) = 1$
- (d) Chapman-Kolmogorov equation:  $\forall s, t \geq 0 \ \forall x, y \in \S$ :

$$p_{s+t}(x,y) = \sum_{z \in \mathcal{S}} p_s(x,z) p_t(z,y)$$

Given a transition function, we can construct  $\mathbb{P}^x$  as follows:

Finite dimensional distributions,  $0 < t_1 < \cdots < t_n$ :

$$\mathbb{P}^{x}(X(t_{1}) = x_{1}, X(t_{2} = x_{2}), \dots, X(t_{n}) = x_{n}) \coloneqq p_{t_{1}}(x, x_{1}) \cdot p_{t_{2} - t_{1}}(x_{1}, x_{2}) \cdot \dots \cdot p_{t_{n} - t_{n-1}(x_{n-1}, x_{n})}$$

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measurable w.r.t.?

We will see while constructing, that this part is tricky, might just  $be \le 1$ 

extend to full time  $\mathbb{R}_+$  by the Kolmogorov(-Daniell) extension theorem, where the consistency,  $\mathbb{P}^x(X(t) = y) = p_t(x, y)$ , relations follow by the Chapman-Kolmogorov equation.

Example.  $0 < t_2 < \cdots < t_n$ 

$$\sum_{x_1 \in \mathcal{S}} \mathbb{P}^x \left( X(t_1) = x_1, \dots, X(t_n) = x_n \right) = \mathbb{P}^x \left( X(t_2) = x_2, \dots, X(t_n) = x_n \right)$$

We are still in the same setting, therefore we use the same  $t_1$  as previously

By modeling we often think at the basic biological/ physical properties of the system  $\rightarrow$  typically we have transition rates, because for  $x \neq y$ :  $\begin{cases} p_{\epsilon}(x,y) = O(\epsilon), & p_{\epsilon}(x,x) = 1 - O(\epsilon) \\ p_{0}(x,y) = 0, & p_{0}(x,x) = 1 \end{cases}$ 

**Definition 1.3.** For a Markov chain X we define the <u>transition rates</u> from x to y  $(x \neq y)$  by:

$$\tilde{q}(x,y) \coloneqq \frac{d}{dt} p_t(x,y) \mid_{t=0}$$

**Definition 1.4.** A Q-matrix (or generator) is a collection of numbers  $\{q(x,y): x,y \in \mathcal{S}\}$  s.t.:

- (a)  $q(x,y) \ge 0 \forall x \ne y$
- (b)  $\sum_{y \in \mathcal{S}} q(x, y) = 0$ , Notation:  $c(x) := q(x, x) = \sum_{y \in \mathcal{S} \setminus \{x\}} q(x, y)$

c(x) is the rate of leaving site x

#### Warning

It is not automatic that transition rates gives a Q-matrix For instance:  $\tilde{q}(x,y) = \infty$ 

For **finite** state space,

Markov chain  $\iff$  Transition function  $\iff$  Q-matrix Goal: Under which condition is the equivalence still true?

# 1.2 Examples

**Example** (Discrete to continuous MC). <u>Given:</u> Markov chain Y in discrete time  $t \in \mathbb{Z}_+$ , i.e. transition matrix  $P = (P(x, y))_{x,y \in \mathcal{S}}$ 

$$\mathbb{P}(Y(n+1) = y \mid Y(n) = x) = P(x, y)$$

Consider a Poisson process(PP) with intensity 1, at each event time of the PP, there will be a jump of the continuous time MC X. The jumps follow the discrete time MC Y  $\Longrightarrow p_t(x,y) = \sum_{n \geq 0} \frac{e^{-t}t^n}{n!} P^n(x,y)$  in this case Q = P - 1

**Example** (Finite S). 
$$p_t(x,y) = \left(e^{tQ}(x,y)\right) := \sum_{n>0} \frac{t^n}{n!} Q^n(x,y)$$

**Example** (Birth and death processes).  $S = \{0, 1, ..., \}, X(t) = Population size at time t. Then$ 

- $q(k, k+1) = \rho_k, k > 0$
- $q(k+1,k) = \lambda_k, q \ge 1$
- $q(k,k) = -\rho_k \lambda_k$  with  $\lambda_0 = 0$
- which implies  $q(k, l) = 0 \forall |k l| \ge 2$

Depending on the choice of  $\rho_k, \lambda_k$  it is possible that the chain goes to  $\infty$  in finite time.

Here will be one of the main problems: If Q is finite / a normed operator, everything is well defined. Otherwise we have to use a different definition of the exponetial . . .

Start of lecture 02 (10.10.2024)

# 1.3 From MC to Q-matrices

#### 1.3.1 MC $\rightarrow$ Transition functions

**Theorem 1.5.** Let X be a Markov chain. Then  $p_t(x,y) := \mathbb{P}^x(X(t) = y)$ .

- (a)  $p_t(x,y)$  is a transition function
- (b)  $p_t(x,y)$  determines uniquely  $\mathbb{P}^x$

*Proof.* (a): Positivity and normalization follows by the properties of  $\mathbb{P}^x$ .

