# Probability and Stochastic Processes

Master in Actuarial Sciences

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# Continuous time inhomogeneous Markov chains



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### Continuous time homogeneous Markov chains

#### Definition: continuous time inhomogeneous Markov chain

Markov process, with countable state space S, in continuous time, such that there exists a probability function  $p_{ii}(s, t)$  such that

 $p_{ii}(s,t) = P(X(t) = i | X(s) = i), \quad \text{for all } 0 \leq s \leq t.$ 

#### Remark

For each pair (s, t), with  $0 \le s \le t$ , the matrix

 $P(s,t) = \left[ p_{ij}(s,t) \right]_{i,j \in S}$ 

is stochastic, and  $p_{ii}(t, t) = \delta_{ii}$ , *i.e.* P(t, t) = I.

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### **Chapman-Kolmogorov equations**

Chapman-Kolmogorov equations

$$p_{ij}(s,t) = \sum_{k \in S} p_{ik}(s,u) p_{kj}(u,t), \qquad 0 \leqslant s \leqslant u \leqslant t.$$



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### Matrix of transition rates

• We will assume that the probability functions  $p_{ij}(s, t)$  are differentiable.

#### Definition: transition rates

For all  $i, j \in S$ , the transition rate, transition intensity or force of transition from state i to state j is

$$\left. q_{ij}(s) = \left. rac{\partial}{\partial t} p_{ij}(s,t) 
ight|_{t=s}$$

#### Remarks

• Then, for all  $t, h \ge 0$ 

$$P(X(s+h) = j | X(s) = i) = p_{ij}(s, s+h)$$
  
=  $p_{ij}(s, s) + q_{ij}(s)h + o(h)$  as  $h \to 0$   
=  $\delta_{ij} + q_{ij}(s)h + o(h)$  as  $h \to 0$ 

- Hence  $q_{ij}(s)$  is the (instantaneous) transition rate of the process from state i to state j
- The matrix

$$\mathbf{Q}(s) = \left[q_{ij}(s)
ight]_{i,j\in S}$$

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is the transition rate matrix, or generator, of the process.

### The forward differential equations

#### Theorem

$$rac{\partial}{\partial t} p_{ij}(s,t) = \sum_{k \in S} p_{ik}(s,t) q_{kj}(t)$$

under the initial conditions  $p_{ij}(s, s) = \delta_{ij}$ . In matrix form

$$\begin{cases} \frac{\partial}{\partial t} \mathbf{P}(s,t) &= \mathbf{P}(s,t) \mathbf{Q}(t) \\ \mathbf{P}(s,s) &= \mathbf{I} \end{cases}$$

#### Proof

$$p_{ij}(s,t+h) = \sum_{k \in S} p_{ik}(s,t) p_{kj}(t,t+h) = \cdots = \sum_{k \in S} p_{ik}(s,t) [q_{kj}(t)h + o(h)] + p_{ij}(s,t)$$

- As in the time homogeneous case, the FDE decouple so that there is a separate set of simultaneous equations in p<sub>ij</sub>(·, ·), for all j ∈ S, for each initial state i ∈ S.
- Further, again for each *i*, by summing these equations over all *j*, we see that one of the is always redundant.

Image: Image:

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### The backward differential equations



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### **Holding times**

#### Theorem

The holding time in any state  $i \in S$  of a time inhomogeneous Markov process with transition rate matrix Q(s) is exponential distributed with mean

 $\frac{1}{-\int_s^t q_{ii}(u)du}$ 

This is to say

$$p_{\overline{ii}}(s,t) = e^{\int_s^t q_{ii}(u)du}$$

where

$$p_{\overline{ii}}(s,t) = P(X(u) = i$$
for all  $u$  such that  $s \leqslant u \leqslant t | X(s) = i)$ 

is the probability that the process starting in state i ate time s remains in state i ate least until t.



Integrated form of the Kolmogorov backward equations

$$p_{ij}(s,t) = \sum_{k \neq i} \int_{0}^{t-s} p_{ii}(s,s+w) q_{ik}(s+w) p_{kj}(s+w,t) dw, \quad j \neq i$$
$$p_{ii}(s,t) = \sum_{k \neq i} \int_{0}^{t-s} p_{ii}(s,s+w) q_{ik}(s+w) p_{ki}(s+w,t) dw + p_{ii}(s,t)$$

You get the backward equations by dierentiating the integral equations with respect to s.