 $\tau := \inf\{t \in \mathbb{R}_+ \mid X(t) \neq X(0)\}\$  is a.s. positive.  $p_t(x,x) \geq \mathbb{P}^x(\tau > t)$ , which implies

 $1 = p_0(x, x) \ge \lim_{t \downarrow 0} p_t(x, x) \ge \lim_{t \downarrow 0} \mathbb{P}^x(\tau > t) = 1.$ 

We still need to verify the Chapman-Kolmogorov equation: Define

 $Y = 1_{X(t)=y} \implies Y \circ \theta_s = 1_{X(t+s)=y}$ 

$$p_{s+t}(x,y) = \mathbb{P}^{x}(X(s+t) = y) = \mathbb{E}^{x} \left( 1_{X(s+t)=y} \right)$$

$$= \mathbb{E}^{x} \left( \mathbb{E}^{x} (1_{X(s+t)=y} \mid \mathcal{F}_{s}) \right)$$

$$\stackrel{\text{MP}}{=} \mathbb{E}^{x} \left( \mathbb{E}^{X(s) \left( 1_{X(t)=y} \right)} \right)$$

$$= \sum_{z \in S} p_{s}(x,z) \underbrace{\mathbb{E}^{z} \left( 1_{X(t)=y} \right)}_{=p_{t}(z,y)}$$

**(b):** By the Markov property  $\forall 0 < t_1 < \dots < t_n, x_1, \dots, x_n \in S$ 

$$\mathbb{P}^{x}\left(X(t_{1})=x_{1},\ldots X(t_{n})=x_{n}\right)=p_{t_{1}}(x_{1},x_{1})\cdot p_{t_{2}-t_{1}}(x_{1},x_{2})\cdot \cdots \cdot p_{t_{n}-t_{n-1}}(x_{n-1},x_{n})$$

using the Kolmogorov extension theorem  $\mathbb{P}^x$  is uniquely determined by  $p_t$ .

## 1.3.2 Transition functions to Q-matrices

**Proposition 1.6.** Let  $p_t(x,y)$  be a transition function.

- (a)  $\forall t \geq 0, x \in S, p_t(x, x) > 0$
- (b) If  $p_t(x,x) = 1$  for some t > 0, then  $p_s(x,x) = 1 \forall s > 0$
- (c)  $\forall x, y \in S, p_t(x, y)$  is uniformly continuous in t, due to

$$|p_{t+s}(x,y) - p_t(x,y)| \le 1 - p_s(x,x)$$

**Remark.** (a) differs from the discrete time markov chain  $\rightarrow$  no issues of periodicity

(b) The M.C. if it reaches site x, stays forever in x. x is an absorbing state

*Proof.* (a): Since  $\lim_{x\downarrow 0} p_t(x,x) = 1 \implies p_s(x,x) > 0$  for  $s \in [0,t]$  for some small t. Therefore

 $\forall s \in [0,t]: p_{t+s}(x,x) \overset{\text{C.Kolmogorov}}{\geq} p_t(x,x) p_s(x,x) > 0$ , which we can iterate to get the claim.

(b): 
$$p_{s+t}(x,x) = \sum_{z \in S} p_s(x,z) p_t(z,x) \le p_s(x,x) p_t(x,x) + \sum_{\substack{z \neq x \\ =1-p_s(x,x)}} p_s(x,z) \cdot 1$$
 Then

still good, since  $p_s(x, z)$  will be 0

$$p_{s+t}(x,x) \le 1 - \underbrace{p_s(x,x)}_{\substack{(a) \\ \ge 0}} (1 - p_t(x,x)).$$

If 
$$p_{s+t}(x,x) = 1 \implies p_t(x,x) = 1$$
.  
 $\forall v \in [t+s, 2(t+s)] : p_v(x,x) \ge p_{\underbrace{v - (t+s)}}(x,x)p_{t+s}(x,x) = 1 \cdot 1 = 1$ 

(c):

$$\begin{split} p_{s+t}(x,y) - p_t(x,y) &= \sum_{z \in S} p_s(x,z) p_t(s,y) - p_t(x,y) \\ &= \underbrace{(p_s(x,x) - 1) p_t(x,y)}_{\leq 0} + \underbrace{\sum_{z \neq x} p_s(x,z) p_t(z,y)}_{\geq 0} \end{split}$$

Since

$$|p_s(x,x) - 1|p_t(x,y) \le 1 - p_s(x,x)$$

and

$$\sum_{z \neq x} p_s(x, z) \underbrace{p_t(z, y)}_{\leq 1} \leq 1 - p_s(x, x)$$

$$\implies |p_{s+t}(x, y) - p_t(x, y)| \leq 1 - p_s(x, y)$$

**Theorem 1.7.** Let  $p_t(x,y)$  be a transition function.

(a)  $\forall x \in S$  the right derivative

$$\tilde{c}(x) = -\tilde{q}(x,x) = -\frac{d}{dt}p_t(x,x)\mid_{t=0} \in [0,\infty]$$

exists and  $p_t(x,x) \ge e^{-\tilde{c}(x)t}, \forall t > 0$ 

(b)  $I \ tildec < \infty \implies \forall y \neq x$ , the right-derivative

$$\tilde{q}(x,y) \coloneqq \frac{d}{dt} p_t(x,y) \mid_{t=0} \in [0,\infty]$$

exists and  $\sum_{y \in S} \tilde{q}(x, y) \leq 0$ 

(c) If for some  $x \in S$ ,  $\tilde{c}(x) < \infty$  and  $\sum_{y \in S} \tilde{q}(x, y) = 0$ , then  $p_t(x, y)$  is  $C^1$  in time for this x and  $y \in S$ . Moreover, it satisfies the **Kolmogorov-backwards equation**:

$$\frac{d}{dt}p_t(x,y) = \sum_{z \in S} \tilde{q}(x,z)p_t(z,y)$$
(1.1)

*Proof.* (a) Let  $f(t) := -\ln(p_t(x,x)) \ge 0$ . By  $p_{s+t}(x,x) \ge p_s(x,x)p_t(x,x)$  it follows

$$f(t+s) \le f(s) + f(t),$$

the function f is subadditive, which implies

In particular, the limit exists

$$\lim_{t \downarrow 0} \frac{f(t)}{t} = \sup_{t > 0} \frac{f(t)}{t} \in [0, \infty]$$

$$\implies \lim_{t \downarrow 0} \frac{f(t)}{t} = \frac{d}{dt} f(t) \mid_{t=0} = \frac{-\frac{d}{dt} p_t(x, x)}{\underbrace{p_t(x, x)}_{\to 1}} \mid_{t=0} = \tilde{c}(x)$$

$$\implies \tilde{c}(x) \ge \frac{f(t)}{t} = -\frac{\ln p_t(x, x)}{t} \to p_t(x, x) \ge e^{-\tilde{c}(x)t}$$

(b) by (a):  $1 - p_t(x, x) \le 1 - e^{-\tilde{c}(x)t} \le \tilde{c}(x)t$ , which implies

$$\sum_{y \neq x} p_t(x, y) \le \tilde{c}(x)t.$$

$$\implies \tilde{q}(x,y) := \limsup_{t \downarrow 0} \frac{p_t(x,y)}{t} \in [0,\tilde{c}(x)]$$

Let  $\delta > 0, n \in \mathbb{N}$ : Let  $p_{\delta} = \{p_{\delta}(x, y)\}_{x,y \in S}$  be a discrete time chain.

$$p_{n\delta}(x,y) \ge \sum_{k=0}^{n-1} \underbrace{p_{\delta}(x,x)^{k}}_{\ge p_{\delta}(x,x)^{n} \ge e^{-\tilde{c}(x)n\delta}} p_{\delta}(x,y) p_{(n-k-1)\delta}(y,y)$$

$$\implies \frac{p_{n\delta}(x,y)}{n\delta} \ge \frac{p_{\delta}(x,y)}{\delta} e^{-\tilde{c}(x)n\delta} \cdot \inf_{0 \le s \le n\delta} p_{s}(y,y)$$

Take a subsequence of  $\delta \downarrow 0$  such that  $n\delta \to t$  and  $\lim_{\delta \downarrow 0} \frac{p_{\delta}(x,y)}{\delta} = \tilde{q}(x,y)$ 

$$\implies \liminf_{t \downarrow 0} \frac{p_t(x,y)}{t} \ge \tilde{q}(x,y) \cdot 1 \cdot 1$$

which implies that the limit exists.

$$\sum_{y \neq x} \frac{p_t(x, y)}{t} \leq \tilde{c}(x)$$

$$\tilde{c}(x) \geq \liminf_{t \downarrow 0} \sum_{y \neq x} \frac{p_t(x, y)}{t} \stackrel{\text{Fatou}}{\geq} \sum_{y \neq x} \liminf_{t \downarrow 0} \frac{p_t(x, y)}{t} = \sum_{y \neq x} \tilde{q}(x, y)$$

$$\implies \sum_{y \neq x} \tilde{q}(x, y) \leq 0$$

(c) By Chapman-Kolmogorov:

$$\frac{p_{t+\epsilon}(x,y) - p_t(x,y)}{\epsilon} - \sum_{z} \tilde{q}(x,z) p_t(z,y)$$

$$= \sum_{z \in S} \underbrace{\left(\frac{p_{\epsilon}(x,z) - o_0(x,z)}{\epsilon} - \tilde{q}(x,z)\right)}_{=(\star)} p_t(z,y)$$