Write down the integrated form of the backward equation for  $p_{HD}(s, t)$ :

$$p_{HD}(s,t) = \int_0^{t-s} p_{\overline{HH}}(s,s+w)\sigma(s+w)p_{SD}(s+w,t)dw + \int_0^{t-s} p_{\overline{HH}}(s,s+w)\mu(s+w)dw$$

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Integrated form of the Kolmogorov forward equations

$$p_{ij}(s,t) = \sum_{k \neq j} \int_{0}^{t-s} p_{ik}(s,t-w) q_{kj}(t-w) p_{jj}(t-w,t) dw, \quad j \neq i$$

$$p_{ii}(s,t) = \sum_{k \neq j} \int_{0}^{t-s} p_{ik}(s,t-w) q_{ki}(t-w) p_{ji}(t-w,t) dw + p_{ji}(s,t)$$







#### Example

Backward

$$p_{HS}(s,t) = \int_0^{t-s} p_{\overline{HH}}(s,s+w)\sigma(s+w)p_{SS}(s+w,t) dw$$
$$= \int_0^{t-s} e^{-\int_s^{s+w}(\sigma(u)+\mu(u))du}\sigma(s+w)p_{SS}(s+w,t) dw$$

Forward

$$p_{HS}(s,t) = \int_{0}^{t-s} p_{HH}(s,t-w)\sigma(t-w)p_{\overline{SS}}(t-w,t) dw$$
  
=  $\int_{0}^{t-s} p_{HH}(s,t-w)\sigma(t-w)e^{-\int_{t-w}^{t}(\rho(u)+\nu(u))du} dw$ 

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### **Other notations**

#### Other notations

- There are many different notations that can be used in Markov chains
- One that you should be familiar with, beacuse it is commonly used in survival models and life contingencies, is:

$${}_t p_x^{ij} = P(X(x+t) = j | X(x) = i)$$

which denotes the parobability that a life in state *i* at age x is in state *j* at age x + t. In this case  $-\pi$ 

$$_{t}p_{x}^{ii} = P(X(x+\tau) = i, \forall \tau \leq t | X(x) = i)$$

denotes the probability tha a life in state i at age x remains in state i until at least age x + t.

 Under this notation, the transition rates between states i and j at time x are usually denoted by μ<sup>j</sup><sub>x</sub>.



### **Other notations**

#### Example

Consider a illness-death model with 3 states

- $1\,$  Healthy
- 2 Sick
- 3 Death

with transition rates  $\mu_x^{ij}$ , i, j = 1, 2, 3. Show from first principles that the forward differential equation for  ${}_t p_x^{12}$  is

$$\frac{\partial}{\partial}{}_{t}p_{x}^{12} = {}_{t}p_{x}^{11}\mu_{x+t}^{12} - {}_{t}p_{x}^{12}\mu_{x+t}^{21} - {}_{t}p_{x}^{12}\mu_{x+t}^{23}$$

Noticing that

 $_{t+h}p_x^{12} = _tp_x^{11} _hp_{x+t}^{12} + _tp_x^{12} _hp_{x+t}^{22}$  Chapman-Kolmogorov equations

 $_{h}p_{x+t}^{12} = h \mu_{x+t}^{12} + o(h)$  transition rates

 $_h 
ho_{x+t}^{22} = 1 - h \mu_{x+t}^{21} - h \mu_{x+t}^{23} + o(h)$  transition rates

Image: A matrix

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## **Other notations**

### In general

$$\begin{aligned} \frac{\partial}{\partial t} t p_x^{ij} &= \sum_{k \in S} t p_x^{ik} \mu_{x+t}^{kj} \\ &= \sum_{k \neq j} t p_x^{ik} \mu_{x+t}^{kj} + t p_x^{ij} \mu_{x+t}^{ij} \\ &= \sum_{k \neq j} t p_x^{ik} \mu_{x+t}^{kj} + t p_x^{ij} \left( -\sum_{k \neq j} \mu_{x+t}^{jk} \right) \\ &= \sum_{k \neq j} \left( t p_x^{ik} \mu_{x+t}^{kj} - t p_x^{ij} \mu_{x+t}^{jk} \right) \end{aligned}$$



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