For all  $z: \lim_{\epsilon \to 0} (\star) = 0$ Take  $T \subset S$ ,  $|T| < \infty \implies \lim_{\epsilon \downarrow 0} \sum_{z \in T} (\star) \cdot p_t(z, y) = 0$ . Let  $x \in T$ .

$$\sum_{z \notin T} \left| \frac{p_{\epsilon}(x,z)}{\epsilon} - \tilde{q}(x,z) \right| \underbrace{p_{\epsilon}(z,y)}_{\leq 1}$$

$$\leq \sum_{z \notin T} \frac{p_{\epsilon}(x,z)}{\epsilon} + \sum_{z \notin T} \tilde{q}(x,z) \stackrel{\sum q(x,z) = 0}{=} \frac{1}{\epsilon} \left( 1 - \sum_{z \in T} p_{\epsilon}(x,z) \right) - \sum_{z \in T} \tilde{q}(x,z)$$

But 
$$\sum_{x} \tilde{q}(x,z) = 0 \stackrel{\epsilon \downarrow 0}{\rightarrow} -2 \sum_{z \in T} \tilde{q}(x,z) \stackrel{T \uparrow S}{\rightarrow} 0$$

Start of lecture 03 (15.10.2024)

**Remark.** For M.C. also the <u>strong Markov property</u> holds: Let  $\tau$  be a stopping time and  $Y : \mathbb{R}_+ \times \Omega \mapsto Y_t(\omega)^a$  measurable, then

$$\mathbb{E}^x(Y \circ \theta_\tau \mid \mathcal{F}_t) = \mathbb{E}^x(Y_\tau), \mathbb{P} - a.s.$$

on the set  $\{\tau < \infty\}$ .

<sup>&</sup>lt;sup>a</sup>shorthand notation ...

$$\frac{d}{dt}p_t(x,y) = \lim_{\epsilon \downarrow 0} \frac{p_{t+\epsilon}(x,y) - p_t(x,y)}{\epsilon}$$

using  $\sum_{z} p_t(x, z) p_{\epsilon}(z, y)$  instead of  $\sum_{z} p_{\epsilon}(x, z) p_t(z, y)$  we get the **Kolmogorov Forward Equation** 

$$\frac{d}{dt}p_t(x,y) = \sum_{z \in S} p_t(x,z)q(z,y)$$
(1.2)

They are almost always equivalent. In computation one often uses the forward equation, in construction the backwards equation is preffered

## 1.4 From Q-matrix to the Markov Chains

Let  $Q = (q(x, y))_{x,y \in \mathcal{S}}$  be a Q-matrix.

## 1.4.1 The backwards equation

**Proposition 1.8.** Let  $p_t(x,y)$  be a uniformly bounded function of x,y,t, then (a) is equivalent to (b):

- (a)  $p_t(x,y)$  is  $C^1$  in t, satisfies the KBE<sup>a</sup>, with initial condition  $p_0(x,y) = \delta_{x,y}$
- (b)  $p_t(x,y)$  is  $C^0$  in t, and satisfies:

$$p_t(x,y) = \delta_{x,y}e^{-c(x)t} + \int_0^t ds e^{-c(x)(t-s)} \sum_{z \in S \setminus \{x\}} q(x,z)p_s(z,y)$$
 (1.3)

 $^a$ Kolmogorov Backwards Equation

*Proof.* (a)  $\Longrightarrow$  (b): KBE  $\Longleftrightarrow$ 

$$\frac{d}{dt}p_t(x,y) = -c(x)p_t(x,y) + \sum_{z \in S \setminus \{x\}} q(x,z)p_t(z,y)$$

$$\implies \frac{d}{dt} \left( e^{c(x)t}p_t(x,y) \right) = c(x)e^{c(x)t}p_t(x,y) + e^{c(x)t} \left( -c(x)p_t(x,y) + \sum_{z \neq x} q(x,z)p_t(z,y) \right)$$

$$= e^{c(x)t} \sum_{z \neq x} q(x,z)p_t(z,y)$$

$$\stackrel{\int_0^t}{=} e^{c(x)t}p_t(x,y) - p_0(x,y) = \int_0^t e^{c(x)s} \sum_{z \neq x} q(x,z)p_s(z,y)$$

(b)  $\Longrightarrow$  (a): The RHS of the equation is  $C^1$  in t. The initial condition is also satisfied. Taking the derivate yields the  $\frac{d}{dt}p_t(x,y) = -c(x)p_t(x,y) + \sum_{z \in S \setminus \{x\}} q(x,z)p_t(z,y)$  which is equivalent to the KBE.

We use proposition 1.8 to show existence of positive solution of the KBE. Idea: Use a **fixed point argument**:

- $p_t^{(0)}(x,y) = 0 \forall x, y \in \mathcal{S}, t \ge 0$
- For  $n \ge 0$ :  $p_t^{(n+1)}(x,y) = \delta_{x,y}e^{c(x)t} + \int_0^t ds e^{-c(x)(t-s)} \sum_{z \ne x} q(x,z)p_s^{(n)}(z,y)$

Lemma 1.9.  $\forall n \geq 0$ :

- (a)  $p_t^{(n)}(x,y) \ge 0 \forall x, y \in \mathcal{S}$
- (b)  $\sum_{y \in \mathcal{S}} p_t^{(n)}(x, y) \le 1 \forall x \in \mathcal{S}, t \ge 0$
- (c)  $p_t^{(n+1)}(x,y) p_t^{(n)}(x,y) \ge 0 \forall x, y \in \mathcal{S}, t \ge 0$

*Proof.* By induction. n = 0 is clearly satisfied for (a),(b),(c).

(a) is obvious, since RHS are all positive terms.

(b)

$$\begin{split} \sum_{y \in \mathcal{S}} p_t^{(n+1)}(x,y) &= e^{-c(x)t} + \int_0^t ds e^{-c(x)(t-s)} \sum_{z \neq x} q(x,z) \underbrace{\sum_{y \in \mathcal{S}} p_s^{(n)}(z,y)}_{\leq 1} \\ &\leq e^{-c(x)t} + \int_0^t ds e^{-c(x)(t-s)} \underbrace{\sum_{z \neq x} q(x,z)}_{=c(x)} \\ &= e^{-c(x)t} + e^{-c(x)t} c(x) \int_0^t ds e^{c(x)s} &= 1 \end{split}$$

(c)

$$p_t^{(n+2)}(x,y) - p_t^{(n+1)}(x,y) = \int_0^t e^{-c(x)(t-s)} \sum_{z \neq x} q(x,z) \underbrace{\left(p_s^{(n+1)}(z,y) - p_s^{(n)}(z,y)\right)}_{\geq 0} \geq 0 \qquad \Box$$

- $\forall x, y, t \text{ we have } 0 \leq p_t^0(x, y) \leq p_t^1(x, y) \leq \dots \leq 1.$
- there exists a limit, which we denote by  $p_t^*(x,y)$

 $:= \lim_{n \to \infty} p_t^n(x, y)$  is Definition 1.10. Thelimit $p_t^*(x,y)$ calledtheminimal solution of the KBE

**Question:** Is  $p^*$  a transition function?

**Theorem 1.11.**  $p^*$  satisfies:

- (a)  $p_t^*(x,y) \ge 0$
- (b)  $\sum_{y \in S} p_t^*(x, y) \le 1$
- (c) The Chapman-Kolmogorov equation
- (d) satisfies (1.3) and thus the KBE

*Proof.* (a) and (b) follow directly from lemma 1.9.

(d): Take the limits of  $p_t^{(n)}(x,y)$  and use monotone convergence. (c): Define  $\Delta_t^{(n)}:=p_t^{(n+1)}(x,y)-p_t^{(n)}(x,y)\geq 0$ .

Notice  $p_t^*(x,y) = \sum_{n\geq 0} \Delta_t^{(n)}(x,y)$  In lemma 1.12:  $Delta_{t+s}^{(n)}(x,y) = \sum_{z\in\mathcal{S}} \sum_{k=0}^n \Delta_s^{(k)}(x,z) \Delta_t^{(n-k)}(z,y) \text{ which then implies}$ 

$$\begin{split} p_{t+s}^*(x,y) &= \sum_{n \geq 0} \Delta_{t+s}^{(n)}(x,y) = \sum_{z \in \mathcal{S}} \sum_{n \geq 0} \sum_{k=0}^n \Delta_s^{(k)}(x,z) \Delta_t^{(n-k)}(z,y) \\ &= \sum_{z \in \mathcal{S}} \underbrace{\sum_{k \geq 0} \Delta_s^{(k)}(x,z)}_{p_s^*(x,z)} \underbrace{\sum_{n=k}^\infty \Delta_t^{(n-k)}(z,y)}_{p_t^*(z,y)} \\ &= \sum_{z \in \mathcal{S}} p_s^*(x,z) p_t^*(z,y) \end{split}$$

using monotone convergence.

**Lemma 1.12.** 
$$\Delta_{t+s}^{(n)}(x,y) = \sum_{z \in \mathcal{S}} \sum_{k=0}^{n} \Delta_{s}^{(k)}(x,z) \Delta_{t}^{(n-k)}(z,y)$$

*Proof.* Consider the Laplace transform of the equation w.r.t. s, t, since the functions are all positive!

$$\int_0^\infty ds \int_0^\infty dt e^{-\lambda s} e^{-\mu t \Delta_{s+t}^{(n)}} \stackrel{?}{=} \sum_{z \in \mathcal{S}} \sum_{k=0}^n \underbrace{\left[\int_0^\infty ds e^{-\lambda s} \Delta_s^{(k)}(x,z)\right]}_{\Psi_{k,\lambda}(x,z)} \underbrace{\left[\int_0^t e^{-\mu t} \Delta_t^{(n-k)}(z,y)\right]}_{\Psi_{n-k,\mu}(z,y)}$$

Define  $\int_0^t ds e^{-\lambda s} \Delta_s^{(n)}(x,y) = \Psi_{n,\lambda}(x,y)$ . Then for the RHS:

$$= \sum_{z \in \mathcal{S}} \sum_{k=0}^{n} \Psi_{k,\lambda}(x,z) \Psi_{n-k} \mu(z,y)$$

For the LHS:

$$\int_{0}^{\infty} ds ds \int_{0}^{\infty} dt e^{-\lambda(s+t)} e^{-t(\mu-\lambda)} \Delta \underbrace{\sum_{s+t}^{(n)} (x,y)}_{=:u} = \int_{0}^{\infty} ds \int_{s}^{\infty} du e^{-\lambda u} e^{-\mu(u-s)} \Delta_{u}^{(n)}(x,y)$$

$$= \int_{0}^{\infty} du \int_{u}^{\infty} ds e^{-(\lambda-\mu)s} e^{-\mu u} \Delta_{u}^{(n)}(x,y) = \frac{\Psi_{n,\mu}(x,y) - \Psi_{n,\lambda}(x,y)}{\lambda - \mu}$$

$$\stackrel{?}{=} \sum_{k=0}^{n} \sum_{z \in \mathcal{S}} \Psi_{k,\lambda}(x,y) \Psi_{n-k,\mu}(z,y)$$

Another identity:  $\Psi_{n+1,\lambda}(x,y) = \sum_{z \neq x} \frac{q(x,z)}{\lambda + c(x)} \Psi_{n,\lambda}(z,y)$ Define the matrix  $A_{\lambda}(x,y) \coloneqq \frac{q(x,y)}{\lambda + c(x)} \mathbf{1}_{z \neq x}$ , then

 $\Psi_{n,\lambda} = (A_{\lambda})^n \Psi_{0,\lambda}$ 

where  $\Psi_{0,\lambda}(x,y) = \int_0^\infty ds e^{-\lambda s} p_s^{(1)}(x,y)$  and  $p_s^{(1)}(x,y) = \delta_{x,y} e^{-c(x)s}$ . Then  $\Psi_{0,\lambda}(x,y) = \delta_{x,y} \frac{1}{\lambda + c(x)}$ .

$$\begin{split} \frac{\Psi_{n,\mu} - \Psi_{n,\lambda}}{\lambda - \mu} &= \frac{(A_{\mu})^n \Psi_{0,\mu} - (A_{\lambda})^n \Psi_{0,\lambda}}{\lambda - \mu} \\ &= \frac{(A_{\lambda})^n (\Psi_{0,\mu} - \Psi_{0,\lambda})}{\lambda - \mu} \sum_{k=0}^{n-1} (A_{\lambda})^k \frac{A_{\mu} - A_{\lambda}}{\lambda - \mu} A^{n-k-1} \Psi_{0,\mu} \end{split}$$

 $\frac{\Psi_{0,\mu}-\Psi_0,\lambda}{\lambda-\mu}(x,y) = \frac{\delta_{x,y}}{\lambda-\mu}\left(\frac{1}{c(x)+\mu} - \frac{1}{c(x)+\lambda}\right) = \Psi_{0,\mu}(x,y) \cdot \Psi_{0,\lambda}(x,y).$ 

Similarly

$$\frac{A_{\mu} - A_{\lambda}}{\lambda - \mu}(x, y) = (\Psi_{0, \lambda} A_{\mu})(x, y)$$

Then

$$\frac{(A_{\lambda})^{n}(\Psi_{0,\mu} - \Psi_{0,\lambda})}{\lambda - \mu} \sum_{k=0}^{n-1} (A_{\lambda})^{k} \frac{A_{\mu} - A_{\lambda}}{\lambda - \mu} A^{n-k-1} \Psi_{0,\mu} = [(A_{\lambda})^{n} \Psi_{0,\mu}](x,y) + \sum_{k=0}^{n-1} \sum_{z} \underbrace{\left[A_{\lambda}^{k} \Psi_{0,\lambda}\right]}_{=\Psi_{k,\lambda}(x,z)} (x,z) \underbrace{\left(A_{\mu}^{n-k} \Psi_{0,\mu}\right)(z,y)}_{=\Psi_{n-k,\mu}(z,y)} \\
= \sum_{k=0}^{n} \sum_{\varepsilon,S} \Psi_{k,\lambda}(x,z) \Psi_{n-k,\mu}(z,y) \qquad \Box$$

Start of lecture 04 (17.10.2024)

**Theorem 1.13.** (a) If  $p_t(x,y)$  is a non-negative solution of the KBE 1.1 with  $p_0(x,y) = \delta_{x,y} \implies p_t(x,y) \ge p_t^*(x,y)$ 

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Above the second equation is wrong, but the skip from the RHS to the 3rd line is correct

Reminder: For matrices  $X^n - Y^n = (X - Y)X^{n-1} + Y(X - Y)X^{n-1} + \cdots + Y^{n-1}(X - Y)$ , since they

don't commute

(b) If  $\sum_{y \in \mathcal{S}} p_t^*(x,y) = 1, \forall x \in \mathcal{S} \implies p_t^*(x,y)$  is the unique transition function satisfying the

*Proof.* (a): By induction:  $p_t(x,y) \ge p_t^{(n)}(x,y) \forall n$ .

For n = 0  $p_t(x, y) \ge 0 = p_t^{(0)}(x, y)$ .

$$p_t^{(n+1)}(x,y) = \delta_{x,y}e^{-c(x)t} + \int_0^t ds e^{-c(x)(t-s)} \sum_{z \neq x} q(x,z) \underbrace{p_s^{(n)}(x,y)}_{\leq p_s(x,y)}$$

$$p_t \text{satisfies KBE} \atop \leq p_t(x,y)$$

**(b)**:

$$= \sum_{y \in \mathcal{S}} \underbrace{(p_t^*(x,y) - p_t^*(x,y))}_{\geq 0} \geq 0$$

$$= \sum_{y \in \mathcal{S}} p_t(x,y) - \sum_{y \in \mathcal{S}} p_t^*(x,y)$$

$$= 1$$

$$\implies 0 \geq \sum_{y \in \mathcal{S}} (p_t^*(x,y) - p_t^*(x,y)) \geq 0$$

$$\implies p_t(x,y) = p_t^*(x,y)$$

Example 1.14.  $S = \{0, 1, ..., \}, q(i, j) = \begin{cases} \beta_i & j = i+1 \\ -\beta_i & j = i \\ 0 & otherwise \end{cases}$  for some positive  $\beta_i$ . One shows

 $\sum_{j \in \mathcal{S}} p_t^*(i,j) = 1 \forall t, i \in \mathcal{S} \iff \sum_{k \geq 0} \frac{1}{\beta_k} < \infty.$  What happens if  $\sum_{k \geq 0} \frac{1}{\beta_i} = \infty$ ? Consider N(t) = # of jumps before time t, then  $\mathbb{P}(n(t) = \infty) > 0$ .

#### The probabilistic construction 1.4.2

**Lemma 1.15.** Let X(t) be a continuous time M.C. and  $\tau := \inf\{t : X(t) \neq X(0)\}$ . Then there exists  $c(x) = [0, \infty]$  s.t.  $\mathbb{P}(\tau > t) = e^{-c(x)t}$ 

*Proof.* Use Markov property with  $Y = 1_{X(r)=x,0 \le r \le t-s}$ :

$$\mathbb{P}^{x}\left(X(r) = x, s \leq r \leq t \mid \mathcal{F}_{s}\right) = \mathbb{P}^{X(s)}(\tau > t - s) = (\star)$$

$$\mathbb{E}^{x}(1_{\tau > s} \cdot \mathbb{P}^{x}(X(r) = x, s \leq r \leq t \mid \mathcal{F}_{s})) = \mathbb{P}^{x}(\tau > t)$$

$$\mathbb{E}^{x}(1_{\tau > s} \cdot \mathbb{P}^{x}(X(r) = x, s \leq r \leq t \mid \mathcal{F}_{s})) \stackrel{(\star)}{=} \mathbb{E}^{x}\left(1_{\tau > s}\mathbb{P}^{X(s)}(\tau > t - s)\right)$$

$$= \mathbb{P}^{x}(\tau > s)\mathbb{P}^{x}(\tau > t - s)$$

$$\Longrightarrow \mathbb{P}^{x}(\tau > t) = \mathbb{P}^{x}(\tau > s)\mathbb{P}(\tau > t - s)$$

Either  $\mathbb{P}^x(\tau > t) = 0 \forall t \implies c(x) = \infty$  (jump before t) or  $\mathbb{P}^x(\tau > t) > 0 \forall t$ . Define  $f(t) := -\ln \mathbb{P}^x(\tau > t)$  with  $f(0) = 0 \implies f(t) \ge 0$ . and  $f(t) = f(s) + f(t-s) \implies f(t) = k \cdot t.$ 

**Idea:** Divide the waiting times (which are exponetially distributed  $\sim \exp(c(x)), \exp(c(y)), \ldots$ ) and the **jumps**. The jumps are given by a discrete time M.C..

(1) Discrete time Markov chain  $\{Z_n, n \geq 0\}$  with transition matrix:

$$P = (p(x, y))_{x,y \in \mathcal{S}}$$

If c(x) = 0:

$$p(x,y) = \begin{cases} 1 & x = y \\ 0 & x \neq y \end{cases}$$

If c(x) > 0:

$$p(x,y) \begin{cases} \frac{q(x,y)}{c(x)} & x \neq y \\ 0 & x = y \end{cases}.$$

(2) Initial distribution:  $\pi = (\pi(x))_{x \in \mathcal{S}}$  is a probability measure.

 $\overline{(3)}$  Family of independent exponetially distributed waiting timesindependent  $\overline{\tau_{x,i}} \sim \exp(c(x)), x \in \mathcal{S}, i \geq 1$ .

If  $c(x) = 0 \to \tau_{x,i} = \infty$ .

The joint distribution of  $Z_n$  are given by

$$\mathbb{P}(Z_0 = x_0, \dots, Z_n = x_n) = \pi(x_0)p(x_0, x_1) \cdot \dots \cdot p(x_{n-1}, x_n)$$

**Remark.** Careful, the waiting times are only independent given the positions of  $Z_n$ 

$$\mathbb{P}(Z_0 = x_0, Z_1 = x_1, \dots, Z_n = x_n, \tau_0 > t_0, \tau_1 > t_1, \dots, \tau_n > t_n) = \pi(x_0) p(x_0, x_1) \cdot \dots \cdot p(x_{n-1}, x_n) e^{-c(x_0)t_0} \cdot \dots \cdot e^{(-c(x_n)t_n)}$$

where  $tau_k$  is the waiting time of the (k-1)th jump.

Denote by 
$$N(t) := \begin{cases} \min(n \ge 0) \mid \tau_0 + \tau_1 + \dots + \tau_n > t & \sum_{k \ge 0} \tau_k > t \\ \infty & otherwise \end{cases}$$

$$N(0) = 0.$$

This should be k+1, otherwise what does  $\tau_{-1}$  mean?

**Definition 1.16.** We define the continuous time Markov chain X(t) by  $X(t) = Z_{N(t)}$  on  $\{N(t) < \infty\}$ .

**Proposition 1.17.** (a)  $p_t^{(n)}(x,y) = \mathbb{P}(X(t) = y, N(t) < n \mid X(0) = x)$ 

- (b)  $p_t^*(x,y) = \mathbb{P}(X(t) = y, N(t) < \infty \mid X(0) = x)$
- (c)  $\sum_{y \in S} p_t^*(x, y) = \mathbb{P}(N(t) < \infty \mid X(0) = x)$

*Proof.* (b) and (c) follow from (a), therefore only (a) is proofed.

(a): By deriving the same iteration as  $p_t^{(+1)}(x,y) = \dots$  For c(x) > 0 Condition on  $\tau_0, Z_0, Z_1$ :

$$\mathbb{P}(X(t) = y, N(t) < n+1 \mid \tau_0 = s, Z_1 = z, \underbrace{Z_0}_{=X(0)} = x) = \begin{cases} \delta_{x,y} & s > t \\ \mathbb{P}(X(t-s) = y, N(t-s) < n \mid X(0) = z) & s < t \end{cases}$$

$$\mathbb{P}(X(t) = y, N(t) < n+1 \mid X(0) = x)$$

$$= \int_{0}^{\infty} \sum_{z \in \mathcal{S}} \mathbb{P}(X(t) = y, N(t) < n+1 \mid \tau_{0} = s, Z_{1} = z, X(0) = x) \cdot \underbrace{\mathbb{P}(\tau_{0} \in ds, Z_{1} = z \mid X(0) = x)}_{\text{density: } p(x,z)e^{-c(x)s}c(x)}$$

$$= \int_{t}^{\infty} ds \delta_{x,y} \underbrace{\sum_{z \in \mathcal{S}} p(x,z) e^{-c(x)s} c(x)}_{-1} + \int_{0}^{t} ds \sum_{z \in \mathcal{S}} \mathbb{P}(X(t-s) = y, N(t) < n \mid X(0) = z) p(x,z) c(x) e^{-c(x)s}$$

$$= \delta_{x,y} e^{-c(x)t} + \int_0^t ds e^{-c(x)} \int_s^{t-s} \sum_{z \in S \setminus \{x\}} \mathbb{P}(X(t-s)) = y, N(t) < n \mid X(0) = z) q(x,z)$$

$$\stackrel{s \mapsto t - s}{=} \delta_{x,y} e^{-c(x)t} + \int_0^t ds e^{-c(x)(t - s)} \sum_{q(x,z)} \mathbb{P}(X(s) = y, N(s) < n \mid X(0) = z)$$

 $\implies$  Same I.C. and recursion as for  $p_t^{(n)}(x,y)$ .

**Theorem 1.18.** The following are equivalent:

(a) The minimal solution  $p_t^*(x,y)$  is stochastic

- (b)  $\mathbb{P}(N(t) < \infty) = 1 \forall t$
- (c)  $\sum_{n\geq 0} \tau_n = \infty$  a.s.
- (d)  $\sum_{n\geq 0} \frac{1}{Z(n)} = \infty$  a.s.

# List of Lectures

- Lecture 01: Introduction, elementary definitions of continuous time markov chains, examples
- Lecture 02: From a markov chain to the infinitesimal description
- Lecture 03:
- Lecture 04: Role of the  $p_t^{(n)}$